IN FOCUS

EVALUATION OF ACTIVE LABOUR MARKET PROGRAMS

Edited by
GÁBOR KÉZDI
INTRODUCTION

GÁBOR KÉZDI

The employment rate has been very low in Hungary over the past 20 years. In 2010, it lagged behind the OECD average by ten percentage points. The lag had already been of the same magnitude in 1994. While most countries saw their place change between these two years in the ranking, Hungary has been a stubborn laggard for the entire period, being fourth-to-last in 1994 and second-to-last in 2010. Figure 1 shows the employment rate statistics in the OECD countries.

Figure 1: Employment rate (employment over population excluding full-time students) among the 15 to 64 year age group in the OECD countries; 1994 and 2010

Many factors may explain the low performance of Hungary, see, for example Commander–Köllő (2008) and Chapter 5 in this In Focus by János Köllő and Ágota Scharle. Active labour market policies are often viewed as potential remedies. However, it is not obvious whether that view is warranted. Active labour market policies may or may not increase aggregate employment, and even if they do, the magnitude of their effect may or may not be substantial. Appropriate evidence is needed to understand more about those effects.

Active labour market policies aim at helping the unemployed find long-term employment. These policies offer services that are supposed to help people...
search for jobs or acquire skills and knowledge that make them more valuable employees. If effective, these programs can increase aggregate employment and thus decrease the lag of Hungarian employment. If not effective, these programs may be harmful by posing an extra burden on taxpayers (even if financed in part from outside Hungary) and using resources and creative energy that could be used for more productive purposes.

Active labour market policies can be classified the following way by the content of their services.
– Job search help (consulting).
– Public works programs (organized typically by government agencies or municipalities).
– Employment subsidy programs (also known as wage subsidy programs, public subsidies to private firms for employing people from the target population) and programs helping self-employment.
– Training programs (in-class or firm-based training for general or specific skills).
– Complex programs (usually small and narrowly targeted programs that include elements from the above four program types).

Countries differ a lot in their use of active labour market policies (see Figure 2). Active labour market policies are relatively small in the Anglo-Saxon countries and they are substantially larger in continental Europe. Post-communist countries are spread out in the rankings but most are in the lower half. Hungary is below the OECD average both in terms of costs (relative to GDP) and participation (relative to labour force), but it is above most of the Anglo-Saxon countries.

Figure 2: Size of the active labour market programs in the OECD countries, 2009 (costs as a percentage of GDP and the number of participants as a percentage of the labour force)


It may sound surprising but we rarely know the effects of active labour market programs on participants, and we know even less about potential side-effects.
Few programs are followed by impact evaluation studies, few of such studies are accessible to the public, and even fewer provide credible results. As we shall see (Chapter 2 in this In Focus), organizational details, incentives or selection of participants are important determinants of the effects of any particular program. As a result, programs of similar type implemented in different countries at different times in different ways may produce very different results. It would therefore be essential to know about the effects of individual programs.

One reason for the shortage of credible analyses is the lack of adequate data. Experimental evaluations are still rare everywhere, and they are all but nonexistent in Hungary. Data that can meet the strict requirements for informative non-experimental evaluations are also rarely available. Another reason is the often low quality of the evaluation methods. Whether due to inadequate data or inadequate expertise, the ultimate reason for the shortage of credible evaluations is on the demand side. If potential users of evaluation studies want to have credible evaluations, they could get them a lot more often than they may think. Many programs can be evaluated by credible experimental methods, and the data and expertise requirements for informative non-experimental evaluations can also be met for many programs. Most of the time it is a question of resources and priorities. Public access by the scientific community, combined with peer reviews is arguably the best way to impose appropriate quality control over the evaluation studies. Ensuring public access and hiring high-quality reviews are also under the control of potential users of the evaluation studies.

Only credible evaluations are valuable if one is genuinely interested in learning the effects of a program. Using a somewhat strong analogy, we can think of the evaluation of program impacts as similar to learning about the effects of medical treatments. If we were to choose from alternative treatments for our sick child, would we want advice that relies on credible evidence (from, say, proper experiments) on the effects and side-effects? Or would we be comfortable with advice based on “expert opinion” without such evidence? And, would we be fine with opinion affected by the doctor’s personal incentives to choose one treatment over the rest?

The effects and side-effects of active labour market programs may not be matters of life and death. But they are important, too. As we shall see (in Chapter 1 below), the methods of program evaluation are not trivially simple but are not extremely complicated either. There is enough expertise to carry out well-designed experiments and informative non-experimental evaluations even in Hungary. Of course, credible evaluations are costly. Setting up experiments takes time and money, and data collection and expert analysis are costly too. However, in most cases, these costs are negligible relative to the total costs of the programs or their potential effects and side effects. In light of this, it is very surprising to hear the all too often repeated excuse of too little money and time available for proper evaluations. Instead, the right question is why it
makes sense to spend vast sums of money and energy on programs with completely unknown effects?

It is a very positive development that credible evaluations are becoming more common in many countries. Experimental evaluations gain ground, and non-experimental evaluations can use better and better data, often from high-quality administrative sources. These trends signal not only positive developments within the evaluation community and improvements in data quality but the increased demand from stakeholders for credible evidence. There are signs for positive developments within Hungary, too. Only one randomized experiment was conducted in Hungary, and that looked at the effects of changes within the benefit system as opposed to active labour market policies (Micklewright–Nagy, 2010, looked at the effect of increased control over unemployment benefit recipients). Large and high quality administrative data are still not accessible in Hungary. Nevertheless, we start to see studies published and reviewed by the scientific community that are based on high quality analyses within the constraints of the data at hand. Most importantly, there are signs that indicate increased demand for credible evaluations by important stakeholders. Sponsoring this volume is a proof in case.

We hope that, by publishing this chapter in the Hungarian labour Market volume of 2012, we can contribute to the positive developments described above. Our goal is to facilitate credible evaluations of Hungarian labour market programs. The ultimate goal of such studies is to make sure that active labour market programs are used in line with their benefits.

* The In Focus consists of seven chapters. Chapter 1 covers the methodology of program evaluation without technical details. Chapter 2 gives a review of the international evidence on the effects of labour market programs, focusing on the most credible evaluations. Chapters 3 to 5 show the results of recent Hungarian evaluation studies. Chapter 6 presents a small complex program and discusses the possibilities of evaluating it. Finally, Chapter 7 reviews the evaluation studies of the unemployment benefit system and the employment subsidy programs in Hungary.

The chapter covers Hungarian evaluations of all major program types, from job search help through public works programs and complex programs. As the results of Chapter 7 point out, evaluations of Hungarian active labour policies are rare even compared to studies on the Hungarian unemployment benefit system. As a result, we could select from a very limited set of studies. Nevertheless, all studies presented in this chapter represent high quality analyses given the data at hand, and all provide important information on the effects of the Hungarian programs.
The effect of a program can be understood as the outcomes following the program compared to the outcomes that would have occurred in the absence of the program. The effect is therefore defined as the difference between actual outcomes (measured after the program) and hypothetical outcomes, also called “counterfactual” outcomes that would have occurred had the program not taken place. Chapter 1, written by Gábor Kézdi, points out that collecting data from participants alone is never enough to measure the effects of a program. Measuring the counterfactual outcomes is equally important, and that measurement is the fundamental challenge of evaluation studies. Statistical evaluations measure counterfactual outcomes with the help of control groups. Outcomes measured in control groups are supposed to represent outcomes that participants would have achieved in the absence of the program. Qualitative evaluations do not make use of control groups, but they too have to make assumptions, even if very implicitly, about counterfactual outcomes. Comparing to counterfactual outcomes is essential because of the definition of the effects of a program.

The most credible and simplest evaluation method is the randomized controlled experiment. The decision of participation is controlled by the evaluation, and the assignment procedure is independent of any relevant characteristic of potential participants, including their potential outcomes. The inevitable advantages of experimental evaluation are often offset by its practical infeasibility. However, recent examples show that experiments can be designed in more situations than we may think. As a result, experimental evaluations of labour market programs, while still rare, are gaining ground.

Of the non-experimental evaluations, regression-discontinuity can lead to credible estimates of the program effects for part of the target group (individuals around the discontinuity point). Under suitable assumptions, matching and regression methods can estimate the program effects for everyone and specific subgroups as well. In reality, however, those assumptions are often unlikely to be met and thus these results are less credible, in general, than either experimental or regression-discontinuity estimates. It is a shared view within the profession that matching and regression methods are better used in combination, and they lead to more credible results the more successful they are at controlling for pre-program labour market histories, program eligibility criteria, and local labour market conditions for control as well as treated individuals.

Chapter 2 reviews credible evaluations to present international evidence on the effect of active labour market programs (authors are Péter Hudomiet and Gábor Kézdi). The evidence suggests that job search programs are often effective (albeit not always), public work programs are very rarely effective (but there are exceptions), and complex and well-targeted programs can be effective even in situations where traditional programs are ineffective. Perhaps the most important conclusion from the literature is that organizational details are more
important determinants of success than the type of program. This highlights the importance of appropriate evaluation of each and every program in order to give feedback for subsequent modifications or termination of the program.

The international evidence shows that even when effective, active labour market programs are not a panacea to aggregate employment problems. The National Supported Work program, a complex employment subsidy program in the U.S. in the 1970’s, increased participants’ employment rate from 30 per cent to 40 per cent. This is a significant increase, but it is very far from making all participants employed. Active labour market programs can improve considerably in Hungary if the institutional background develops further, programs are designed by incorporating international evidence, and the effects of implemented programs are measured by credible evaluations for providing feedback for further improvements. We can expect positive effects from programs under such conditions. On the other hand, even if done in the best ways, active labour market programs cannot be the stand-alone solution to the problem of the low employment rate in Hungary.

Chapter 3, written by Zsombor Cseres-Gergely, evaluates the effect of the modernization of the Public Employment Services in Hungary that took place between 2005 and 2008. The intervention included national elements (e.g., development of the informational infrastructure), but many elements targeted local offices, including quality assurance systems, staff training, office reconstructions and the introduction of a new model of service. The improvements aimed at making local offices help job search of the unemployed. The evaluation study uses panel data on unemployment offices and a difference-in-differences methodology to look at whether employment probabilities increased in participant offices relative to non-participant offices. The results imply a modest increase in employment probabilities, and the effects are most pronounced among the 20 to 50 year old unskilled unemployed with some labour market experience. Simple back-of-the-envelope calculations with the estimates suggest that, as a result of the modernization program, unemployment duration decreased by one to two months (5 to 10 per cent) in this most affected group.

The ambitious study of Judit Csoba and Zita Éva Nagy, in Chapter 4, looks at the effect of three major labour market programs in a single framework. The evaluated programs include, in principle, all training, employment subsidy and public work programs in Hungary between 2009 and 2010. The authors use administrative data on a sample of registered unemployed people, supplemented with their own questionnaire. They compare participants of the programs to a non-participant but eligible unemployed control group, and they use regression methods to control for observed differences. The administrative data does not provide enough information on pre-program labour market histories, and the treated groups differ from each other and from the control group in terms of many important observable characteristics. These problems limit the
credibility of the estimated effects, but the results are nevertheless informative. The longer-term results indicate that, six to twelve months after the end of the program, participants of public work programs were significantly less likely to find jobs than members of the control group, participants of the training program had a slightly higher success rate, and participants of employment subsidy programs were significantly more likely to find jobs.

The survey data used in the study contained additional interesting information. Members of the control group feel less healthy on average, they are more likely to think that they lack the skills demanded on the labour market, and they are less willing to work in occupations that demand long hours or working in public areas. These differences highlight the limits to comparing participants to members of the control group. These limitations are likely to be relevant beyond this particular study, as they suggest that controlling for “hard” characteristics observable in administrative data are unlikely to capture important aspects of self-selection. Another important result from the supplementary data suggests that at least one third of the jobs created by the employment subsidy program would have been created in the absence of the program as well.

In Chapter 5, János Köllő and Ágota Scharle analyse the effect of public works programs between 2003 and 2008. The effects of these programs can show the expected effects of the “Road to jobs” [“Út a munkához”] program introduced in 2009. In essence, the Road to jobs program is a scaled up version of the previous public works programs. Data from the program indicate that targeting was adequate but the take-up was larger than expected due to incentives built in for municipalities to use money in the central budget to finance public work locally.

The analysis uses administrative data and panel regressions at the level of municipalities to uncover the effects of year-to-year changes in participation in public work programs on subsequent changes in long-term unemployment. The results suggest that public work programs between 2003 and 2008 did not lead to a decrease in long-term unemployment. These results are in line with international evidence as well as previous Hungarian studies on the effects of public work programs. Therefore, while its targeting is adequate, the new Road to work program is unlikely to achieve its goals in decreasing long-term unemployment.

Chapter 6, written by Nándor Németh and Gergely Kabai, describes the “Change of destiny, forming of destiny” program, a small complex program targeting long-term unemployed in small villages in Hungary. In contrast to the other chapters in this In focus, Chapter 6 is not based on a statistical evaluation study. Instead, it is a descriptive study with qualitative evaluations. At the same time, the study follows the spirit of the chapter by asking questions relative to counterfactuals, that is, by contrasting outcomes after the program to outcomes that would have happened in the absence of the program.
The program was launched by the Labour Office of Southern Transdanubia (Dél-dunántúli Regionális Munkaügyi Központ), in 40 villages of three counties, with 200 participants. Program participants took part in an eight-month personalized training and are then employed in subsidized agricultural jobs for a year or two. This is a complex program with elements of training, subsidized employment, and consulting with goals of personality development. The results show that participants of subsidized employment stayed employed after the employment subsidies ran out, and participants formed local communities with purpose. The qualitative analysis indicates positive effects of the program, which would be interesting to evaluate with statistical methods as well.

Chapter 7, the last one in this chapter, reviews the evaluation studies of the unemployment benefit system (“passive labour market policies”) and employment subsidy programs in Hungary. The paper, written by Zsombor Cseres-Gergely and Ágota Scharle, lists all evaluation studies and their references, and it classifies them in many dimensions. These dimensions include methodological criteria and thus help the reader judge the credibility of each study separately.
1. METHODS FOR ASSESSING THE IMPACT OF ACTIVE LABOUR MARKET PROGRAMS
GÁBOR KÉZDI

Introduction

Impact assessment studies\(^1\) use statistical methods to assess the impacts of social programs on a few well-defined outcomes. For labour market programs, the most common outcome variables are employment, earnings and unemployment duration. Impact assessment studies are important in their own right, and their results are essential ingredients of further analyses. It is natural to ask whether programs have the impact they promise, what the magnitude of that impact is, and whether they have negative or positive side-effects. Results of impact assessment studies are used in cost-benefit analyses, which convert the estimated impacts into monetary value and contrast that with the costs of the program.

This chapter gives a short overview of the methods of impact assessment. It introduces the methods without many technical details and Greek letters. The discussion is kept short, but it covers the most often used methods, the conditions of their use, and I discuss their advantages and disadvantages.

Most impact assessment studies focus on direct impacts on program participants. These participants are often named as the “treated” group. In some cases, indirect effects may be important as well. Perhaps the most widely considered potential indirect effects are displacement effects, also known as program substitution effects. These can occur, for example, in an employment subsidy program if program participants are hired instead of non-participants, but the number of jobs remains fixed, and, as a result, non-participants are less likely to find jobs as a result of the program. Other indirect effects may include labour market equilibrium effects that operate through wages (a large employment increase due to a program may put downward pressure on wages) or general equilibrium effects that affect other markets as well (if large programs use resources which would otherwise be used elsewhere). Indirect effects are relatively rarely investigated. The larger part of this methodological chapter focuses on the measurement issues of direct effects, and it covers indirect effects in less detail in a separate section.

The effect of a program is defined as the difference between outcomes following the program and outcomes that would have been realized in the absence of the program. The effect to measure is therefore the difference between actual outcomes and so-called counterfactual outcomes. These counterfactuals are not observable by definition. As a result, it is impossible to assess the effects of a program by looking at the participants only. Take the example of a program that wants to help unemployed people find a job. Suppose that we find that 30

---

1 The economics literature uses the “program evaluation” label for such studies. For many people, evaluations may include analyses that do not make a serious attempt at assessing impacts, and, at the same time, some people restrict evaluations to studies that talk about the costs and benefits of the program as well. In this paper I use program evaluation in the economics terminology, synonymous with impact assessment.
per cent of the participants find a job within six months after the end of the program. This number tells us nothing about the impact of the program. It is possible that this 30 per cent would have found a job without the program as well, in which case the effect of the program is zero. It is also possible that, without the program, no participant would have found a job, in which case the effect of the program is a 30 percentage point increase in the job finding rate. In principle, it is also possible that even more people would have found a job without the program, in which case the effect of the program is negative.

Assessing the impact of a program is impossible without the counterfactual outcomes. These counterfactual outcomes are not observable though. Unfortunately, there is no way to know what would have happened to participants in the absence of the program. One can think of this problem as the unfeasibility of a thought experiment. The thought experiment would make people participate in the program, record their outcomes, roll back time, and then make the same people not participate in the program and record their outcomes.

Statistical impact assessment studies measure counterfactual outcomes by looking at the outcomes of a control group. Members of the control group are supposed to substitute for the participants in the unobservable state of the world in which they would be non-participants. Qualitative program evaluations do not examine control groups. Because of the low number of the cases they look at, statistical analysis of a control group would, in any event, be problematic. On the other hand, qualitative studies make assumptions about the counterfactual outcomes, too, if only in implicit ways. It is important to make those assumptions explicit, because the effect of a program can be understood only in comparison with counterfactual outcomes. The need for counterfactual comparisons is not specific to the method of the impact assessment study. It follows from the definition of the impact of a program.

Statistical impact assessment studies can be distinguished along two dimensions. 1. How is the control group selected? 2. What is the method of comparison between the treatment group and the control group? It turns out that the better the control group (i.e. the more credible the assumption that the control group represents the counterfactual outcomes of the treatment group), the simpler the methods needed for comparing the outcomes.

The design closest to the ideal is the randomized experiment (also known as randomized controlled trial). In these experiments a random rule is used to decide who participates in a program and who does not. Randomized experiments used to be relatively rare in program evaluation for practical reasons. At the same time, their results are exceptionally credible and widely accepted. For this reason, as stakeholders have started demanding credible results, randomized experiments have seen a recent increase in impact assessment studies worldwide.

All other methods are labelled together as non-experimental methods. The so-called natural experiments do not use an explicit random rule for deciding
who participates in the program and who does not, but they assume that there is some factor in the determinants of participation that can be viewed as random. The essence of natural experimental evaluations is finding and isolating that factor and comparing the outcomes of the participants to control group members whose participation is decided upon that particular factor. Another method is regression-discontinuity design that makes use of specificities of the assignment rule. Regression-discontinuity can be used if there is a threshold value of some variable (say, age), that leads to a jump in participation (because, say, nobody can participate above the threshold). Matching evaluations and regression analyses compare the outcomes of treated and control individuals that are similar in their observable characteristics. Matching methods explicitly pair people for comparison, while regression methods attempt to do the same in a more implicit fashion. An important subset of the matching and regression analyses is the difference-in-differences (diff-in-diiffs) method that uses information on the outcome variables from before the program.

There are several excellent papers on the methods of statistical program evaluation. Most of these are more technical than this introduction. The most widely known of the papers are Heckman, LaLonde and Smith (1999), Imbens and Wooldridge (2009) and DiNardo and Lee (forthcoming).

Randomized experiments

Randomized experiments, also known as randomized controlled trials, start with a pool of potential participants and split this pool into a treatment group and a control group. The rule that assigns potential participants into the two groups is random. The meaning of randomness is sometimes the subject of rather abstract discussion, but the practical requirement here is simple: the rule should be completely independent of outcomes of the individuals. There are more complicated designs with more groups or certain overlaps, but they are not covered here.

In many randomized experiments, not everybody ends up participating in the program from the group that was selected to be treated. Some of the people that were selected to participate may decide not to participate or drop out during the program. In such cases comparing the outcomes of the original treatment group and the control group does not measure the effect of the treatment itself because part of the original treatment group ended up not receiving the full treatment. Instead of the treatment effect, such comparisons measure the effect of being assigned to the treatment group. This effect is called the “intention to treat effect” in contrast to the “treatment effect” itself. Actually, intention to treat effects are more relevant for program cost-benefit analyses than the treatment effects themselves. Incomplete participation in the treatment in the randomized experiment shows that incomplete participation is a feature of the program itself that should be taken into account in the cost-benefit analyses.
the same time, the effect of the treatment itself is often an interesting question in its own right. Incomplete participation makes the eventual treatment group systematically different from the control group, even if the original treatment group was very similar to the control group because of randomized assignment. As a result, a simple comparison of the eventually treated group to the control group leads to biased estimates of the treatment effect. Fortunately, the treatment effect can be estimated in such cases using the econometric technique of instrumental variables, under certain assumptions.

Randomized experiments are the standard methods of scientific inquiry from physics through medical research to psychology. Besides its sound methodological credibility, a major advantage of the experimental method is its simplicity. Simplicity helps in communicating the results to the stakeholders and the general public and leaves less room for manipulating the results. Their disadvantage is that randomization has to be built into the design of the program itself, which requires close collaboration between researchers and program administrators. Program administrators often resist accommodating the needs of program evaluation in general and randomized selection in particular. However, the needs of credible evaluation are necessary to take into account if one is genuinely interested in the impact of a program. The most successful demonstration programs (programs that are implemented in order to show the impacts for present and future stakeholders) are all evaluated by randomized experiments. There are powerful movements in the research community that advocate randomized program evaluations, such as the J-PAL group (http://www.povertyactionlab.org).

Natural experiments

In principle, natural experiments are similar to randomized experiments, with the important distinction that assignment to treatment group and control group is not the result of a randomized algorithm but some other mechanism. That mechanism qualifies for natural experiment if the researcher assumes that the assignment mechanism is practically random from the viewpoint of the outcome variables. Typically, though, the mechanism that is assumed random is only one of the many factors that influence participation in the program. Simple comparison of the treatment group to the control group is therefore not appropriate. Instead, the researcher has to isolate the effect of this particular mechanism on the outcome variable. The econometric technique used for this isolation is the method of instrumental variables. While instrumental variable estimations based on natural experiments have produced important results in labour economics, their practical applicability has been limited in impact assessment studies.

Regression-discontinuity

Discontinuity means a sudden increase or decrease in a variable at a particular point. In impact assessment studies, the regression-discontinuity design is ap-
plicable if the likelihood of participation shows a sudden increase or decrease at a threshold value of a variable. In the so-called “sharp” case there is a threshold that completely determines participation in the program: everybody participates from one side of the threshold value and nobody participates from the other side. Frequent examples include compulsory programs with age or income thresholds or restrictions on the length of unemployment (the program is restricted to a group defined by the thresholds but everybody in that group participates).

The basic idea is the following. People on the two sides of the threshold are very similar if they are close to the threshold value (e.g., those who just turned 25 are very similar to those who will be 25 within a short time). Their outcomes would also be very similar in the absence of the program. Importantly, this is true even if the underlying variable itself (e.g., age) is strongly related to the outcome variable (e.g., employment). The requirement here is one of “continuity”: age can have a strong effect on employment, but, in the absence of the program, there should be no sudden increase or decrease in the employment probability at the particular threshold value (age 25). As a result, the employment prospects of those who have just turned 25 and those who will be 25 within a few months should be very close in the absence of the program. If, therefore, we observe significant difference in the outcomes at the age threshold, we can safely attribute that to the program. The method can identify the effect of the program for those who are in the immediate neighbourhood of the threshold value. The effect may of course be different for people far away from the threshold value, but the regression-discontinuity design cannot identify the effect for them.

In a sharp regression-discontinuity design, the probability of participation changes from zero to one as one passes the threshold. In more complicated cases the change in the participation probability is smaller. Frequently, the age threshold is a requirement but the participation is not compulsory among the age-eligible people: in this case the jump in the participation probability is from zero to a non-zero number less than one. It may also happen that the threshold is not prohibitive so that people from both sides can participate, but there is a sudden significant change in the probability (the fraction of people who participate). These are called “fuzzy” regression-discontinuity designs. Fuzzy designs require instrumental variable methods to identify the effect of the program. These methods can isolate the effect of the threshold on program participation and then on the outcome. The identified effect is local in one more sense compared to the case of sharp design. The fuzzy regression-discontinuity design allows for measuring the effect for people that are around the threshold and would change their participation status if they crossed the threshold.

Figure 1.1 shows a hypothetical sharp regression-discontinuity design with an age threshold. In this setup nobody participates in the program if above
the threshold and everybody participates if below the threshold. The outcome variable is earnings. The continuous line shows the observable average of earnings as a function of age, while the dashed line shows the counterfactual average that would be observable in the absence of the program. Average earnings and age are positively related in a continuous fashion in the absence of the program. The sudden jump at the age threshold indicates that the effect of the program is positive and significant. We can estimate this effect by comparing the average earnings of those who are just below the age threshold and those who are just above.

Many researchers argue that regression-discontinuity design gives the most credible assessment of program impacts from among the non-experimental methods. The logic of the method is simple and intuitive, but the practical implementation is not without difficulties. The analysis has to balance two opposite problems when deciding whom to compare with whom. On the one hand, the closer the compared individuals are to the threshold value the more credible the comparison. On the other hand, the more we restrict the comparison groups the smaller the number of observations that we can use for the comparison, which decreases the precision of the results. The other caveat is that regression-discontinuity identifies the effects locally as described above.

Despite those caveats, regression-discontinuity is a powerful method to give credible results for at least a subset of the potential program participants. It has become very popular not only in impact assessment studies but in other areas of labour and education economics. The method was first used by Thistlewaite and Campbell (1960), but Angrist and Levy (1999) was perhaps the most influential study that popularized the method. The most important questions of identification and estimation are covered by Hahn, Todd and Van...
der Klauuw (2001), and the most recent methodological surveys are Imbens and Lemieux (2008) and Lee and Lemieux (2010).

Matching and regressions

Matching and regression are the most widely used non-experimental methods in impact assessment studies. They attempt at handling potential biases due to non-random program participation by “controlling” for “confounding” factors. Controlling means restricting the treated versus control comparisons to individuals that are the same in terms of those factors. Confounding factors are variables that are thought to affect program participation and may be related to program outcomes in their own right, too. Comparing the outcomes of treatment and control individuals that are different in those variables would not identify the effect of the program because the differences in outcomes may be due to their differences in the confounding factors as well as the effect of the program. For example, if more motivated people are more likely to participate in the program, but more motivated people would be more likely to find jobs in the absence of the program, then comparing the job finding rates of program participants to non-participants cannot tell us the effect of the program. (More precisely, the comparison gives an “upward biased estimate” of the effect of the program, meaning that the true effect is smaller than the estimated effect.) The example highlights the major problem with these methods: Motivation is one of those potential confounding variables that are hard to measure, but it is necessary to control for all confounding variables, including hard-to-measure ones like motivation, in order to avoid bias.

In order to focus on the more technical questions, assume for now that all potential confounders are measured appropriately and can be used as control variables. Also assume that within groups of people that share the same control variables we find both participants and non-participants. In this case, within those groups, program participation can be considered random for the purpose of impact assessment analysis. Within those groups, therefore, simple treatment-control comparisons identify the effect of the program just as in the case of randomized experiments. The assumption that all potential confounding factors are observed and controlled for is called “unconfoundedness” or “ignorable treatment”. The assumption that we can find both participants and non-participants within groups of people that are identical in terms of the confounders is called the “overlap” or the “common support” assumption.

Matching methods carry out the comparisons in a very intuitive way. They take one or more program participants and match them with one or more non-participants that have the same control variables. The treatment versus control comparison then is carried out within these matched pairs or groups. If the overlap assumption is satisfied, the comparison is always feasible. If the unconfoundedness assumption is satisfied, the outcomes of the matched con-
trol person or persons can be used as the counterfactual outcomes for the treated person or persons. There are various matching methods that differ in the way they use the control variables and the algorithm they use to find matched pairs or groups.

There are two ways to use the control variables in matching. The first one is “exact” matching, simple in principle but rarely used in practice. It looks for matches that have the exact same control variables. Exact matching is feasible if we have relatively few variables that are categorical in nature, and we have a large enough sample. One can also make categories out of continuous variables and do exact matching on those categories. However, exact matching suffers from the “curse of dimensionality”, the problem that it is extremely difficult to find exact matches when the number of control variables is large (high-dimensional). That is a serious problem because the number of potential confounding variables is large in most applications, and thus many variables have to be controlled for in order to satisfy the unconfoundedness assumption.

Matching on the “propensity score” offers a solution to the curse of dimensionality. The propensity score, introduced by Rosenbaum and Rubin (1983), (1984), is the probability of program participation as a function of all the control variables, estimated for each individual (participants and non-participants). The method uses the one-dimensional propensity score in place of the many control variables. Obviously, two people with the same control variables have the same propensity score. But people can have the same propensity score with different combinations of the control variables. Less obvious but also true is that, from the point of view of program evaluation, comparing two people with the same propensity score is practically the same as comparing two people with exactly the same control variables even if they actually differ in those variables. As a result, matching on the propensity score yields the same results as exact matching, at least in theory. The reason is that the propensity score contains all the information in the control variables that are relevant for program participation. Any remaining differences in the control variables are irrelevant from the point of view of program participation and thus do not cause any problems for estimating the effect of the program.

Matching on the propensity score solves the curse of dimensionality because it reduces the potentially high-dimensional set of confounding variables to a single-dimensional variable that is also bounded (it is a probability and is thus between zero and one). When implemented, the propensity score is the predicted left hand-side variable in a probability model (typically probit or logit). Perhaps more intuitively, it is a combination of the control variables, like some kind of a weighted average. In a sense, variables that are more important in predicting program participation receive larger weights, and those that are less important receive smaller weights. Higher values on some variable can compensate for lower values on other variables and yield the same propensity score.
The propensity score is a continuous variable in principle. Two people are unlikely to have the exact same propensity score if the dimension of underlying control variables is high. Matching on the propensity score therefore means finding matches with a similar but not necessarily the same score. Various methods are used to find matches with a similar propensity score.

The most intuitive method is nearest neighbour matching. It takes treated individuals one by one and pairs them with the control individual with the closest value propensity score. This is a one-to-one matching. Typically, matching is done "with replacement", which means that control individuals can be used to match with multiple treated individuals (controls are replaced to the pool of potential matches after they are matched to the treated individual). It is also common to specify a maximum distance and leave treated individuals unmatched if no match is found within that interval.

In many cases, the control group is many times larger than the treatment group. In such cases nearest neighbour matching is inefficient because it leaves a lot of potentially useful control individuals unused. Many-to-one methods may be more efficient in such cases. One of them specifies an interval around the propensity score of each treated individual and matched all control individuals that fall within that interval. A symmetric procedure, applicable if the treatment group is larger than the control group, matches treated individuals within an interval around each control. The fourth group of methods yields many-to-many matches. One such method specifies intervals of the propensity score and compares average outcomes of all treated individuals in the interval to the average outcome of all control individuals within the interval.

In ideal circumstances and infinitely large samples, the different matching methods should yield very similar results. In small samples and if the unconfoundedness assumption is invalid, they can be very different. Choosing among the matching methods is not easy therefore if they yield different results. In such cases there is always a danger of manipulation, conscious or unconscious: Researchers may favour results that confirm their prior expectations ("nice" or "meaningful" results). The problem is obvious: it is very possible that the unexpected estimates are the ones that are closer to the truth. The problem is aggravated by the many modelling choices some matching methods allow for (size of the intervals, potential weighting within intervals etc.). In order to minimize the role for manipulation, the profession treats matching results as credible only if results from many different matching methods are presented and they all yield similar results.

Regression methods are alternatives to matching. They use some additional technical assumptions, but they are easier to estimate and are somewhat less subject to potential manipulation. The unconfoundedness assumption is needed for the regression, too. In regression models, in contrast to matching models, we have to specify a functional relationship between the expected value of
the outcome variable and the confounder variables. The usual choice is a linear relationship. That is less restrictive than generally thought as it allows for nonlinear transformations of the control variables (higher-order polynomials, interactions, splines etc.). However, the choice of functional form introduces flexibility that can lead to manipulation of the results similarly to the choice of the details of the matching methods.

The other difference is that regression models do not explicitly require the overlap assumption (the assumption that for any combination of confounding factors that are observed for treated individuals we can find control individuals with the same confounding factors). Regressions can be estimated without technical problems even if there is a range of the value of the right hand-side variables that correspond to treated or control individuals only. Matching would not be able to compare such individuals to anyone and thus they would be left out. But they may influence the regression estimates. That is not a problem in principle if the functional form assumption behind the regression is correct. However, since the functional form is an assumption that is hard to test, the influence of those observations may very well bias the results.

It is therefore advised to restrict the regression analysis to treatment and control observations that share the same value range of the control variables. With such preparations and using flexible functional forms, regression estimates for the effect of the program yield similar results to matching estimates in general. Matching models are surveyed by Imbens (2004), Caliendo and Kopeinig (2008) and Imbens and Wooldridge (2009), the latter covering regression models in great detail as well.

Matching and regression methods have been the choice of the vast majority of ex-post program evaluations. Randomized experiments require close collaboration with the program administrators, and regression-discontinuity design is applicable under rare and lucky situations. Matching and regression methods are feasible in many more situations. Unfortunately, however, matching and regression methods do not necessarily yield credible estimates for the effect of programs. The profession is therefore sceptical of the results of such methods (LaLonde, 1986, contributed significantly to that scepticism). In order to circumvent scepticism, evaluations that use regression or matching methods have to provide many robustness checks and extra evidence to support their analysis.

In principle, the condition for credible results from matching or regression analyses is simple: we have to control for all confounding variables that affect program participation and program outcomes at the same time. That is, of course, easier said than done. But decades of experience in evaluating active labour market programs produced some general guidelines for the kinds of information that always needs to be controlled for (see, for example, Heckman, LaLonde and Smith, 1999). We can summarize those guidelines in three rules. 1. Control and treated individuals should belong to the same labour market. 2. Members of the control
group have to satisfy all the eligibility criteria of the program, in a similar way to participants. 3. The analysis should control for detailed individual labour market histories. In the end, evaluations using matching or regression methods have to say something about why it is that some individuals participate in the programs while others not even though they share the same characteristics. This last requirement is often referred to as the need for “exogenous variation” in participation.

The importance of labour market history and difference-in-differences analysis

Controlling for labour market histories is important because they contain the most important information about individuals’ productivity and labour supply characteristics, which are in turn the most important determinants of the outcome variables. These labour market histories should contain past values of the outcome variable (employment, earnings, employment durations), as far back in the past as possible.

The simplest method to control for pre-program labour market histories is the difference-in-differences (diff-in-diffs or DID) method. In its simplest form, diff-in-diffs means measuring the change in the outcome variable from before the program to after the program for each individual (the “diffs”), and then comparing the average of these changes in the treatment group to the average of the changes in the control group (the “diff” in the diffs). This simple method meets the unconfoundedness assumption if treatment and control individuals who have the same previous value of the outcome variable would end up with the same outcome variable later, on average, in the absence of the program. If this assumption is true, evidence for significant difference between treated and control outcomes after the program are evidence for the effect of the program.

It is of course unlikely that the assumption behind the simplest diff-in-diffs model is satisfied. Longer labour market histories and other potential confounder variables may also be necessary to control for. If we think that we have all those confounding variables observed, we can easily embed the diff-in-diffs logic in a matching or a regression context. All that is needed is to include past value (or, even better, a series of the past values) of the outcome variable among the control variables. Matching on the propensity score uses those control variables in the estimation of the propensity score, while regressions use them directly. Heckman et al (1997) provide a convincing argument for the need of combining the diff-in-diffs approach with matching.

If the other conditions in the guidelines are not met (same labour market and every individual meeting the eligibility criteria), the diff-in-diffs logic can increase extra bias in the analysis if it is based on short labour market history. The reason is a phenomenon that is called Ashenfelter’s dip after its discoverer (Ashenfelter, 1978). The typical active labour market program aims at helping
people with labour market disadvantages. Being unemployed during a reference period before the program is usually required in order to be eligible for the program. As a result of the eligibility rule, all participants were unemployed during the reference period. However, if we do not restrict members of the control group to those that were unemployed during the same reference period, diff-in-diffs can lead to a severe overestimation of the employment effect of a program.

In order to see this, assume that members of the control group are from the same target group of disadvantaged people as members of the treatment group, but control group members do not satisfy the program eligibility criteria. Because they are from a disadvantaged group, many of them were unemployed during the reference period, but some were probably employed (say, 20 per cent). If nothing significant happens to the labour market, the fraction of employed people would remain the same among them (20 per cent). The participants are from the same target group, but none of them were employed during the reference period. If the program had no effect, the employment rate among them would converge to the employment rate of the entire group in the long run (20 per cent). If the diff-in-diffs comparison compares long-run outcomes to before-program outcomes measured in the reference period, it would show a positive increase (from 0 per cent to 20 per cent) in the treatment group, compared to a zero increase in the employment rate in the control group. A simple diff-in-diffs comparison would attribute this increase in the employment probability of 20 per cent to the effect of the program, even though the program had no effect.

In the jargon of econometrics, the problem is that the eligibility criteria make the group of participants a selected sample of the target group. The target group has a low but non-zero employment probability. The selected sample has zero employment probability at baseline by construction. Figures 1.2 and 1.3 show examples for Ashenfelter’s dip.

Figure 1.2: Ashenfelter’s dip for the treatment group in a standard training program (reproduced from Ashenfelter, 1978)
The first of the two figures, reproduced from Ashenfelter (1978), shows yearly nominal earnings of participants of a standard training program. The program started in 1964 and participants had to be unemployed at the starting date. The program ended in 1965. There is a significant increase in earnings from that point. The dip is in there right before the program, with a break in the trend and even a slight decrease in nominal earnings.

The second example is more striking. It shows the dip for an experimental control group. The outcome variable is again nominal earnings, monthly earnings in this case. Members of this experimental control group satisfied all the eligibility criteria for the program, that is why they show the dip. Month zero on the figure denotes the start date of the program, and the eligibility criteria included unemployment in the previous six months.

Ashenfelter’s dip is a rather general phenomenon for active labour market programs. It is problematic for program evaluation if it affects the treatment group but not the control group. There are two kinds of “remedies” to this problem. One trivial way out is making sure that members in the control group are subject to the same eligibility criteria as members of the treatment group. The other one is avoiding diff-in-diffs-type comparisons to recent outcome variables and, instead, taking significantly longer labour market histories into account.

**Indirect effects**

The question analysed so far was the effect of the program on its participants. The social benefits and costs of programs include the effects on non-participants too. Take the example of an active labour market program that increases the employment prospects of its participants. From a social point of view, it is important whether that increase is due to creation of new jobs or filling a fixed number of jobs with program participants instead of non-participants.
Indirect effects are the effects of a program on non-participants. Potential indirect effects include so-called displacement or substitution effects, when program participants take jobs that would otherwise be filled with other people, thereby displacing them. Partial equilibrium effects occur when the direct effect of the program leads to a significant increase of effective labour supply in a labour market. Such an increased labour supply can affect wages (a large increase in employment may put downward pressure on wages), which may feed back to the labour supply of other people, etc., until a new equilibrium is reached. General equilibrium effects occur when the program affects other markets as well, for example, by using resources that would be used elsewhere otherwise.

Indirect effects cannot be measured by simply comparing outcomes of participants and non-participants even in randomized experiments. Moreover, indirect effects may distort the proper measurement of the direct effects. Take the example of an employment subsidy program that increases the probability of employment of participants by ten percentage points. Assume, however, the worst: All of that increase was achieved by displacing other people without any new job creation. The net employment effect is therefore zero per cent in the society. Take an experimental control group that would be ideal for any impact assessment study. It may be, however, that the control group includes people who were displaced by the program participants, i.e. who would have found a job in the absence of the program but did not find one because of it. If the program or the control group is large enough, we can expect many such people to be included in the control group. The effect of the program on the average employment probability in the control group is, therefore negative. If we compare the employment of the treatment group to the employment of the control group, the estimated effect of the program would be more than 10 per cent.

This latter problem is not easy to handle. If the treatment and control groups are small relative to the size of the labour market, the bias is likely to be insignificant. Ignoring potential indirect effects when estimating the direct effects is not a grave mistake in such cases, and most studies do just that. If the program is very large, though one has to deal with the problem, and that’s not easy. One solution is comparing treated and control people across labour markets when estimating the direct effect in order to avoid the bias outlined in the previous paragraph. But that may introduce other problems, as it fails to satisfy the requirement for comparing people within the same labour market in the case of non-experimental evaluations. One strategy that circumvents all these problems would randomize treatment not across individuals but across labour markets. In such a design, the program is implemented in randomly selected local labour markets, and outcomes from control labour markets are used to evaluate the net effects of the program. Blundell et al. (2004) use a similar, although non-experimental design to evaluate the direct and indirect effects of a comprehensive job search program in the U.K. Their evidence clearly shows the lack of indirect effects.
That finding is not uncommon in the literature: however worrisome indirect effects may be in principle, they may be less of a problem in practice. But that is not always the case. Employment subsidy programs are sometimes prone to produce displacement effects. Large programs can affect equilibrium wages, which may create partial equilibrium effects.

There are programs that, by their sheer magnitude, are very likely to exert indirect effects on other markets, creating general equilibrium consequences. Unfortunately, statistical methods are not useful to measure such effects. The thought experiment to measure would require recording outcomes on various markets in an economy in the absence of the program and compare them to outcomes after the program. This thought experiment cannot be made operational in any way (except for fundamentally problematic cross-country comparisons). The fact that they are impossible to measure does not mean that one should not think about such effects when the programs are very large. Albrecht, van der Berg and Vroman (2004) analyse the equilibrium effects of an ambitious Swedish training program using a calibrated macro-model, and they present qualitative evidence on effects on other markets (the demand for teachers in the program is likely to have led to a decrease in high quality schoolteachers in public education).

Conclusion

This chapter gave a brief and not very technical introduction to the methods of impact assessment studies, with a special focus on the evaluation of active labour market programs. Much of the chapter focused on direct effects on program participants and it covered indirect effects briefly in a separate section. Besides introducing the methods, the chapter highlighted that not all methods are equally credible.

Randomized controlled experiments are the most credible for assessing the impact of a social program. They are also the simplest to analyse, leaving less room for ex post manipulation than other methods. In the context of evaluating social programs, randomized experiments have been somewhat rare, in part because of practical problems. However, those problems can be more often overcome than many researchers and program administrators think. Randomized controlled experiments have been applied in the most successful demonstration studies, and they are becoming more and more common for testing new programs and assessing older programs worldwide.

Of the non-experimental methods, regression-discontinuity design offers the most credible identification when it is applicable. An important disadvantage of regression-discontinuity is that it identifies the effect for a subset of the participants (those who are around the threshold). But, for that subset, the identification is credible.

In principle, matching and regression methods are more flexible non-experimental methods, and they can identify the effects for many groups of par-
participants and non-participants. However, their application rests on the unconfoundedness assumption, which is fundamentally untestable. Of the two, matching has the advantage of being free of any functional form assumption. On the other hand, they often require large samples, and they offer many modelling choices that may lead to publication biases. Regression models are simpler to estimate, but they are not free from modelling choices either, and they require functional form assumptions that are hard to test. An important advantage to matching models relative to regressions is that they make the need for comparable individuals explicit. When evaluating active labour market programs with results that need to make the unconfoundedness assumption, guidelines help for informing as to which factors one should control for. Economists also converged to the view that it makes a lot of sense to combine matching and regression methods, especially if pre-program outcomes are also taken into account as in diff-in-diffs models.

As it should be clear for the reader at this point, the methodological requirements for statistical impact evaluation studies are rather strict. One often hears opinions stating that statistical studies that want to meet these strict methodological criteria are ones from the many possible approaches, and they may be substituted by more qualitative analyses that could produce equally valuable results.

Such statements confuse two arguments, one of which is true and one, as I will argue, is false and dangerous. Statistical impact assessment studies that want to meet the strict methodological criteria cannot uncover all the evidence that may be useful in understanding the impact of a program. Qualitative evidence is very often needed to complement statistical analyses. Quite often, statistical evaluations cannot be carried out in a sound way. In those cases some qualitative evidence is usually more informative than nothing.

Not all methods are created equal, however. Strict scientific criteria are necessary to differentiate between sound analysis and opinion. Without methodological criteria, how could we tell which result is credible and which is not? By the authority of the analyst? Or whether the results match the prior opinion of administrators or whoever commissioned the study? The essence of scientific inquiry is to set up rules that need to be met in order for results to be credible. The fact that most program evaluation studies cannot meet all the rules does not mean that they should not aspire to do that.

It is important to know the effects of social programs that spend taxpayers’ money on people in need for help. The methods outlined above require statistical training, but they are not extremely complicated. While the most credible evaluations require substantial amounts of money and organizational input, these resources are negligible in comparison to the budget of the programs and their potential social impact. There is little excuse not to aim for sound impact assessment studies.
2. INTERNATIONAL EVIDENCE ON THE IMPACT OF ACTIVE LABOUR MARKET PROGRAMS
PÉTER HUDOMIET & GÁBOR KÉZDI

Introduction

This chapter provides an international overview of the impact of labour market programs. We emphasize the results of evaluation studies that use credible identification strategies. We show examples of the four program types (training, wage subsidy, employment services and public works) and we compare their effectiveness to each other.

First, we analyse three complex and targeted programs, the Job Corps and the National Supported Work programs from the United States and the New Deal for Young People from Britain. Then we describe the large scale national programs of Sweden, Denmark and Switzerland. We present evidence from training programs in a separate section. We describe the labour market programs of post-socialist countries in the last section.

This chapter builds on a previous paper of ours (Hudomiet and Kézdi, 2008) published in the online journal “Kormányzás, Közpénzügyek, Szabályozás” [Governance, Public Finance and Regulation] in Hungarian. In that paper, we overview more evaluation studies and more programs, and we describe them in more detail. The conclusions of that paper are, of course, the same.

Well-designed complex programs

First, we analyse three programs that combine multiple program types and add additional elements to them, such as stipends, counselling, etc. These programs are typically well organized, small and relatively expensive. All three programs are designed in a way that facilitates their evaluation. Partly for this reason and partly because of the quality of the evaluation studies, we have a good understanding of their impacts.

Job Corps (United States)

The Job Corps program was launched in 1964, and it continues to this day. Its target group is 16–24 year old young people, typically high school dropouts, who are unemployed or employed at low wages. The Job Corps is a six month intensive training program. One part of the curriculum focuses on general skills, while the other part is flexible. Counselling and placement assistance are essential parts of the program. These build on the established network of Job Corps. Participation in the program is free of charge, and the majority of the participants live in dormitories where they are provided with additional benefits such as meals and sports facilities. Practically, participants in the dormitories are also under 24 hours of supervision.
The Job Corps is a federally initiated and federally financed program. The first extensive studies in the eighties found the program largely successful (LaLonde, 1995). The success led to a large increase of the budget that turned Job Corps into a rather large scale program. It covers 60,000 young Americans per year, with an overall budget of about $1.5 billion.

More recent evaluations use data from a randomized controlled experiment called the National Job Corps Study. The data collection started in the mid-nineties, and the first results were published between 2001 and 2003. The results created a lively discussion both among academics (mostly economists) and policy makers. The data covers 9000 participants and a randomized control group of 6000. Individuals were followed for four years following the completion of the program. Attrition from the survey is not negligible, which, as we shall see, may distort the results.

Using the survey data, Gritz and Johnson (2001) found that program participants’ weekly earnings were $20–$25 higher four years after the program than earnings of the control group. They found that participants of vocational training earned $40–$50 more in a week than the control group, while those who finished high-school during the program earned $60–$70 more than the control group. They also found that earnings of other groups did not change significantly due to the program. Using the same data, McConnell and Glazerman (2001) made a cost-benefit analysis. The benefit part included increased earnings of the participants, decreased receipt of transfers and other social expenditures and decreasing involvement in crime. The preferred estimate of the study suggests that the social benefit of the program exceeded its cost, but a large part of the benefit was non-monetary. The analysis is based on the assumption that the earnings premium of participants would never decline. In a specification where the authors assume that the earning premium would drop to zero after four years, the estimated rate of return of the program is zero.

Schochet, Burghardt and McConnell (2006) show that the early evaluations based on the National Job Corps Study overestimated the benefits of the program. Instead of survey data, they used more reliable administrative data from the Social Security Administration. This data also covers a longer time horizon and it does not suffer from survey attrition. The results show that the earnings premium of program participants was much smaller than indicated by the previous studies, in large part because the premium strongly declined after four years. The preferred estimates of the study indicate a net positive effect on participants but a net negative effect on society. They also indicate that the program was most beneficial for the older 20–24 year old group where the net social benefit was positive, too.

Important reasons for the initial success of the Job Corps program may have been its small scale and careful targeting. It makes perfect political sense that, after the early successes, the program was enlarged to offer opportunities to
more and more people. It seems, however, that as the program grew in size, targeting became less successful and the program lost efficiency. In this light, it is remarkable that the program is still socially beneficial for a subset of the target group.

National Supported Work (United States)

The National Supported Work (NSW) program is perhaps the most widely analysed among the active labour market programs. The NSW was designed for demonstrative purposes. The goal of the program was to equip hard-to-employ people to obtain and hold normal, unsubsidized jobs. The target groups were AFDC recipient women (poor single mothers), former drug users, ex-inmates, high school dropouts and the very-long term unemployed. Joblessness for a certain amount of time was a criterion in all cases.

The NSW was an employment subsidy program with additional elements. Participants were guaranteed a job for 9–18 months with a private employer. During this time the program covered all labour costs. The program followed the individuals throughout their participation in the program and held weekly group meetings with them. Initial wages paid to participants were lower than the market rate, but they rose relatively rapidly. The job was guaranteed during the program, but once the program was over, participants were left on their own to keep their jobs, now as regular employment, or find a new job. Women were typically employed in unskilled clerical jobs in services while men typically worked in construction.

The demonstration feature of the program meant that the criteria for credible evaluation were built into the design of the program. Applicants were randomly selected into a treated group and into a control group, and detailed information was collected about them.

The impact of the program was a significant increase in employment prospects. LaLonde (1986) showed that average earnings of treated women exceeded average earnings of the control group by $850 a year after the program, and the difference prevailed for subsequent years. The mean effect for men was similar, $880, but higher heterogeneity of wages made the point estimate less reliable. The differences between participants and the control group were mostly due to differences in the employment rates (non-employed individuals were entered with zero earnings into the figures above). Figure 2.1 shows the evolution of employment rates on a monthly basis.

Employment rates were the same (and very low) before the program. 90 per cent of the treatment group started employment with the program. Their employment rate declined throughout the program, and it reached 65 per cent at the end of the subsidized period. Once the program was over, employment of the treatment group fell to 40 per cent. In the 24th month, that is, 12 months after the program ended, the employment rate of the treatment group was rough-
ly 10 percentage points higher than the employment rate of the control group. The medium-run effect of the program, thus, was about a 10 percentage point increase (almost a third in relative terms) in the probability of employment.

**Figure 2.1:** Employment rates among treated and control women in the NSW before the experiment (months -12–0), during the experiment (months 0–12) and after the experiment (months 12–26)

Source: Replication of Figure 1 of Ham and LaLonde (1996) p. 182.

*Figure 2.1* points at three other issues that are interesting. First, the employment rate was zero only at the beginning of the program and a couple of months earlier. It was positive before and after this period even in the control group. This is a nice illustration of Ashenfelter’s dip that has already been introduced in the first chapter. In order to be eligible for the program, individuals had to be out of work for a couple of months. The same was not required for the preceding time periods, and some people from the target group were indeed then employed. It is thus natural that some people in the control group, after realizing that they would not participate in subsidized employment, find regular non-subsidized jobs. Their non-zero employment rate is analogous to the non-zero employment rate long before the program. Zero employment at the beginning of the program creates a “dip”. By this argument, the dip should be symmetric in the experimental control group, and in many cases it is indeed symmetric. That is obviously not the case here.

The second remarkable feature of the figure is the strong asymmetry in Ashenfelter’s dip. The employment rate of the control group jumps to significantly higher than its pre-program level: to 15 per cent four months following the start of the program and 20 per cent in the 12th month at the termination of the program, compared to the 5 per cent level 12 months before the start of the program. Many factors may explain this significant increase. Most likely among them is a general improvement in economic conditions. That can also
explain the fact that employment in the control group kept increasing after the end of the program. Due to this trend in economic conditions the employment prospects of the target group would have improved significantly even in the absence of the program. Changes in the economic conditions do not invalidate the analysis here because the target and the control groups were affected by the same conditions. On the other hand, such potential changes highlight the need for a properly chosen control group.

The third interesting feature of the figure is a drop in the employment rate of the treatment group after the end of the program, a further decrease and then a moderate increase. The reason for the drop is that many program participants were not offered the option of staying with the employer they were assigned to, and they could not immediately find alternative jobs for themselves. Some of them, though, eventually managed to return to employment.

The results of the NSW program evaluation are quite optimistic. The program positively affected the employment prospects of the participants, especially for AFDC recipient women. According to the estimates of Ham and LaLonde (1996), the program increased the duration of employment spells, indicating that the increased employment prospects were due to increased skills that could lead to more stable jobs for many participants. The results also highlight what a successful program can do. The estimated impact is a 10 percentage point increase of a baseline 30 per cent employment probability. This is a significant increase but far from what full employment would require.

New Deal for Young People (United Kingdom)

The United Kingdom has operated active labour market programs since the 1980s to reduce the unemployment rate of the young. The New Deal for Young People program was introduced in 1998 in order to help the disadvantaged youth find regular employment. It was built on the experience of the previous British programs as well as the Job Corps. The New Deal for Young People is a complex program. Participation is mandatory for every 18–24 year old who has been unemployed for six months (noncompliance leads to the withdrawal of other social transfers), but certain groups can join the program even earlier.

The program consists of three phases. The first phase is called the Gateway period. During its four months, everyone is linked to a personal consultant who tries to provide help and incentives to find a job. People can participate in short term training as well, in subjects such as computer usage. Those who fail to find a job during this period enter the second phase, in which they are offered four options: subsidized employment in the private sector, full time training, subsidized employment in the non-profit sector, and public work in environmental projects. In certain cases people can also stay in the first phase and continue searching for unsubsidized jobs. The second phase typically lasts.
for six months, after which people enter the third phase. The third phase is similar to the first one, with job search help.

Because participation in the program is mandatory, it is not easy to find an appropriate control group for evaluation purposes. However, certain features of the program can help identify its impact. The first feature is the age restriction. The target group is the 18–24 year old population, and those who are 25 years old or older are excluded. It seems a reasonable assumption that the 24 and the 25 year old population are more or less similar, and a regression discontinuity design can exploit this characteristic of the program. The second feature is that while the program started in the country in April, 1998, it was launched three months earlier in 12 locations. With the help of some assumptions and appropriate matching methods, this three-month period can be used to estimate the effect of the early launches.

De Giorgi (2005) used the regression discontinuity design in his analysis with the threshold of age 25. It is a sharp design as participation in the program is mandatory for those below the age threshold and satisfying the eligibility criteria, while those who would be eligible otherwise but are 25 already cannot participate. The sharp feature of the regression-discontinuity lends itself to a quite simple and precise statistical method to analyse the local effect of the program on the 24 year old population. De Giorgi (2005) found that the program raised the employment rate by 6–7 percentage points, which is a quite remarkable increase for a large program. However, regression discontinuity can identify the effect on the 24 year old but not on other age groups. Moreover, the estimates are biased if there is a displacement effect of the program, i.e., if program participants found jobs at the expense of the 25 year-old population. If they are significant, displacement effects make estimated effects larger than the true situation because the employment chances of the 24 year old are compared to the decreased employment rate of the 25 year old instead of the appropriate counterfactual that would occur in the absence of the program. As we shall see later, there is no evidence that the New Deal for Young People program would have any displacement effects.

Blundell et al. (2004) and Van Reenen (2001) analysed both the direct and the indirect effects of the Gateway phase of the program. Both papers used the pilot program in which 12 regions started the program earlier than the rest of the U.K. Comparing early starter regions to appropriately chosen control regions can identify both the direct and the indirect effects of the program. These papers found that the Gateway phase lead to an increase in the employment rate in the target group by 4–5 percentage points. The indirect effect of the program, at least as measured for in 1998, was negligible. This finding reinforces the results of De Giorgi (2005).

In the second phase of the New Deal for Young People program, participants could choose from four options. Dorsett (2004) used administrative data and
propensity score matching to compare the effectiveness of these four different options. The study aimed at estimating the relative effect of the four New Deal options, i.e. how people who chose a particular option would have done had they chosen another option. He found that the private sector wage subsidy programs worked considerably better compared to the other three options that were otherwise quite similar to each other.

Beyond the quantitative evidence, the program seems successful and enjoyable based on the opinion surveys. The most successful part seems to be the first Gateway phase, which is also significantly cheaper than the second phase. In the second phase, the private sector wage subsidies seem to work the best, although due to the limited number of offers only a smaller fraction of the participants could choose this option.

Large national programs

While the active labour market programs in the United States and in the United Kingdom focus on specific problems or specific age groups, countries in continental Europe usually have large scale national programs. They contain a big set of program types and participation is relatively open. These national programs are not complex and the target groups typically receive only one type of treatment at any one time. These programs are usually not targeted and the treatment is not individual specific but general.

Sweden

The Swedish national programs are perhaps the best documented, this is our reason to describe them in detail. We will briefly talk about the programs in Denmark and in Switzerland afterwards.

By the end of the 1990s, active labour market programs of Sweden were among the largest in the world. In 1997 when the unemployment rate exceeded 10 per cent, 4.5 per cent of the population participated in various labour market programs, with monetary costs of 3 per cent of the GDP (Sianesi, 2002a). Since then these numbers have declined, but Sweden still spends more on these programs than most other European countries. Sweden provides a textbook example of the large scale national programs.

The registered unemployed are automatically and continuously assisted in job search. Besides giving information about vacancies at local employers, they are also provided with services to help job search and boost motivation. One feature of the Swedish system is that these services are not even considered to be labour market programs since they are part of the unemployment benefit system. As a consequence, all other labour market programs are compared to this reference program, unlike in other countries where the counterfactual state of a program does not involve official help in job search. The unemployed are also informed about training programs, employment subsidies and public
work programs. Another important feature of the Swedish system is that participation in these programs is monetarily incentivized in a way that makes them more beneficial than regular unemployment.

Larrson (2000) studied a typical Swedish training program for the 18–24 year old group. The program was very large with approximately 200,000 participants. The author argued that many people participated in the program because of the monetary benefits only, and this feature had negative effects on the success of the program. The evaluation method was a propensity score matching method with a large set of variables. Her results suggested that the training program actually decreased both the employment rate and the earnings of the training group in the short run (1 year), while the effect was zero in the long run (2 year). Larrson (2000) also looked at a wage subsidy program where her estimates were slightly positive but very imprecise so that the point estimates were not statistically significant.

Regner (2002) analysed the effect of vocational and general training programs using data from 1987–1992. He used a version of the difference-in-difference method. Similarly to Larrson (2000) he did not find a positive treatment effect. Frederiksson and Johansson (2003), using regression based methodology and data from 1993–1997, found negative effects: both the training and the employment subsidy programs decreased the chance of finding jobs later. They speculate that the negative effect is a consequence of decreasing geographical mobility of the treatment group.

Although the majority of the papers find zero or negative effects for the Swedish programs, the effects of some programs on some demographic groups may be positive. Sianesi (2001), using the propensity score matching method, looked at those who became unemployed for the first time in their careers. She found that the wage subsidy programs significantly increased the employment rate in this group (by as much as 25 percentage points). Similarly to other studies she found the other program types ineffective.

Most evaluation studies argue that the Swedish labour market programs are ineffective partly because of the incentives in the system. The main message is that if organizers provide monetary incentives for participants and anyone can join a program then the program can easily turn ineffective as people may not use the services in the right way. As we will see later, this mechanism is supported by findings from other countries as well. Another explanation for the failure of the Swedish programs is their lack of targeting.

Denmark

According to evaluation results, the Danish programs were more successful. Denmark is among the countries that spend the most on active labour market programs by international comparison. From the mid 1990’s, the unemployed in Denmark had to participate in a labour market program every year (later
every half year) in order to remain eligible for UI benefits, Jespersen et al. (2004) analysed the effect of classroom training and on-the-job training programs on earnings and employment prospects. They followed the treatment group and a control group for six years after the program and they used a propensity score matching method with a large set of variables. According to their results, the most successful programs were the private on-the-job training programs that turned positive after one year and led to a 10 per cent wage increase for the treatment group after the third year. The second most successful programs were the classroom training programs that turned positive in the second year with an 8 per cent wage gain. The public on-the-job training program also had a positive effect after four years at a 5 per cent level.

Geerdsen and Holm (2004) analysed the employment effects of Danish training programs and employment subsidy programs. Using data from 1995–1998 and a regression based method they found that program participants increased their job search effort even before the programs started. They found that the success of these programs was partly coming from these early effects. An explanation for this is the mandatory nature of the programs. People who do not want to participate in labour market programs will increase their effort to find a job just to avoid participation. Similar effects have been documented in other mandatory programs in other countries as well. These results suggest that compulsory programs have an indirect but definitely positive effect on labour market outcomes.

Switzerland

Switzerland reformed its unemployment insurance system in 1997 and increased the importance of its active labour market programs. These programs are organized by individual cantons but the regulation is federal. Participation is mandatory for roughly 15 per cent of the unemployed. The largest programs are the training programs that are typically composed of short classroom training, but there are employment subsidy programs in both the private and the non-profit sector as well.

Lalive et al. (2002) used regressions to analyse the effect of the programs on the duration of unemployment spells. He found the programs largely ineffective. The only subgroup with partial success was the immigrants for whom the wage subsidy programs seemingly worked.

Frolich and Lechner (2004) used regression-discontinuity design to analyse the effects of employment subsidy programs at private and non-profit firms. Importantly for their analysis, participation in such programs is mandatory in some cantons but not in others. The identification strategy used discontinuity at canton borders: participation is mandatory for unemployed people living on one side of the border but it is voluntary for unemployed people on the other side. The method can identify the treatment effects for those living...
close to canton borders and would participate in the program if it is mandatory but not if voluntary. The results suggest that the employment subsidy programs had a substantial effect (15 percentage points) on the employment rate. While these results are not entirely in line with the findings of Lalive et al. (2002), both studies find that the wage subsidy programs might work in Switzerland.

Training programs

*Job Training Partnership Act (United States)*

The United States has organized federal training programs since the 1960s. The literature usually refers to these programs by the law that enacted them. The first law was called Manpower Development and Training Act (MDTA) in 1962. This was replaced by the Economic Opportunity Act in 1964. The Job Corps program that we have been discussing earlier was initiated within this act. The next law was the Comprehensive Employment and Training Act (CETA) from 1973 to 1982, then the Job Training Partnership Act (JTPA) from 1982 to 1998. The most recent law is called the Workforce Investment Act (WIA) and has been in effect since 1998.

These laws determine the institutional framework and the goals of the different sub-programs. Although the programs are organized federally, the actual training and related services are provided by more than 600 private and public institutions locally. As a result, the programs show significant geographic heterogeneity. The programs can be differentiated by the type of service and by groups they target. Participation in the programs has been increasing; and at present one million Americans are involved.

Within the JTPA a controlled experiment was carried out involving 16 training centres between 1987 and 1989. This experiment was called the National JTPA Study. The long-run results of the experiment were published in a 1996 GAO report. The most important findings were the following. Overall, the programs had a short-run positive effect on both employment and earnings, but the long-run effects were smaller and not statistically different from zero. The effects were more positive for women, and there were little if any effects on young participants regardless of their gender.

*Figures 2.2 and 2.3* show male and female earnings from three years before the program through the 5 years following the program. The earnings of the non-employed are counted as zeros. The figures show the short-run positive effects and their decrease over time. They also show a marked decrease before the program (especially among men) and marked increase after the program both for participants and the control group: this is Ashenfelter’s dip discussed above. At the end of the 5th year the earnings difference between participants and the control groups is not statistically significant.
Heckman et al. (2000), however, are critical about these conclusions. They point to important biases due to the program substitution and attrition from the program. The first means that some people in the control group actually received treatment through other programs. The second point means that some members of the randomly selected treatment group did not receive treatment in the end. For this reason the controlled experiment identifies the intent-to-treat effect and not the actual effect of the training on program participants. The authors claim that the effect of these two biases is substantial.
According to their estimates, 27–40 per cent of the control group received some treatment. At the same time, only 49–60 per cent of the treatment group received the treatment. These numbers are very far from the theoretical 0 and 100 per cent. When the authors estimate the treatment effects using these treatment probabilities, they found substantially stronger positive impacts in all groups, young or adults, men or women.

Eberwein et al. (1997) aimed at uncovering the mechanism behind the positive employment effect of the programs. Their question was whether employment increased because employment spells became longer or unemployment spells became shorter. Longer employment spells would be evidence for increased productivity, while decreased unemployment spells would suggest more efficient search. They found that the duration of the employment spells did not increase significantly, but the length of the unemployment spells shortened. Ham and Lalonde (1996), using the same techniques, analysed the effect of the NSW program and found the opposite effect. According to their results, the NSW program increased the duration of the employment spells while it did not affect the unemployment spells. Taken together, these two studies suggest that the complex and targeted NSW program likely increased the productivity of the participants, but the JTPA training programs only helped people find jobs more easily and quickly.

Worker Profiling and Re-employment Services (United States)

Black et al. (2003) analysed the Worker Profiling and Re-employment Services (WPRS) program that has taken place in Kentucky since 1994. The program consisted of a 4 to 6 hours long counselling session and consequent training. The program started as a mandatory program for the long-term unemployed, but it soon turned out that they did not have enough capacities. For this reason the program become mandatory only for specific target groups. Based on a priori estimates the organizers profiled the unemployed into groups based on the expected duration of their unemployment. From these estimated durations a few groups were created, and the treatment groups were selected based on which group people belonged to (the group with the highest expected durations received treatment first, the second highest next, etc.). For those who were selected, participation was compulsory. The number of the groups was small, and selection to the program was often randomized in order to break ties. These randomized selection events created an experiment that helped identify the effect of the program.

The authors found that participants found jobs in a significantly faster way. This led to lower social costs and an increase in the employment rate of the target group. They also found a remarkable result: the majority of the increase in the employment rate could be attributed to those who left the program (by finding a job) before it actually started. The most important effect of the program, thus,
was not due to the enhanced job search skills of the participants. Instead, the program made some people look for a job in order to avoid participating in it.

Knowledge Lift (Sweden)

The Knowledge Lift, a program “unprecedented in its size and scope” (Albrecht et al., 2004) was launched in 1997 by the dramatic extension of the existing Swedish training programs. It aimed at raising the general skill levels of the undereducated Swedish adults in Swedish and English languages, maths, etc. The program ran with more than 200,000 participants until 2002, a number that is remarkably high even in Sweden. As a comparison the total number of students in Swedish high schools is 300,000.

Albrecht et al. (2004) estimated the direct effect of the program by propensity score matching method, and they assessed the general equilibrium effect by a calibrated macroeconomic model. Because of the long history of labour market programs, most of the long-term unemployed had already participated in some of those programs before the Knowledge Lift. This fact made the evaluation difficult as the authors restricted the analysis to treated and control individuals without such history. The results of the study suggest that the Knowledge Lift did not increase the earnings of any age groups or any gender, but it helped the employment prospects of young (25–40 year old) males. According to the general equilibrium analysis, the program led to a shift in labour demand for skilled occupations. Another important effect is that the program created a large demand for teachers that led to significant shortages in high schools.

Evidence from post-communist countries

Active labour market programs are significant in most of the post-communist countries of Central and Eastern Europe. In terms of total spending relative to the total cost of unemployment benefits, they are comparable to other continental European countries. Relative to GDP, though, their active labour market programs are smaller. Programs in Central and Eastern Europe are typical examples of the large scale national programs. At the beginning of the 1990s they were mostly targeted to the unemployed who had lost their jobs due to the structural transformation of the former socialist economies. Later the programs started focusing on minorities, on the disabled, on the young and on other special groups. All types of active labour market programs have been widespread in post-communist countries, including general and vocational training programs, wage subsidies, assistance of the self-employed and public work programs.

An important conclusion of the program evaluation literature is that the effectiveness of the programs largely depends on details of the organization, incentives and other regulatory issues. Since unemployment was a new phe-
nomenon, it took a considerable amount of time to build up appropriate institutions and streamline procedures. Consequently, the early programs in the region were often even less effective than their Western European counterparts.

Unfortunately, similarly to most continental European countries, the active labour market programs in Central and Eastern Europe were not evaluated by credible studies in the 1990s. There are signs of progress after 2000 but randomised experiments are still very rare, and the quality of the available data is very weak for non-experimental evaluations. Nevertheless, we describe some studies not because of the credibility of the evaluations but because of the relevance of the programs for the Hungarian experience.

**East Germany**

Active labour market programs immediately appeared in East Germany after the unification of the country in 1990. The largest programs were the vocational training programs, but there existed general training programs, wage subsidy programs and public works programs as well. At the beginning of the 1990s, vocational training programs focused on retraining the unemployed for other occupations. Participants of these programs received a stipend that was larger than the unemployment benefit, and participation counted as employment for the renewal of unemployment benefits.

*Bergemann et al. (2005)* used a propensity score matching method with data including long labour market histories. They analysed programs that ran at the beginning and at the end of the 1990s. They found that training programs had zero or at most a marginally positive effect on employment rates. According to their results, the small positive effects on employment came from decreased duration of unemployment spells, without any effect on employment spells. In other words, participants could find jobs a little easier, but they could not hold on to their jobs longer. This suggests that the programs did not increase the skill level of the treatment group. Instead, it increased participants’ efforts or efficacy in job search. At the end of the 1990s when programs shrank in size but became more complex and more costly, the analysis found a weak positive effect on the duration of employment spells as well.

**Poland**

The early Polish programs were described by *O’Leary (1997)* and *Puhani (1998)*. Poland had all the major types of active labour market programs. At the beginning of the 1990s, retraining programs were the largest. Similarly to the early East German programs, participants in Poland received a stipend that was higher than the unemployment benefit. The wage subsidy programs subsidized all wage and social security costs of employment up to 150 per cent of the average wage in the country. Beyond training and wage subsidy, there were programs to assist the self-employed and public works as well.
Several studies analysed the effects of these programs, for example O’Leary (1998a), Puhani (1998) and Kluve et al. (2004). They all used non-experimental methods (propensity score matching method or regression models), but data quality in all cases was rather weak. The results of these studies are often contradictory, except that they all agree that the public works programs were ineffective in Poland. The more reliable papers found that the wage subsidy programs were also unsuccessful. Some of the evaluations showed short run positive effects for retraining programs, but medium and long-run effects were not analysed due to the unavailability of data.

A Hungarian study

O’Leary (1998b) analysed the labour market programs of Hungary in the mid-1990’s. Hungary is also among the countries that have offered a wide range of program types. The popular retraining programs of the 90s, similarly to other countries in the region, offered a stipend that was higher than the unemployment benefits. The wage subsidy programs were less generous than elsewhere, with employers obliged to make up for at least half of the labour costs. Due to the quality of the available data the results of the evaluation study are not particularly credible. The study suggests that self-employment assistance increased the employment rate of the participants but it decreased their earnings. The training programs may have had some small positive effects, while the wage subsidy program decreased the employment rate but increased earnings.

Romania

Active labour market programs grew rapidly in Romania after a reform in 1997. The largest were the training programs, both general and vocational, but all other program types were offered as well.

Planas and Benus (2006) analysed programs between 1999 and 2002 using a propensity score matching method. Their data were richer than the Polish and Hungarian data, with information about the labour market histories of individuals. Unfortunately, their sample was rather small which made their estimates imprecise. They found that the programs assisting small businesses increased the employment rate of the participants but they did not influence their earnings. The results suggest that the employment and relocation services were also beneficial, the training programs had a small and statistically insignificant effect, and the public works programs were rather detrimental.

Slovakia

The most reliable analysis of the active labour market programs of Slovakia from the 1990s is by Ours (2000). Slovakia also used all the major program types, but their size distribution was different from other post-communist countries. The largest programs, both by the number of participants and by their costs,
were the wage subsidy programs, followed by public works programs. Training programs were smaller than those.

The wage subsidy programs commenced in 1991. They were open for the registered unemployed, and they supported employment in profit-oriented private firms. These criteria became less restrictive over time. In 1992 they eliminated the profit-seeking requirement and in 1994 fresh graduates from the schooling system were also let in. Public works programs, or as they were called, publicly useful jobs, were originally run by the public administration and the non-profit sector. In 1992 Slovakia reorganized this system. Public organizations were excluded from these programs, and private firms were allowed to enter as long as the subsidized jobs were considered publicly useful. The programs grew significantly after 1995, with many large-scale construction projects (highways, water dams etc.). Retraining programs in Slovakia were relatively small. They were also very short, with a duration of two months on average. Most training programs were organized by the private sector, and they focused on vocational training.

Ours (2000) analysed the effects of the programs on the duration of the employment and unemployment spells. Contrary to findings from other countries in the region, he found the public works programs quite effective. According to his results, these programs reduced the unemployment spells and they increased the length of the employment spells. It seems, therefore, that the Slovak model for public works programs worked, perhaps because of the participation of private firms. The results suggest that participants of these programs could acquire skills that were valued on the labour market, unlike in typical public works that do not increase participants’ skills. Ours (2000) found that the most popular wage subsidy programs were ineffective, and the training programs were effective only by decreasing the length of the unemployment spells. The author is somewhat sceptical about this last result. According to his argument, the duration of unemployment may have dropped because some people joined the private sector training programs only when they were offered future employment to begin with.

Conclusion

This study gave an international overview of active labour market programs. Because credible identification of the effects of these programs is challenging, our overview was very selective by concentrating on the papers we judged the most credible. One of the most important results of our survey is that the quality of the organization and other details matter more for the effects of a program than the type of the program. Because details of any particular program are very important, it is important to evaluate each and every program to give feedback to policy makers and the general public.

The evidence suggests that complex and well-targeted programs, like the Job Corps, the New Deal for Young People or the NSW, can have positive results,
even for groups in which labour market programs usually do poorly. An example for this is the long-term unemployment of the young. The example of the Job Corps is also interesting because it demonstrates how a successful small scale program can lose efficiency when it grows very fast.

In continental Europe, however, complex and targeted programs are rare. Instead, these countries have large scale national programs that offer a wide range of program types to almost all unemployed. Participation in these programs is usually incentivized by stipends that are larger than the unemployment benefit. Our survey suggests that these programs have very little or zero impact on the employment and earnings of the participants.

We surveyed examples of ineffective programs where participation is helped by monetary incentives. There are reasons to believe that such incentives can be responsible for the negative results even if the content of the program is meaningful. Other examples showed that mandatory programs, when refusal of participation leads to the withdrawal of social benefits, can be very effective. The mandatory nature of programs can help even if the content of the program itself is not particularly useful because it gives strong incentives to people to search for jobs.

Training programs are relatively expensive and their effects are questionable if they are offered within the large scale national programs. The targeted American training programs are more successful to help finding jobs, although they may not enhance skill acquisition that could make participants more productive once employed. Mandatory training programs may be successful, but as we argued above, it may very well be because of the mandatory nature of the program that makes many people intensify their search effort in order to avoid participation.

Wage subsidy programs show mixed results, but the more successful ones increase the employment prospects of participants. Unfortunately we know very little about the indirect effects of these programs on other workers who might be crowded out of the labour market by the program participants. Public works programs are almost always ineffective and may be detrimental to participants’ future employment by decreasing their search effort and locking them into geographic areas. The only one positive example we are aware of is from Slovakia, but those programs were run by private institutions and they were more similar to the wage subsidy programs than classic public work programs.

Overall, the results suggest that the active labour market programs in continental Europe, including the former socialist countries, are not effective. The effects can be improved with more careful design of the institutional details and the incentives, better targeting and complex solutions. Proper incentives for job search are essential, and credible evaluation of the programs is necessary for feedback for improvements. These improvements can make active labour market programs help the employment prospects of the long term unemployed. At the same time, even the best designed programs cannot produce miracles.
3. GREASING THE WHEELS OF THE LABOUR MARKET?

Impact estimation of modernising the public employment service (project HRDOP 1.2)\(^1\)

**ZSOMBOR CSERES-GERGELY**

Introduction

The Public Employment Service (PES, in Hungary: *Állami Foglalkoztatási Szolgálat*) is an important player in the Hungarian labour market as well with a budget of around HUF 20 billion per year (of the GDP) and serving – depending on the labour market conditions – 450–600 thousand registered jobseekers (about 11–15% or the active population).\(^2\) It has absorbed more than HUF 10 billion in order to modernise its operation, yet we know little about the effectiveness of its operations and none about the effect of this modernisation. The study in this chapter attempts to quantify this effect using econometric techniques.

The aim of a PES in general is to facilitate the match of the demand and supply side of the labour market using specific tools – in terms of the matching theory of (*Blanchard and Diamond*, 1989) and (*Pissarides* 2000), this amounts to helping the operation of the matching technology. Although the need for public funding in this area was supported since the post-second world war period (*Baldwin*, 1951) to today (*OECD*, 2006), the toolbox has undergone several changes. In industrialised countries and most importantly in Europe, the role of the PES is not limited to the mere matching of jobseekers and job offers. The PES offices are the main vehicle of delivering labour market policy measures, the most important tools of activation (*OECD*, 2007), both a cause and result of the high proportion of the less educated among clients. The PES is thus a central institution in the so-called flexicurity framework as it greatly facilitates the transition between labour market states (*Wilthagen*, 2008).

The Hungarian PES (HPES) plays the role of both an authority and a supporting organisation. In line with its previous strategy, its latest articles of incorporation declare that its main role is the delivery of active and passive labour market measures. However, being the primary delivery network of employment policy with the potential of relatively rapid and direct intervention, it almost always had to attend different duties as well, such as the administration of temporary jobs, administering a large part of the rehabilitation process of disabled workers or assisting public employment (parts of these have been institutionalised by the new articles of incorporation in 2011). In addition to the diversity of duties, only limited resources are available to the HPES, barely

---

\(^1\) This work is an advanced version of the qualitative part of the evaluation study delivered by Budapest Institute and IFUA Horváth & Partners Ltd. in fulfilment of an order within a framework-agreement by the National Development Agency. The evaluation project was led by Ágota Scharle and its full final report can be downloaded in Hungarian from the web address [http://www.budapestinstitute.eu/dontestamogatas/prj/Az_Allami_Foglalkoztatasi_Szolgatal_modernizaciojanak_ertekelese](http://www.budapestinstitute.eu/dontestamogatas/prj/Az_Allami_Foglalkoztatasi_Szolgatal_modernizaciojanak_ertekelese).

I would like thank the constructive comments that Ágota Scharle and Gábor Kézdi made on the manuscript and the opportunity of thinking about the problem together. I would like to thank the support and the precise work of the people at the National Employment Office. I also appreciate the apt research assistance given by Bálint Szőke to this version of the paper.

2 Act 2008. CII. on the Central Budget of the Hungarian Republic. The Public Employment Office was subsequently renamed the National Employment Office in 2011.
sufficient to run basic operations. As part of the austerity measures, the number of employees in local offices had already began to shrink in 2006 and neither did they increase following the onset of the 2008 crisis. While in 2006 one officer attended an average of 206 clients, this has increased to 273 by 2009 (figures from direct HPES communication).

The modernisation process started in 2002 and is still ongoing. Given its importance both in terms of employment policy and the size of the development effort, one might assume that its monitoring has attracted as much attention as did its delivery, but this is not so. The current paper aims at providing a quantitative assessment of the potential impact of the program, based on the best data available for analysis. The question I would like to answer here is a very simple one from the point of view of economics: did the modernisation project contribute to making the operation of the PES more efficient so that the chance of its clients finding jobs increases? In what follows, I use a difference-in-differences (DiD) strategy to estimate such a possible effect of the developments between 2004 and 2008 using aggregate data relating to the local PES offices. First I briefly describe the development program itself, then move on to the theoretical and methodological considerations and the data used for the analysis. Thirdly I describe the estimation results and finally put them in context and provide conclusions.

The modernisation of the Hungarian Public Employment Service – a summary

The EU-funded modernisation of the Hungarian PES started before Hungary joined the EU in 2002 and is still ongoing. Its main aim was to carry out a general reform of operations to boost its performance in improving clients’ re-employment potential. At the end of the 1990s, the operation of local PES offices were characterised by neglected interiors, out-dated IT infrastructure, officers lacking a general overview of competency areas and clients being not only served by but also dependent on officers and interested mostly in collecting unemployment insurance and benefit payments. A Phare “twinning” project (with a budget of around HUF 1.2 billion) was started in 2002 to prepare accession to the European Union through improving upon these unfavourable circumstances, followed by the HRDOP 1.2 measure (with a budget of HUF 9.3 billion) and the still on-going SROP 1.3.1 project. These efforts have touched upon all mentioned areas in 20, 60 and another 60 local offices, respectively as well as in the National Employment Office, the methodological and coordination centre of the PES. Here I shall look at the middle of this process, the HRDOP 1.2 measure.

The aims of the development process are mapped onto projects – often overarching actual measures or programs – whose combined effect is what I con-
sider here as the intervention to be analysed. The total of 89 projects in the HRDOP 1.2 measure consisted of the following elements:

– **Introduction of the new service model (NSM)** on the participating local offices, including profiling of the clients. The essence of the NSM is that clients are profiled and sorted on the basis of their individual characteristics so that they can be served with personalised services. International experiences show that efficient profiling and a proper match of services can shorten the length of the unemployed status and thus be an effective device in lowering the unemployment rate as such.

– **Remodelling the participant local offices, installation of self-help terminals (M-points)** This development was based on the international experience that a more open, client-centred interior makes a less official impression and presents the place as a service provider rather than a rigid governmental office, which in itself facilitates the interaction with clients. Self-help terminals can be useful for all jobseekers, but after proper profiling, a large share of the services provided by the PES can be directed to this channel.

– **Introduction of a quality assurance system in the participating local offices.** The HRDOP 1.2 measure included the introduction of the Common Assessment Framework, CAF in order to support the operation of the NSM through the increase of overall efficiency of operations.

– **Training of staff at the participating local offices.** This part of the project connects to the NSM and CAF in order to educate staff members about their operation and enable them to successfully adopt them.

– **Introduction of an integrated IT system (IR).** This development is aimed at supporting the operation of the whole network of local offices by providing an integrated information backbone to services dependent on data managed or processed centrally. It supports administration, serves as a basis of a performance monitoring system and provides data for statistical analyses.

– **Other types of activities, such as research indirectly related to the operation of the PES.**

Given the above main interventions, we can expect effects in relation to all offices on the one hand (in the case of the IR) and in relation to participant offices on the other (such as the self-help terminals, remodelling, the introduction of NSM and CAF). Although the progress of the program can be monitored in the case of all projects, we can measure the effect on the re-employment chances of clients only in cases where the interventions were specific to participating local offices. The reason for this is that there is no good reference point for developments that affect all local offices uniformly and therefore an impact cannot be reliably estimated. Although not helpful for measurement, the presence of such interventions does not create a problem either: it has affected all offices uniformly and most importantly, it was completed only by the end of the period at which we are looking. We also have to note that given the

---

3 Profiling is essentially the prediction of the length of the unemployment spell. Unemployed persons with a longer and shorter expected unemployment spell might require very different assistance and be capable of very different levels of self-help.
multiple links between interventions and that all of them were completed in the participating local offices, we cannot measure their individual effect, but only the totality of them.

The principle and model for impact estimation

The current analysis aims at estimating the program effect on the level of establishments using data on operation before and after the program period, in line with the suggestions of (Nagy, 2006). Being interested in the actual outcome of the program, I estimate the average treatment effect on the treated (ATT). Because the analysis is focussed on labour markets in which the PES offices are located, this measurement provides us with an estimate of the net effect of the program on the unemployed, that is the combined effect of direct effect and indirect effects. It does not include the possible displacement effects on people outside the unemployment registry or people in neighboring labor markets. Given however that the program effects are freely available to everyone and without extra obligations, those unaccounted indirect effects are likely to be modest.

In what follows, I estimate program effects in a difference-in-differences (DiD) framework corrected with linear regression, first applied directly to the affected groups of offices, then using matching to homogenise them. These total four versions of the estimates can be used as a cross-check on the one hand (similarly to the dated but comprehensive evaluation of active labour market programs in Hungary (O’Leary, 1998)), but on the other hand, they also deliver information on the contribution of other factors to the outcomes of the program.

In order to get rid of the time-invariant effects possibly correlated with program participation, I have written the estimating equation in differences-form:

\[ \Delta Y_{it} = \tau + \delta p_i + \beta X_{it} + u_{it}, \]

where \( p_i \) is an indicator of program participation, \( \tau \) is a constant measuring the autonomous rate of change in this, \( X_{it} \) is a set of variables indicating relevant observable characteristics of the local offices, while \( u_{it} \) summarises characteristics that are not correlated with these observables. \( Y_{it} \) is an outcome indicator, which I chose to be the re-employment rate of clients. The relationship is defined over PES offices observed in different time-points, \( i \) being an index for an office, \( t \) being an index for a specific month. The difference (\( \Delta \)) operator takes time-difference of a variable between the same month in both the “before” year and “after” year. For example, if \( Y_{it} \) is the re-employment rate of the registered unemployed at office \( i \) in January 2008, then \( \Delta Y_{it} \) is the difference between this and the re-employment rate in January 2004 at the same office. Our interest centres on \( \delta \), the coefficient on the \( p_i \) indicator for program-participation, which delivers the program effect in this context. One can show that
the equation in this form is a direct implementation of the DiD idea, generalised to the multiple-regression case.

First I estimate this equation using OLS based on the assumption – supported by program design – that the participant and nonparticipant groups are similar. Variable \( \Delta X_{it} \) ensures that we take into account the differences developing over time between the participant and non-participant group, and thus we do not confuse these with the effect attributable to the program. This estimation strategy runs into difficulties if there are differences between participant and non-participant groups that are correlated with the \( \Delta X_{it} \) variables or with the indicator of participation. In order to treat this, I first perform propensity-score matching, which amounts to predicting program participation using such detailed set of pre-participation variables and use the predicted propensity to find observations for every program participant that is close to it in some way (Rosenbaum and Rubin, 1983). The individual program effect can be calculated from the difference of the differences in the actual and the counterfactual outcome: \( (\Delta Y_{it} - \Delta Y_{it}^*) \). Averaging office-level differences gives us a reliable DiD estimate of the ATT under the working assumptions (Heckman, Ichimura, and Todd, 1998). After calculating simple averages, I control for time-variation in characteristics as in the simple OLS case. Here I am not matching on original re-employment rates, but on residuals from a first-step regression similar to the one used in simple OLS estimation but without the program-participation indicator. Even though the estimation is not complicated, the matching introduces hidden nonlinearity, therefore the straightforward way of calculating standard errors for the estimator would be misleading. In order to handle this situation correctly, I calculate and present bootstrap standard error estimates.

The lack of correlation between observed and unobserved effects is an assumption that one cannot prove, only argue for. We shall see that although the arguments are valid, we have a reason to be careful and observe the possibility of selective sorting into the participant group. The amount of the inconsistency in such cases depends greatly on the size and direction of effects governing such selection.

Data, preliminary results and the estimation method

This study uses data primarily from the IR developed within the framework of the HRDOP 1.2 measure itself, which contains individual data on the registered unemployed from 2000-on. I was granted permission to use these data aggregated at the level of the local offices and this aggregation was performed within the Employment Office based on the required rules. Individual data on registered unemployed in the IR of the PES contains information on sex, age, education, the occupational code of the previous job as well as an indicator of disability. Aggregate indicators calculated from these data play the role
of X variables, characterising the PES offices (using levels of their post-program values) on the one hand as well as the role of controlling for initial observable differences between participant and non-participant groups (using levels and some interactions of levels of their pre-program values) in the matching process on the other. The indicators are all defined as the share of a particular type of registered client within all registered clients.

In the case of the registered unemployed staying in touch with the local PES office, we know the direction of exit at the end of the registered status – and we can base the measure of efficiency on this information. The possible directions are the following:

1. Employment (open market).
2. Public works.
3. Supported employment (various forms of wage subsidy).
4. Training.
5. Not known due to lack of cooperation with the PES.

We can use one of the above as the indicator of efficiency to look at the program effect from different angles. Given that the primary goal of the PES is facilitation labour market match, the most directly relevant measure of efficiency is the share of clients exiting the registry towards unsupported employment on the open labour market, the rate of re-employment. Exit rates were calculated not only for different directions but also for different subpopulations too, defined over individual characteristics such as age, education or disabled status. Looking at these exit rates too enables us to assess the heterogeneity in the impact of the program, if there is any.

A great advantage of the database I use is that it is coming from the administrative records of the PES and it is thus a complete account of the events happening to the registered unemployed. This nevertheless has drawbacks too. Even though contact and thus reporting is required in principle and loses the right for financial benefits administered by the PES without it, there is no real penalty if this is not the case either due to no initial eligibility in the first place or due to having exhausted such benefits (the punishment is that it is not possible to re-claim the benefit for 3 months). In relation to the current analysis, this means that we have reliable information on the direction of exit only for those eligible for benefit. In terms of measurement, this means that we are able to measure the direction of exit and thus the aggregate outflow rate only with error. Moreover, this error can be correlated with the factors determining the chance of exit and I have no outside information to assess its size. If the measurement error is present, it lowers the outcome variable by not counting every successful exit to the open labour market, but we have no reason to suppose that the reverse can happen.

In order to apply the DiD strategy here, we have to chose an appropriate before and after period. Considering that the HRDOP 1.2 measure was rolled...
out between the second half of 2004 and the first half of 2008 and also that the
effects of the economic crisis were very apparent in the third quarter of 2008, I
chose the first 6 months of 2004 to be the before and that of 2008 to be the after period. To assess the program effects fully – following the advice of (Nagy,
2006) and the evidence presented in (Card, Kluve and Weber 2010) – one
should ideally follow and observe program participants for years after the end
of the program. Unfortunately this is impossible due to both the very asym-
metric impact of the economic crisis and the continuation of the development
through the SROP 1.3.1 project, basically eliminating the control group.

There are 158 local PES offices in the analysis – only those present both in
the first half of 2004 and 2008. I have omitted the one on Haller Street (spe-
cialising in helping homeless people) and the one on Andrásy Street (special-
ising in helping higher education graduates) in Budapest. I have omitted those
two offices as well where all of the Phare developments were completed. If we
are looking only at the formal definitions, we can consider offices modernised
during the HRDOP 1.2 measure as participants and those not modernised
in either during the Phare project or during the HRDOP 1.2 measure as non-
participants. However, as only 7 offices adopted the CAF during the Phare
project and the rest (13) did so only during the HRDOP 1.2 measure, I also
consider these as participants in this exercise. The end result is that out of the
total 158 offices, I have 71 participants and 85 non-participants as their controls.

The information I obtained regarding the selection of participating offices
was not free from controversy. On the one hand, we know on the basis of pre-
liminary information that participants were selected from smaller and larger
towns in every county, providing a degree of uniform randomness. On the
other hand, interviews made during the evaluation exercise showed that par-
ticipation was guided to some extent and offices in worse shape had a higher
chance to become participants. Because it proved to be impossible to collect
information on the condition of the buildings or anything similar, we can only
use the characteristics of the clients as a proxy.

Figure 3.1 shows monthly exit rates from the unemployment register be-
tween 2004 and 2008. Based on this evidence, the two major exit paths are 1) unsupported employment, with around 4 percent rate by the end of the peri-
od, 2) Not known due to lack of cooperation with the PES, with an average of
around 8 percent. Besides the slight increase in exit to unsupported employ-
ment, we can observe a much stronger decrease in the exit rate to a not known
state (already present from 2000 on, not visible on the graph), highlighting the
importance of autonomous changes. Exit rates towards all destinations appear
to show seasonal cyclicality. Figure 1 shows that exit rates grow particularly
strongly during the summer and decrease during the winter – the reason for
this is partly that seasonal jobs are offered mostly during the summer and sub-
sidies are made available during the spring, building up capacity by the summer.

5 These are located in Baja,
Orosháza, Barcs, Ózd, Hatvan,
Balassagyarmat and Esztergom.

88
Figure 3.1: Exit rates between 2004 and 2008 (average, all offices)

The large and trending decrease in the rate of exit towards an unknown state makes it very likely that there is indeed measurement error present in the indicator of the exit route and so it also affects the aggregate variable created from it. Given that the modernisation is likely to have a positive effect on efficiency a negative correlation between participating in the modernisation program and the measurement error in the outflow rate is likely to arise, leading to the overestimation of the program effect (by usual omitted variable arguments). Although we can be sure that such a distortion exists, I suspect that its size is likely to be small.

Table 3.1 shows re-employment rates in the pre- and post-program period based on office-level data, weighted by the number of the locally registered unemployed. Re-employment rate has increased greatly from 2004 to 2008. Program participants observed a 1% point increase, while the same was 0.8 % point in the case of the control group – following the DiD, the program-effect is the difference between these two numbers, 0.23 % point. This number is not small in relation to the overall re-employment rate, but is not significantly different from 0.

Table 3.1: Changes of average re-employment rates at the PES offices by HRDOP 1.2 program participation status

<table>
<thead>
<tr>
<th>Year</th>
<th>Participant?</th>
<th>Difference</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>No</td>
<td>Yes</td>
</tr>
<tr>
<td>2004</td>
<td>0.0414</td>
<td>0.0386</td>
</tr>
<tr>
<td>2008</td>
<td>0.0496</td>
<td>0.0491</td>
</tr>
<tr>
<td>Difference</td>
<td>0.0082</td>
<td>0.0105</td>
</tr>
</tbody>
</table>

Notes: Without participants of the Phare program; averages are weighted by the number of clients registered with the local office.

Source: Own calculations using data aggregated from the IR of the PES.
Aggregating data to the level of the whole country, we can observe trends relating to the registered unemployed. Figure 3.2 shows that the composition of the clients has changed over time – this is one of the external effects we have to control for during estimation. The graph shows the proportion of vulnerable groups, likely to have difficulty with re-employment: those without maturity exam, above 50, labour market entrants and disabled persons (counting them multiply, hence proportions add up to more than 100).

Figure 3.2: Average composition of the local PES offices between January 2004 and December 2008

The most pronounced change is the growth of the share of the 50+ among the registered clients. Their share was a mere 15% in 2000, which grew by 5% points in 10 years, partly explained by the rise in retirement age, partly by the autonomous increase in their level of education. The share of those without a maturity exam decreases slowly but steadily, showing a strong seasonal pattern: it decreases rapidly during the summer months providing seasonal jobs, but decreases during the winter.

When using a DiD strategy, it is very important to have very similar participants and non-participants on average so that the latter form a valid control group. Table 3.2 shows the average of indicators of offices’ characteristics in the beginning of 2004, just before program participation. There are two types of indicators: one set includes the characteristics of the registered unemployed, the second includes their exit rates towards different outcomes.

Participating and non-participating local offices appear very similar: there is no real difference either in re-employment chances. The main difference is that there are almost twice as many clients registered with participating offices on average than in the case of non-participants whereas the share of better
educated clients is larger in the latter case (with very low absolute shares). Not only means, but also the spread of the indicators are very similar (not shown in the table), therefore the requirement of using only observations whose characteristics are actually comparable (staying on the common support) during the DiD analysis is not particularly demanding, as we shall see when performing matching.

**Table 3.2: Main observable characteristics of participating and non-participating local PES offices in January 2004**

<table>
<thead>
<tr>
<th></th>
<th>Participant</th>
<th>Non-participant</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Average number of clients of the local office</strong></td>
<td>1531</td>
<td>3090</td>
</tr>
<tr>
<td><strong>Re-employment rate in the given subgroup of the registered unemployed</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Age: 15-25</td>
<td>4.3</td>
<td>4.1</td>
</tr>
<tr>
<td>Age: 26-50</td>
<td>4.4</td>
<td>4.1</td>
</tr>
<tr>
<td>Age: 50+</td>
<td>3.1</td>
<td>2.9</td>
</tr>
<tr>
<td>Education: without maturity exam (including lower secondary vocational education)</td>
<td>3.9</td>
<td>3.6</td>
</tr>
<tr>
<td>Education: with maturity exam</td>
<td>4.9</td>
<td>5.0</td>
</tr>
<tr>
<td>Education: higher education</td>
<td>5.5</td>
<td>5.3</td>
</tr>
<tr>
<td>Education: lower secondary vocational education</td>
<td>5.1</td>
<td>4.8</td>
</tr>
<tr>
<td>Disabled</td>
<td>2.5</td>
<td>2.6</td>
</tr>
<tr>
<td>Not labour market entrant</td>
<td>4.5</td>
<td>4.2</td>
</tr>
<tr>
<td><strong>The share of the given subgroup among all the registered unemployed</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Age: 15-25</td>
<td>14.7</td>
<td>14.5</td>
</tr>
<tr>
<td>Age: 50+</td>
<td>19.4</td>
<td>19.3</td>
</tr>
<tr>
<td>Education: without maturity exam (including lower secondary vocational education)</td>
<td>75.9</td>
<td>77.5</td>
</tr>
<tr>
<td>Education: higher education</td>
<td>3.2</td>
<td>2.4</td>
</tr>
<tr>
<td>Not labour market entrant</td>
<td>7.3</td>
<td>6.8</td>
</tr>
</tbody>
</table>

Source: Own calculations using data aggregated from the IR of the PES and *TSTAR* data from the HCSO

I work with aggregate data during estimation, in which observations appear more than once and this has a direct effect on the calculation of standard errors. In order to get rid of seasonal effects and increasing efficiency at the same time, I use observations for 6 months for each office in the period before and after the program, respectively. This way every observation contributes 6 times to the estimation, and the final estimate will be an average of the monthly effects. Since there is a high degree of autocorrelation between the time-periods, I calculate clustered standard errors to take account of this. Aggregation of units with different numbers of observations in them creates a well-known form of heteroskedasticity, therefore I weight the regressions by the number of registered individuals.

Explanatory variables in the model of program participation include the 2004 January values of the variables characterising local labour markets in the
parametric estimating equations, as well as levels, squares and cross-products of outflow rates towards unsupported employment and unknown direction. I have calculated z-statistics using the bootstrap method, with 100 replications. I have used the PSMATCH2 Stata module for matching (Leuven and Sianesi, 2003). I experimented with different averaging methods such as 1:1, k-nearest neighbour, kernel and local linear matching.

Estimating results

I start presenting results with estimated coefficients from simple OLS regression of the differenced estimating equation, using the method explained earlier, including restriction to the common support obtained from the participation equation in the matching estimator. Estimates related to the re-employment chances of an average registered unemployed person are shown in Table 3.3, the program effect being the coefficient on the participation indicator in the first row. The results from the most simple specification (1) simply echoes the results seen in Table 3.1, indicating a program effect of 0.16% point, or 5% (the numerical difference is due to the slightly different conditions of the estimation). However, this estimate is not significant at conventional significance levels. This can be due to – besides the lack of certain control variables – the negative bias caused by measurement error. Specification (2) includes separate indicators for all months to filter out the effect of the time of measurement. Although it increases explanatory power to 14%, neither the estimate of the program effect nor its precision has changed. Specification (3) includes even more information, most importantly shares of registered clients with a particular characteristic: age, education, labour market entrant status. Besides the rise in explanatory power, we observe an increase in the program effect to 0.3% point and an improvement in precision that makes the estimate significant. The size of the effect is close to the one obtained with matching (see later), but is somewhat larger than the raw estimate. I have experimented with further specifications, in particular with the inclusion of characteristics of local labour markets, but these brought little gain in precision, so specification (3) has remained my preferred one.

After the completely parametric estimates, I turn to matching to take into account possible initial differences between the local offices, which I have so far assumed away. Table 3.4 shows estimated program effects from simple DiD matching with various methods. Program effects are positive in all cases, but they are somewhat smaller in magnitude than raw effects in the case of averaging methods using all data (such as the kernel and local linear methods).

The next step is to combine matching and parametric estimation that is the control for both initial differences and changes in characteristics during the program period using the two-step method outlined earlier. Based on earlier results, I use the kernel method in matching and include the parametric generation in the bootstrap procedure used for the z-statistics.
Table 3.3: Results from DiD OLS regressions in various specifications

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>HRDOP 1.2. participant</strong></td>
<td>0.0016</td>
<td>0.0016</td>
<td>0.0030*</td>
</tr>
<tr>
<td></td>
<td>(0.431)</td>
<td>(0.433)</td>
<td>(0.0952)</td>
</tr>
<tr>
<td>Age: 15–25</td>
<td></td>
<td></td>
<td>-0.000919</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(0.991)</td>
</tr>
<tr>
<td>Age: 50+</td>
<td></td>
<td>0.0716</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.225)</td>
<td></td>
</tr>
<tr>
<td>Education: without maturity exam (including lower secondary vocational education)</td>
<td>0.0259</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.703)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Education: higher education</td>
<td>0.367**</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.0123)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Not labour market entrant</td>
<td></td>
<td>-0.188***</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.0039)</td>
<td></td>
</tr>
<tr>
<td>Disabled</td>
<td></td>
<td>-0.154**</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.0264)</td>
<td></td>
</tr>
<tr>
<td>Constant</td>
<td>0.00852***</td>
<td>0.00874***</td>
<td>0.00653***</td>
</tr>
<tr>
<td></td>
<td>(6.36e-09)</td>
<td>(1.06e-09)</td>
<td>(0.00859)</td>
</tr>
<tr>
<td>N (on common support)/All observations</td>
<td>834/948</td>
<td>834/948</td>
<td>834/948</td>
</tr>
<tr>
<td>R²</td>
<td>0.02</td>
<td>0.14</td>
<td>0.22</td>
</tr>
</tbody>
</table>

* Standard errors in parentheses.
** Significant at 10% level, *** significant at 5% level, **** significant at 1% level.

Table 3.4: Raw program effects estimated with matching

<table>
<thead>
<tr>
<th></th>
<th>Nearest neighbour</th>
<th>5 nearest neighboursa</th>
<th>Local linearb</th>
<th>Kernalc</th>
</tr>
</thead>
<tbody>
<tr>
<td>Estimated program-effects</td>
<td>0.0039</td>
<td>0.0059**</td>
<td>0.0019</td>
<td>0.0016</td>
</tr>
<tr>
<td>Bootstrap z-statistics</td>
<td>1.41</td>
<td>2.03</td>
<td>0.69</td>
<td>0.71</td>
</tr>
</tbody>
</table>

a Five nearest neighbours matching (with replacement).
b Local linear estimation uses tricube kernel with the default bandwidth of 0.8.
c Kernel estimation uses Epanechnikov kernel with the default bandwidth of 0.06.

Results from controlled matching are shown in Table 3.5 in the same structure adopted in Table 3.3. It is worth noting that the smallest difference is brought about here by controlling for seasonality, after which the program effect increases to 0.5% point and is significant at the 10% level. This value decreases slightly but not significantly after including the composition of the client pool of the local offices. For the same reasons as earlier, the most credible and thus preferred results come from specification (3). Replacing raw numbers with those coming from a multivariate DiD procedure combined with matching has thus small but significant net effects in the end, benefitting from correcting for both initial differences and those developing over time.
Table 3.5: Program effect calculated with matching on residuals obtained from DID OLS regressions

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Estimated program-effects</td>
<td>0.0016</td>
<td>0.0051*</td>
<td>0.0048*</td>
<td>0.0043</td>
</tr>
<tr>
<td>Bootstrap z-statistics</td>
<td>(0.71)</td>
<td>(1.81)</td>
<td>(1.76)</td>
<td>(1.55)</td>
</tr>
</tbody>
</table>

Notes: Matching is performed using the kernel averaging method. Regression pre-processing uses specification (3) from Table 3.3. *Significant at the 10% level.

Working with numbers aggregated over the whole client pool, I could not so far look at the heterogeneity of the program effect. Given that some parts of the program targeted some types of clients, this way we actually obtain some information that is possible to relate to specific parts of the program. One example of this is self-help terminals which are more targeted on the better educated clients: obtaining a positive program effect of the latter makes it more likely that elements targeted at them could have worked better. The NSM on the other hand is more likely to benefit the less able, where we can apply the same argument. Another dimension of the heterogeneity of the program effect is the direction of exit. It might be the case that overall, the NSM is more efficient in directing clients towards training, but not so effective in directing them towards employment. In order to take a look at the effect on different groups of clients and with regard to different outcome indicators, I have replicated the analysis for all combinations of these using different populations and outcome indicators. The former included groups defined by characteristics listed in Table 3.2, the latter included exit towards the directions discussed in the data section, that is towards the open labour market, public works, other active programs, training and exit to unknown direction.

The main conclusion from this exercise is that the program helped open-market employment the most: two out of three significant effects are estimated in this case. The 0.3% point average estimate comes from few large (and significant) and many smaller (and less significant) results. This can be a result of a measurement problem, given that these groups are small, but also that the program actually had a more pronounced effect in the case of the affected group. While in the case of open market employment, we do not see a significant effect in the case of the young and the 50+, the effect for the prime-age group is well above the average at 0.38%point. There is no real difference in terms of educational attainment, but these coefficients are rather imprecise. Finally, the effect for those already on the labour market is significantly larger than the average. Other coefficients are not significant at conventional levels, except for those with higher education towards ALMPs, where the program effect is negative and significant. If this effect is real, it can be attributed to the better information provided and the selection mechanism put to work and can suggest that less participation in ALMPs might be appropriate for this group.
Conclusions

This study has evaluated the effect of the 2004–2008 phase of the modernisation of the Public Employment Service in Hungary on exit chances from the unemployment registry. Results suggest that the program had a positive effect on exit to unsupported open-market employment. If I control for initial differences between participant and non-participants local offices in the composition of clients as well as its changes over time, I can conclude that the total of the interventions carried out in the HRDOP 1.2 measure had a statistically significant positive effect on re-employment chances. Analysis of subgroups revealed that the program effect was strongest in the case of prime-age workers.

The final numerical results include several corrections and are slightly larger than the one obtained from averaging raw numbers. Re-employment chances have risen from 3.86% to 4.91% in the case of participant local offices, including also effects distinct from the program itself. The impact of the program is calculated to be 0.3–0.48% point. In the first half of 2008, the number of registered unemployed was 450 thousand, of which 263 thousand were registered with program participant local offices and 5% of these become employees on the open labour market in the next month. Results show that approximately 800–1200 of them became employed as a result of the development program. We can express this result also in terms of an effect on the length of the unemployment episode. The approximately 5% exit rate measured in 2008 means that an average unemployed person spends 100/5=20 month as a registered unemployed person (assuming a constant hazard of exit), which we can calculate to be 100/(5–0.3)=21.3 to 100/(5–0.48)=22.1 months in the counterfactual case had the program not being rolled out. This means that the length of the unemployment spell was shortened by 1.3–2.1 months by the program for clients registered with the participating local offices.

Because the modernisation of the PES can be considered as a labour market program, one might want to ask the question how the benefits from the modernisation effort compare to costs and to alternative programs. Had the program an effect that lasted forever, its cost can be shown to be equal to an annual HUF273 million, or a yearly HUF1038 thousand, a monthly HUF86 thousand per capita cost. Comparing this to monthly costs of training programs and subsidies for self-employment, being a monthly HUF101 and HUF177 thousand per capita respectively, this is similar, but somewhat smaller amount.
4. THE EVALUATION OF TRAINING, WAGE SUBSIDY AND PUBLIC WORKS PROGRAMS IN HUNGARY

JUDIT CSOBA & ZITA ÉVA NAGY

This study explores the impact of labour market training, wage subsidy and public works programs in Hungary using multivariate analyses and a control-group design. The evaluation of active labour market policies was first carried out in Hungary in 1992–1993 upon the initiative of the Japanese Program of the ILO (Godfrey, Lázár and O’Leary, 1993). Following this the Hungarian public employment service has been assessing the results of active labour market programs on an annual basis since 1994. This uses a tailored monitoring system to measure the aggregate outcomes of labour market programs after they have ended. The monitoring system was reviewed and new methods were introduced in 2009 (Tajti, 2009). These – similarly to the previous period – cover the participants of wage subsidy and school leaver and business start-up schemes; however the largest group, those in public works programs, are not included.

Information in this system came from two main sources before 2009: on the one hand from the records of the employment services, and on the other hand from the responses to a postal questionnaire sent to participants three months following the end of the program. Furthermore, data have been available on employees in unsubsidised employment from the records of the National Tax and Customs Administration (NAV) and the unified employment database (UED) of the National Employment Service (NES) since 2009. However, the database of the NES does not allow us to control for selection bias nor to calculate an aggregate employment effect. Information on the largest active programs, the participants of public works schemes are still not available. This prompted us to turn to new methods.

Participants, methods and research questions

This evaluation focuses on the three active labour market policies with the largest number of participants over the past 10 years, as well as the largest share of the budget allocated to active labour market policies. The study explores the operation and impact of training, wage subsidy and public works from multiple perspectives. In 2009 73.8% of the decentralised Employment Sub-fund of the Labour Market Fund was spent on these measures.

In accordance with the generally accepted methodology of program evaluations, in addition to the participants of active programs we included a matched control group that had similar characteristics but did not participate in any active labour market policies (for a discussion of counterfactual analysis see Chapter 1 of In Focus).
Monitoring questionnaires for training programs were self-report, and for all other programs employers provided the required information in Hungary until 2009. Given that the information came from different sources – for some programs the employers and for other programs the employees – their comparability is questionable. This study collected data from job seekers and program participants, regardless of their current status.

We employed a two-stage stratified sampling to select participants. First, we selected the 18 small regions to be involved in the study using the CSO’s complex deprivation/development indicator of small regions. Then the Public Employment Service selected the participants in the active program and control groups from the records of the job centres in the 18 small regions. The list was then forwarded to the research coordinators in each local job centre who contacted the individuals – current or past job seekers – for consent to take part in the study (individuals who were no longer registered clients received a postal letter).

Data collection was carried out by independent researchers who were not affiliated with the employment service. Interviews took place at the participant’s chosen location between August 15 and September 30, 2010. The number of participants was 1,041 in the active program group and 1,068 in the control group. Thirty-eight percent (n = 394) of the active program group participated in training, nearly 10% (n = 100) in wage subsidy and 52% (n = 547) in public works.

One of the main methodological components of the research (and also one of its main challenges in terms of implementation) was the longitudinal nature of data collection. We studied the period between September 1, 2009 and February 28, 2010 and the sample included those who participated in active programs in this six-month period. Those who did not take part in any programs made up the control group. Changes were followed through four time points: pre-intervention, intervention, exit and at the time of actual data collection. In-depth analysis of the 12-month period prior to data collection was carried out.

The four time points allow multiple comparisons and a detailed follow-up. Apart from the difference between the entry and exit status, the changes within the 12-month period, their timing, seasonality, their duration and the direct impact of active policies can be analysed. In addition to the four fixed data collection points, additional points and intervals can be chosen because there is information on job finding. Therefore the dynamics and characteristics of employment can also be analysed for example taking into account the cyclical nature of the jobs market or other factors (such as seasonal cycles, economic and political factors, changes in the legislation etc.) within the 12-month period. The flexible timeframe and the four data collection points allow a pre- and post comparison and thus a more thorough program evaluation.

The scope of the questions was significantly expanded. Apart from the net employment effect of the programs and other standard indicators of program evaluation, detailed information was collected on:

1 This is a random sampling method that allows statistical generalisation of results because, in principle, there is no selection bias. We used data on participants and job centres in our analysis.
2 There were a total of 18 job offices in the sample from five territorial development categories. Within each development category we selected the three or four small regions that had the best development indicators. This was justified because we aimed to measure the effectiveness of active measures, so we selected small regions that provided the most favourable conditions within each category.
3 Data was collected on differences in income and occupation at the start, during and after the end of the program that could indicate an improvement, worsening or stability of status.
– the targeting of each program (the socio-demographic and employment characteristics of participants and those left out from the programs),
– the expectations of the participants towards the programs,
– “what happened” in each program,
– what were the objective changes in the participants’ situation after the program (for example changes in the job role, income etc.),
– what other, indirect effects did the programs have (for example in terms of quality of life, work-related skills, future expectations and work-related plans),
– the respondents’ subjective evaluation of active labour market programs and services.

Comparison of program participants and the control group

To assess the impact of active programs the study has compared the total sample of program participants (n = 1,041) and control group (n = 1068) in three main areas: firstly the most important socio-demographic and social characteristics – termed “entry differences”. It was necessary to explore these because they might have an impact on the outcomes (selection bias). Secondly, it was assessed to what extent the members of the control group had these entry characteristics and what other factors they mentioned to explain their passive status. Finally, the analysis of the changes observed during the 12 months of the study aimed to explore the impact of the active programs in contrast to no intervention.

Entry differences and the targeting of the programs

There were marked differences between the groups in terms of socio-demographic characteristics (Table 4.1). These indicators might act as pre-selection criteria and might ultimately have an impact on the outcomes of active programs (selection bias).

Table 4.1: Main socio-demographic characteristics of program participants and the control group

<table>
<thead>
<tr>
<th>Active program (mean)</th>
<th>Gender (%)</th>
<th>Place of residence (%)</th>
<th>Development/deprivation level (mean)</th>
<th>Family size (mean persons)</th>
<th>Distribution of the sample between the active programs and the control group (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>female</td>
<td>male</td>
<td>Age (mean)</td>
<td>county centre</td>
<td>other town</td>
</tr>
<tr>
<td>Participated in active program</td>
<td>57</td>
<td>43</td>
<td>37</td>
<td>20</td>
<td>44</td>
</tr>
<tr>
<td>Training</td>
<td>57</td>
<td>43</td>
<td>35</td>
<td>30</td>
<td>50</td>
</tr>
<tr>
<td>Wage subsidy</td>
<td>72</td>
<td>28</td>
<td>38</td>
<td>43</td>
<td>44</td>
</tr>
<tr>
<td>Public works</td>
<td>54</td>
<td>46</td>
<td>39</td>
<td>9</td>
<td>40</td>
</tr>
<tr>
<td>Control group</td>
<td>54</td>
<td>46</td>
<td>41</td>
<td>31</td>
<td>43</td>
</tr>
<tr>
<td>Total</td>
<td>55</td>
<td>45</td>
<td>39</td>
<td>26</td>
<td>43</td>
</tr>
</tbody>
</table>

The share of women in active labour market programs (57%) is slightly higher than in the control group (54%). Therefore active labour market measures – at
least in principle – (might) considerably contribute to reducing the employment disadvantages of women.

The average age of participants in active labour market measures is substantially lower (by 4.2 years) than the average age of the control group. This result confirms the widely known disadvantage of older workers even in subsidised employment schemes. The average of the control group is closest to that of participants in public works projects (difference of 2.6 years) that suggests that out of the active measures public works projects are the most inclusive for the older generations, however they are still less likely to be involved in these.

There are also substantial differences in terms of education level. Out of all the groups included in the research, the education level of the control group was the second lowest. Participants in training programs and wage subsidy schemes had spent the longest time in education (the control group left school at the age of 18.7 on average, while the recipients of wage subsidies left school at at 20.7 years, training participants at 19.8 years and those in public works at 17.8 years).

While the majority of those in public works programs and in the control group had vocational qualifications, participants from wage subsidy schemes and training programs had graduated from secondary school and had a baccalaureate in addition to their vocational qualification (Figure 4.1).

Figure 4.1: Highest education level by status groups (percent)

The difference in the education levels at baseline is particularly relevant because – according to a number of previous studies – this alone, regardless of the effect of the program can influence the probability of different outcomes. Galasi, Lázár and Nagy (2003) argued that education level has a large impact on the success of participants. A higher education level increases the probability of job-finding, with the exception of the baccalaureate alone without any higher qualifications.

At first glance this study also predicts a higher job finding rate for those with a higher education level which could lead to the conclusion that active labour market measures are more effective for this group. Nevertheless estimating
more complex relationships using a logit model later will suggest that education level alone is not a significant factor.

According to our hypothesis, the likelihood of a successful employment outcome might be also related to the labour market status prior to the program. The correlation between program entry, education level and previous activity might suggest that the employment prospects of active program participants are overall considerably worse than in the control group (at the beginning of the intervention 4% of the active labour market program participants and 21% of the control group had worked in unsubsidised employment (Table 4.2).

**Table 4.2: Number of participants in unsubsidised jobs by highest education at the beginning of the intervention, September 2009 (per cent)**

<table>
<thead>
<tr>
<th>Education level</th>
<th>Participants in active labour market measures</th>
<th>Control group</th>
</tr>
</thead>
<tbody>
<tr>
<td>Less than eight grades</td>
<td>0</td>
<td>5</td>
</tr>
<tr>
<td>Eight grades</td>
<td>20</td>
<td>17</td>
</tr>
<tr>
<td>Vocational training school</td>
<td>24</td>
<td>41</td>
</tr>
<tr>
<td>Vocational secondary school</td>
<td>11</td>
<td>14</td>
</tr>
<tr>
<td>High school</td>
<td>26</td>
<td>12</td>
</tr>
<tr>
<td>Post-secondary vocational program</td>
<td>4</td>
<td>6</td>
</tr>
<tr>
<td>College or more</td>
<td>15</td>
<td>6</td>
</tr>
<tr>
<td>Total</td>
<td>4</td>
<td>21</td>
</tr>
</tbody>
</table>

The data highlight the variability of previous labour market activity by education level both within and between groups. It can be assumed that individuals with at least a higher secondary education (baccalaureate) are not only more likely to find employment on the open labour market but they are also more likely to be involved (and more quickly) in active labour market measures.

The family structure of the control group and participants of public works programs is more traditional than in other participant groups (38 and 39 per cent respectively are married or live with a partner, while this is only 29% among training participants and 33% among the beneficiaries of wage subsidies). The average family size in the control group was smaller than among program participants (3.1 compared to 3.4 persons). The members of larger families are more likely to be involved in active labour market measures than job seekers from smaller families (the average family size for participants receiving wage subsidy is 3.1 persons, 3.3 for those in training and 3.6 for participants of public works). Single mothers – who make up nearly a quarter of the sample – and young families with small children are particularly motivated to secure an income.

The residents of the most disadvantaged small regions were more likely to be involved in active labour market programs than those in more developed regions, nevertheless jobseekers living in smaller localities were less likely to ac-
cess wage subsidies and training opportunities than would be expected based on their other characteristics (Table 4.1).

Whether the participants of active labour market programs received any unemployment or inactivity related assistance significantly influences the probability of finding a job. Galasi, Lázár and Nagy (2003) suggested that people who previously had received benefits were around 40 percent less likely to find a job than people who never received benefits before. Our findings confirm this. Job seekers who are not receiving unemployment benefits are twice as likely to find a job than the recipients of unemployment benefits. Consequently those who have ever been in subsidised employment before (mainly in public works) are half as likely to find open employment as the people who have never been involved in these active labour market measures.

The majority of participants of active labour market measures in this study had previously received unemployment-related assistance or took part in other active programs (Table 4.3). Only 29% of those entering a training program were registered unemployed who did not receive any unemployment-related assistance. The same number was 26% in the wage subsidy schemes, 17% in public works programs, and 23% in the control group. The control group and the public works group had the highest level of inactivity-related income. As a result they had relative income security and thus were least motivated to find a job on the open jobs market.

Table 4.3: Participants and control group according to labour market status prior to the program

<table>
<thead>
<tr>
<th>Status</th>
<th>Training</th>
<th>Wage subsidy</th>
<th>Public works</th>
<th>Control group (March 2010)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Worked in a non-subsidised job</td>
<td>4</td>
<td>6</td>
<td>3</td>
<td>5</td>
</tr>
<tr>
<td>Worked in a subsidised job</td>
<td>2</td>
<td>10</td>
<td>7</td>
<td>1</td>
</tr>
<tr>
<td>Participated in training</td>
<td>3</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Registered unemployed</td>
<td>88</td>
<td>83</td>
<td>88</td>
<td>90</td>
</tr>
<tr>
<td>Out of which without assistance</td>
<td>29</td>
<td>26</td>
<td>18</td>
<td>23</td>
</tr>
<tr>
<td>Other inactive</td>
<td>3</td>
<td>1</td>
<td>1</td>
<td>5</td>
</tr>
</tbody>
</table>

The longest jobless period was on average 15 months among people entering wage subsidy schemes, 18 months for training participants, 27 months in the control group and 35 months among the participants of public works programs. Considering the length of unemployment, the control group would be very motivated to get involved in active labour market programs because they have been out of work for a long time. Only participants of public works programs had longer jobless cycles, nevertheless – as will be shown later – the majority of them is not trying to access active labour market programs or find lawful employment.

When analysing socio-demographic indicators, the question of counter-selection should be addressed too: is there a counter-selection and if so, what fac-
tors influence this for each active labour market policy. Data estimated using a logit model shows the following specific entry selection criteria. The individual’s locality plays a significant role: people who live in “other towns” were 1.7 times more likely to enrol in a training program than those who live in a village (the situation of people who live in county centres is not significantly different from those who live in villages). Education level is also significant: training programs seem to be targeted at individuals with a baccalaureate – grammar school graduates were 1.83 times and secondary school graduates with a vocational qualification were 1.49 times more likely to be involved in training compared to those with the highest level of education. In comparison to people with eight years of general education these numbers are 3.04 and 3.74. In terms of regional development/deprivation, training is significantly more common in areas with an average or above average development level than in the most deprived areas (2 and 1.65 times).

For wage subsidy the main selection criteria are gender, education level and per capita income, however the level of regional development is also a significant factor. Women are 1.8 times more likely to participate in a wage subsidy program than men. In terms of education level, the advantage of college and higher vocational qualification is the greatest factor, they are 4.1 times more likely to receive a wage subsidy than people with only eight years of general education, while the same number is 3.3 for people with a university degree.

Participation in public works was influenced by the individual’s locality, the development level of the small region, ethnic background and previous work history. The Roma unemployed were 1.8 times more likely to be involved in public works than the non-Roma unemployed. Village dwellers were twice as likely to enrol in public works than residents of towns, and seven times more likely than city dwellers. Public works is primarily targeted at lower educated people and the least developed areas. Compared to the least developed small regions, the probability of being in public works was 0.5 in “less developed” and “developed” areas, and was even lower in the remaining two categories.

Data suggest that the unemployed who lived in county centres were 3.3 times more likely to be in the control group than the unemployed in villages (or in other words: not participating in any active labour market programs and having no income). The unemployed who lived in the most developed small regions were 6.3 times more likely to be in the control group than those in the least developed small regions, 0.6 for a non-Roma versus a Roma, and 0.6 for a graduate versus someone with eight years of general education.

These findings about the selection into participant and control groups suggest that active labour market measures are after all targeted at those with multiple disadvantages (although different groups of them). While selection into the control group was more likely among the non-Roma unemployed with a vocational qualification or a degree who lived in the more developed areas, the Roma unemployed and those with only eight years of general education were primarily targeted at.
small regions (whose employment prospects are in principle better), regional disadvantages, being lower on the hierarchy of localities and educational disadvantages increased the likelihood of participation in active labour market measures. However there were some fundamental differences between the different measures in terms of entry characteristics. Therefore, it can be assumed that training was more powerful in reducing disadvantages due to its more favourable entry characteristics (small regions of average or above average development level, medium-sized towns, grammar school or baccalaureate with vocational qualification) than for example public works where the main entry characteristics (least developed regions, villages, low education level) forecast the lower probability of successful labour market integration.

**Subjective status assessment and work motivation**

There were other differences between the control group and active labour market program participants that were due to subjective rather than objective factors. These differences are presented in *Table 4.4*.

<table>
<thead>
<tr>
<th>Subjective reason</th>
<th>Training</th>
<th>Wage subsidy</th>
<th>Public works</th>
<th>Control group</th>
</tr>
</thead>
<tbody>
<tr>
<td>Health problem that limits the range of potential jobs</td>
<td>18</td>
<td>14</td>
<td>21</td>
<td>28</td>
</tr>
<tr>
<td>Too old</td>
<td>14</td>
<td>12</td>
<td>18</td>
<td>22</td>
</tr>
<tr>
<td>Too young, not enough experience</td>
<td>18</td>
<td>15</td>
<td>11</td>
<td>10</td>
</tr>
<tr>
<td>Ethnicity</td>
<td>3</td>
<td>1</td>
<td>13</td>
<td>5</td>
</tr>
<tr>
<td>Does not have enough money to buy adequate clothes</td>
<td>3</td>
<td>0</td>
<td>5</td>
<td>4</td>
</tr>
<tr>
<td>Housing problems</td>
<td>1</td>
<td>0</td>
<td>4</td>
<td>4</td>
</tr>
<tr>
<td>Has been out of work for long/has never had a job</td>
<td>11</td>
<td>14</td>
<td>9</td>
<td>15</td>
</tr>
<tr>
<td>Outdated qualification</td>
<td>6</td>
<td>4</td>
<td>8</td>
<td>11</td>
</tr>
<tr>
<td>Outdated knowledge</td>
<td>4</td>
<td>3</td>
<td>6</td>
<td>11</td>
</tr>
<tr>
<td>Stopped working during maternity leave</td>
<td>12</td>
<td>4</td>
<td>8</td>
<td>12</td>
</tr>
<tr>
<td>Village locality, lack of jobs</td>
<td>21</td>
<td>13</td>
<td>39</td>
<td>22</td>
</tr>
<tr>
<td>Too expensive to commute</td>
<td>26</td>
<td>16</td>
<td>34</td>
<td>27</td>
</tr>
<tr>
<td>Lack of adequate public transportation</td>
<td>10</td>
<td>5</td>
<td>18</td>
<td>8</td>
</tr>
<tr>
<td>They are less likely to hire unemployed</td>
<td>11</td>
<td>4</td>
<td>10</td>
<td>16</td>
</tr>
<tr>
<td>Too many unemployed in the area</td>
<td>54</td>
<td>30</td>
<td>54</td>
<td>55</td>
</tr>
<tr>
<td>Could not sell property to move away</td>
<td>12</td>
<td>8</td>
<td>19</td>
<td>14</td>
</tr>
</tbody>
</table>

People in the control group were considerably more likely than active labour market measure participants to view their health status, their age and the lack of employment opportunities, the time spent out of work, outdated qualification and the rejection of the environment as a result of their unemployment as an obstacle. The subjective assessments also emphasise the regional disadvantages and the lack of work-related skills (knowledge and work experience) but
health status is considered more important – nearly one third of respondents mentioned this as the reason for not looking for work.

Apart from the data presented in Table 4.4 there is information on the intensity of job search after participation in an active labour market measure. Twenty-six percent of the control group clearly expressed their unwillingness to work and even to participate in active labour market programs. The main reason for non-participation (50% of respondents mentioned this) was that they were not offered the possibility to take part in an active measure. Nevertheless, it should be highlighted that it was only in a minority of cases that participation in an active labour market policy was initiated by job centre advisors. Participation was initiated by the unemployed themselves in 65% of the cases in training, 67% in wage subsidy and 55% in public works. For each measure less than one third of the participants were referred by the employment service. Therefore passivity results from other factors, not only the fact that they were not informed of the opportunities or offered participation.

Investigating (further) the reasons for staying away from the labour market both among the participants of active labour market measures and the control group, we sought the individual’s own explanation of their passivity, why they are not looking for work on the open labour market.

Men were deterred from the open labour market by their health status and alternative sources of income, while women mentioned child care responsibilities, increased expenses as a result of taking up work, less flexibility in terms of time use and higher expectations. Participants of training, wage subsidy and public works programs mentioned similar factors as disadvantages of active labour market programs. Only a smaller proportion of respondents said that illegal employment was one of the factors that kept them away from work (24% said that they have ever worked illegally and 6% said that they were working illegally at the time of responding), the majority of them in the control group. Nearly a quarter of those not giving any reasons for staying out of work were employed illegally.

There were also differences in terms of expectations towards work. Participants of active labour market programs were “less picky” and the control group had higher expectations and stricter conditions towards potential jobs: they were more likely to reject outdoors work (71% vs 59% among participants), work that is potential harmful to health (68% vs 65%), were less likely to accept 12-hour working days (56% vs 60%), and shift work (53% vs 56%). Interestingly there were no differences between the groups in terms of accepting part of the salary paid directly in cash (and without paying tax or other contributions), or if they could work part-time or from home – nearly two thirds were willing to accept these options. Almost 60% would be willing to pay for equipment or work wear if they had the opportunity to work. A similar percentage (over 70%) would reject undeclared employment, if they could not take annual leave, if they could not go on sick leave and unpaid overtime.
The results suggest that there are substantial differences between the two groups in terms of their attitude and expectations towards work. The participants of the active labour market programs seem more open and flexible, while the control group is more attached to traditional forms of employment both in lawful and undeclared work. This has implications for the job prospects of both groups and highlights the necessity of differential treatment.

Program evaluation – a comparative analysis of active labour market policies

Changes in the status as a result of participation in active labour market programs (ALMPs) was measured at four time points: before intervention (baseline), at the beginning of the intervention, at exit from the active measure (and linked to this the outcome indicator), and at data collection (at the end of a 12-month observation period). Data from the first two stages (baseline and intervention) have already been discussed. The results as measured by the outcome indicators and the impact of the programs will now be summarised, highlighting and comparing the similarities and differences between the four measures.

Immediately after the intervention (at exit) participants of the wage subsidy schemes were the most likely to find unsubsidised work on the open labour market (72 per cent) (Table 4.5). These were followed by training participants (12 per cent). The employment rate of the control group and public works participants – both at 5% – seems to support the previous argument that the probability of open employment is similarly low in the two groups with less favourable socio-demographic indicators, less motivation and worse employment prospects.

<table>
<thead>
<tr>
<th>Status</th>
<th>Training</th>
<th>Wage subsidy</th>
<th>Public works</th>
<th>Control group (March 2010)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Employed, not subsidised</td>
<td>12</td>
<td>72</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>Employed, subsidised</td>
<td>4</td>
<td>0</td>
<td>7</td>
<td>1</td>
</tr>
<tr>
<td>In training</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>Unemployed</td>
<td>83</td>
<td>25</td>
<td>87</td>
<td>89</td>
</tr>
<tr>
<td>Other inactive</td>
<td>2</td>
<td>3</td>
<td>1</td>
<td>5</td>
</tr>
</tbody>
</table>

Note: For ALMPs N = 839, for the control group N = 1061.

Thus far monitoring studies have measured the effectiveness of programs with the number of participants who found paid employment within six months from the end of the program. Although this study has adopted a longer timeframe (12 months), data from the first six months was compared to findings of other studies (see Table 4.6).
Table 4.6: Comparison of the number of participants taking up employment within six months in the two studies (per cent)

<table>
<thead>
<tr>
<th>ALMP</th>
<th>1997 (Galasi, Lázár and Nagy, 1999)</th>
<th>2010 (this study)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Training</td>
<td>45</td>
<td>33</td>
</tr>
<tr>
<td>Wage subsidy</td>
<td>71</td>
<td>77</td>
</tr>
<tr>
<td>Public works</td>
<td>25</td>
<td>23</td>
</tr>
<tr>
<td>Control group</td>
<td>n. d.</td>
<td>7</td>
</tr>
</tbody>
</table>

The findings of the study by Galasi, Lázár and Nagy (1999) – that used the records of the Hungarian Employment Methodological Centre with nearly 5,000 participants in 1997 – and this study using survey methods are by-and-large similar.

Considering the differences in the methods between the two studies (one was based on the secondary analysis of data from a large administrative database and the other on data collection using survey methods on a sample of participants) this is noteworthy and reassuring. Nevertheless, it also suggests that the efficiency of active labour market policies – regardless of any changes in the labour market or external circumstances – has been fairly constant over the past 13 years.

The comparison of the status of the ALMP participants and the control group at the end of the study (at the 12th month) also indicates the more favourable situation of ALMP participants. Nineteen percent worked in un-subsidised jobs as opposed to 11% of the control group. Sixty-three percent of ALMP participants were registered unemployed in comparison to 80% of the control group.

There was a substantial difference between ALMP participants and the control group in terms of participation in subsidised employment at the end of the 12-month period: this was only 4% in the control group while 15% of ALMP participants were once again among the beneficiaries. A considerable number had already participated in subsidised employment at least once (mainly public works). Only one per cent of the control group were involved in a wage subsidy scheme and 4% were in a public works programs at the end of the 12-month period. At the same time, they were significantly more likely to receive welfare benefits (34% as opposed to 20% among former ALMP participants) or unemployment benefits (23% in the control group and 20% in the ALMP group), and they were twice as likely to be inactive (4% vs 2%).

At the end of the analysis the question as to what influenced successful job finding on the open labour market should be addressed. Did the individual active labour market programs contribute significantly to this success? Using a logit model we found that two variables were particularly important in predicting a successful employment outcome. These were: status variable that showed participation in a given ALMP during the intervention period, and

---

5 In our logit model the dependent variable was non-subsidised work at the time of measurement. The explanatory variables were type of active measure (or control group), type of locality, the development level of the region, gender, age, education, marital status and ethnic background of the respondent, the length of time spent in employment within the total work history, labour market characteristics at the beginning of the intervention period (working on the open market – all else; in a subsidised job – all else; claiming unemployment benefit – all else; parental leave – all else).
unemp(1) that represented the ratio of the duration of unemployment within the total work history (see Annex 4, Table 4A1).

A further analysis of the variables with a significant effect within the model reveals that training participants were 1.82 times and recipients of a wage subsidy 20.24 times more likely to be in employment on the open labour market. The outcomes of those in public works programs are even less favourable than those of the control group: the likelihood of them finding employment was 0.26.

The “raw effects” of each active labour market program, summarised in Table 4A1 (which includes the effects of all variables and not only those of ALMPs), were compared to the results of the logit model adjusted for the effect of active labour market programs. Effects were expressed as the group-wise odds ratio between the participant and control groups. The comparison of results suggests that, despite the different methods of data analysis, the un-adjusted and adjusted odds were relatively similar (the unadjusted odds ratio of wage subsidy was 19.04 and the adjusted 20.24, while both odds were 1.82 for training relative to the control group). In public works the unadjusted odds ratio of finding employment relative to the control group was 0.37 and the adjusted odds ratio was 0.26. This difference can indicate on the one hand a selection bias or alternatively a negative effect on employment outcomes (for example through a stall effect – this is discussed later in the study).

The variable unemp(1) highlighted again the already known negative relationship between the length of unemployment and the probability of finding employment.

Our results are similar to the findings of Galasi et al. (2003) and also to those reported by Kluve (2010): the outcome is largely explained by the type of ALMP the individual has participated in (or the absence of ALMP). This also means that the differences in the entry characteristics do not have an effect of their own but they are accumulated through selection into a program.

During the 12-month period of data collection in this study – considering the whole of the period (Table 4.7) – 41% of ALMP participants found employment. Unlike the general practice, in this time period only lawful employment was considered in order to give an insight into the movements between “re-employment”, subsidised employment and active labour market programs. (This indicator includes very short-term employment too.) On average it took 2.1 months for ALMP participants to find work and 24% found employment on the open labour market. Thirty-eight percent of those who have taken part in training found work and this took on average 2.3 months. Two thirds of those who found work (24.4% of total training participants) secured a job on the open labour market and the rest were divided between wage subsidy schemes (66%) and public works (33%). Eighty percent of those previously involved in a wage subsidy scheme found work out of which 76% were on the open labour market and less than one percent in public works. On average it took 0.5 month
for them to find employment. From those in public works 32% found employment in an average of 2.8 months. However only 6% of these were employed on the open labour market.

Table 4.7: The number of those in employment at the end of the 12-month observation period (per cent)

<table>
<thead>
<tr>
<th></th>
<th>Employed</th>
<th>Not employed</th>
<th>Missing status</th>
<th>If employed, how long did it take to find first “regular” employment?</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>total subsidised</td>
<td></td>
<td></td>
<td>0-3 months</td>
</tr>
<tr>
<td>Training</td>
<td>38</td>
<td>14</td>
<td>60</td>
<td>1</td>
</tr>
<tr>
<td>Wage subsidy</td>
<td>80</td>
<td>4</td>
<td>20</td>
<td>-</td>
</tr>
<tr>
<td>Public works</td>
<td>32</td>
<td>26</td>
<td>62</td>
<td>6</td>
</tr>
<tr>
<td>Total ALMPs</td>
<td>41</td>
<td>17</td>
<td>56</td>
<td>3</td>
</tr>
<tr>
<td>Control group</td>
<td>37</td>
<td>5</td>
<td>63</td>
<td>-</td>
</tr>
</tbody>
</table>

* Data includes unsubsidised employment, public works, wage subsidy, business start-up support and agricultural producers.

Thirty-seven percent of the control group found work – including any type of employment – during the 12-month observation period. Thirty-two percent of this was open employment.

The methodology of the study also allows us to consider changes longitudinally rather than at discrete data collection points. Figure 4.2 illustrates this.

**Figure 4.2: Status of ALMP participants and the control group between March and September 2010 (per cent)**

*Figure 4.2* shows the second half of the 12-month observation period (from February to August) in which anyone from the control group had the possi-
bility to take part in ALMPs as well – this was an exclusion criterion only in the first six months of the study. The figure allows us to follow the dynamics of changes month by month and also illustrates the variety and changes of outcomes other than open employment.

The figure shows that those who had previously participated in ALMPs were more likely to be among the beneficiaries again even in the second – relatively short – part of the observation period than people from the control group. Previous ALMP participants were found in wage subsidy schemes, training and public works in fairly large numbers. Maybe they were participating in a different type of program, however it can be concluded that they were making better use of the opportunities offered by the labour market and the employment service than members of the control group. Another notable difference is that the people in the control group were more likely to move to an economically inactive status than participants who had previously been involved in ALMPs.

In program evaluation – in addition to the employment outcomes – it is also important to consider to what extent active labour market policies contributed to maintaining or improving the situation and social status of those involved. To assess this effect, changes in the participants’ job roles and type of work contract were analysed.

The findings of the analysis are briefly summarised here. Participants in training and public works programs generally maintained their status following intervention and there was even evidence of a small improvement. The status of those in wage subsidy schemes remained largely the same. There is also a minor status improvement in the control group, however its scale lags behind the changes among training and public works participants.

Another finding was that the share of fixed-term contracts increased considerably among all employment contracts. Wage subsidy programs appear the most stable because the share of permanent contracts was highest among their former beneficiaries (nearly two thirds), and there was only a 9% decline in the number of permanent contracts compared to the longest job ever held by the respondent. Among public works participants barely every tenth worker has the chance for a stable and long-term job.

During the intervention period – apart from the above changes – the wage expectations grew substantially in each ALMP group while there was no change in the control group.

Active labour market policies can also have an impact on participants’ employability and skills, self-esteem, feeling a useful member of society and social networks. These “soft” indicators are not widely accepted in program evaluations because there are no standard measurements for these. Nevertheless it is useful to offer a brief insight into these more “indirect” outcomes of ALMPs.

The subjective assessment of programs was carried out among those who participated in ALMPs and were successful in finding a job: 32% said that they
would have not found a job without the program while 28% thought that they would have been able to secure a job without it. Particularly, those in training thought they would have been able to find a job without it while public works participants were more likely to attribute their success to the program.

According to our findings, ALMPs also altered some of the self-concepts held by the individual participants. Participants of training and wage subsidy schemes were more likely to report a significant improvement in their employability. Nearly two thirds of those who reported a deterioration were from the control group. In terms of self-esteem, changes were similar: the highest improvement was reported by training participants (21%). Participants of wage subsidy schemes were the least likely to report a significant improvement in self-advocacy and work motivation (13% for both). The biggest improvement in professional and theoretical knowledge was – not surprisingly – reported by training participants: 61% said that the program was “good”. According to respondents programs were not very helpful in developing contacts and social networks for job search.

Data suggest that one of the value added elements of ALMPs can be personal development and the strengthening of self-esteem. Participants most often mentioned the importance of respect, a feeling of being useful and actively contributing to the family income.

An important finding however, was that individuals from the control group who succeeded in finding a job on their own were much more positive about the changes than ALMP participants who considered job creation and continued employment as part of the programs.

Characteristics and outcomes of active labour market policies

The following section considers each of the three active labour market policies and examines the characteristics, the processes and factors behind the effects presented earlier.

Training

According to data 38% of participants found a job between the end of training and the end of the data collection period. On average this took 2.3 months from the end of the training program. In the first three months 29% of respondents, in months four to six an additional two per cent and in months 10 to 12 only 0.3% found work (Table 4.7).

Program evaluations often put forward a number of misgivings with respect to the effectiveness of training / re-training programs (see the Chapter 2 in this In Focus as well as Frey, 2008). According to earlier program evaluations, many of the participants in training programs are people with higher than average education level and skills sought after on the labour market who would most likely find work without this (Frey, 2008, Tajti, 2009). Our results confirmed
this finding. Training is not the most adequately targeted active labour market policy. It is not helping those without qualifications to gain a qualification but rather it is assisting in the re-training of people who already have a qualification. Sixty-two percent of training participants already had a vocational qualification (the same level was 68% in wage subsidy schemes and 47% in public works) and the number of early school leavers was not higher than the average among them (24% of the total sample and 23% of training participants were early school leavers). It is also known that just over a quarter of previous qualifications were never used!

Fifty-two per cent of participants received a re-training allowance once, while 13% twice or more often – most likely they participate in successive training programs to secure an income.6

The large majority of respondents (92%) participated in their preferred training. Typically participants were self-referred (65%) and only 30% of the participants were referred by the employment service. It is interesting to note that 28% of those referred by the employment service found a job on the open labour market, while this was only 13% for those who referred themselves. So why did participants enrol in training? Most people mentioned interest, the hope of finding work, however 60% also mentioned that they were paid to participate.

About a fifth of labour market training programs do not provide a new qualification but develop new skills and abilities that improve the employability of the participants.7 The majority of training programs are in the areas of information technology, languages, administration, retail, welding and cutting and social care.

The majority of training participants (78%) received a national vocational qualification, 6% received a vocational diploma (not included in the register of national vocational qualifications), 12% received a special vocational qualification (such as ECDL, language examination, non-professional driving licence etc.), and two per cent received other qualification. Furthermore two per cent of participants dropped out of training. Results show that there was no significant relationship between the type of training and the probability of re-employment (the only exception was the special vocational qualification that was significantly less likely to be associated with successful job search).

Out of the 38% in employment only 24% were on the open labour market. A further nine per cent of training participants were employed in a wage subsidy scheme and five per cent in a public works programs. The effect of training on the probability of re-employment is strongest in the first three months and is also noticeable for a further three months, however it becomes negligible six months after the closure of the program.

To understand the effect of training on employment, we should also consider the characteristics of participants who succeeded in returning to the labour

6 Researchers created the term trainfare based on the concepts of work-fare and welfare. It describes the phenomenon when participants enrol in subsequent training programs in order to qualify for the training allowance that becomes a main source of income (Jordan, 1996).

7 Such as language courses, computer courses, driving lessons and other courses developing competences.
market. The effect of the following variables on re-employment was examined: status at the beginning of the observation period, education level, age, marital status, gender, ethnic background, type and development of place of residence. The main characteristics of training were also taken into account, such as the duration, the type of qualification, type of referral, and whether it was the preferred training option of the respondent.

Results show that three factors had a significant effect on the probability of a successful employment outcome. On the one hand the respondent’s place of residence, and the development level of the small region on the other, were important. Compared to a respondent living in a county centre, the odds of job-finding were significantly higher for somebody living in a small town (2.27) and significantly lower for a village resident (0.642). Therefore, training might be more important and effective in helping job seekers to find work in medium-sized localities where there are potential job vacancies, however there might be a shortage of qualified workforce (even due to the drain effect of larger cities) than in small localities or cities. It might also be argued that the organisers of training initiatives are more aware of the needs of the local labour market. The results also show that the effect of training on the probability of re-employment is smallest in the most developed areas: the odds of a successful employment outcome here are 0.113 compared to the least developed areas. The effect of training on employment outcomes is similar to the general trends observed within all active labour market policies: there are particular patterns within the five development categories. The effects are most favourable in the less and more developed areas, however they are negligible in the least developed, developed and most developed areas.

The third contributing factor to the probability of re-employment was the duration of training. Similarly to earlier research (see Chapter 2 of In Focus, O’Leary, 1998b) the findings of this study also confirmed that short training courses (one to six months) are much more effective than long courses (see Table 4.8).

8 This effect was strengthened by the fact that people with higher levels of education were more likely to enrol in longer – especially six to nine months – training courses. The only exception was skilled workers who were significantly more likely to be involved in training courses longer than nine months.

Table 4.8: The effect of the duration of training on the probability of re-employment

<table>
<thead>
<tr>
<th>Duration of training</th>
<th>Percentage of respondents in unsubsidised jobs at time of data collection</th>
</tr>
</thead>
<tbody>
<tr>
<td>Less than 3 months</td>
<td>25</td>
</tr>
<tr>
<td>4-6 months</td>
<td>27</td>
</tr>
<tr>
<td>7-9 months</td>
<td>14</td>
</tr>
<tr>
<td>Longer than 9 months</td>
<td>10</td>
</tr>
</tbody>
</table>

The evaluation of training programs by the participants was not very favourable. Forty-five per cent of participants said that they would have been able to find a similar job without training (the same number was 21–28% for other active labour market policies). If we accept the assessment of the participants
then it can be concluded there is a large deadweight loss associated with current labour market training policies.

Respondents said that the training was useful at a personal level; two thirds thought that the new knowledge was slightly or very useful and would happily take part in training again: 27% said they would do it once more, 39% said they would be willing to do it twice or even more times. The most popular courses were language, IT, accountancy, and machine-operating. In terms of motivation to enrol in a training program and the willingness to participate again by 62% of the respondents, it might be argued that this is not only the result of the new skills / knowledge but also due to the motivating effect of the income that training participants received which was significant. This suggests that participants were motivated to “learn a living”.

The effect on employment outcomes of training is distorted by changes in the attitude of participants: they were looking for different, higher prestige jobs than they would have done without training. According to our results there was no skill enhancement effect among participants. Expectations regarding potential jobs – for example wage expectations or whether they would be willing to take up undeclared employment – changed substantially (improved). However this was also the case for other active labour market policies (therefore it might be argued that this is an effect of the intervention itself, the fact that people were offered support, rather than a training-specific issue).

Wage subsidy

In wage subsidy schemes 80% of the participants found a job between the end of the intervention and the end of the observation period. Out of this figure, 77% found one within six months which increased to 80% by the end of the twelfth month.

The primary aim of wage subsidy schemes within active labour market policies is to help job seekers into the open labour market and provide a sustainable employment outcome following the end of the subsidy. The follow-up of these programs within the monitoring system over the past years showed a very favourable picture and indicated a very high (continued) employment ratio (Frey, 2008, Tajti, 2009). For example in 2009 over two thirds of recipients (75.6%) of a wage subsidy were still employed six months following the end of the scheme (and over the required duration) (Tajti, 2009).

In this study the average duration of unemployment before entering a wage subsidy scheme was 9.7 months – 23 months for social assistance recipients, 10 months for registered job seekers, and the shortest time (one month) among former training participants. Those who spent three months or less without work before the wage subsidy scheme were significantly more likely to find unsubsidised employment than those who were jobless for longer. In this respect, wage subsidy policies support the employment of the short-term unemployed.
In terms of the typical routes into wage subsidy schemes, three groups were identified among the job seekers using a clustering method.\(^9\)

The largest group (active jobseekers – 44 people) was characterised by self-referral without previous participation in wage subsidy schemes or work experience at their chosen workplace. Typically they were not hired for new positions and it was observed only in this group that other employees lost their job because the employer hired subsidised workers. The second group (those using their social networks to return to work – 22 people) was made up – among others – of participants who had already received wage subsidy in the past. Their personal relationship with their future employer is key in their employment which is also indicated by the fact that they had not worked at the chosen workplace before and they were hired for a newly created position. The smallest group (returnees – 12 people) comprised of people who had already been employed by the same employer at least once but sometimes even twice. Nevertheless, most respondents said that they were hired for a newly created position.

According to Gerfin-Lechner and Steiger (2002) as cited by Hudomiet and Kézdi (2008) wage subsidy schemes often have a greater effect on the situation of women than that of men, although the reasons are unclear. The effect was particularly large for the long-term unemployed for whom the wage subsidy scheme increased the probability of employment by 13% 18 months following the end of the scheme. For the short-term unemployed the estimated effect was not significant.

In this study the percentage of men and women in the scheme was 28% and 72% respectively – men were more likely to say that they would have been able to find a job without the subsidy (61%) than women (50%). Their expectations were confirmed by the actual outcomes, 74% of men were in unsubsidised employment after the end of the scheme as opposed to 68% of women, while seven per cent of men and 13% of women were registered job seekers. In conclusion it might be argued that contrary to the international literature, women were more likely to be involved in wage subsidy schemes, however men had more favourable employment outcomes.

Over one third of respondents (38%) said that they received a wage subsidy for up to six months. The average duration of subsidised employment was 8.3 months.\(^10\) Women are not only more likely to be involved in a scheme, they also receive the subsidy for longer than men: on average for 8.5 months compared to 7.8 months. The mode employment duration was nine months.

The analysis reveals no relationship between successful employment and socio-demographic factors (age, marital status, place of residence and development level, education). What are the significant factors? According to some it is the relationship with the employer. Wage subsidy schemes were often criticised in this study because they do not create new jobs but rather they create incentives for re-employing existing workers for a higher wage. A

\(^9\) The analysis was carried out using \(k\)-means clustering.
\(^10\) The monitoring studies carried out by the PES collected data from employers to evaluate the efficiency of wage subsidy programs. Our research aimed to involve participants in the survey. One of the disadvantages of this is that participants couldn’t always differentiate the different phases of the program. They struggled to distinguish the subsidy period from the continuation of employment although our questionnaire clearly asked about the situation after the end of the program regarding the continuation of employment.
few respondents mentioned that whole teams (for example sewing factories) were “re-employed”.

To check the accuracy of these claims and to try to uncover the real effect respondents were also asked whether they started their subsidised employment alone or as part of a group of workers. Seventy-seven per cent said they started their job alone and 23% reported that they were part of a group.

There was a highly significant relationship between group re-employment and the willingness of employees to continue the employment of workers without the subsidy. Seventy-four per cent of those in a group said that they would have been employed without the subsidy as opposed to 48% of those employed alone.

The extent of the substitutions effect – or termed differently the hypothetical “re-employment” – was measured using further variables. First, respondents were asked where they heard about the opportunity. Fifty-four per cent of recipients of wage subsidy reported that they first heard about the scheme in the job centre and 23% said that they were informed by their employer. This gives an upper value for the extent of re-employment in the sample. The second question to measure the extent of re-employment was about who initiated employment. In 20% of the cases this was the job office, in 23% the employer directly and in 53% of the cases the job seeker (out which 9% had already participated in a wage subsidy scheme).

Thus in 23% of the cases employers provided information and in 23% of the cases initiated subsidised employment but only 9% of participants had previously participated in a wage subsidy scheme. Sixteen per cent of respondents had worked for the same employer before and three per cent said that this was their third job with the same employer. It can be assumed that nearly 20% of participants were re-employed in the wage subsidy program. There was evidence that the respondent had previously worked for the same employer and participated in a wage subsidy program before. Therefore it is justified to argue that approximately 20% of program participants were simply re-circulated into the labour market.

Another highly contested feature of wage subsidy programs is whether they actually contribute to the creation of jobs. The monitoring studies of the employment service asked employers whether they would have created the same jobs in the absence of the subsidy. This is the so-called deadweight loss of the active labour market policy and according to the monitoring studies it was consistently around 20–25% each year.

To measure deadweight loss this study asked participants of wage subsidy programs: 53% said that they would have been employed without the subsidy. The logit model looking at the effect of deadweight loss highlights an important factor: there was no significant difference between the probability of finding employment after the program among those who said they would have got
the job without the subsidy and those who thought the subsidy was necessary. Therefore it seems that there is a large deadweight loss at the start of the programs, at the selection of participants, however there is no deadweight loss in terms of the long-term employment effect and it makes no difference whether they would have got the job with or without the subsidy.

Examining deadweight loss from a different perspective, the data show that 70% of job seekers were first-time wage subsidy recipients and had not worked for the same employer before. Therefore it could be assumed that these were new positions that were created using wage subsidy. However the possibility that workers were made redundant to hire somebody using wage subsidy cannot be ruled out. Nine per cent of the respondents – and one person who was re-employed by the same employer – said that their job had been filled by somebody else and that person had been made redundant before they started working for the employer.

It is useful to consider the substitution effect and the deadweight loss of wage subsidy programs because the average unit cost of wage subsidy was 853,200 forints per person in the first half of 2009 in contrast to 173,200 forints for training (Tajti, 2009, p 17).

Hudomiet–Kézdi give an overview of the attempts to measure the impact of active labour market policies in Chapter 2 of In Focus and in their 2008 paper. They also argue that international experience suggests that wage subsidy programs have large negative externalities, highlighting the substitution effect in particular. They argued in Chapter 1 that wage subsidy programs in the United States are generally considered successful, while wage subsidy programs in Northern Europe are regarded as non-successful. The picture in Western Europe and the post-communist countries is mixed. This study could not measure the substitution effect but it can be assumed that it is present in the wage subsidy programs. In conclusion, with regard to the substitution effect and deadweight loss, up to 60–65% of jobs created using wage subsidies are new positions.

Our results also suggest that the employment effect of wage subsidy programs is very good in comparison to other active labour market policies, however this is moderated by their strong substitution effect and deadweight loss. If these are taken into account the effects are only slightly better than in training programs, nevertheless wage subsidies are still considerably more effective than public works programs. Nevertheless recipients of wage subsidy have more favourable socio-demographic characteristics and the unit cost of the program is higher than in public works programs.

Public works

Looking at the employment effect of public works – the combined ratio of open employment and subsidised employment – the results are surprising. Within the first three months following the end of public works 21% of the partici-
pants found work, within four to six months a further 8%, two per cent within seven to nine months and one percent within ten to 12 months. This meant that 32% of the participants worked for some time either in open employment or in another subsidised job, mainly public works after leaving the program.

The effectiveness and the impact of public works programs can be assessed in the context of their objectives. None of the regulations and documents setting out the aims of public works programs makes an explicit reference to the objective of open employment. Therefore we only used one indicator to measure the effectiveness of public works – participation in the open labour market. The analysis of other effects and outcomes can also indicate whether the program fulfils its objectives, and its effectiveness.

Participants in public works programs had less favourable demographic characteristics and lower levels of education – as shown in previous sections – than registered job seekers. Older, less educated groups were significantly over represented among public works participants. Probably related to this, the type of work typically carried out in public works is unskilled physical labour. According to data from the monitoring evaluations between 2001 and 2006 77–80% (Frey, 2008), and according to this study 78% of the jobs were unskilled positions related to community infrastructure. Related problems have been well documented by a series of research studies over recent years and several alternative models were put forward, however the type of work carried out in public works programs has basically remained unchanged for the past 20 years (Csoba, 2010a, 2010b).

The effectiveness of a program is enhanced by the attitudes of participants – whether they are motivated to take part and accept the aims of the program as their own. The element of coercion is considerably stronger in public works than in wage subsidy programs and this also impacts on its effectiveness. Sixty-five per cent of training participants and 67% of wage subsidy program participants said that they volunteered to participate, the same number was 55% for public works and in 27.9% of the cases unemployed participants were approached by the job office.

Mainly men with upper secondary education or lower rely on this employment opportunity, although over half of those with a vocational qualification would also be willing to take part again. Over three quarters of women with a vocational qualification or a baccalaureate said that they were likely to participate again in public works. Participants aged 25 years or under were more optimistic in terms of open employment prospects than those aged 26 years or over who would be more willing to participate in public works again. In general, older generations are more willing to accept public works as a substitute for open employment.

There is a significant relationship between the average duration of participation in public works and the level of regional development. In the most devel-

11 The study looked at the previous regime of public works that included municipal public works (organised by local councils), communal public works (organised by the PES) and centrally organised public works. In the end we did not consider the three types separately because participants could rarely tell which type they were involved in. They identified themselves as "public workers" regardless of the type of the program.

12 At the time of the research municipal public works were regulated by Article 36 of Act 3 of 1993 on Social Administration and Social Provisions. The aims and the subsidies available for communal public works were set out Article 16/A of Act 4 of 1991 on the Promotion of Employment and Provision of Unemployment Assistance and Article 12 of Mol regulation no. 6/1996. (16. 07). Centrally organised public works projects were regulated by Government regulation no. 49/1999 (26. 03).
in focus

oped areas this was 8.6 months, 8.4 months in the areas of average development and 7.9 months in the least developed areas. The duration of employment is also related to the position filled by the worker: in non-administrative white collar positions this was 9.7 months, in administrative white-collar positions 8.5 months and in blue-collar positions it was 6.9 months.

The effect of subsidised employment is significantly reduced if the same job seekers are admitted to the programs more than once. The ratio of re-employment is highest in public works programs, more than half of the job seekers had already worked in the same job. As a result of recurrent employment the worker considers the public works agency as a long-term stable employer which can hinder efforts to find a job elsewhere. Nearly one third of the sample took part in public works programs on a recurring basis (more than twice). Nearly one third of respondents (28%) had already worked in their current position more than once. This is particularly characteristic of men who regularly return to public works as if it was a seasonal job. Four per cent of the sample participated in public works 10 times or more; the average number was 2.78–3.2 for men and 2.4 for women. In line with the recurring nature of public works, 80% of participants were planning to take part in similar programs again.

There is a significant relationship between the effect of the program and the age of participants. For the under 25 group there is a stronger transition effect than in the older age group. Younger people are more likely to find work than older people who participated in similar active labour market programs. People aged 36–45 years, although they make up the largest proportion of participants in public works, seem to rely on returning to the program for temporary income. There are very few people aged over 45 years involved in public works and they are less likely to find a job.

The probability of unsubsidised employment is also influenced by the place of residence: compared to county centres, public works is 1.47 times more likely to lead to employment in small towns and 2.4 times more likely in villages. Therefore personal networks and the fact that new employees are often hired as public workers at the beginning due to the lack of resources have a considerable impact on the transition effect of the program.

The probability of open employment increases as the sector of the employer moves away from the local council status. The average ratio of open employment was 5% – if the employer was a local council this was only 3% and at non-profit organisations it was 5%. However, if the participant was employed by a state-owned company the job finding rate increased to 8% and if it was a large state-owned company it reached 17%.

Public works in companies that operate under market or quasi-market conditions is significantly more effective in helping people find employment than the public sector where the emphasis is on temporary work and income rather than on the transition effect.
It is assumed that some of the participants of public works programs remain in open employment for a couple of months following the end of the program (6% of those lawfully employed), however the open employment effect of the program declines after the third month and by the twelfth month hardly reaches 3% and the share of public works increases again (at the 12th month point of the observation this stood at 38%).

Within six months following the end of the program, participants of public works were the most likely to develop a “survival strategy” and 24% reported working illegally for various lengths of time.

Those who became economically inactive after taking part in a public works program did not want to return to subsidised employment again. It seems that public works and inactivity-related benefit receipt were considered mutually exclusive rather than alternative options by the respondents. Those who managed to secure an inactivity-related benefit did not wish to return to public works. Public works are not attractive enough in terms of income potential, prestige or any other factor to motivate inactive benefit recipients to return to work.

The more often people return to public works the less motivated they become to find a job on the open labour market. Seventy-three percent of first-time public works participants said that they were likely to participate again, while the same number was 77% for those who participated twice and 93% for people who participated in public works three times. The numbers of participation in public works is strongly related to the probability of re-employment. Among those who were involved in public works five or six times the job finding rate on the open labour market is very low. Public works programs create dependency and a loss of self initiative.

The lock-in effect develops during participation in the third public works program and leads a situation that is less favourable than the individual’s initial situation in terms of employment prospects (Hudomiet and Kézdi, 2008, Scharle, 2011).

Considering another aspect of this phenomenon it confirms our hypothesis that it is the third public works program that locks in participants. Eighteen per cent of those who participated once said that they were not planning to take part again, and the same number was 14% for those who participated twice and only five per cent for those who participated three times. Therefore those who are at the beginning of their “public works career” and not yet considering subsidised employment as a way to extend eligibility for unemployment-related benefits were more likely to reject this rather unfavourable job opportunity than those for whom this short-term employment and income opportunity had already become part of their survival strategy.
Conclusion

This study has examined the impact of three active labour market policies – wage subsidy, training and public works – using multivariate analysis with a control group design. The study considered a broad range of outcomes and looked at the changes longitudinally. The sample was selected using a two-stage stratified sampling method and consisted of 1,041 participants in active labour market policies and 1,068 job seekers not receiving any interventions in the control group.

In terms of the targeting of programs, the findings of the study suggest that ALMPs generally target disadvantaged job seekers compared to the control group: regional disadvantages, localities at the lower end of the hierarchy and educational disadvantages increase the probability of participation in an active labour market program. However there are differences between the participants in each program that might have implications for their effectiveness.

At the end of the study 19% of ALMP participants were in unsubsidised employment compared to 11% of the control group. However the probability of re-employment on the open labour market – a successful employment outcome – was significantly different by type of program. Training participants were twice as likely as the control group to find a job, while beneficiaries of wage subsidy programs were 20 times more likely. However, participants of public works programs were considerably less likely – one fourth as likely – to find work than the control group.

Training is not the most adequately targeted active labour market measure because over two thirds of participants had a formal vocational qualification thus it did not help participants to gain their first vocational qualification. Furthermore it only helped one in four participants to find work on the open labour market. This might be related to the fact that over half of the participants considered training not as a direct route into work but rather as an income opportunity.

Taken literally, the results suggest that wage subsidy programs tend to enhance the employment of the short-term unemployed. On the other hand, the program has a large substitutions effect and deadweight loss: only up to two thirds of the jobs are newly created as a result of the subsidy and in over half of the cases the participants would have also been hired in the absence of the subsidy. Therefore the finding that the participants of wage subsidy programs were 20 times more likely to find work than the control group probably massively overestimates their real effect.

Public works programs had a strong lock-in effect: the job finding rate dropped sharply after the end of the programs and nearly half of the participants were involved in the same public works program more than once.

The regression analyses suggest that training and wage subsidy programs had a direct positive effect (even conditional on the characteristics of the participants). The direct effect of public works is less favourable than the outcomes.
in the control group. These programs on the one hand cannot compensate for the unfavourable entry characteristics of participants and on the other hand, through their lock-in effect, they reduce the probability of open labour market integration.

The study also provided interesting information on the subjective status assessment and work motivation of job seekers. Participants of active labour market programs were more open, flexible and pro-active than those in the control group, while they were more attached to the more traditional forms of lawful employment. These characteristics can also influence the effectiveness of active labour market policies.

In conclusion, out of the active labour market programs public works definitely does not have a positive effect while the effect of training and particularly wage subsidy programs can be positive. The real effect of the program is more modest than the employment ratio after the end of the program indicates because the majority of participants would have been employed also in the absence of the subsidy and not more than two in three posts were newly created thanks to the wage subsidy program. The evaluation of active labour market programs was further complicated by the fact that there were significant differences between the participants and the control group that could not be fully taken into account in the analysis.

Appendix 4

Table 4A1: Factors influencing the probability of open employment.
Dependent variable: individual takes up work on the open labour market (rather than any other type of employment or status)

<table>
<thead>
<tr>
<th>Explanatory variable*</th>
<th>b</th>
<th>Standard error</th>
<th>Wald-test</th>
<th>Degree of freedom</th>
<th>Significance</th>
<th>Exp(b)</th>
</tr>
</thead>
<tbody>
<tr>
<td>ratio</td>
<td>-0.014</td>
<td>0.006</td>
<td>5.732</td>
<td>1</td>
<td>0.017</td>
<td>0.986</td>
</tr>
<tr>
<td>education</td>
<td>9.076</td>
<td>6</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>education(1)</td>
<td>0.056</td>
<td>0.508</td>
<td>0.012</td>
<td>1</td>
<td>0.912</td>
<td>1.058</td>
</tr>
<tr>
<td>education(2)</td>
<td>0.178</td>
<td>0.493</td>
<td>0.131</td>
<td>1</td>
<td>0.717</td>
<td>1.195</td>
</tr>
<tr>
<td>education(3)</td>
<td>0.054</td>
<td>0.52</td>
<td>0.011</td>
<td>1</td>
<td>0.918</td>
<td>1.055</td>
</tr>
<tr>
<td>education(4)</td>
<td>0.435</td>
<td>0.506</td>
<td>0.741</td>
<td>1</td>
<td>0.389</td>
<td>1.545</td>
</tr>
<tr>
<td>education(5)</td>
<td>0.163</td>
<td>0.585</td>
<td>0.077</td>
<td>1</td>
<td>0.781</td>
<td>1.176</td>
</tr>
<tr>
<td>education(6)</td>
<td>0.863</td>
<td>0.536</td>
<td>2.592</td>
<td>1</td>
<td>0.107</td>
<td>2.371</td>
</tr>
<tr>
<td>Development</td>
<td>4.971</td>
<td>4</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Development(1)</td>
<td>-0.363</td>
<td>0.265</td>
<td>1.876</td>
<td>1</td>
<td>0.171</td>
<td>0.696</td>
</tr>
<tr>
<td>Development(2)</td>
<td>0.27</td>
<td>0.294</td>
<td>0.841</td>
<td>1</td>
<td>0.359</td>
<td>1.31</td>
</tr>
<tr>
<td>Development(3)</td>
<td>-0.231</td>
<td>0.303</td>
<td>0.582</td>
<td>1</td>
<td>0.445</td>
<td>0.794</td>
</tr>
<tr>
<td>Development(4)</td>
<td>-0.13</td>
<td>0.301</td>
<td>0.186</td>
<td>1</td>
<td>0.666</td>
<td>0.878</td>
</tr>
<tr>
<td>status4</td>
<td>120.148</td>
<td>3</td>
<td>0</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>status4(training)</td>
<td>0.597</td>
<td>0.213</td>
<td>7.852</td>
<td>1</td>
<td>0.005</td>
<td>1.817</td>
</tr>
<tr>
<td>status4(wage subs)</td>
<td>3.008</td>
<td>0.39</td>
<td>59.402</td>
<td>1</td>
<td>0</td>
<td>20.243</td>
</tr>
<tr>
<td>status4(publ empl)</td>
<td>-1.324</td>
<td>0.336</td>
<td>15.517</td>
<td>1</td>
<td>0</td>
<td>0.266</td>
</tr>
<tr>
<td>Explanatory variable*</td>
<td>$b$</td>
<td>Standard error</td>
<td>Wald-test</td>
<td>Degree of freedom</td>
<td>Significance</td>
<td>Exp($b$)</td>
</tr>
<tr>
<td>-----------------------</td>
<td>-------</td>
<td>----------------</td>
<td>-----------</td>
<td>-------------------</td>
<td>--------------</td>
<td>----------</td>
</tr>
<tr>
<td>Age group</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Age group(1)</td>
<td>-0.335</td>
<td>0.276</td>
<td>5.833</td>
<td>4</td>
<td>0.212</td>
<td>0.715</td>
</tr>
<tr>
<td>Age group(2)</td>
<td>-0.396</td>
<td>0.311</td>
<td>1.481</td>
<td>1</td>
<td>0.224</td>
<td>0.763</td>
</tr>
<tr>
<td>Age group(3)</td>
<td>-0.656</td>
<td>0.331</td>
<td>3.925</td>
<td>1</td>
<td>0.048</td>
<td>0.519</td>
</tr>
<tr>
<td>Age group(4)</td>
<td>-0.946</td>
<td>0.439</td>
<td>4.651</td>
<td>1</td>
<td>0.031</td>
<td>0.388</td>
</tr>
<tr>
<td>Residence</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>residence(1)</td>
<td>0.46</td>
<td>0.308</td>
<td>2.223</td>
<td>1</td>
<td>0.136</td>
<td>1.583</td>
</tr>
<tr>
<td>residence(2)</td>
<td>0.416</td>
<td>0.319</td>
<td>1.709</td>
<td>1</td>
<td>0.191</td>
<td>1.516</td>
</tr>
<tr>
<td>Gender</td>
<td>0.229</td>
<td>0.175</td>
<td>1.709</td>
<td>1</td>
<td>0.191</td>
<td>1.257</td>
</tr>
<tr>
<td>d4 (marital stat)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>d4(1)</td>
<td>0.22</td>
<td>1.186</td>
<td>0.034</td>
<td>1</td>
<td>0.853</td>
<td>1.246</td>
</tr>
<tr>
<td>d4(2)</td>
<td>0.083</td>
<td>1.202</td>
<td>0.005</td>
<td>1</td>
<td>0.945</td>
<td>1.087</td>
</tr>
<tr>
<td>d4(3)</td>
<td>0.75</td>
<td>1.164</td>
<td>0.415</td>
<td>1</td>
<td>0.519</td>
<td>2.116</td>
</tr>
<tr>
<td>d4(4)</td>
<td>1.329</td>
<td>1.246</td>
<td>1.137</td>
<td>1</td>
<td>0.286</td>
<td>3.776</td>
</tr>
<tr>
<td>d4(5)</td>
<td>0.232</td>
<td>1.409</td>
<td>0.027</td>
<td>1</td>
<td>0.869</td>
<td>1.261</td>
</tr>
<tr>
<td>d4(6)</td>
<td>0.25</td>
<td>1.182</td>
<td>0.045</td>
<td>1</td>
<td>0.833</td>
<td>1.284</td>
</tr>
<tr>
<td>d4(7)</td>
<td>-0.105</td>
<td>1.266</td>
<td>0.007</td>
<td>1</td>
<td>0.934</td>
<td>0.901</td>
</tr>
<tr>
<td>d4(8)</td>
<td>0.109</td>
<td>1.278</td>
<td>0.007</td>
<td>1</td>
<td>0.932</td>
<td>1.116</td>
</tr>
<tr>
<td>u10 (Roma)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>u10(1)</td>
<td>0.175</td>
<td>0.332</td>
<td>0.277</td>
<td>1</td>
<td>0.599</td>
<td>1.191</td>
</tr>
<tr>
<td>u10(2)</td>
<td>-0.353</td>
<td>0.533</td>
<td>0.437</td>
<td>1</td>
<td>0.509</td>
<td>0.703</td>
</tr>
<tr>
<td>Open empl(1)</td>
<td>-0.522</td>
<td>0.312</td>
<td>2.793</td>
<td>1</td>
<td>0.095</td>
<td>0.593</td>
</tr>
<tr>
<td>Subsidised empl(1)</td>
<td>-0.355</td>
<td>0.428</td>
<td>0.686</td>
<td>1</td>
<td>0.407</td>
<td>0.701</td>
</tr>
<tr>
<td>unemployed(1)</td>
<td>0.436</td>
<td>0.314</td>
<td>1.925</td>
<td>1</td>
<td>0.165</td>
<td>1.547</td>
</tr>
<tr>
<td>Unemployment benefit(1)</td>
<td>0.127</td>
<td>0.273</td>
<td>0.216</td>
<td>1</td>
<td>0.642</td>
<td>1.135</td>
</tr>
<tr>
<td>Child care allowance(1)</td>
<td>0.89</td>
<td>0.825</td>
<td>1.165</td>
<td>1</td>
<td>0.28</td>
<td>2.435</td>
</tr>
<tr>
<td>Constant</td>
<td>-3.111</td>
<td>1.886</td>
<td>2.72</td>
<td>1</td>
<td>0.099</td>
<td>0.045</td>
</tr>
</tbody>
</table>

*Explanatory variables included in the analysis:
ratio: ratio of the duration of unemployment within total work history
Education(.): respondent's highest education level;
Development(.): development level of the small region;
status4(.): respondent's status at time of intervention (training, wage subsidy, public works);
age group(.): respondent's age;
residence(.): respondent's place of residence;
gender: respondent's gender;
d4 (marital stat)(.): respondent's marital status;
u10(.): respondent's Roma ethnic background;
open empl(1): employment status at the beginning of the observation period: previously in open employment;
subsidised empl(1): employment status at the beginning of the observation period: previously in subsidised employment;
unemployed(1): employment status at the beginning of the observation period: previously unemployed without receiving any unemployment-related benefit;
unemployment benefit(1): employment status at the beginning of the observation period: previously unemployed, receiving unemployment-related benefit;
child care allowance(1): employment status at the beginning of the observation period: maternity leave
5. THE IMPACT OF THE EXPANSION OF PUBLIC WORKS PROGRAMS ON LONG-TERM UNEMPLOYMENT

JÁNOS KÖLLŐ & ÁGOTA SCHARLE

The Road to Work Program (“Út a Munkához Program”) was launched in 2009 with the objective of providing income and an employment opportunity to unemployed low education level workers living in the most disadvantaged small settlements, or in other words to provide work incentives and improve the employability of the long-term unemployed. This study examines the impact of public work programs on the employment opportunities of the long-term unemployed before the launch of Road to Work Program. As will be shown later, Road to Work Program was virtually the extension of existing public works programs therefore the same effects can be expected as those presented here.

After an overview of the labour market and institutional context, the study will present the development of public works in the 2000s and the Road to Work Program in more detail. It will address the allocation of additional resources to different groups of workers (targeting) and the number of unemployed workers and local councils – including disadvantaged micro-regions – that were reached by the program (take-up). A separate section will discuss the experiences of public works programs in Hungary. The last section will focus on the key question of the study, the impact of public works programs between 2003 and 2008.

The study is based mainly on administrative data and analyses the impact of public works programs on the development of long-term unemployment at the settlement-level. This chapter is a substantially shortened version of the full research report prepared by the Budapest Institute and Hétfő Research Institute (Budapest Intézet, 2011).

Labour market and institutional context

Unemployment rose very rapidly during the early years of regime change. Long-term unemployment rose steadily within the working age population and stood at just over three per cent by the 1990s. Over half of the unemployed took longer than a year to find employment. Both the number of short- and long-term unemployed was declining until 2001, but then started growing again. The relative ratio of short- and long-term unemployment remained relatively stable until 2008 when the crisis started to deepen. Since then the ratio of short-term unemployment has somewhat increased. The rate of long-term unemployment (within the working age population) again approached three per cent in 2009 (Figure 5.1).
The majority of the long-term unemployed are older workers, low-educated workers and people living in remote settlements with poor public transportation links due to cuts in transport services (Galasi-Nagy, 1999, Köllő, 2009). Consistently low employment levels can be linked to a number of issues – inherited economic structure, the post-communist transition, demographic trends and government policies might have been contributing factors. The global crisis in 2008 made things worse but internal structural problems and processes that had originated much earlier are a more likely explanation for the low employment level. It is not surprising that the rapidly changing economy during the years of the post-communist transition did not offer many employment opportunities to masses of the low-educated workforce. However, persistent long-term unemployment is not an inherent part of the market economy – it is explained by the characteristics of Central East European economies. One of the contributing factors is most probably the weak education system (public and vocational education) that is responsible for the low adaptability of the workforce (Commander and Köllő, 2008, Kézdi, Köllő and Varga, 2009). Other important factors include the lack of small and family businesses and the extensive red tape that hinders their expansion. In Western Europe the employment rate of the low-educated is similar to that of graduates, but they are typically employed by small businesses. This sector was disrupted by the socialist economies and cannot be restored overnight – even in Poland where the SME sector is growing rapidly thanks to the capital investments of returning migrants, it falls behind the extent of the SME sector in similarly developed South American economies (Maloney, 2004). In addition, the Hungarian regulatory environment is not particularly helpful in fostering the long and difficult process of small business development.
Economic and employment policy decisions with a direct effect on labour supply and demand also contributed to stagnating employment. There is an unhealthy balance on the labour market that is maintained by the mutually reinforcing effects of low employment, welfare policies, high taxes and low tax revenues. As a result of the generous welfare policies designed to mitigate tensions in the early years of regime change, currently the main source of income for nearly one fourth of the working age population is some form of welfare payment with the majority of recipients being economically inactive and absenting themselves from the labour market for a long period or permanently (Cseres-Gergely and Scharle, 2008). The ex-post evaluation of the increase to the minimum wage in 2001–2002 showed that it had no significant impact on labour supply, however it clearly reduced employment in labour-intensive sectors (Benedek et al., 2006, Kertesi and Köllő 2004). In some areas it is not the excessive government intervention but rather inaction that hinders employment. These areas include the reluctance to combat gender and ethnic discrimination and delaying the adoption of public health policies to improve the health status of the population. Finally, based on the rather sporadic data and empirical studies active and passive labour market policies aimed at reducing unemployment did not prove effective in increasing employment (Bódis et al., 2005, Bódis and Nagy, 2008, Cseres-Gergely and Scharle, 2010, Fazekas, 2001, Csák, 2007, Nagy, 2008).

The system of unemployment assistance, including both the insurance-based and time-limited unemployment benefit and the means-tested social assistance paid to the long-term unemployed, underwent numerous changes over the past 20 years (for more on this see Budapest Intézet, 2011, and Duman and Scharle, 2010). There are a number of conditions attached to the payment of unemployment assistance in most countries, including Hungary. As part of a workfare reform in 2000 a new rule was introduced that required the long-term unemployed claimants of social welfare assistance to take up public works for up to 30 days before the assistance could be paid. The only exception was when the local council or job centre was unable to organise public works programs. Claimants were required to cooperate with the local council or a designated service (typically the family counselling service or the job centre), sign on as unemployed, visit their advisor on a regular basis, report any changes that may affect eligibility and take part in employability programs.

The reform in 2000 expanded the range of activities that could be carried out in public works programs and allocated central government subsidies for financing public works. The new rules were based on the principles of workfare and aimed to provide “work instead of benefits”. It was hoped that the new system would help to reduce benefit fraud and reduce long-term unemployment (Fazekas 2001, Duman and Scharle, 2011). The focus of this study, the Road to Work Program, is not substantively different from the system created
in 2000 in terms of its objectives or main components. The difference was a substantial increase of the budget available for public works programs managed by local councils.

The Road to Work Program

The Road to Work Program was launched by the Hungarian government in 2008. The main objectives of the program were to improve the labour market situation of benefit claimants, reduce work disincentives of the benefit system and increase employment level (Szűcs, 2009). It also aimed to create public sector jobs to provide work opportunities for the long-term and improve joint working between social and employment services. Some further benefits of the program were also envisaged by the Government, such as reducing the number of working-age people claiming social assistance, improving the time-use of job centres, enhancing the efficiency of partnerships between local councils, family counselling services and the public employment service, increasing the number of work-ready jobseekers, better targeting of employment assistance and improving the employment situation in the most disadvantaged small regions.

The Road to Work Program was introduced in two phases in 2009. Social assistance claimants were allocated into two groups: those able to work and those not able to work. The first group would either participate in public works or receive income replacement. The central government substantially increased the budget available to local councils for public works programs. There were also further incentives for local councils to make the most of this: the rate of government co-financing was 95% for public works programs but only 80% for income replacement (reduced from a previous level of 90%). The income replacement is a fixed-sum payment and somewhat less than the average amount of social assistance. This rule entered into force in January 2010, thus in 2009 all claimants were receiving the regular social assistance according to the previous regulation (Csoba, 2010a; Frey, 2010).

The Road to Work Program limited the possibilities of local councils because it no longer allowed the involvement of a broad range of long-term unemployed only those receiving the income replacement. At the same the available budget increased. Previously the budget was allocated by the ministry responsible for local councils based on the needs forecasted by local councils at the beginning of each year. On the contrary the Road to Work Program was based on normative post-financing which meant that the State Treasury reimbursed 95% of the expenses related to the employment of public workers to local councils at the end of each month (they received a 50% discount on contributions paid by employers). For related program components and details on management see Budapest Intézet (2011). The rapid expansion of the Road to Work Program is illustrated by Figure 5.2. The figure clearly shows that the total number of participants increased substantially by the end of the year.
Figure 5.2: Introduction of the Road to Work Program – number of people claiming social assistance, income replacement and participants of public works programs, monthly average, 2009, thousand people

Source: Regular social welfare assistance and income replacement claimants Employment Office (claimant count on the closing day); municipal public works participants: Ministry of National Resources.

Figure 5.3 shows that there was no clear upward trend in the number of registered unemployed prior to the launch of the Road to Work Program: the number of claimants and registered unemployed only started to increase as a result of the global crisis. Nonetheless the number of long-term unemployed receiving assistance already showed an upward trend throughout 2008. There have been three kinds of public works programs in operation in Hungary: the “municipal” public works programs, the “communal” public works programs and the public works programs organized by the National Employment Service (ÁFSZ) program.

Figure 5.3: Average headcount of unemployment assistance claimants and public works participants

According to available administrative data on average 14,000 to 16,000 people were employed in one of the three types of public works programs each year before to the introduction of Road to Work. Thus each program provided employment to approximately 10–12% of the long-term unemployed (20–22% in total). In the first year of the Road to Work program, the average headcount in communal public works jumped to 60,000 (while it shrank to 5,000 in municipal public works) that made up 27% of income replacement claimants. Together with those in municipal public works the number of workers increased by 50% compared to the previous year. It should be noted that the level of employment in public works was only extraordinary in a sense that local councils have never been allocated so much funding for this from the central government’s budget. The level of employment measured as the percentage of benefit claimants was similarly high in 2003 as well: then 15% of benefit claimants were employed in municipal public works and a further 10% in the community (Figure 5.4).

**Figure 5.4: Ratio of participants in municipal and communal public works within the total eligible benefit claimants, 1996–2009**

Source: Average headcount in October for municipal public works (Labour Market Review, 2010, Table 5.13). Average annual headcount for communal public works (Fazekas, 2001; based on data provided by the National Audit Office (2007) and the Hungarian State Treasury, 2001 and 2007 figures estimated on the basis of data provided by the State Treasury, 2008 figure calculated using settlement data on the number of days eligible for reimbursement. Both data sets are expressed as the ratio of eligible benefit claimants (income replacement assistance, social welfare assistance and income replacement) using the average October headcount (The Hungarian Labour Market, 2010, Table 5.13): ratio of municipal public works participants/(municipal + community + benefit claimants); ratio of communal public works participants/(community + municipal + benefit claimants).

Regional disparities in targeting and take-up

Considering the regulatory background of the Road to Work Program, we can expect that its targeting is probably adequate and it reaches the most disadvantaged long-term unemployed with the worst employment prospects. However, it might create problems if local councils decided to employ more educated jobseek-
ers who are more likely to have the necessary skills, or if due to inadequate management capacities programs fail to reach those living in remote villages with no jobs. Uptake at the national level is very high, but due to the decentralised nature of the program implementation, there might be major regional disparities.

Table 5.1 shows the number of settlements that had municipal public works programs between 2003 and 2009. After the introduction of the state subsidy in 2000 it took a few years for the program to take off. While in 2003 less than half of the local councils arranged public works, in 2005 this was more than 85%. This then remained by-and-large unchanged until 2008 and increased slightly after the launch of the Road to Work Program. In 2009 nearly all local councils had public works programs. A more detailed analysis shows that uptake increased most by local councils with high long-term unemployment and where there was a history of public works (Budapest Institute, 2011). In localities further away from the capital the number of participants in public works programs increased more, however the most remote villages were less able to take advantage of the increasing funding opportunities. There was no significant relationship between the percentage of the Roma population within the settlement and changes in the uptake.

Table 5.1: Share of local councils that organise public works programs, 2003–2009

<table>
<thead>
<tr>
<th>Local council</th>
<th>2003</th>
<th>2005</th>
<th>2007</th>
<th>2008</th>
<th>2009</th>
</tr>
</thead>
<tbody>
<tr>
<td>Village with a population of less than 50</td>
<td>0.00</td>
<td>0.53</td>
<td>0.53</td>
<td>0.67</td>
<td>0.84</td>
</tr>
<tr>
<td>Village with a population of 50–99</td>
<td>0.13</td>
<td>0.73</td>
<td>0.73</td>
<td>0.80</td>
<td>0.91</td>
</tr>
<tr>
<td>Village with a population of 100–149</td>
<td>0.14</td>
<td>0.81</td>
<td>0.81</td>
<td>0.86</td>
<td>0.91</td>
</tr>
<tr>
<td>Village with a population of 150–499</td>
<td>0.25</td>
<td>0.85</td>
<td>0.85</td>
<td>0.87</td>
<td>0.96</td>
</tr>
<tr>
<td>Village with a population of 500–4,999</td>
<td>0.44</td>
<td>0.87</td>
<td>0.87</td>
<td>0.88</td>
<td>0.97</td>
</tr>
<tr>
<td>Town / village with a population of 5,000–9,999</td>
<td>0.71</td>
<td>0.92</td>
<td>0.92</td>
<td>0.92</td>
<td>0.97</td>
</tr>
<tr>
<td>Town with a population of 10,000–19,999</td>
<td>0.70</td>
<td>0.84</td>
<td>0.84</td>
<td>0.84</td>
<td>0.99</td>
</tr>
<tr>
<td>Town with a population of 20,000–49,999</td>
<td>0.71</td>
<td>0.98</td>
<td>0.98</td>
<td>0.98</td>
<td>1.00</td>
</tr>
<tr>
<td>Town with a population of over 50,000</td>
<td>0.90</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
<td>1.00</td>
</tr>
<tr>
<td>Total local councils</td>
<td>0.39</td>
<td>0.85</td>
<td>0.85</td>
<td>0.87</td>
<td>0.96</td>
</tr>
</tbody>
</table>

Note: Villages with a population of less than 50 inhabitants make up nearly one third of all municipalities, and villages with 500 to 5,000 inhabitants make up nearly two thirds (without population weights).
Source: own calculations based on data from the Hungarian State Treasury and Tstar.

The targeting of the Road to Work Program was analysed using quasi-panel monthly data generated from individual-level data held by the Employment Office. The data-set was generated using individual data and a time-series of group-level observations; “real” panel data follows individuals over time, however the data did not allow this type of analysis. Our groups were based on gender, level of education and age group.

Our regression model analysed the relationship between take-up and other characteristics within groups and the month of observation. The results in Ta-
ble 5.2 show that the Road to Work Program did reach its target groups at the individual level: low educated job seekers were more likely to participate. Age had no significant effect on take-up aside from the fact that in the age group under 25 years, involvement was significantly more likely. This might not be a good thing if they do not gain work experience in public works that can also be useful elsewhere but they have less time and motivation to look for work on the open labour market.

Table 5.2: Probability of public works take-up (social assistance recipients from the register of Employment Office, 2009)

<table>
<thead>
<tr>
<th>Education Level</th>
<th>Co-efficient</th>
<th>Standard error</th>
</tr>
</thead>
<tbody>
<tr>
<td>Less than 8 years of general education and aged over 35 years</td>
<td>-0.003</td>
<td>0.0023</td>
</tr>
<tr>
<td>Less than 8 years of general education</td>
<td>0.002</td>
<td>0.0019</td>
</tr>
<tr>
<td>Eight years of general education</td>
<td>0.006***</td>
<td>0.0013</td>
</tr>
<tr>
<td>Lower secondary</td>
<td>0.005***</td>
<td>0.0011</td>
</tr>
<tr>
<td>Upper secondary</td>
<td>0.002**</td>
<td>0.0010</td>
</tr>
<tr>
<td>Education level not known</td>
<td>-0.054***</td>
<td>0.0015</td>
</tr>
<tr>
<td>Under 25</td>
<td>0.003**</td>
<td>0.0011</td>
</tr>
<tr>
<td>25–29 years</td>
<td>0.001</td>
<td>0.0011</td>
</tr>
<tr>
<td>30–34 years</td>
<td>0.000</td>
<td>0.0011</td>
</tr>
<tr>
<td>Over 55</td>
<td>0.001</td>
<td>0.0011</td>
</tr>
<tr>
<td>January – March</td>
<td>-0.002</td>
<td>0.0014</td>
</tr>
<tr>
<td>May</td>
<td>0.063***</td>
<td>0.0017</td>
</tr>
<tr>
<td>June</td>
<td>0.088***</td>
<td>0.0017</td>
</tr>
<tr>
<td>July</td>
<td>0.050***</td>
<td>0.0017</td>
</tr>
<tr>
<td>August</td>
<td>0.053***</td>
<td>0.0017</td>
</tr>
<tr>
<td>September</td>
<td>0.065**</td>
<td>0.0017</td>
</tr>
<tr>
<td>October</td>
<td>0.050***</td>
<td>0.0017</td>
</tr>
<tr>
<td>November</td>
<td>0.035***</td>
<td>0.0017</td>
</tr>
<tr>
<td>December</td>
<td>0.020***</td>
<td>0.0017</td>
</tr>
<tr>
<td>Constant</td>
<td>-0.002</td>
<td>0.0015</td>
</tr>
</tbody>
</table>

Source: own calculations based on data from the Employment Office.

The experiences of public works programs in Hungary before 2008

In a review of the international experiences of the effectiveness of active labour market policies Kluve (2010) argues that public works programs are rarely effective and in some cases they can be detrimental for future employment prospects – if they lack even a minimal positive impact that could compensate for the reduced time available for job search by job seekers (see also Chapter 2 of In Focus).

Our understanding of the effectiveness of active labour market policies in Hungary is limited. The validity of program evaluations has been challenged by the fact that participants of public works programs are a special group and it is very difficult to find adequate control groups to measure the effect of non-intervention (on the importance of this see for example Chapter 1 of In Focus).
There have been no experimental studies and the validity of non-experimental studies can often be questionable. Nevertheless this section will review research on the effectiveness of public works programs before 2008. There seems to be a general agreement in the literature that public works failed to achieve its main objectives and improve the job prospects of workers in public works programs.

Many reports present only unadjusted job-finding rates. Although these data are informative, they do not allow us to separate the program’s real effects from the participant and context effects. The ÁFSZ monitoring studies measure the job-finding rate three months after the end of the main programs. These studies in 2007 found that after training nearly 40% and after participation in wage subsidy schemes, over 60% of jobseekers found work (ÁFSZ, 2007). Frey (2008) argued that the job-finding rate among participants of public works programs was virtually zero (0.1–0.3 per cent) in counties with high unemployment, and even in more developed counties this rate was around 5% between 2001 and 2005. Csoba (2010a) showed that the exit rate from long-term unemployment as a result of finding a job was only 1.4% in 2009.

Galasi, Lázár and Nagy (1999), in their study that also takes into account selection, found a weak positive effect of training and a negative effect of public works on job finding. Galasi, Lázár and Nagy (2003) showed that public works participants were less likely to find a job than participants in other active labour market programs, but the difference was partly explained by participant characteristics. Fazekas (2001) studied the first results of the introduction of public works using data recorded by local councils, interviews and data from administrative sources in 2001. He found that the number of participants in municipal public works fell short of expectations and job finding among the long-term unemployed did not increase. Nevertheless the program provided an opportunity to offer employment for people who were not eligible for unemployment assistance. According to Fazekas (2001) 11.4% of those claiming regular social assistance were recruited to public works using this option between May 1, 2000 and October 31, 2001.

The National Audit Office investigated the use of resources allocated for public works in 2002 and 2007. In 2002 18% of the financial resources were included in the investigation using site visits and data from the unemployment register. They concluded that there was a lack of coordination between the different types of public works and their efficiency was rather poor. There was limited use of employment plans to facilitate return to the open labour market and public works programs were largely led by the short-term financial interests and workforce needs of local councils. In terms of the allocation of resources, no consideration was given to efficiency and all program proposals received funding until the budget was exhausted. The second report from 2007 was based on site visits between 2003 and 2006 and a survey on the perception of public works. This report reiterated the conclusions of the previous
report and also highlighted deficiencies in the monitoring and evaluation of programs whether they delivered the planned outcomes (employment, education etc.). It also concluded that public works failed to considerably improve the employment prospects of participants. The survey however, found that public works were socially accepted and the jobs carried out were considered useful by the majority of respondents.

Bódis and Nagy (2008) suggested that the differences in the administration of benefits between local councils persisted based on a survey carried out in the summer of 2007. The survey examined the assessment of eligibility and considered whether claimants were offered the possibility to take part in public works at 44 municipalities and local job centres. Eighty per cent of participating local councils designated the employment service to monitor the compliance of claimants, in the rest of the cases the designated authority was the family counselling service but benefit claimants were still required to cooperate with the public employment service too. The findings of this study, similarly to some earlier case studies, suggested that, in the absence of incentives and professional inspection, adherence to policies was very uneven: authorities involved in the administration of benefits could create their own rules and based on personal characteristics, their legal and human rights awareness and local interests decided on the level of cooperation expected from benefit claimants and the use of sanctions (Bódis and Nagy, 2008, Szalai, 2004–2005).

Firle and Szabó (2007) examined the job finding rate among benefit claimants and public works participants based on individual panel data from the CSO’s Labour Force Survey between 2001 and 2004. They found that public works participants were six to nine per cent less likely to find (non-subsidised) employment in the following quarter than a comparable group – in terms of age, education level and family situation – of unemployed people. To date this is the only research using a large sample to estimate the individual-level effect of public works in Hungary. Similarly to other program evaluations discussed here, the results of Firle and Szabó (2007) are more informative than a simple comparison of exit rates, however it does not necessarily reflect the real effect of the program because non-experimental studies usually need to control for a wider range of variables (for example long-term employment history).

The effect of public works on job finding – a settlement-level analysis

Based on international experiences and program results in Hungary so far we can hypothesise that public works does not contribute to the reduction of long-term unemployment in the short run. Longer term effects are also unlikely to be positive, but we cannot assess those, due to the relatively recent nature of these programs.
Before the settlement-level analysis, it is worth considering national rates. According to the CSO’s Labour Force Survey very few workers – approximately one to five per cent – with short fixed-term contracts in any type of public works program (municipal, communal and national) find unsubsidised employment from one quarter to the next (Figure 5.5). Approximately 20–40% of those employed in public works in a given quarter will (again) become registered unemployed in the following quarter.

**Figure 5.5: Exit rate from subsidised employment compared to baseline rates**

*by labour market status*

Note: The flows are consistent with the population flows and do not include new entrants.

Source: Calculations by Zsombor Cseres-Gergely based on the CSO Labour Force Survey (Budapest Institute, 2011).

Our study measures the effect of public works programs implemented between 2003 and 2008 using a settlement-level analysis. It seeks to answer the question of whether long-term unemployment declined more (increased less) in settlements where there was an expansion of public works programs (relative to long-term unemployment) than in settlements where public works did not expand or expanded less. If public works programs really improve participants’ employment prospects then we should find a negative relationship, namely with the expansion of public works we should see a decline in long-term unemployment with time.

Impact evaluations of employment policies generally compare participants’ employment odds with the counterfactual odds, notably those that would have characterised them in the absence of the program. The counterfactual odds are usually measured with a carefully selected control group. The comparison is typically between individuals in the intervention group and individuals in the control group. In this case, to measure the effect of public works we should consider the long term employment odds of program participants with the employment odds of long-term unemployed people who were not involved in
the program but are similar to participants in all other aspects. To achieve an adequate sample size and to measure participation more accurately the participant and the control group should be selected from the register of the Employment Office. However it was not possible to obtain individual data in time for the analysis and the measurement issues of individual data would have posed more difficulties. Therefore we decided to do settlement-level analysis instead of an individual analysis.

Unemployment data was obtained from the central register of the Employment Office, data on program participation came from the municipal public works database of the State Treasury. The analysis also used the CSO Labour Force Survey as well as the CSO’s settlement-level indicators database (expanded by the IE HAS). The settlement-level estimates were based on a panel dataset generated from these, in which each observation represents yearly data from settlements in Hungary. The data necessary for the analysis were only available for the years 2003 to 2008, therefore all estimates are for this period.

Using the panel datasets we employed panel regressions to estimate the relationship between trends in public works and long-term unemployment. The first type of regression tested the levels with fixed settlement and year effects; the second type of regression analysed changes between years. The two regression formulas are as follows:

\[
LU_{it} = \beta P_{it} + \alpha SU_{i,t-1} + \gamma X_{it} + c_i + d_t + u_{it}
\]

\[
\Delta LU_{it} = \beta \Delta P_{it} + \alpha \Delta SU_{i,t-1} + \gamma \Delta X_{it} + \Delta d_t + \nu_{it},
\]

where:
- \(i\) is the settlement,
- \(t\) is the year of observation,
- \(LU\) is the rate of long-term unemployment within the working age population,
- \(SU\) is the rate of short-term unemployment within the working age population.
- \(P\) is the share of public works participants within the total number of public works participants and social assistance recipients.
- \(X\) is the vector for settlement-level control variables,
- \(c_i\) denotes fixed effects of the settlement,
- \(d_t\) denotes fixed effects of the given year,
- \(u_{it}\) denotes the unobserved time-varying heterogeneity within settlements,
- \(v_{it}\) is the equivalent of this in the differential equation.

Both \(u_{it}\) and \(v_{it}\) can be auto-correlated, therefore in the basic models we estimated clustered standard errors and we also repeated all estimations using generalised least squares that takes autocorrelations into account to estimate parameters. Both regressions were also estimated with the lagged rate of public works (\(P_{i,t-1}\)) instead of the real time rate (\(P_{it}\)) in the right side of the equation.
The rate of long-term unemployment was defined in our analysis as the total number of jobseekers who had been registered for 180 days or longer and participants in public works, divided by the total working age population. The rate of short-term unemployment is the total number of public works participants and jobseekers registered for less than 180 days divided by the total number of the working age population. The rate of public works participants was calculated dividing the number of public works participants with the total number of social assistance recipients (income replacement assistance) and public works participants.

These regression equations can also be seen as variations of the difference in differences (DiD) technique: we compared long-term unemployment rates of settlements along the expansion of public works. However, while in DiD models there is a pre- and post treatment measurement, in our model we compare data from settlements with different levels of public works expansion. The comparison is based on changes in participation rates over time.

The regression outputs with the estimated coefficients are presented in Table 5.3 (see next page). The results clearly show that public works do not reduce long term unemployment and they may even increase it slightly in the short term. The results of the equations estimating the short-term relationships show that if the share of public works participants among the total unemployed increases by one per cent in a given settlement then the rate of long-term unemployment in the working age population increases by 0.016 percentage points within the same year. The results of the models estimating lagged effects are more ambiguous; depending on the parameters this coefficient is not significant (zero), positive or negative. Considering that all coefficients are very small, it can be concluded that increased participation in public works does not lead to a perceptible decline in long-term unemployment in subsequent years either.
Table 5.3: Municipal public works and long-term unemployment in Hungarian settlements, regression estimates (2002–2008)

<table>
<thead>
<tr>
<th></th>
<th>Estimated on levels</th>
<th>Estimated on differences</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>given year</td>
<td>previous year</td>
</tr>
<tr>
<td><strong>Baseline model</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Public works ($P_{it}$)</td>
<td>0.0098***</td>
<td>–</td>
</tr>
<tr>
<td></td>
<td>(0.0018)</td>
<td></td>
</tr>
<tr>
<td>Public works, lagged ($P_{it-1}$)</td>
<td>–</td>
<td>0.0045***</td>
</tr>
<tr>
<td></td>
<td>(0.0022)</td>
<td></td>
</tr>
<tr>
<td>Short-term unemployment, lagged ($SU_{it-1}$)</td>
<td>0.1217***</td>
<td>0.1252</td>
</tr>
<tr>
<td></td>
<td>(0.0098)</td>
<td>(0.0116)</td>
</tr>
<tr>
<td>Other explanatory variables ($X_{it}$)</td>
<td>YES</td>
<td>YES</td>
</tr>
<tr>
<td>Settlement fixed effects</td>
<td>YES</td>
<td>YES</td>
</tr>
<tr>
<td>Year fixed effects</td>
<td>YES</td>
<td>YES</td>
</tr>
<tr>
<td>Autoregression ($\rho$)</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Observations</td>
<td>16,610</td>
<td>12,726</td>
</tr>
<tr>
<td>Number of settlements</td>
<td>2,994</td>
<td>2,878</td>
</tr>
<tr>
<td>Number of years</td>
<td>6</td>
<td>5</td>
</tr>
<tr>
<td><strong>Model incorporating autocorrelation</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Public works ($P_{it}$)</td>
<td>0.0169***</td>
<td>–</td>
</tr>
<tr>
<td></td>
<td>(0.0026)</td>
<td></td>
</tr>
<tr>
<td>Public works, lagged ($P_{it-1}$)</td>
<td>–</td>
<td>0.0028</td>
</tr>
<tr>
<td></td>
<td>(0.0026)</td>
<td></td>
</tr>
<tr>
<td>Short-term unemployment, lagged ($SU_{it-1}$)</td>
<td>0.1833***</td>
<td>0.1842***</td>
</tr>
<tr>
<td></td>
<td>(0.0102)</td>
<td>(0.0121)</td>
</tr>
<tr>
<td>Other explanatory variables ($X_{it}$)</td>
<td>YES</td>
<td>YES</td>
</tr>
<tr>
<td>Settlement fixed effects</td>
<td>YES</td>
<td>YES</td>
</tr>
<tr>
<td>Year fixed effects</td>
<td>YES</td>
<td>YES</td>
</tr>
<tr>
<td>Autoregression ($\rho$)</td>
<td>0.3787</td>
<td>0.3412</td>
</tr>
<tr>
<td>Observations</td>
<td>13,116</td>
<td>9,848</td>
</tr>
<tr>
<td>Number of settlements</td>
<td>2,904</td>
<td>2,732</td>
</tr>
<tr>
<td>Number of years</td>
<td>5</td>
<td>4</td>
</tr>
</tbody>
</table>

*a The number of public workers divided by the total number of public workers and benefit recipients.

*b Number of those who have been registered unemployed for less than 12 months divided by the total number of working age people.


Dependent variable: Number of those who have been registered unemployed for 12 months or over divided by the total number of working age people.

Standard errors in parentheses. ** Significant at the 10 per cent level, *** five per cent level, *** one per cent level.
Conclusion

The results suggest that the *Road to Work Program* reached the most disadvantaged long-term unemployed with the worst employment prospects. Although occasionally local councils tried to compensate for the sharp drop in their resources with the *Road to Work Program*, this did not worsen the targeting significantly – this would have been indicated by the increased involvement of more capable and educated public workers.

The *Road to Work Program* had a very high take-up and it significantly increased in 2009. In the year of its introduction nearly all settlements and over a quarter of the eligible long-term unemployed participated in public works for some time. The program grew faster than expected probably as a result of the uncapped budget and the total absence of professional control. Local councils could use the additional resources provided by the *Road to Work Program* within broad limits and with favourable conditions to expand public services or reduce wage costs. There were no inbuilt mechanisms that would have limited the increase in spending either to ensure fiscal prudence or to allocate funding for other employment policy objectives.

The *Road to Work Program* has not been evaluated yet, but its effects will probably be similar to the public works programs of previous years. The literature and our analysis of settlement-level data clearly show that public works programs in Hungary did not bring about a reduction in long-term unemployment. It seems evident that public works schemes do not improve the employment prospects of participants. Therefore it is uncertain how the *Road to Work Program* could achieve its primary objective of improving the labour market situation of benefit claimants and thus increasing the level of employment.
6. The Implementation of a Complex Labour Market Program and Its Local Effects in the South-Transdanubian Region

Gergely Kabai & Nándor Németh

Introduction

The South-Transdanubian region (apart from a few urban areas) is an area lagging behind on a range of economic and social indicators. One of the most important indicators is the high rate of long-term unemployment in the hamlets that are characteristic settlements of this area. To address long-term unemployment in these small settlements the (then) South-Transdanubian Regional Employment Service (STRES) launched the complex labour market program *Life Changing – Life Shaping* (in Hungarian: *Sorsfordító – sorsformáló*) in March 2009. Initially the program was implemented in six localities in Tolna County, but later it was expanded to Somogy and Baranya counties and currently it covers over 30 localities. The participants in the *Life Changing Program*, over 200 former long-term unemployed, received training and were employed mainly in vegetable and fruit growing. They are employed by local councils, cooperatives and agricultural businesses. “Life shaping” was a work experience scheme using wage subsidies to support graduates with an agriculture-related degree.

The *Life Changing – Life Shaping* Program is a complex labour market program.1 Its main characteristics are that it is a long term scheme (it runs for two years or longer) and it can offer assistance using the whole range of active labour market policies including paid training, employment and mentoring.

This study presents the *Life Changing – Life Shaping* labour market program, highlights its good practices, compares it with other similar initiatives and also uncovers some of its weaknesses. The second part of the study highlights the results of the program using concrete examples from participating settlements and also considers what effects could be contributed to the program.

The study is based on a research project in which we interviewed staff from the relevant employment services who were responsible for coordinating the program, colleagues at Diófa Consortium who were responsible for the management and implementation of the program and other stakeholders. The interviews aimed to explore the weakness of the program and any issues with management and implementation, and also future opportunities and plans. We also reviewed – mainly publicly available – documents to collect information about the program and its implementation. Furthermore, we followed the implementation of the program for nearly 18 months: we participated (and made a presentation) at the evaluation conference in the spring of 2011 and half a dozen other program events including meetings, workshops and prod-

---

1 See the *Introduction* and Chapter 2 of *In Focus* on the types of active labour market programs in general and on complex programs in particular.
uct launch events. These provided valuable information that would have been impossible to obtain from other sources.²

**Background of the Life Changing – Life Shaping Program**

The program was officially launched in March 2009; however preparations had started much earlier. The key individuals of the programs had been discussing and planning an innovative program specially designed to address the employment problems and challenges of the South-Transdanubian hamlets. The long preparatory phase was also necessary because some of the staff of the regional employment services questioned the feasibility of the program; they doubted whether the long-term unemployed who already had lost some of their work capacity and had no or little previous agricultural experience would be suitable participants for an agricultural/horticultural project. Finally the director of STRES was won over to support the program in 2008 and he also persuaded his colleagues who initially had misgivings about the initiative.

Two forerunners of the Life Changing Program should also be mentioned here that had a key role in shaping the scheme. Successful, although slightly different horticultural programs had been running for more than 10 years in two localities of Tolna County: Belecska (Tamási Small Region) and Kisvejke (Bonyhád Small Region). In Belecska vegetable and fruit growing is part of a social employment program organised by the local council. The example of Belecska demonstrated to the organisers of the Life Changing Program that low educated long-term unemployed people can be successfully employed in horticulture and it might be worth expanding this. In the Kisvejke area there are a number of farmers who had achieved considerable international market share through their sale cooperative and are realising high profits. The example of Kisvejke demonstrated that fruit growing is viable in the region and with adequate knowledge this can be really successful. Therefore the first phase of the Life Changing Program was implemented around these settlements where it was less risky to test the feasibility of the ideas in practice.

A direct predecessor of the scheme was the Herbal Network project financed from Interreg in the South-Transdanubia Region in 2006–2007. The project, jointly implemented by partners from Croatia and Hungary, aimed to set up a business cluster to collect, produce and distribute medicinal herbs that would become a major source of jobs for people living in these disadvantaged rural areas. One of the aims was to create an economic development and agricultural integration model. In the end, the project which also provided training and mentoring, had limited results, but it convinced SRES staff and other stakeholders about the viability of a job creation project based on the region’s agricultural heritage. The results also inspired the design of the Life Changing Program.³

By the second half of 2008 the basis of the Life Changing Program was laid down. The aim was to create a labour market and employment program that

---

² This study is based on Kabai (2010).
³ For more information on the Herbal Network Program see: http://www.bmmk.hu/herbal-netw/interregnyito.html
was more sustainable than previous initiatives, addressed local challenges and used local resources to tackle the complex issue of rural job creation.

Program structure and participants

The long term objective of the *Life Changing* Program was to provide jobs for people living in rural areas who had been excluded from the labour market, and equip them with the necessary skills to search and take up employment independently. The training provided as part of the scheme would give them the necessary knowledge and experience to earn their living or additional income from growing and processing agricultural produce.

The structure of the program is simple. The long-term unemployed take part in a 1000-hour agricultural training program and then do work experience in a local council scheme or with local producers for one or two years. The program provides a wage subsidy (but its rate might vary according to the phase and employer). As far as possible the participants also undergo personal development to create an adequate attitude to work. The aim is that following the program as many of them as possible find a job on the primary labour market or convert their work experience job into long-term employment. The essence of the “life shaping” component was to provide paid work experience for eight agriculture graduates aged under 30 years for three years and thus launch them in their careers.

Municipalities aimed to become – at least partly – self-sufficient and produce goods that they could either use in their own kitchens or for welfare purposes and any surplus could be sold and thereby produce an income.

The agricultural producers in the program could employ trained workers for up to 12 months with a wage subsidy and use this time to reduce the burden or develop their business.

The program was not managed by the public employment service – unlike many similar programs – but an organisation, Diófa Consortium, set up specifically for this purpose at the beginning of 2009.

Initially the Consortium was made up of the following organisations:

– The Szekszárd-based Agrokonzult Training and Project Consultancy Ltd was responsible for the overall management of the Consortium and the program;

– The Szekszárd-based Ministry of Agriculture Agricultural Training Centre, Csapó Dániel High School and Agricultural College were responsible for organising training, and later they also took over the management of the Consortium;

– The local councils of Udvari and Závod were responsible for finding and coordinating work placements in the Tamási and Bonyhád small regions;

– The South-Transdanubian Regional Resource Centre public interest company was responsible for administration, monitoring and fund-raising activities with the Consortium (*Programterv*, 2009).
### Table 6.1: Settlements taking part in the Life Changing Program and the initial number of participants

<table>
<thead>
<tr>
<th>Settlement</th>
<th>Number of participants*</th>
<th>Note</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Phase 1 (spring 2009 to fall 2010)</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>1 Udvari</td>
<td>5</td>
<td>Tamási Micro-Region, municipal and business employment</td>
</tr>
<tr>
<td>2 Szakadát</td>
<td>3</td>
<td>Tamási Micro-Region, municipal employment</td>
</tr>
<tr>
<td>3 Belecska</td>
<td>12</td>
<td>Tamási Micro-Region, municipal employment (one of the participants passed away and the number of participants dropped to 11)</td>
</tr>
<tr>
<td>4 Kisviejke</td>
<td>9</td>
<td>Bonyhád Micro-Region, participants employed by agricultural producers. (The number of participants increased slightly in the Kisviejke area later on.)</td>
</tr>
<tr>
<td>5 Závod</td>
<td>5</td>
<td></td>
</tr>
<tr>
<td>6 Lengyel</td>
<td>6</td>
<td></td>
</tr>
<tr>
<td><strong>Phase 2 (January 2010 to fall 2011)</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>7 Nemesdéd</td>
<td>20</td>
<td>Marcali Micro-Region</td>
</tr>
<tr>
<td>8 Gyulaj</td>
<td>10</td>
<td>Dombóvár Micro-Region, municipal employment</td>
</tr>
<tr>
<td>9 Döbrököz</td>
<td>10</td>
<td></td>
</tr>
<tr>
<td>10 Gerényes</td>
<td>11</td>
<td>Sásd Micro-Region</td>
</tr>
<tr>
<td>11 Alsómocsolád</td>
<td>9</td>
<td></td>
</tr>
<tr>
<td><strong>Phase 3 (April 2010 to January 2012)</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>12 Hegyszentmárton</td>
<td>20</td>
<td>Sellye Micro-Region</td>
</tr>
<tr>
<td>13 Csányoszró</td>
<td>10</td>
<td></td>
</tr>
<tr>
<td>14 Drávasztára</td>
<td>4</td>
<td></td>
</tr>
<tr>
<td>15 Sellye</td>
<td>6</td>
<td></td>
</tr>
<tr>
<td>16 Somogyjád</td>
<td>8</td>
<td>Kaposvár Micro-Region</td>
</tr>
<tr>
<td>17 Osztopán</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>18 Juta</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>19 Alsóbogát</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>20 Eddie</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>21 Nagybajom</td>
<td>6</td>
<td>Csurgó, Kaposvár and Marcali Micro-Region</td>
</tr>
<tr>
<td>22 Szenyér</td>
<td>6</td>
<td></td>
</tr>
<tr>
<td>23 Berence</td>
<td>8</td>
<td></td>
</tr>
<tr>
<td>24 Fadd</td>
<td>20</td>
<td>Tolna Micro-Region</td>
</tr>
<tr>
<td>25 Regőly</td>
<td>3</td>
<td>Tamási Micro-Region</td>
</tr>
<tr>
<td>26 Kisszékely</td>
<td>4</td>
<td></td>
</tr>
<tr>
<td>27 Kalaznó</td>
<td>3</td>
<td></td>
</tr>
<tr>
<td>28 Nagykónyi</td>
<td>5</td>
<td></td>
</tr>
<tr>
<td>29 Udvari</td>
<td>7</td>
<td></td>
</tr>
<tr>
<td>30 Szárazd</td>
<td>3</td>
<td></td>
</tr>
<tr>
<td>31 Varsád</td>
<td>3</td>
<td></td>
</tr>
<tr>
<td>32 Kocsola</td>
<td>3</td>
<td>Dombóvár Micro-Region</td>
</tr>
<tr>
<td>33 Szakcs</td>
<td>3</td>
<td></td>
</tr>
<tr>
<td><strong>Total</strong></td>
<td>229</td>
<td></td>
</tr>
</tbody>
</table>

*Initial or planned number of participants.

Source: Based on the project documentation of Life Changing – Life Shaping regional labour market program (http://www.ddrmk.hu/?id=20&subid=0&parent=2).
Figure 6.1: Flow chart of the Life Changing Program

DDRMK
Diófa Consortium

LIFE CHANGING

Phase 1
2009–2010/2011

6 villages,
40 participants

Selection of participants (long-term unemployed)
by municipalities and employment offices by
personality tests and health examinations

Training program: An 1000-hour training tailored
to local circumstances and future subsidised job

Subsidised employment for 1 or 2 years, in jobs
provided by the municipalities or agricultural
producers

Goal: participants should stay employed at their
job of subsidised employment (without further
subsidies) or find another job

Phases 2 & 3
2010–2012

27 villages,
60 + 129
participants

LIFE SHAPING

subsidised
employment of 8
young high-skilled
agricultural
professionals,
2009–2012

3 years of subsidised
employment in order to experience
many agricultural
jobs. Selection of
the most suitable
position

Personality enhancement
and mentoring

Goal: participants should acquire
necessary work experience and
should stay employed at their job
of subsidised employment without
further subsidies
The Consortium was later expanded to include the Municipality of Fadd and as a sixth member the Institute of Medicinal Herb Research Ltd as well. The local council of Fadd provided work placements while the research institute contributed to the program by providing expert advice and buying the medicinal herbs produced in the program (Programterv, 2010a).

The program aimed to take into account local characteristics and opportunities as much as possible. Therefore training was provided locally in most settlements so that participants did not have to commute. More importantly production was also set up taking into account the local context and opportunities. Where fruit production was more viable, training was based on that, where there was demand for vegetable growing then training was set up accordingly (although not only fruit and vegetables but also medicinal herbs and small animal husbandry were possible options). Taking into account local characteristics was one of the main pillars that ensured the efficiency of the program by not setting out to “simply” provide assistance but to produce value locally.

One of the main advantages of the Life Changing Program was its flexibility – apparent at all levels: training, finances, mentoring and organisational structure. One of the consequences of this flexibility was that the management and coordination of the program was much more labour-intensive than other employment schemes. To sustain a unique system tailored to specific needs required numerous negotiations, discussion and a lot of attention – a continuous challenge to the management organisation.

Also as a result of its flexibility the Life Changing Program was not a well-defined and mature initiative at the outset (that turned out to be one of its main advantages). The management of the program learnt along the way, the original structure had a few failures that required correction. Shortly after the launch of the program it emerged that the long-term unemployed living in the small villages are very self-contained and their training needed more complex approaches than originally planned. Addressing this personal development was added to the program and was led by qualified psychologists. Also as a result of the introvert nature of participants, the training had to be reviewed continuously. After a couple of initial pilots it became clear that only locally provided training could be really effective in their case. It also emerged that training should be hands-on and practice oriented because the knowledge transferred by theoretical training was not easily accessible for them. Due to the need for a special approach, the trainers received further training during the implementation of the program to make sure they could effectively deal with the unique challenges associated with the target group.

There were also problems in the first phase of the program that nobody could foresee. For example there were conflicts between the participants of the Road to Work Program (see Chapter 5 of In Focus) and the participants of Life Changing in a number of localities because public works participants were somewhat
jealous of Life Changing participants whose job was secured for two years. This was particularly the case where Life Changing participants were required – against the aims of the program – to undertake municipal maintenance jobs or public works participants had to take part in agricultural production. These difficulties were addressed by clearly separating job roles. Sometimes these problems were aggravated by local councils that often treated participants as “usual” public works participants in the absence of any prior experience with similar schemes. In comparison to other assistance schemes, one of the key features of this program was that it provided job seekers with long-term employment for approximately two years (although this is not at all uncommon among complex employment schemes). Long-term participation was also thought to increase the likelihood of re-employment on the open labour market. Although post-program employment prospects are also closely related to the selection of participants the management of the program aimed to involve participants with the right attitude and skills.

Participants had to go through a selection process. First, the local job centre together with the local council identified potential participants from the given locality who were invited to apply for the program. First they had to go through a medical examination (people with long-term health condition and people with alcohol addiction were not eligible), and an assessment of their mental and cognitive status, motivation and social circumstances. Each locality was allowed to recruit a certain number of participants, usually no more than 10 persons (see Table 6.1).

The selected participants then had to take part in training that lasted eight to nine months and, as far as possible, delivered locally or at a nearby locality. This comprised 200 hours theoretical and 800 hours practical training and it was very flexible – just like the rest of the program. The training was tailored to local characteristics in all cases: the curriculum was tailored to participants’ future area of work. For example for individuals that were going to be employed in fruit growing, most of their training focused on this subject. Trainers were also from the local area (as far as possible) to make sure that had an adequate knowledge of the local circumstances and could build trust in participants. Teaching did not aim to provide a deeper understanding of underlying issues but was practice-oriented and taught participants how to carry out certain tasks without necessarily understanding why it was done in a certain way. Work placements were provided by local councils and local businesses. At the end of the training all participants were required to sit an exam.

Following the training participants started a work placement that lasted up to two years. Placements were provided by local councils or agricultural businesses and producers. Workers always received the statutory minimum wage and were employed full time. Similarly to the training phases, changes in the participants’ personality and attitudes were continuously monitored.

---

4 Settlements that joined the program later could receive the subsidy for one year only.
One of the major challenges in both public and private sector employment was that it was difficult to provide participants with adequate work during the autumn and winter seasons. Municipalities tried to deal with this by involving participants in community jobs. For businesses this was more difficult and workers had to take leave during the winter period and make up for this time between spring and autumn (the Labour Code allows to average working hours over a longer period of time). This caused difficulties where the workers did not have the right attitudes and were unwilling to work more than eight hours a day during the summer.

Participants finished the program after the one- or two-year work placement. As was mentioned earlier, the aim of the program was that many of the participants could retain their job long term. In the business sector this was possible if the individual performed well throughout the year and the business could keep them on without the wage subsidy. In the public sector the aim was to create sustainable schemes that would employ a larger number of people. A third option for participants – now equipped with adequate qualifications and work experience – was to start looking for a job or start their own production business.

The aforementioned mentoring activity can be potentially very important for the outcomes of the program and the satisfaction of participants. Therefore the mental well-being of workers and psycho-social assistance were important elements of the program. Qualified mentors were part of the selection process as well to make sure those with the most appropriate attitudes were selected as participants. Later on designated mentors were in regular contact – in person and even over the phone – with the participants. Apart from work-related issues they also offered assistance in other areas such as debt management, budgeting advice etc.

The selection of settlements in the program happened through various channels. On the one hand organisers contacted villages with high unemployment where there was still a living tradition of agricultural production that could be effectively used in the program. Whether they decided to take part in the program depended on the openness of the local government or businesses. The opportunity was typically very well received also due to the personal networks and lobby activities of coordinators. On the other hand some (although not too many) localities put themselves forward to take part in the program.

**Budget structure of the program**

Due to its complex nature, *Life Changing* was slightly more costly than other labour market programs. The total cost of the program is expected to be around 660 million forints during the three years. Funding comes from the decentralised employment sub-fund of the Labour Market Fund and various SROP funded sources. Wage subsidies and training make up the largest part...
of the cost (approximately 572 million forints), while the management of the program costs approximately 90 million forints.

The rate of wage subsidies paid to businesses and local councils varied depending on the locality and the program phase. In most cases it was 100% but for those who joined the program later at the third phase it was only 50% of the statutory minimum wage and contributions (Programterv, 2010b).

According to 2010 calculations the cost of an eight- or nine-month training for one participant was no more than 500,000 forints. During the training period participants were receiving training assistance that amounted to 93,000 forints per month (a total of 837,000 forints during the nine-month period). The majority of the workers were employed by local councils which meant a 100% wage subsidy in the program. In 2010 this was 96,000 forints per month – for a 12-month employment period this totaled 1,152,000 forints (ibid idem). The per capita cost of the program excluding other costs (such as transport) was 2,489,000 forints at 2010 prices. Considering the total cost of the program (660 million forints) the per capita cost was 2.8 million forints.

With regards to the cost of the program it should be mentioned that participants would have been receiving social assistance had they not been in the program; this reduces the real cost of the program by approximately 100–150 million forints. Furthermore some of the costs of the program are returned to the state budget through taxation (for example income tax paid by the management, trainers etc.). If these factors are taken into account the program is comparable to an average-cost complex labour market program. Nevertheless the costing and the cost-efficiency of the program would require further investigation that is only possible after its closure.

Comparison of Life Changing and other labour market programs

County (later regional) employment services have had the possibility to initiate and implement their own labour market programs using active labour market policies (for example training, wage subsidy). Job centres have ample autonomy in the design and implementation of these programs; therefore there has been a wide variety of such programs over recent years.

So far there was only one comprehensive study on complex labour market programs that was conducted by the HAS Institute of Economics in 2007 (Fazekas et al., 2007). The study also examined locally initiated labour market programs like (Simkó, 2007). The results of this study allow us to compare the Life Changing – Life Shaping Program with other, similar initiatives, identify its unique features and better assess its efficiency.

The main characteristics of complex labour market programs can be summarised as follows. Joint working of the job centre and external organisations is an important part of the programs. Labor market programs are generally longer (up to three years), take into account individual needs and character-
istics and use a combination of employment services and active labour market policies. The unit cost of labour market programs is generally high due to their complexity but the programs might be more effective in achieving their objectives thanks to the combination of different types of assistance (Fazekas et al., 2007).

The comparison of different labour market programs is made difficult by the absence of clear categorisation. To resolve this we adopt the categories used by Fazekas et al. and attempt to classify the program in these. Considering that the life changing and life shaping components of the program are fundamentally different, these should be analysed separately as well. The comparison uses the 149 complex labour market programs implemented in Hungary between 2000 and 2006.

The duration of complex labour market programs implemented between 2000 and 2006 was between six and 36 months, on average around 24 months. The life changing component lasted three years which puts it among the longer programs. The differences in the program objectives do not allow us to directly compare the number of participants and this is made even more difficult by differences in the geographical scope of the programs (i.e. municipal, county and regional programs). The average number of participants in the 149 programs included in the study was 180 people which is similar to the number of participants in Life Changing.

The average per capita cost of the programs in the study was 453,00 forints. As has been shown the Life Changing Program was considerably more expensive than this, the average per capita cost was nearly three million forints. This is only comparable to the cost of small-scale intensive mentoring programs that also had two to three million average per capita costs. (Taking into account the rate of inflation since the previous study Life Changing was somewhat cheaper than these programs.)

The life shaping component can be compared to graduate programs. These programs ran on average for 2.5 to three years with around 20–30 participants. Their primary aim was typically to keep young graduates in the local area or in some cases to retrain them to improve their employment prospects. The life shaping component was a unique initiative in that it provided work placements in agriculture. There was a similar program in Heves County in 2000–2002 that provided work experience for 12 young agriculture graduates in agricultural business development with the longer-term aim of setting up their own business. All 12 participants found long-term employment after the program. As has been shown, the life shaping component is also expected to be similarly successful.

It is a lot more challenging to compare the life changing components because potentially there are many similar programs. It can be equally regarded as a program aimed at the low-educated, those aged 40 years or over or even the re-
integration of disadvantaged people. Although there are countless programs in these categories (these groups are the most common target groups of complex labour market programs), they might not be comparable meaningfully.

Looking at the objectives of the programs it might be argued that Sorfordito was innovative in that it offered long-term employment in agriculture that no other program did. Horticulture-related training and longer-term employment appeared in other programs (for example Bácsalmás, Jánoshalma 2003–2006), but these were smaller and did not focus on production.

Most labour market programs are based on training and/or wage subsidy but mentoring and personal development are also common elements. In this comparison the Life Changing Program was not at all unique. However it should be highlighted that mentoring in this program was much more effective than in other programs. Fazekas et al. (2007) argued that different advisory services and personal and skills development training that should have been a key part of the programs, were often implemented inadequately. Job centres usually used their own staff to provide services that had negative implications for the efficiency of programs. Nevertheless, relatively few job centres took advantage of the possibility to purchase external services or outsource program management to ease pressure on internal capacities (Simkó, 2007). To the contrary, the Life Changing Program provided seemingly efficient psycho-social development throughout its duration.

The Life Changing – Life Shaping Program is clearly unique in a sense that both its components are fully linked to agriculture and aims to provide long-term jobs in this sector. Its effective mentoring provision also differentiates it from many other programs. Its complexity is not without examples and similar programs have been implemented efficiently before. It should be highlighted that the Life Changing Program, compared to other initiatives was somewhat more costly and this was mainly due to extended and combined use of multiple active labour market policies.

Views about the program, the feedback and problems so far

Before presenting a more systematic evaluation of the program effects, it is worth considering the feedback and views about the program that might give an indication of its potential success.

The program ended in November 2010 in the settlements around Kisvejke that had been involved in its first phase. A total of 20 workers participated in the micro region. Initially it was expected that approximately half of them would be able to find a job locally, however this did not happen. Only a couple of people could find a job and local agricultural businesses would only employ workers on a seasonal basis. It is unlikely that other agricultural businesses involved in the program would provide many jobs to former participants and therefore only a few of them will break the cycle of unemployment.
With regards to employment after the program, the organisers were also faced with some unforeseen challenges: some of the agricultural producers are, to a large extent, unfamiliar with lawful employment because they have always employed undeclared workers. It is likely that some of the participants will find undeclared work after the end of the program.

Most local councils are aiming to maintain the achievements of the program over the long term and provide jobs to as many workers as possible; the scheme might become the foundation for social land programs and local social economy projects.

The majority of workers participating in the program are satisfied. The satisfaction of 57 workers who joined the program in its second phase was surveyed in the autumn of 2010. Seventy per cent of respondents said that their financial situation improved thanks to the regular monthly income from the program. Eighty-six percent indicated that they would be willing to take part again and 52 out of 57 said that they would like to remain in their current job (it is interesting to note that four respondents were undecided on this question because they had achieved a higher income as undeclared workers previously). The majority of respondents (46 persons) were 80 per cent or more satisfied with the program. The fact that no workers rated their attitude lower than three (on a scale of five, 5 indicating the best possible score) is quite telling about the mentoring activity (Anonim elégedettségi ... 2010).

There were other unforeseen results as well. Most importantly real communities formed among the workers that had the effect of linking them outside work. This is extremely important in the life of these settlements which are facing multiple difficulties.

The graduate component of the program also seems successful based on the feedback. With this organisers aimed to revive the tradition of a once very popular graduate program in agriculture in Hungary and provide an example to follow for agricultural businesses. The program provided wage subsidies for eight agriculture graduates for three years. Despite various difficulties, in the end all participants found a job that was right for them and they are likely to stay in their job long term (the young people are mainly employed by agricultural companies based in Szekszárd).

Until recently the Life Changing Program had very limited national publicity; employment policy makers and the press did not show much interest. This changed in the autumn of 2010 and since then the program has had a considerable amount of media attention.

The management of the program would like to turn the results and the experiences into a “best practice” model. This is based on two principles: on the one hand they are aiming to create a program that is not in conflict with the private sector, and on the other hand it places great emphasis on community development because experience showed that the success of the program
largely depends on this. These principles were embraced by other initiatives: a pilot public works program is being launched in various regions at the time of the publication of this study, based at least partly on the example of the Life Changing Program. Furthermore, the organisers’ aims are linked to the objectives of the Government’s rural development policy thus it is hoped that the experiences accumulated in the program will not be lost.

In many settlements the program ended in September 2011 and some problems were already visible that might hamper the sustainability of the results. First and foremost, the shortage of land; some local councils that are planning to continue production do not have enough land to expand. Generally local councils own so little land that this issue might require government intervention. If decision makers would like to maintain the successful results achieved by the programs then some of the state-owned land will have to be handed over to local councils involved in agricultural production. (Or alternatively local landowners might concede a small part of their estate to local councils.)

One of the program objectives was to help previously unemployed participants to set up their own business. Unfortunately those who had such plans came to realise that the administrative and financial burdens were insurmountable. This could only be resolved by simplifying existing regulations that might encourage more people to start their own business.

Another obstacle to self-employment in agriculture is the uncertainty in the sale of produce. The renewal and strengthening of cooperatives in the future might address this problem (there have been some policy measures to improve this situation). Similarly to local councils, potential producers are also affected by the shortage of land. The size of gardens often does not allow a profitable production but without any capital these producers are unable to purchase or rent land. Again, government intervention might be necessary to address this issue. Furthermore it has been proposed that land sharing might allow those without land to start production.

For the sustainability of results the lack of funding was also a major problem. There were not many grant programs since the first half of 2010 for about a year and that also hindered the development of the Life Changing Program. Hopefully this will change in the future; different stakeholders of the program (primarily local councils, management organisations etc.) are putting great efforts into mobilising external resources that might help to sustain the results.

Results of the Life Changing Program – a local case study

The full evaluation of the program will only be possible after its end in early 2012. The most important result indicator will be the percentage of participants with a regular job after the end of the wage subsidy, either as a continuation of their work placement in the same job or elsewhere including successful self-employment. This indicator will need to be compared to a re-employment
rate that we would have found in the absence of the program. This counter-
factual result could be measured using a control group of unemployed people
from settlements that are similar to those in the program. Such program eval-
uation is yet to be carried out.

Feedback so far and the general results suggest that the implementation of
the centrally designed program plan depended on the local context and dif-
fered from village to village, so results will probably differ too. Using a case
study design this chapter presents the potential results of the program, under
what conditions and through which mechanisms these might prevail.

The case study summarises the experiences of Gyulaj, a village in Tolna Coun-
ty. The village has been presented in more detail in our volume on the evalua-
tion of local economic development programs in the IES Book Series (Kabai–
Németh, 2010b), thus this chapter focuses on more recent trends since the
publication of the earlier paper and explores the impact of the program on the
village. Neither the history of the village nor its current social situation will
be presented in detail here.6

In parallel with the socio-economic trends in rural Hungary, Gyulaj has
slowly but steadily become increasingly poorer for the past 60 years. Apart from
the many problems – the nationalisation of the forest that was a vital part of
the local economy, the creation of a collective farm, then the negative impact
of the land restitution, the gradual dismantling of local services – the biggest
problem is that the village does not have a through road. Its only good quality
road links the village with Dombóvár therefore its spatial relations are limited
and that is thought to be the main obstacle of development.

There were various attempts to tackle this issue after the change of regime
that all aimed to build a new through road towards the village of Szakály. The
finally happened in 2000 but the new road did not meet expectations: the
construction was financed by various public funds and the new road was too
narrow which makes it impossible to run a regular bus service there. Furth-
ermore the road goes through private property which generated numerous legal
disputes. Therefore Tamási (and Szekszárd) are only accessible for Gyulaj res-
idents with a car although access to these settlements was the main rationale
of the new road. Therefore Gyulaj still depends on Dombóvár; its spatial rela-
tions are as limited as ever.

There was another significant capital investment in the village during the
last decade: the primary school was refurbished with a total budget of 160
million forints financed by the South-Transdanubian Regional Operational
Program. Certainly this is an important development in its category and un-
doubtedly it will have a positive effect that local children can attend a state-
of-the-art school. In parallel with the development of the infrastructure, the
school is also aiming to reform the curriculum to tailor it to the needs of the
mainly multiple disadvantaged pupils.

6 The Gyulaj-born author, Fer-
enc Bali Pap published a longer
piece on the societal situation
of the village (Bali Pap, 2011).
Since the dissolution of the collective farm in the early nineties, there has been only one actor that can play a significant role in the development and prosperity of the village, and the improvement of living conditions: this is the local council and its primary school. The Catholic Church effectively abandoned the village around the early 1990s and there have been no significant religious activities in Gyulaj since then. There are no local charities in the village; regional or national charities are also absent. The presence of public institutions is rather limited too. Since 1990 the village had a general practitioner for only short periods of time; currently the GP from the neighbouring village visits the village on certain days. This has resulted, indirectly, in the deterioration of the health of Gyulaj residents. Today two thirds of the village’s residents are Roma – living in poverty, the majority of them are out of work and they do not have access to the necessary health care services. Nevertheless there is a full time health visitor in Gyulaj. The police – similarly to the general practitioner – “left” the village in the early nineties and Gyulaj has had its own police officer again for two years. Gyulaj has very high levels of crime, anything left unattended is likely to disappear, and empty houses are sooner or later burgled. This hampers the little entrepreneurial drive that is left in a few people in or around Gyulaj.

Good quality land around the village is concentrated in the hands of a few individuals, the majority of whom are not from Gyulaj and have hardly any relationship with the village. Agriculture is no longer part of the local culture in the village.

Nowadays the residents of Gyulaj hardly own any land, only three or four families have 30–40 hectares supplemented with some small areas of rented land. The other families either do not own any land at all or only one or two hectares or a larger garden in the village. The land is not used to grow any produce (other than wheat). Furthermore, it might be argued that most Gyulaj residents would not be capable of starting agricultural production on their own because the younger generations lack the necessary skills.

In summary, human resources in the village are generally poor and many individuals are in a poor physical condition. If there was a sudden surge in the demand for unskilled labour in the village, it is unlikely that this would be satisfied hiring local unemployed because probably only a few of them have their full work capacity.

The Life Changing Program in Gyulaj achieved results in these areas and proposed a viable development model. As has been mentioned earlier, the local council is the only organisation that can effectively do something to improve the employment prospects of the local population. The most influential figure in the local council is the mayor in such small settlements – Gyulaj has 1,000 inhabitants. If the mayor can understand local challenges and act then the village has the chance to develop. However, if the local leadership is not adequate...
then the whole village is likely to stagnate. A number of case studies and local experience suggest two further conditions: first, the mayor needs a network of local allies – headed by the district notary and the management of local institutions – that is supportive of the development; and second there needs to be a general consensus behind the developments, any internal tensions (social, political) can undermine the initiative.

After the transition to a market economy there were no positive changes in the life of the community in Gyulaj, the social and economic situation of the village continued to deteriorate and deprivation became permanent. This is illustrated by trends in two commonly used regional development indicators: per capita income and unemployment rate (Figures 6.2 and 6.3). While the per capita income in the small region is around 75% of the national average, in Gyulaj it was under 40% between 2000 and 2008. Only two of the smallest hamlets with 100–200 inhabitants are “poorer” than Gyulaj in the Dombóvár Small Region: Jágó and Lápafő. The unemployment rate shows a similar trend: there was a clear deterioration until the mid-2000s then it stabilised at a level over 2.5 times above the national average. Similarly to income, Gyulaj has one of the highest unemployment levels in the small region.

**Figure 6.2: Per capita average income in the settlements of Dombóvár Small Region, as a percentage of the national average, 1990–2009**

*National average = 100. The figure also includes the average for Dombóvár Small Region, Tolna County and the South-Transdanubia region as a percentage of the national average income. Settlements are ranked from the highest rate to the lowest as follows: Dombóvár, Kaposszülné, Kaposszekcső, Csikostóttsós, Attala, Döbrököz, Dalmand, Kurd, Szakcs, Nak, Várong, Kocsola, Csibrák, Gyulaj, Jágónak, Lápafő. Source: Based on the Regional resource map of the IE HAS.*
As has been pointed out above the only actor that in our view is capable of taking significant action to improve the employment situation and the quality of life of the disadvantaged population is the local council. Therefore it is important to consider the measures that were available to them to influence the local jobs market – also as an employer – before the Life Changing Program. As in many similar villages these were mainly related to public works and subsidies. The local village council tried to make use of the opportunities of public works policies to provide – at least temporarily – work opportunities to as many people as possible. An overview of the public works programs in the village over the past few years might give a good indication of the opportunities that would characterise the labour market policies of the local council in the absence of the Life Changing Program.

Between 2006 and 2008 a total of 50 people were employed in different forms of public works in the village (Table 6.2); most of them in jobs related to the statutory services provided by local councils (mainly water management, rainwater drainage, maintenance of the public cemetery and street cleaning). Thanks to the Road to Work Program in 2009 the number of people in public...
works increased to 69 but their tasks did not change substantially compared to previous years as is highlighted by Table 6.3.

**Table 6.2: Public works in numbers, Gyulaj 2006–2008**

<table>
<thead>
<tr>
<th>Year</th>
<th>Municipal public works</th>
<th>Communal public works</th>
<th>Public works organized by PES</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>participants work hours (total)</td>
<td>participants work hours (total)</td>
<td>participants work hours (total)</td>
</tr>
<tr>
<td>2006</td>
<td>35 15,792</td>
<td>2 2,016</td>
<td>14 8,484</td>
</tr>
<tr>
<td>2007</td>
<td>38 15,802</td>
<td>3 3,024</td>
<td>17 6,468</td>
</tr>
<tr>
<td>2008</td>
<td>29 11,319</td>
<td>7 5,040</td>
<td>16 10,332</td>
</tr>
</tbody>
</table>


**Table 6.3: Tasks carried out in public works and number of participants in Gyulaj, 2009**

<table>
<thead>
<tr>
<th>Task</th>
<th>Number of workers necessary to carry out task (full-time equivalent)</th>
<th>Person-hours used (total)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Social housing maintenance</td>
<td>2</td>
<td>1,512</td>
</tr>
<tr>
<td>Water management, rainwater drainage</td>
<td>12</td>
<td>8,064</td>
</tr>
<tr>
<td>Maintenance of public cemetery</td>
<td>5</td>
<td>5,376</td>
</tr>
<tr>
<td>Maintenance of local roads, parks and public spaces</td>
<td>22</td>
<td>14,784</td>
</tr>
<tr>
<td>Street cleaning</td>
<td>4</td>
<td>2,688</td>
</tr>
<tr>
<td>Management of public works projects</td>
<td>3</td>
<td>3,528</td>
</tr>
<tr>
<td>Provision of social care</td>
<td>4</td>
<td>2,688</td>
</tr>
<tr>
<td>Culture and sport related activities</td>
<td>1</td>
<td>672</td>
</tr>
<tr>
<td>Administrative support</td>
<td>2</td>
<td>1,344</td>
</tr>
<tr>
<td>Implementation of national and ethnic minority rights</td>
<td>1</td>
<td>1,008</td>
</tr>
<tr>
<td>Tasks related to education</td>
<td>2</td>
<td>3,024</td>
</tr>
<tr>
<td>Tasks related to the maintenance of council buildings</td>
<td>8</td>
<td>5,376</td>
</tr>
<tr>
<td>Tasks related to public works</td>
<td>2</td>
<td>1,344</td>
</tr>
<tr>
<td>Total</td>
<td>68</td>
<td>51,408</td>
</tr>
</tbody>
</table>


The public works plans of the village for 2009 and 2010 give us an idea of what would have happened and what the situation would be like without the Life Changing Program. Considering the management and financing of public works policies, they would have provided only short-term results without any long-term effects and sustainability.

Apart from increased funding, three changes in public works policies in 2009–2010 were important for Gyulaj: first, unemployed persons aged under 35 years who had dropped out of primary school were required to return to school to complete primary education. This affected 13 people in Gyulaj. Pri-
mary education forms the basis of any further education or training, therefore this was a useful measure.

The other important change was – potentially – the provision of long-term employment up to 12 months in the Road to Work Program. Local implementation plans had to be reviewed and submitted each year which involved the uncertainty of changing financial frameworks and conditions. One of the implicit aims of the Road to Work Program was to allow village councils to create “real” jobs locally and promote community production and the development of social economy. Policy makers did not want to “penalise” settlements where the local council was slower to change; therefore tasks that had been typically carried out in public works projects remained eligible for funding in the Road to Work Program. It was envisaged that the system would gradually shift towards job creation over three years. It is not known how they thought to implement this because the program was cancelled halfway through, however it is certain that a major review of existing legislation would have been required to allow the startup of social cooperatives and community production in villages with the potential to employ a large number of unskilled job seekers. Even with these changes it is unlikely that Gyulaj could have developed a production system similar to its current one using previous financing schemes.

Finally, the third important change brought about by the Road to Work Program in Gyulaj’s public works projects was the possibility to purchase a limited amount of equipment and work wear (Table 6.4). However the amount allowed for this would not be sufficient to purchase enough equipment for the cultivation of the current 3–4 hectares of land that is continuously increasing.

Table 6.4: Planned measures for public works in Gyulaj, 2009

<table>
<thead>
<tr>
<th>Measure</th>
<th>Expenditure (HUF, 000)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Occupational health and safety training</td>
<td>60</td>
</tr>
<tr>
<td>Occupational health assessment</td>
<td>50</td>
</tr>
<tr>
<td>Work wear, safety gloves</td>
<td>476</td>
</tr>
<tr>
<td>Tools and equipment</td>
<td>480</td>
</tr>
<tr>
<td>Purchase of tangible assets</td>
<td>150</td>
</tr>
</tbody>
</table>


At the same time the Life Changing Program brought about strategic change in the village. Gyulaj became involved in the program with 10 participants at the beginning of 2010 thanks to the efforts of the local mayor who was familiar with the program in Beleska (Kabai–Németh, 2010a) and wanted to organise and set up similar municipal agricultural production in Gyulaj. The main features of the Life Changing Program in Gyulaj were presented in our previous study (Kabai–Németh, 2010b); the subsidy period ended on the 30th of September 2011, at the time of writing this paper. During the two years the
program grew and provided the village’s kitchen with vegetables and could sell some of the produce (ground paprika and medicinal herbs).

Community production in Gyulaj follows the model developed in Belecska, similarly to the majority of participating villages where the main producer is the local council. The majority of Gyulaj residents participating in the program – similarly to other villages – are probably not capable of starting their own business just yet and the external conditions are not favourable either – legislation, tax rules, market access – the role of the local council seems vital.

Probably one of the most important factors in the success of the Life Changing Program is that it allowed local governments to set up their own production. Before the launch of the program the business sector was unable to solve the employment problems of the village and the local council was the only actor that was capable of bringing about any change in this. However this was only possible with the local council setting up a system that is sustainable or at least viable in the long-run: based on real activity, work and production. Furthermore it allows paying a decent wage and enhancing the cohesion of the local community. Workers have individual responsibility but they can also have an individual sense of achievement. According to the initial experiences the Life Changing Program succeeded in creating such a system.

The planners of the program adequately recognised that in small settlements like Gyulaj, only the local council can tackle unemployment at least by improving the employability of local residents with impaired work capacity. In our view, the Life Changing Program laid down the foundations for the implementation of a comprehensive village development strategy. Community production is not simply public works but a social enterprise linked to the local economic development strategy; the creation of a third sector business. Therefore its frameworks are completely different from that of public works. There is a need for equipment and specialist knowledge in the social economy as well as the market production taking into account community interest. There is a qualitative difference between earlier public works projects and the Life Changing Program.

Partly due to the local results of the Life Changing Program, the primary school launched a two year agricultural vocational course for year 9–10 in 2011. Pupils do work experience on a communal allotment while their parents, relatives and neighbours do similar work on municipal allotments. This initiative is the innovation of the new school headmaster and indicates the aspiration to revive agricultural traditions in the life of the village. By establishing community production and involving the primary school in the scheme Gyulaj might become a model settlement in South-Transdanubia or even nationally. The Life Changing Program has undoubtedly contributed to this by providing the initial impetus and the organisational framework.

The recognition of the village is indicated by the fact that Dombóvár Small Region was included in the public works model program (Start work scheme).
of the Ministry of Home Affairs largely thanks to the community production in Gyulaj. This might open up further opportunities for the village: the possibility to employ further workers, purchase equipment and the processing of fresh vegetables and fruit planned to be launched in the next couple of years might bring about significant changes in production as well.

Gyulaj also decided to join the program “Nyúl-unk a munkáért!” aiming to re-introduce small animal husbandry in the village that might be a further step towards strengthening the local community and promoting the labour market re-integration of participating families.

The need and the results also attracted community activity and the attention of independent non-profit organisations. The Hungarian Maltese Charity Service became active in Gyulaj and the village was invited to take part in a community building project supporting long-term local development in settlements with high levels of social exclusion.

These are obviously just the first steps of what – in our view – could be a regeneration in Gyulaj (and other similar settlements) providing the opportunity for a better life to its inhabitants. This will be a long process and will require decades of continuous and committed work, government support (not only financial), European Union funding and significant grass-roots activity.

Conclusion

The Life Changing – Life Shaping complex labour market program was launched in two small regions of Tolna county by the South-Transdanubia Regional Employment Service in the spring of 2009. The program built on the once successful agricultural tradition of the region and provided long term unemployed participants with tailored training for eight months and work placement for one to two years with local councils or agricultural businesses using wage subsidies. In the three counties of South-Transdanubia nearly 40 settlements and over 200 workers participated in the program.

The complexity of the Life Changing – Life Shaping Program can be captured in the simultaneous use of multiple active labour market policies: training, training assistance, work experience, wage subsidy and mentoring. The participating local councils could start their own production and apart from providing jobs to local residents, this could save a significant amount of money. The program had an impact on local development; a good example of this is the case of Gyulaj.

The Life Changing Program had a number of positive side effects as well. Participants built strong social networks that bring people together outside work too. The majority of participants went through real personal development which significantly improved their lifestyle and future employment prospects. The graduate component of the program was no less successful. The eight gradu-
ate trainees in the agricultural sector did very well in their job and they all had the possibility to remain there after the end of the three-year subsidy period. The *Life Changing* Program ends in 2012, therefore its overall assessment is not yet possible. It is hoped that as many workers as possible can stay in their job long term. It is promising that some of the local councils are already making plans to maintain the achieved results after the subsidy ends. Hopefully these will be successful and results will be sustained over the long term.

The content and approach of the *Life Changing* Program, together with the experiences gained through its implementation make it suitable for a national roll-out. This could significantly improve the employment situation in small settlements of rural Hungary.
7. EVALUATING THE IMPACT OF HUNGARIAN LABOUR MARKET POLICIES

ZSOMBOR CSERES-GERGELY & ÁGOTA SCHARLE

This chapter of In Focus reviews evaluation studies of active labour market policies and unemployment benefits in Hungary. Somewhat unusually we focus on what was evaluated and especially how, rather than on the outcomes. We also briefly consider some obstacles to fulfilling the methodological requirements discussed in Chapter 1 of In Focus, and their consequences for Hungarian empirical investigations. The review also provides some basis for reconstructing the evidence base available to the previous governments when they made decisions regarding employment policies. We cannot claim to have uncovered all the information accessible to politicians, but we can safely say that at least this evidence was available to them.

Selection criteria for the studies

We confine our analysis to two types of studies: evaluations of wage subsidies, and those of unemployment benefits. Wage subsidies deserve special attention for several reasons. First, the few existing studies from Central and Eastern Europe show either a neutral or a negative effect, while in developed countries this type of program has a mostly positive impact on employment (Kluve, 2010). Second, these programs often have indirect effects which may override their positive effects: e.g. the subsidised employer could employ people without the subsidy, or it puts other employees at a disadvantage by taking on the subsidized unemployed person. Finally, these programs are quite costly, thus even if their net impact turns out to be positive, the subsidy might not be cost-efficient. The reason for including unemployment benefits is much simpler: these are the most frequently evaluated programs in Hungary.

The review covers evaluation studies in the narrow sense as well as reports where the title or the abstract makes an explicit reference to policy evaluation. Thus, we did not exclude studies which aim to analyse the impact of a program but are lacking in terms of methodological rigour as these may still offer important insights on program outcomes.

We know of no archive either public or private which provides a complete collection of evaluation studies. Therefore, we used online search engines to collect the relevant literature.¹ This method restrains our findings, because such evaluations are typically prepared when commissioned by a government body (such as a ministry or the National Development Agency) or when the program promises to be interesting from an academic point of view. In the former case, results are not necessarily made public.²

¹ On 15 September 2011, a web search under the “tanulmányok” (research studies) label in the document archive of the new government homepage yielded zero results (http://www.kormany.hu/hu/dok?type=411#Document-Browse).
² The Introduction of In Focus discusses the importance of public access to evaluation reports.
Our list is therefore not complete, and it is most likely selective. Well-written papers motivated by scientific interest are likely to be overrepresented in it.³ The possibility of a contractor restricting access to its paper (due to unfavourable results which could e.g. divert funds from its program) can also cause bias. Finally, the same reason can introduce significant bias in the timing and subject of such evaluations.

**Dimensions of the review**

*Appendix 7.1* lists the papers reviewed. We categorize unemployment benefits based on whether they pertain to insurance based benefits or social assistance. Studies analysing wage subsidies are classified into monitoring reports and econometric evaluations.

We highlighted those characteristics of the studies which can determine the expected quality of their contents. They are listed in a table in *Appendix 7.2*: size, structure, and information-content of the database, observation period, identification strategy, identified effect(s), success criteria, and indirect effects (if estimated in the paper). This part of the review only covers studies that actually measure program effects, i.e. monitoring reports are excluded.

**The demand for evaluations**

The list of studies surveyed in the review shows that during the period of roughly 20 years, evaluations of main employment policy instruments were scarce, and were mostly written after 2000. The reason behind this may be partly methodological – since high-performance desktop computers only became widely available around this time – and partly political as the demand for evaluations rose after Hungary’s accession to the EU (Váradi, 2012). However, since we know that sufficient computing power had been available e.g. in the CSO (Központi Statisztikai Hivatal, Hungarian Central Statistical Office), and most analyses would not require exceptional capacities (sample sizes were not extraordinary and methods did not require much processing power), we are inclined to believe that the definitive reason in this case is on the demand side.⁴

Finally, it is important to note that the majority of these studies have been written by a handful of experienced, senior members of the Hungarian academic community. Even though an evaluation study needs serious thought and thus, senior experience also, we believe the above circumstance is due to the previously described lack of demand. It is quite understandable in the case of a small market if studies which are expected to have scientific value are carried out by the most experienced and well-connected researchers. But if the demand for evaluation studies in Hungary were in line with international trends, we should be able to find works of younger researchers. This suggests that policymakers of this period did not consider program evaluation a conventional tool of routine use in their decision making, neither before nor after the launch of

---

³ One concern may be the so-called publication bias: authors (and journals) tend to like to publish statistically significant results that conform prior expectations.

⁴ See e.g. Scharle (2008) on the decision making process in Hungarian employment policymaking compared to the British practice.
policy interventions. Nevertheless, the appearance of a few papers written by younger researchers around the end of the 2000s is promising /a hopeful sign.

The data sources

Most of the studies reviewed rely on the administrative data of the unemployment register. This is because alternative data sources, e.g. labour force surveys do not contain information on participation in employment policy programs; therefore, their use would greatly restrict or completely rule out evaluations (except maybe in the case of unemployment benefits). However, since the register reveals nothing of the period following exiting the programs, using it as the sole source of data also significantly restricts research. Up until 2009, the monitoring of active programs provided this sort of surplus information, but only in a very particular form: exiting workers (or their employers, in the case of wage subsidies) were surveyed 3 months after leaving the programs. Thus, information on after-program status is very limited – practically non-existent on individual workers for wage subsidies. In light of the findings of Card et al. (2010), this suggests that using these data, the majority of program effects can only be estimated with a bias. To reduce these restrictions, some studies relied on supplementary data collected in a separate survey. That cannot provide a permanent solution to the lack of data because of the costs, but it helps create a more realistic view of program effects by enabling their estimation farther from the ending of the program.

The CI. law of 2007 allows studies to surpass the limitations of the unemployment register by anonymously linking other administrative data sources (e.g. on social security contributions), and to follow the subsequent work history of participants or control for previous employment spells. We have not yet found such a paper however, maybe due to the shortness of the period and the frequency of policy changes. The National Employment Office (NEO) introduced a reform in active program monitoring in 2009 to exploit the possibility that data from the register and social security contributions can be linked at the individual level. This could have improved the quality of the monitoring, but we have not yet seen its use in evaluations, and it is unrelated to the studies reviewed here. Administrative data (e.g. on tax contribution) could have also helped evaluation reports in the essential issue of making detailed work history (among other observable characteristics) controllable. Lately, the NEO collects data on the registered unemployed for four years retrospectively, but this is unavailable for those who participate in one of the programs but are not entitled to benefits. Therefore, supplementary data is needed in their case.

Identification and estimation methods

Most evaluations do not make a formal distinction between identification and estimation, nor do they systematically discuss identification conditions and
the implications when these are unmet – though several studies touch on these issues when discussing the estimation methods. The applicable identification strategies (see Chapter 1 of In Focus) are mostly determined by data availability. Small sample size makes parametric methods the most viable option. Controlling for heterogeneity is restricted by the number of observable individual characteristics. Since the register only collects data necessary for administration, it does not contain information on other members of the household, which is a serious limitation. This is why the studies working with supplementary data can be more reliable than those relying solely on the register. Matching methods are rarely used in our sample of studies, despite the fact that matching is considered superior to parametric models when comparing treated and untreated individuals having similar observable characteristics.

We have found few attempts to control for unobserved heterogeneity; these include the modelling of selection, exploiting changes to administrative rules; and a genuine experiment in one case (see Appendix 7.2). Hungarian labour market policies change very often, but evaluations of the above sort have not yet accompanied these. Quasi-experimental situations arise because newly introduced nationwide programs are grandfathered (i.e. only affect new entrants). The lack of experiments makes selection modelling hard to verify and leaves it to rest on disputable assumptions. Therefore, a rise in the number of experimental studies would make a big difference. If the demand for evaluations increased, this may generate an increase in the number of studies exploiting policy changes in space or time. We expect that such evaluations would generally outperform those that only use observable characteristics.

Publication of estimations and attention to detail

The presentation of estimation methods and results is quite varied. Publishing marginal effects or at least their average has now become common in international practice, but not in the studies we have reviewed. Thus, estimates from nonlinear models are impossible to compare with previous research for reference. Some studies do not even present regression tables, though fortunately these are rare. Almost every paper interprets treatment effects in the form of average treatment effect or average treatment effect on the treated.

Further analysis

In order to evaluate a program completely and conduct a cost-benefit analysis, indirect or unintended effects also need to be measured in addition to direct impact. Such are substitution effects, which arise when a subsidised program participant replaces a non-subsidized worker who therefore loses his/her job. Few studies examine these program effects, but at least some do. Evaluations would be even more comprehensive if (1) treatment distribution was estimated, (2) treatment heterogeneity was measured by observable characteristics, or (3)
program impacts on other relevant parts of the population were assessed. Indirect effects should be of primary concern to institutions directly involved with the programs, since substitution effects can completely override the positive effects. Also, various features of the distributions can identify how and why the observed effect arises. The fact that analyses rarely investigate these suggests that, in most of the cases, contractors were simply not concerned by these issues.

Conclusions

Based on the evaluations reviewed above, we come to a number of conclusions which promote an optimistic outlook.

1. Though the number of evaluation reports on the employment policy instruments in question is rather small, quite a few of these can adequately fulfil its purpose. In other words, many of them can be used to determine how a given program has affected its participants. However, there remains some room for improvement, e.g. it would be useful if future evaluations considered comparability with similar studies (Hungarian or international) an explicit priority.

2. It seems that the lack of methodological skills needed for evaluations does not hinder this kind of research. Considering the trends of the past, there are more than enough experts who would be able to conduct several analyses on a regular basis. However, it is undeniable that the integration of evaluation studies in labour market policy routine requires more experts of the same quality. Most university programs in Hungary do not provide for meeting this demand.

3. Data availability may also cease to be a problem if policymakers show interest in program effects and contribute to making the required data anonymous. Besides, there is great potential in routinely collecting data from development projects and in small alterations of program design that would facilitate identification strategies. Short, well-organized questionnaires can yield valuable additional information which can even be linked to administrative data later. Finally, it is possible to put cut-off points deliberately into a program, or conduct experiments (e.g. by randomly including or dropping regions) which are not too costly but effectively solve several methodological problems.

4. Apparently, the demand from policymakers falls below the possibilities. First, this damages the programs themselves, since their efficiency might not improve as much as it could, with the help of evaluations. Second, in the present legal environment, academic interest is not sufficient for conducting evaluations, because only politicians can initiate the anonymity status of the required administrative data. Thus, unless policymakers show more explicit interest in evaluations, they will not have the possibility to choose from a wide range of products. In order for this to change, data accessibility policies must be expanded, and policymakers must turn their attention toward this area of research. There is some reason to expect an improvement, signalled by the establishment of the new Government Centre for Impact Assessment charged
with such a mission. A single government institution, however, cannot make up for the sustained work effort of a competitive research community motivated by academic interest. These two activities are complements rather than substitutes. Therefore, the increase in government efforts calls for increasing support for external research activities as well.

Appendix 7.1. List of papers reviewed

Out of work benefits

Unemployment insurance benefits


Social benefit and unemployment assistance


165

Review article on unemployment insurance and assistance


Employment incentives
Monitoring reports of the ÁFSZ (PES)
Statistical data on the operation of the “major” ALMP instruments on the website of the ÁFSZ (PES): http://www.munka.hu/engine.aspx?page=afsz_stat_fobb aktív eszközök

Summaries and reviews of ALMP efficiency


Evaluation reports
Csoba, Judit, Nagy, Zita Éva and Szabó, Fanni (2010): Aktív eszközök, munkaerő-piaci programok kontrollosoportos, többváltozós értékelése [Evaluation of Active Labour Market Programs with Control Groups – see revised version in this volume].

Galasi, Péter and Nagy, Gyula (2005): Az aktív programokban résztvevők állásba lépési esélyei és az aktív programok időtállamát meghatározó tényezők a Monitoring adatállománya alapján [Determinants of Transition to Work Probabilities of Active Program Participants on the Basis of “Monitoring” data].


Galasi, Péter, Lázár, György and Nagy, Gyula (1999): Az aktív foglalkoztatóspolitikai eszközök hatásságát meghatározó tényezők [Determinants of the Ef-


## Appendix 7.2.

**Table 7A1: Out of work benefits**

<table>
<thead>
<tr>
<th>Impact of the receipt of unemployment insurance benefits on re-employment</th>
<th>Target group</th>
<th>Database</th>
<th>Sample size</th>
<th>Observation period</th>
<th>Identification method</th>
<th>Quantitative findings</th>
</tr>
</thead>
<tbody>
<tr>
<td>Galasi and Nagy (2002a)</td>
<td>UI recipients</td>
<td>UI register new entrants in 1 Jan – 15 March 2000 Excluded voluntary quits and severance pay recipients.</td>
<td>31,031 control 27,947 treatment group</td>
<td>9–12 months</td>
<td>Quasi experiment: new claimants after 1 Feb are entitled to UI for a period 25% shorter (worst case); Kaplan Meier survival functions, for treated and control group, right censored, by sex and four subgroups by prior employment spell, which determined length of UI entitlement.</td>
<td>No effect. Control group exit rates are even higher for some of the subgroups which may be explained by the higher share of recalled workers in January claims (control group).</td>
</tr>
<tr>
<td>Bódis and Micklewright and Nagy (2004) = Micklewright and Nagy (2010)¹</td>
<td>Entered UI register and entitled to 75–179 days of UI benefit.</td>
<td>Interview surveys and PES registers Entrants between 26 May and 26 July 2003</td>
<td>479 w aged &lt;30 1,037 men(longer entitlements excluded to control for the intro of a new incentive)</td>
<td>4–6 months following entry to register</td>
<td>Experiment. Treatment: 4 visits to PES and questions on job search in 3 month (control: no visit in 3 months). Right censored (excl. exhausters) Conditional prob of exit to job or ALMP, proportional hazard with treatment dummy and controls for indiv. char. and local u.</td>
<td>Hazard ratio for women over 29 is 1.43 (43% over control group’s)</td>
</tr>
<tr>
<td>Köllő and Nagy (1996)²</td>
<td>UI recipients re-entering employment 4 subgroups: (a) job losers with &lt;181 days in UI (b) Job losers with 180+ days in UI, (c) voluntary quits, (d) recalled workers</td>
<td>interview survey of re-employed + PES register of UI recipients; exits from UI register to a job between March 20 and April 20, 1994</td>
<td>9,420 divided into 4 subgroups. (a) 3,839, (b) 3,092, (c) 383 (d) 2,106 Q: tested selectivity of non-response (18%) Weighting observations with the inverse of the predicted non-response rate does not affect the results.</td>
<td>NA, less than 270 days (max duration of UI)</td>
<td>OLS on log(w1/w0) – ΔlogW, dep on individual and job characteristics, controlling for local u. Subgroups justified by Chow tests of pooling restrictions; parameters are jointly significant, heteroscedasticity is rejected; Ramsey test for omitted variables not rejected for (a) and (d).</td>
<td>The median unemployed lost 5.2% percent in real terms. Duration of UI spell: compared to a spell lasting for six months the new wage is estimated to be 5% higher if completed duration was 0–3 months, and almost 5% lower if the spell lasted for one year.</td>
</tr>
<tr>
<td>Köllő (2001)</td>
<td>UI recipients</td>
<td>interview survey of re-employed (+ PES register of UI recipients) exits from UI pool between 18 March and 7 April 2001</td>
<td>1994: 8,549 (238,841) 2001: 8,339 (105,924) (excluding those exhausting UI during period observed and recalled workers)</td>
<td>9–12 months</td>
<td>multinomial logit (1) stays in UI, (2) exit to new job, (3) exit to old job, (4) exit to unknown job, controls for indiv char, past lm experience (e, u), tests robustness with alternative specifications. Cross section.</td>
<td>Remaining benefit on exit to new job: -0.182 Entitlement: for upper secondary &amp; graduates: if E/UI &lt; 50, odds of exit is 1.56 times higher towards end of eligibility.</td>
</tr>
<tr>
<td>Impact of the amount of unemployment insurance benefits on re-employment</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>-----------------------------------------------</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Micklewright and Nagy (1995)</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Target group: UI recipients</td>
<td>Database: UI register new entrants in Dec 1992 and Jan 1993</td>
<td>Sample size: 50,441 control, 30,270 treatment group</td>
<td>Observation period: 3–19 months</td>
<td>Identification method: Quasi experiment: after 1 Jan, 1st phase of UI is shorter (1/4 of old system) but replacement rate is higher (75 vs 70%); Kaplan Meier survival functions and hazards, for treated and control group, right censored, by sex and four subgroups by prior employment spell, which determined length of UI entitlement.</td>
<td>Quantitative findings: No effect. Treatment group exit rates are higher for some of the subgroups, but this is most likely due to the higher share of recalled workers in January claims (the treatment group in this case).</td>
<td></td>
</tr>
<tr>
<td>Wolff (2001) I.</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>UI recipients</td>
<td>Database: UI register new entrants in Dec 1992 and Jan 1993</td>
<td>Sample size: 18,995 w, 7,031 (control), 12,914 m, 5,397 (treatment)</td>
<td>Observation period: 3–19 months</td>
<td>Identification method: Quasi experiment, Kaplan Meier survival as in M&amp;N1995, but only for a sub-sample considered less likely to be recalled workers based on previous job history.</td>
<td>Quantitative findings: No effect</td>
<td></td>
</tr>
<tr>
<td>Wolff (2001) II.</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>UI recipients, excl. older workers</td>
<td>Database: UI register new entrants in Dec 1992 and Jan 1993</td>
<td>Sample size: 13,121 (control), 10,373 (treatment) m aged below 55, 6,162 (control), 5,047 (treatment) w aged below 50</td>
<td>Observation period: 3–19 months</td>
<td>Identification method: Quasi experiment, data as in M&amp;N1995, but using variation in entitlement and replacement rates. ML estimate of semi-parametric continuous duration model, tests alternative specifications.</td>
<td>Quantitative findings: No robust effect for men, small robust effects for women: entitlement effect: job hazard is 53% higher than base (over 270 days) in the last 30 days. For women &lt;30: Elasticity wrt UI –0.35, wrt wages 0.31</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Impact of unemployment assistance on re-employment</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Firle and Szabó (2007) I.</strong></td>
</tr>
<tr>
<td>Target group: Exited UI</td>
</tr>
<tr>
<td><strong>Firle and Szabó (2007) II.</strong></td>
</tr>
<tr>
<td>Target group: Non-employed (excluding those not seeking a job because are in full time/ill/disabled/caring for family member) aged 18–62</td>
</tr>
<tr>
<td>Target group</td>
</tr>
<tr>
<td>--------------</td>
</tr>
<tr>
<td>Galasi and Nagy (2002b)</td>
</tr>
<tr>
<td>Micklewright and Nagy (1998)</td>
</tr>
</tbody>
</table>

1 The Hungarian LFS is a rotating panel where an individual may be included for a maximum of 6 consecutive quarters.
2 Success criterion: wage gain (compared to the average rise in wages for the UI pool in the same period).
Table 7A2: Wage subsidies

<table>
<thead>
<tr>
<th>Type of program</th>
<th>Target group</th>
<th>Database</th>
<th>Sample size</th>
<th>Observation period</th>
<th>Identification method</th>
<th>Quantitative findings</th>
</tr>
</thead>
<tbody>
<tr>
<td>O’Leary (1998) (also in O’Leary and Nesporova (2001))</td>
<td>Wage subsidy paying up to 50% of the wage bill up to one year. Employment must be sustained for an identical period after exit from program. (Provides a similar evaluation on training programs and public works too.)</td>
<td>Longer term registered unemployed</td>
<td>Survey data collected following-up supported individuals and a randomly selected control group. Treated: exit from program: Q2 1996, observed: up to Q2 1997 Control: entered registered status in Q2 1995</td>
<td>Whole sample: 9,219 treated: 1,131, control: 3,338 (training: 2,543; public works: 1,140; self-emp: 1,067)</td>
<td>OLS on exit with control group. Matched pairs, interaction terms in linear OLS. Personal and regional characteristics used in OLS and matching.</td>
<td>Effect on employment probability: 17–24% points if unadjusted/unmatched, 0 to -6% point with controls. No effect on earnings. Also significant parameters on individual characteristics</td>
</tr>
<tr>
<td>Galasi and Nagy (2005)</td>
<td>As in O’Leary (1998)</td>
<td>Long term registered unemployed</td>
<td>Survey data collected following-up supported individuals and a randomly selected control group. Treated: exit from program: between September 2009–February 2010</td>
<td>Treatment group: 1,041; control group: 1,068</td>
<td>logit on exit with control group. Personal and regional characteristics used in logit estimation.</td>
<td></td>
</tr>
<tr>
<td>Csoba, Nagy and Szabó (2010)</td>
<td>As in O’Leary (1998), but support payable up to 100% of the wage bill (new regulation)</td>
<td>Longer term registered unemployed</td>
<td>Survey data collected following-up supported individuals and a randomly selected control group. Treated: exit from program: between September 2009–February 2010</td>
<td>6 months</td>
<td>logit on exit with control group. Personal and regional characteristics used in logit estimation.</td>
<td>No marginal effect, significant positive effect on program parameter (odds ratio compared to the control group: 24)</td>
</tr>
</tbody>
</table>
### Galasi and Nagy (2008)

<table>
<thead>
<tr>
<th>Type of program</th>
<th>Target group</th>
<th>Database</th>
<th>Sample size</th>
<th>Observation period</th>
<th>Identification method</th>
<th>Quantitative findings</th>
</tr>
</thead>
<tbody>
<tr>
<td>As in O’Leary (1998), but also looking at public works and training</td>
<td>Registered unemployed</td>
<td>Individual data snapshots (sampled from PES register) matched with monitoring data. Sampling: June 2005–January 2006</td>
<td>N = 351,787 (7.6% of which in one of the three types of programs)</td>
<td>6 months</td>
<td>discrete time duration model of hazard to exit towards ALMP registry drop out hazard (Jenkins logit)</td>
<td>UI recipients have 33% higher probability of participation than those who get no subsidy, social benefit recipients: 50% less</td>
</tr>
</tbody>
</table>

1. Success criteria: exit to employment.
2. Continued employment with the same employer.
3. Participated in one of the programs analysed.
4. Wage if employed and use of UI.
5. Six months, three months if labour market entrants.
8. REFERENCES


FREY, MÁRIA (2010): A foglalkoztatási törvényben rögzített és az ÁFSZ által működtetett, továbbá az ezekben kívüli szabályozott és bonyolított aktív munkaerő-piaci eszközök értékelése a 2004–2009 közötti időszakban [Evaluation of active labour market measures regulated by the Employment Act and managed by the


LALonde, R. J. (1995): The Promise of Public Sector-Sponsored Training Programs. Journal of Economic Per-


Programterv (2010a): Program documentation for the regional labour market program Life Changing – Life Shap-