

## **Training for the Urban Unemployed: A Reevaluation of Mexico's Probecat**

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### Abstract

For many years the Government of Mexico has implemented a large training program for the unemployed. The program has been evaluated twice with similar methodologies. These two evaluations yield encouraging results in that the program apparently reduces unemployment and increases earnings. This paper suggests that both evaluations may suffer from inappropriate controls for the endogeneity of program participation. Using the availability of the program at the state level as a determinant of individual participation, the paper uses the data of the second evaluation to indicate that Probecat does not decrease unemployment, nor does it increase wages.

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## I Introduction

Since 1984, the Government of Mexico has implemented a training program (Probecat) for the unemployed living mostly in urban areas. The program has been evaluated twice, first by the World Bank using data from 1992 (Revenge, Riboud, and Tan, 1994), and next by the Mexican Ministry of Labor using data from 1994 (STPS, 1995)<sup>2</sup>. Both evaluations use longitudinal data, comparing a sample of Probecat participants (the treatment group) with a sample of unemployed individuals from Mexico's urban employment survey (the control group). The second evaluation closely follows the method used in the first evaluation. The two evaluations yield positive results. Controlling for other characteristics such as education, family situation, and past professional experience, program participants find employment faster than non-participants and they earn higher wages. These encouraging results have been used to support the extension of the program, which now serves more than 500,000 beneficiaries per year (period from 1996 to 1998), as compared to only 50,000 in the first decade of the program (period from 1984 to 1993).

This paper reassesses the results of past evaluations of Probecat. We suggest that past evaluations suffer from inappropriate controls for the endogeneity of program participation. For example, past evaluations conclude from Cox regressions that the time necessary to find employment after the training is reduced for program participants, but this may be due to sample selection since program participants may also be those most eager to work. We argue that the matching technique used in past evaluations of Probecat are not sufficient for taking care of this type of problem. Following Ravallion and Wodon (1998), we use an alternative econometric method for evaluating the impact of Probecat. Specifically, the availability of Probecat at the state level is used as an instrumental variable to control for the endogeneity of program placement. We then find that Probecat has no impact on unemployment and on wages.

The disappointing results of Probecat in terms of raising wages and employment should not be surprising. Most retraining programs in OECD have been found to have limited impacts, and when programs have been found to have some impact, this impact tends to vanish after a few years (Dar and Gill, 1998). The fact that Probecat may not be beneficial in the medium to long run for participants does not mean that it should be suppressed. The program could be considered as providing temporary safety nets (through the minimum wage stipend) rather than training. Or it could be improved so as to provide training with longer lasting effects. But in order to motivate an inquiry into how to improve the program, it must first be recognized that contrary to what the result of past evaluations, the program may not be satisfactory.

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<sup>2</sup> An earlier evaluation was attempted by Carlson (1991), but this evaluation does not consider the problem of endogeneity of program placement or selection bias which is crucial for good evaluations.

The structure of the paper is as follows. After a description of the program and of the changes made over the years in section 2, we review the results of past evaluations in section 3. We present our own new results in section 4. A conclusion with policy implications follows.

## **II Program description**

The Mexican Job Training Program for Unemployed Workers (Programa de Becas de Capacitación para Desempleados — Probecat) was established in 1984 in response to rising unemployment and deteriorating living standards in the aftermath of the 1982 economic crisis. Despite a comparatively low open unemployment rate (6 percent in four major cities in 1984), Mexico suffered (then as now) from chronic underemployment (for a discussion of unemployment and underemployment data in Mexico, see for example Fleck and Sorrentino, 1994). Moreover, there was a recognition of the existence of shortages of adequately trained labor in selected growing sectors of the economy. Thus the stated objectives of the program were to improve the productivity of unemployed workers in order to help them find employment. The program has become massive in recent years. In the first decade of its existence (1984-1993), Probecat provided training to roughly 50,000 people per year. But this rose to 199,000 people in 1994, 412,000 people in 1995, and about 550,000 people per year between 1996 and 1998.

PROBECAT is administered by the Mexican Ministry of Labor (*Secretaría de Trabajo y Previsión Social, STPS*). The program provides publicly funded job training and a subsistence allowance during the training period to participating unemployed workers. Initially, in 1986, Probecat provided training in high schools and other training centers. This first modality of the program is referred to as school-based training (*Cursos Escolarizados*). Later, to strengthen the link between the training provided under the program and the actual needs of the productive sector in the economy, a new module of so-called in-service training was added (*Cursos Mixtos*). Under this module, local employers provide training while the workers' stipend is paid by the Government. Upon completion of the training, the employers are required to hire at least 70 percent of the trainees. A third module of the program consisting of training for the self-employed (*Programa de Iniciativas Locales de Empleo y Ocupación Temporal, PILEOT*) was created in 1995 in response to rapidly increasing unemployment after the 1994 financial crisis.

For the school-based and in-service modules, program beneficiaries are selected from the unemployed workers who register in the State Employment Service offices. The applicants' job skills and interests are evaluated against the needs of the local market. Apart from basic requirements for all, the selection procedure gives variable weights to different criteria. Only individuals with a total composite score exceeding a threshold level are eligible to participate. Participants can obtain training only once. The training lasts for two to three months on average,

and participants receive stipends equivalent to the value of the minimum wage, plus transportation costs to the training site and basic health insurance coverage while on training. More details on the eligibility and features of each of the three modules are provided in Table 1. It is worth mentioning that the share of in-service training within the Probecat beneficiaries increased from about five percent in 1987-1992 to 20 percent in 1993, but dropped to 13 percent in 1996 as a result of the rapid expansion of the new PILEOT modality created in 1995.

### **III Past evaluations**

As mentioned in the introduction, two main evaluations of Probecat have been conducted so far. The first evaluation was prepared by the World Bank (Revenga, Riboud, and Tan, 1994). This evaluation used data from a survey administered in 1992 to Probecat trainees from the 1990 cohort. Data was gathered on unemployed individuals in the 1990-91 National Urban Employment Survey (Encuesta Nacional de Empleo Urbano, ENEU) to construct the control group. The second evaluation was conducted in 1995 by the Ministry of Labor with a similar methodology (STPS, 1995) and data for employment in 1993-94. Both studies seek to evaluate the impact of Probecat according to two main indicators: (i) the time to find a first job after the training and (ii) monthly earnings. The studies also contain information on hours worked and hourly wages, as well as cost-benefit simulations, but these will not be discussed here. Whereas Probecat participants are drawn only from the school-based modality for the 1992 study, the 1995 study evaluates the impact of both the school-based and in-service modalities.

Both studies use the post-training labor market experiences of randomly selected program participants who graduated in the previous year, and they compare these experiences to that of a control group. For this control group, the studies use panel data on unemployed individuals drawn from the quarterly urban employment survey (ENEU) covering the main urban areas of Mexico (Probecat is now available in rural areas, especially through the PILEOT modality, but this was not the case before). The ENEU uses a quarterly rotation system so that each rotation group of individuals remains in the survey for five consecutive quarters. The choice of the quarterly data in the ENEU matches that of the survey of the Probecat participants. The ENEU survey includes detailed information on employment status, monthly earnings, and hours worked per week. The Probecat survey administered to program participants was designed to match the questions in the ENEU, so that the information for the two groups are to a large extent comparable. That is, in addition to standard socio-demographic characteristics, the Probecat survey elicited information on the current or last job since completing Probecat, the first job after the training, the second job after the training, and the last job prior to participating in the program.

The two studies acknowledge that there may be unobservable characteristics influencing program participation and outcomes, and that these characteristics may be different between the treatment group (Probecat participants) and the control group (ENEU unemployed workers). To control for selection bias, the studies use two different approaches. The first approach is used for analyzing the length of the employment search, and the second for monthly earnings.

For the analysis of the length of the employment search, the studies rely on probit regressions for the probability of participating in Probecat in order to construct the control group. In the 1992 study, the individuals below a certain probability (0.6) of participating are eliminated from the sample. It is less clear what exactly is done in the 1995 evaluation, but the principle is the same. Next, using the (slightly) reduced sample of pooled treatment and control groups members, Cox proportional hazards model are estimated to assess the impact of training on the time necessary to find employment. In 1995 for example, the regressors include dummy variables for the individual's age, the level of education, marital status, household structure, prior work experience, and characteristics of employment before being unemployed. The dependent variable is the number of months needed for finding employment, counted from the first month when the individual enters the control group (ENEU) or completes the training for participants.

For the analysis of earnings, both studies use Heckman's (1979) sample selection model. In the 1992 study, the first equation refers to the logarithm of monthly earnings, and the second equation to the probability of participating in the Probecat training. In the 1995 study, while the participation equation is the same, the first equation measures the difference in the logarithms of earnings before and after program participation with the logarithm set to zero in case of unemployment (none of the studies actually provides the participation equations).

Overall, the results obtained in the two studies are encouraging for the program. In 1992, Probecat is found to reduce the length of unemployment for men, but not for women. In the 1995 study, both the school-based and in-service modules reduce the time needed to find employment for both men and women. As for monthly earnings, the 1992 study finds positive impacts for men, but again not for women. The 1995 study finds positive impacts for men, and negative impacts for women. Both studies also conduct cost-benefit analyses. The program costs are calculated using administrative data from the state and national program offices. Direct costs include the remuneration for instructors, the costs of training materials and facilities, and administrative costs for program operation. The training stipends paid to the program participants during the training are not included in the costs because they are considered as safety net transfers. The treatment of indirect opportunity costs for participating in the program differ, but overall the two studies conclude that the program is cost effective and performing relatively well.

While both studies are carefully implemented, several critiques can be made as to the methodology used. Manski (1996) mentions a few. First, in using the unemployed individuals in the ENEU to form the control group, it is assumed that none of the ENEU individuals have benefitted from the program. This is not the case since every individual in the ENEU has some probability of having participated in Probecat. Fortunately, given that the program was small until 1993, only a very small minority of the individuals in the control group are likely to have participated in the program (this would not be true for future evaluations). Next, Manski points out that the combination of two random samples (Probecat trainees and ENEU unemployed individuals) is not a random sample, so that in the absence of the standard properties for the residuals, the results of regressions may not yield consistent parameter estimates, especially since the models used are sensitive to the assumption of bivariate normality (Goldberger, 1983). In the absence of better data, not much can be done about this using standard techniques. Third, Manski argues that there is no theoretical proof that matching methods do indeed provide for a solution to the sample selection problem, although this could be debated. Finally, Manski notes that in the estimation of earnings, while participation in Probecat is controlled for, the sample selection bias resulting from the decision to work (which originally motivated the Heckman model) is not account for, which is recognized by the authors of the 1992 study.

In our own evaluation of Probecat to be presented in the next section, we do not solve all of the above problems, but we do try to solve some. Consider first the analysis of the duration of unemployment. The method used to control for endogeneity in past evaluations is rough because the matching is imprecise. Typically, when using matching procedures, one matches every participant with one (or perhaps a few) non-participants by minimizing the distance between the probability of participation of each participant and that of his match among the non-participants (Rosenbaum and Rubin, 1983, 1985). Here the procedure must be reversed because there are more participants than non-participants in the pooled sample, but the logic is the same: every non-participant should be matched on an individual basis with one (or very few) participants. This is apparently not what was done in past evaluations of Probecat. The matching was apparently not done on an individual basis. Instead individuals with low probabilities of participating were excluded from the pooled sample. There is in fact evidence in both the 1992 and 1995 studies that there remains a problem of sample selection after the frequency matching since the coefficients of the inverse Mill's ratios in the Heckman regressions used for earnings tend to be statistically significant. This problem is not recognized in presenting the Cox regressions.

Consider next the earnings regressions. As noted by Manski, while the Heckman procedure allows in principle to control for the endogeneity of program placement, there is no

control for the endogeneity of the decision to work. The procedure that will be proposed in the next section does provide a way to control for both types of endogeneity in the estimation.

## IV Alternative evaluation

### IV.1 Participation equations

To evaluate the impact of Probecat, we use the same data as that used in the 1995 study, but with an alternative methodology. Following Ravallion and Wodon (1998), we use the program availability in a geographic area as a determinant of program participation at the individual level, assuming that program availability does not influence outcomes (unemployment duration and earnings) conditional on individual participation. To measure geographical availability, we use one thousand times the number of Probecat participants in a given state (Mexico is a federal entity with 32 states) as a proportion of the urban population in that state (up to recently, the program was targeted mainly to urban areas, and this holds for the 1994 data). We use the same measure of program availability at the state level for the two modalities (school-based and in-service) because we do not have data on the availability of each modality separately.

As noted by Ravallion and Wodon (1998), if the program availability at the state level is to be used as an instrumental variable for determining program participation, apart from individual level variables, it is important to include in the regressions state level variables which may affect program participation as well as outcomes. A full set of state dummies cannot be included because in this case the program availability at the state level would be collinear with the state dummies. But state variables can be included. We used thirteen such variables as controls. These are not shown in the regression Tables. They include primary and secondary schooling and spending indicators, population density, shares of the urban and indigenous populations, state-level income, and variables related to wealth and consumption (cars, water, and electricity).

To tackle the problem of sample selection for the participation in the program, we first run probit regressions for men and women separately to analyze the determinants of participation in the two modalities of Probecat available at the time of the survey (school-based and in-service). Denote by  $P1_{ij}$  the probability of participating in the school-based modality, and by  $P2_{ij}$  the probability for individual  $i$  living in state  $j$  of participating in the in-service modality.  $X_{ij}$  is a vector of individual level variables, and  $Z_j$  is a vector of state level variables for state  $j$ . The relative availability of Probecat in state  $j$  is denoted by  $AP_j$ . We estimate:

$$P1^*_{ij} = \gamma_{p1}'X_{ij} + \delta_{p1}'Z_j + \mu_{p1}AP_j + \mu_{p1}AP_j^2 + \varepsilon_{p1ij} \text{ with } P1_{ij} = 1 \text{ if } P1^*_{ij} > 0 \text{ and } 0 \text{ if } P1^*_{ij} \leq 0 \quad (1)$$

$$P2^*_{ij} = \gamma_{p2}'X_{ij} + \delta_{p2}'Z_j + \mu_{p2}AP_j + \mu_{p2}AP_j^2 + \varepsilon_{p2ij} \text{ with } P2_{ij} = 1 \text{ if } P2^*_{ij} > 0 \text{ and } 0 \text{ if } P2^*_{ij} \leq 0 \quad (2)$$

The results of these probits are given in Table 2. Individuals between 15 and 55 are more likely to participate in the program than younger and older individuals, which corresponds to the eligibility rules. Better educated individuals also tend to participate more (in comparison with the excluded category in the survey which corresponds to the illiterate and those not having completed primary school), again because having completed primary school is a requirement. Being married has an impact on participation only for women in the school-based modality. There is also evidence that individuals with several workers (apart from the individual considered in the sample) have higher participation rates. Having previous work experience is negatively correlated with participation (maybe because these individuals need less training), but having worked in the last year is positively correlated (maybe because these individuals remain actively seeking employment). Having worked in firms with other workers (instead of having been self-employed) facilitates participation, as does the fact of having worked a large number of hours in the last year. Higher incomes in the last year also influence positively the probability of participating. Finally, individual participation is positively correlated with the program availability at the state level. This is important because it confirms that state-level availability of the program is a valid instrumental variable. (There is no need to discuss here the impact of the other thirteen state-level variables included in the regressions but not shown in the Tables)

#### ***IV.2 Impact of Probecat on the length of employment search***

We now consider the impact of Probecat on the length of the employment search. In the terminology of survival analysis, the survivor function  $S(t)$  represents the length of unemployment after training ( $t$  is measured in months). Given  $S(t)$ , the hazard function  $\lambda(t)$  denoting the chance of becoming employed (or the risk of remaining unemployed) at time  $t$  among the individuals who are not yet employed at that time is  $\lambda(t) = -d(\log S(t))/dt$ . The survivor curve can be specified as a function of program participation, individual characteristics, and state characteristics, so that  $\lambda = \lambda(t; X, Z, P1, P2)$ . In Cox's proportional hazard model, we have:

$$\lambda(t; X, Z, P1, P2) = \lambda_0(t) \exp(\gamma' X_{ij} + \delta' Z_j + \mu_1 P1_{ij} + \mu_2 P2_{ij}) \quad (3)$$

Cox proposed a partial maximum likelihood estimation of this model in which the baseline function  $\lambda_0(t)$  does not need to be specified. The relative chance of being employed (or risk of remaining unemployed) of two individuals can then be compared. Consider two identical individuals, except for their participation in the school-based and in-service Probecat modalities. At any given time  $t$ , the ratio of the hazard rates for the two individuals, also referred to as the relative risk ratio, is  $\exp(\mu_1/\mu_2)$ . If  $\mu_2$  is larger than  $\mu_1$ , all other things being equal, the individual



having received the in-service training has a higher probability of finding employment than the individual having received the school-based training. If both  $\mu_1$  and  $\mu_2$  are positive, the individuals receiving any one of the two forms of training are likely to find employment before the individuals who did not receive any training.

The results of the Cox regressions are given in Table 3. The first columns under the “naïve” heading were obtained by using the hazard function  $\lambda_0(t) \exp(\gamma'X_{ij} + \delta'Z_j + \mu_1P1_{ij} + \mu_2P2_{ij})$  where  $P1_{ij}$  and  $P2_{ij}$  denote the actual participation in the two programs. The columns under the heading “control” were obtained by using instead  $\lambda_0(t) \exp(\gamma'X_{ij} + \delta'Z_j + \mu_1IP1_{ij} + \mu_2IP2_{ij})$ , where  $IP1_{ij}$  and  $IP2_{ij}$  denote the index values obtained from the estimation of the probits. The index values are the right hand sides of the probit equations less the residuals (not the expected probabilities of participation which can be computed using the normal distribution).

If Probecat participants are among the more dynamic individuals who are willing to make sacrifices in order to be trained and to find employment, we would expect that in comparison with naïve estimates, the impact of Probecat would be smaller when suitable controls are introduced for the endogeneity of program participation. This is exactly what we observe. In the naïve estimates, because we use a model which is very similar to that used in the 1995 study, we get coefficients estimates which are fairly close to those obtained in that study (see Table 3). These naïve estimates indicate that training reduces the length of unemployment for both men and women. The impacts are apparently statistically significant at the 10 percent level for the school-based modality, and at the 5 percent level for the in-service modality. Yet, once we use the index values from the probits instead of the actual program participation indicators, these positive impacts vanish. We even observe negative impacts in the case of men for the school-based training, although these are not statistically significant at the 5 percent level.

The results obtained for the school-based modality may not sound too surprising given the short period of time during which individuals receive training. But how can it be that despite the requirement for employers to hire 70 percent of the trainees in the in-service modality, we do not observe any positive impact on employment after controlling for the endogeneity of program placement? It must be that without the stipend (wage subsidy) provided by the Government, firms participating in the in-service modality would have hired the same workers anyway. This is referred to as a deadweight loss in the literature. In OECD countries, the impact of such dead weight losses has been shown to represent from 40 to 90 percent of all hires (Foley, 1992)

### IV.3. *Impact of Probecat on earnings*

Consider now the impact of Probecat on monthly earnings. For this, we use a standard Heckman sample selection model, but in a different ways than in past evaluations of Probecat. Denote by  $\log w$  the logarithm of the expected wage for an individual. This wage is non zero if and only if it is larger than the individual's reservation wage (otherwise, the individual would choose not to work). Denote the unobserved difference between the individual's expected wage and his reservation wage by  $\Delta^*$ . The individual's expected wage is determined by a number of individual (essentially the individual's education and past experience) and state variables. The difference between the individual's expected wage and his reservation wage is determined by the same variables, plus the number of children, the fact of being a household head, and the fact of being married. If we split the individual level variables into those which influence both expected earnings and the reservation wage (vector E) and the demographic variables which influence the reservation wage only (vector D), the standard Heckman model is:

$$\Delta_{ij}^* = \phi_{\Delta}'E_{ij} + \pi_{\Delta}'D_{ij} + \eta_{\Delta}'Z_j + \alpha_{\Delta}P1_{ij} + \chi_{\Delta}P2_{ij} + v_{ij} \text{ with } \Delta_{ij} = 1 \text{ if } \Delta_{ij}^* > 0, \text{ and } 0 \text{ if } \Delta_{ij}^* < 0 \quad (4)$$

$$\text{Log } w_{ij}^* = \phi_w'E_{ij} + \eta_w'Z_j + \alpha_wP1_{ij} + \chi_wP2_{ij} + \kappa_{ij} \text{ with } \text{Log } w = \log w^* \text{ if } \Delta=1 \text{ and } 0 \text{ if } \Delta=0 \quad (5)$$

The above model controls for the endogeneity of labor force participation. We estimated this model first with the actual values of the participation dummies  $P1_{ij}$  and  $P2_{ij}$  (this is the naïve estimation in Table 4), and next with the index values from the probit regressions (this is the control estimation in Table 4). By estimating the program participation equations first, and then using the standard Heckman model, we are able to control for both sources of bias at once (note that all the coefficients of the inverse Mills' ratios are statistically significant in Table 4, which highlights the importance of controlling for endogenous participation in the labor force). Our parameter estimates are in principle consistent, although they may not be efficient because the first probit equation is estimated separately rather than with the Heckman model. How do the results of the naïve and control estimates compare? There are less differences than with the Cox model. In both the naïve and control estimations, the impacts of Probecat are negative instead of being positive. While the levels of significance differ according to the method, these results at the least shed doubts on the positive impact of Probecat observed on earnings in past evaluations.

## V Conclusion

During the last 15 years, Probecat has been implemented by the Mexican Government as a training program for the unemployed. Past evaluations of the program suggest that it is effective in reducing the length of unemployment and increasing earnings for participants. But it could be

that these results were obtained because of inadequate consideration of the problem of sample selection. Using the data of the last evaluation of Probecat conducted by the Mexican Ministry of labor, we have proposed another methodology for assessing impacts. According to our results, Probecat does not have large positive effects for participants. From a policy point of view, this finding suggests that the program may not be delivering its promise. From a scientific point of view, our results point to the sensitivity of evaluation results to the methodologies used to generate the results, which is of concern when these results are used for policy recommendations.

New initiatives have been taken in recent years to try to improve the functioning of Probecat. This includes the implementation of the Pileot modality since 1995, which is targeted at economically disadvantaged communities. The only requirement for participation in Pileot is that the applicant must have basic literacy and numeracy skills, and be unemployed or under-employed. The modality provides training for individuals and groups who intend to engage in self-employment or community-based productive activities. The training contents are demand-driven. It will be important to evaluate this new modality rigorously, and to test the sensitivity of the evaluation results to the assumptions made and the techniques used in the evaluation process.

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**Table 1: Modalities of Probecat’s main training modalities**

	<b>School-based training</b>	<b>In-service training</b>	<b>Pileot</b>
<b>Eligibility rules</b>	Unemployed candidate registered with SES, aged 16 to 55, having completed primary school and having at least 3 months of experience.	Unemployed, registered with SES, aged 18 to 55, having completed primary school (this can be waived by firm). No prior experience required.	Unemployed aiming at self-employment, aged 16 to 55, literate, no upper secondary schooling. Special module for community activities.
<b>Training provider</b>	Training Schools/Centers.	Participating firms.	Training Centers/Instructors
<b>Training duration</b>	1 to 6 months.	1 to 3 months.	1 to 3 months.
<b>Benefits received</b>	Training, minimum wage, transportation costs, health insurance under IMSS.	Training, minimum wage, transportation costs, health insurance under IMSS.	Same, plus a set of tools for self employment module.
<b>Training costs</b>	Probecat program.	Firm (cost of instructors, equipment, and materials.)	Probecat program.
<b>Training content</b>	Set by training provider with little customization.	Set by firm. Mostly hands-on training.	Set by the training provider with customization.
<b>Placement</b>	Required to register with SES. No placement.	Firms required to employ 70 percent of the trainees.	No particular follow-up.

Source: Created by the authors from information provided by the Ministry of labor (STPS).

**Table 2: Participation regressions (probits) for Probecat by modality**

	Escolar				Mixto			
	men		women		men		women	
	dF/dx	P> z	dF/dx	P> z	dF/dx	P> z	dF/dx	P> z
Aged between 15 and 25	0.183	0.004	0.288	0.243	0.283	0.997	0.973	0.000
Aged between 16 and 55	0.229	0.000	0.419	0.084	0.378	0.997	0.984	0.000
Primary education completed	0.158	0.000	0.174	0.058	0.902	0.997	0.068	0.146
Secondary or post-primary training	0.321	0.000	0.317	0.000	0.325	0.997	0.021	0.593
Higher level	0.248	0.000	0.318	0.001	0.627	0.997	-0.075	0.030
Household head	0.080	0.010	0.153	0.033	0.000	0.606	-0.008	0.796
Married individual	0.024	0.404	0.104	0.043	0.001	0.092	0.012	0.564
Household with two workers	0.050	0.022	0.029	0.505	0.000	0.416	0.053	0.009
Household with 3+ workers	0.042	0.078	0.027	0.601	0.000	0.915	0.042	0.078
Previous professional experience	-0.068	0.103	-0.067	0.366	-0.004	0.035	-0.217	0.000
Working experience in past year	0.108	0.006	0.115	0.061	0.000	0.891	0.021	0.434
Past year in firm 1-15 workers	0.103	0.002	0.243	0.000	0.001	0.309	0.026	0.433
Past year in firm 16-100 workers	0.035	0.355	0.156	0.055	0.003	0.041	0.094	0.048
Past year in firm 101-250 workers	0.121	0.006	0.056	0.609	0.001	0.301	0.123	0.072
Past year in firm 251+ workers	0.005	0.897	-0.048	0.535	0.001	0.303	0.159	0.002
Worked 35-48 hours in past year	0.087	0.000	0.084	0.075	0.000	0.451	0.006	0.763
Worked 48+ hours in past year	0.075	0.000	0.091	0.046	0.000	0.550	0.037	0.076
Income past year	0.000	0.000	0.000	0.001	0.000	0.664	0.000	0.911
Income squared past year	0.000	0.019	0.000	0.114	0.000	0.394	0.000	0.752
Program availability	0.107	0.000	0.092	0.014	0.003	0.005	0.155	0.015
Program availability squared	-0.004	0.001	-0.004	0.233	0.000	0.004	-0.018	0.020
Number of children			0.025	0.463			0.017	0.229
Number of children squared			-0.002	0.617			-0.003	0.198
Number of observations	2160		947		2160		947	
Pseudo R <sup>2</sup>	0.297		0.233		0.351		0.331	

Source: Authors' estimation from pooled Probecat and ENEU panel surveys (1993-94). Thirteen state-level control variables are included in the regression but not shown in the Table. See text for more details.

**Table 3: Cox regressions for number of months unemployed**

	Naïve				Control			
	men		women		men		women	
	Coef.	P> z	Coef.	P> z	Coef.	P> z	Coef.	P> z
<b>Program impact in 1995 study</b>								
Probecat – Escolar	0.107	0.115	0.361	0.002				
Probecat - Mixto	0.476	0.000	0.776	0.002				
<b>New estimates of impact</b>								
Probecat – Escolar	0.121	0.094	0.221	0.055	-0.156	0.063	0.360	0.074
Probecat - Mixto	0.436	0.000	0.742	0.000	-0.023	0.091	0.045	0.284
<b>Other variables in estimation</b>								
Aged between 15 and 25	1.031	0.000	0.664	0.203	1.363	0.000	0.388	0.546
Aged between 16 and 55	1.032	0.000	0.705	0.179	1.376	0.000	0.264	0.693
Primary education completed	0.054	0.672	0.726	0.001	0.362	0.031	0.569	0.016
Secondary or post-primary training	0.096	0.427	0.403	0.059	0.510	0.007	0.113	0.679
Higher level	0.000	0.997	0.434	0.049	0.370	0.045	0.045	0.868
Household head	0.221	0.004	0.204	0.170	0.285	0.001	0.082	0.630
Married individual	0.073	0.323	-0.292	0.012	0.126	0.090	-0.350	0.007
Household with two workers	0.035	0.540	0.506	0.000	0.059	0.323	0.511	0.000
Household with 3+ workers	0.132	0.037	0.505	0.000	0.170	0.009	0.511	0.000
Previous professional experience	0.041	0.769	-0.186	0.273	-0.096	0.502	-0.208	0.260
Working experience in past year	0.436	0.000	0.124	0.354	0.523	0.000	0.029	0.850
Past year in firm 1-15 workers	-0.028	0.748	0.024	0.871	0.094	0.339	-0.169	0.417
Past year in firm 16-100 workers	-0.099	0.327	0.172	0.334	-0.008	0.935	0.099	0.629
Past year in firm 101-250 workers	-0.159	0.262	0.060	0.814	-0.006	0.967	0.037	0.890
Past year in firm 251+ workers	-0.140	0.131	0.017	0.922	-0.097	0.303	0.111	0.524
Worked 35-48 hours in past year	0.078	0.131	0.154	0.135	0.135	0.022	0.088	0.425
Worked 48+ hours in past year	0.045	0.375	0.112	0.252	0.091	0.097	0.057	0.601
Income past year	0.000	0.066	0.000	0.196	0.000	0.653	0.001	0.046
Income squared past year	0.000	0.234	0.000	0.532	0.000	0.300	0.000	0.225
Number of children			0.043	0.587			0.013	0.872
Number of children squared			-0.014	0.268			-0.010	0.419
Number of observations	2160		947		2160		947	
Log likelihood	-13407		-4200		-13412		-4217	

Source: Authors' estimation from pooled Probecat and ENEU panel surveys (1993-94). Thirteen state-level control variables are included in the regression but not shown in the Table. See text for more details.

**Table 4: Heckman regressions for wages and employment**

	Naïve				Control			
	men		women		men		women	
	Coef.	P> z	Coef.	P> z	Coef.	P> z	Coef.	P> z
<b>Logarithm of wage</b>								
Probecat - Escolar	-0.020	0.763	-0.083	0.460	-0.204	0.000	-0.080	0.421
Probecat - Mixto	-0.115	0.269	-0.300	0.021	-0.021	0.082	-0.032	0.432
Aged between 15 and 25	-0.070	0.698	-0.237	0.619	0.224	0.259	0.006	0.992
Aged between 16 and 55	0.159	0.376	-0.074	0.876	0.507	0.012	0.224	0.699
Primary education completed	0.049	0.684	-0.026	0.899	0.337	0.017	0.010	0.961
Secondary or post-primary training	0.182	0.107	0.117	0.552	0.562	0.000	0.169	0.418
Higher level	0.309	0.007	0.316	0.121	0.662	0.000	0.410	0.054
Previous professional experience	0.107	0.331	-0.287	0.036	0.164	0.136	-0.241	0.083
Working experience in past year	-0.199	0.012	0.054	0.645	-0.141	0.080	0.041	0.731
Constant	5.517	0.000	6.637	0.000	5.754	0.000	6.284	0.000
Lambda	-0.918	0.022	-1.138	0.041	-0.914	0.021	-1.175	0.044
<b>Probability of working</b>								
Probecat – Escolar	0.018	0.842	0.211	0.046	0.075	0.312	0.291	0.004
Probecat – Mixto	0.207	0.308	0.455	0.001	-0.038	0.061	0.033	0.407
Aged between 15 and 25	0.892	0.000	0.541	0.187	1.073	0.000	0.116	0.819
Aged between 16 and 55	0.810	0.000	0.482	0.245	0.937	0.000	-0.047	0.927
Primary education completed	0.041	0.813	0.574	0.003	0.241	0.274	0.469	0.014
Secondary or post-primary training	-0.043	0.796	0.191	0.286	0.106	0.637	-0.007	0.971
Higher level	-0.083	0.620	0.420	0.024	0.049	0.821	0.218	0.253
Previous professional experience	0.108	0.426	0.175	0.182	0.068	0.611	0.122	0.343
Working experience in past year	0.561	0.000	0.058	0.599	0.523	0.000	0.017	0.875
Household head	0.165	0.063	0.312	0.009	0.174	0.048	0.198	0.069
Married individual	0.039	0.655	-0.145	0.069	0.070	0.421	-0.167	0.013
Number of children			-0.023	0.635			-0.028	0.492
Number of children squared			0.000	0.978			0.002	0.674
Constant	0.700	0.299	-2.573	0.004	-1.001	0.324	-1.636	0.085
Number of observations	2167		948		2160		947	
Log likelihood	-3044		-1332		-3020		1328	

Source: Authors' estimation from pooled Probecat and ENEU panel surveys (1993-94). Thirteen state-level control variables are included in the regression but not shown in the Table. See text for more details.