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# ABSTRACT

## "For One More Year with You": Changes in Compulsory Schooling, Education and the Distribution of Wages in Europe<sup>\*</sup>

Using data from 12 European countries and the variation across countries and over time in the changes of minimum school leaving age, we study the effects of the quantity of education on the distribution of earnings. We find that compulsory school reforms significantly affect educational attainment, especially among individuals belonging to the lowest quantiles of the distribution of ability. There is also evidence that additional education reduces conditional wage inequality, and that education and ability are substitutes in the earnings function. These results do not support an elitist public education policy – more education to the brightest – and suggest that investing in the less fortunate, either because of poor labour market fortune or because of poor talent, could pay off both on efficiency and on equity grounds.

JEL Classification: J24

Keywords: education reforms, returns to education, quantile treatment effects

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#### Introduction

Does education affect earnings? This question has attracted enormous attention among labour economists, as reviewed by Card, 2001. By and large, the empirical literature has focused on the mean returns to education, with substantial effort devoted to the identification of a causal relationship. Less has been done to investigate how additional education affects the distribution of earnings. Does education reduce (conditional) wage inequality? Are the returns to education heterogeneous and is this heterogeneity correlated to ability?

These are important policy questions. If education reduces the dispersion of earnings, and equality is valued by the policy - maker, then additional schooling can be a powerful tool to combat inequality. It is well known that individual ability is strongly affected by genetic and environmental factors (see Cunha and Heckman, 2007). If education and ability are substitutes in the production of human capital and earnings, then additional investment in the former can contribute to reducing the differences generated by the latter (see Ashenfelter and Rouse, 1998).

How education and ability interact in the generation of earnings and human capital has important implications for optimal education policy. For instance, De Fraja, 2002, shows that optimal public policy is more elitist than market provision in the following sense: the difference in educational attainment between bright and less able children is greater than it would be if education were only provided privately. In this case, redistributive education policies that target the less able are bound to have a substantial cost in terms of efficiency. His results, however, require that education and ability are complements in the generation of human capital and earnings.

The paper addresses these questions by investigating the relationship between the quantity of attained education and the conditional distribution of (gross) hourly earnings in a unique sample of 12 European countries, which we have constructed by pooling together information drawn from three different surveys. Our empirical methodology is an instrumental variable approach to the endogeneity of education in a quantile regression

framework, as in recent work by Chesher, 2003<sup>1</sup>, and Ma and Koenker, 2006<sup>2</sup>. We identify the causal effects of education on earnings by using the country and time variation provided by the compulsory school reforms implemented in Europe after the end of the Second World War<sup>3</sup>.

The exogenous variation provided by minimum school leaving age laws has been used in the empirical literature since Angrist and Krueger, 1991, to identify the causal relationship between education and earnings. Since these laws have been targeted at the less educated component of the population, who typically belongs to the lower quantiles of the distribution of earnings, their use in the current context prompts the question whether the changes in compulsory education observed in Europe after the last war have been particularly beneficial to the targeted population or have spread instead their effects to the population at large.

When we treat education as exogenously assigned to individuals, we find that one additional year of schooling increases conditional wage inequality both for males and for females, in line with previous findings in the US (Buchinsky 1994) and Europe (Martins and Pereira, 2004). In our approach, we allow earnings to depend on both luck and ability, while education depends only on ability. This makes education endogenous in the earnings equation. This makes education endogenous in the earnings equation. When we allow for endogeneity, we find instead that conditional (gross) wage inequality is reduced by marginal increases in education for all ability levels.

Focusing on the mean quantile treatment effect, we find that assigning an extra year of education to males in the sample marginally reduces the estimated 90-10 conditional log wage differential by 0.99; the reduction is 1.34 in the case of females. By conditioning on selected quantiles of the conditional distribution of earnings, we investigate how the returns to education vary as we move from the bottom to the top quantile of the distribution of ability. Our key finding is that returns decrease with ability, which points to substitutability

<sup>&</sup>lt;sup>1</sup>In this approach the structural model has a triangular structure, both in the observables and in the latent variables.

<sup>&</sup>lt;sup>2</sup>Ma and Koenker, 2006, consider estimation and inference in a parametric version of the structural model developed by Chesher, 2003.

<sup>&</sup>lt;sup>3</sup>Moretti and Lochner, 2004, Lleras Muney, 2005, and Oreopoulos, 2006 use a similar approach by exploiting regional variation within a single country.

in the relationship between education and ability.

Our empirical estimates also show that the effect of compulsory schooling laws on educational attainment is statistically significant for all but the top deciles of the distribution of male education (all but the very top for females). As expected, the size of this effect declines as we move from the bottom to the top quantile. The statistically significant effect of compulsory school reforms on individuals with higher educational attainment can be intepreted as suggesting that better educated individuals react to increases in compulsory schooling by raising their own attainment, in an effort to maintain their educational advantage over the less educated, who are more directly affected by the reforms.

Overall, these results do not lend support to the elitist education policy discussed by De Fraja, 2002. They are in line with the findings of Ashenfelter and Rouse, 1998, in the US context and suggest that education policies which target the less fortunate and / or less talented group in the pursuit of equality of opportunity are not necessarily inefficient. Substitutability also indicates that since ability and parental background are closely intertwined, education policy can contribute to undo the differences generated by the latter.

The paper is organized as follows: Section 2 reviews the empirical literature and Section 3 presents the empirical model. Our identification strategy is described in Section 4. Next, we turn to the data in Section 5 and to the results in Section 6. Conclusions follow.

#### 2. Review of the Literature

By using quantile regression techniques one can trace the entire conditional wage distribution and examine how the shape of this distribution is affected by schooling, age and experience. In his pioneering work on the impact of education on the distribution of US wages, Buchinsky, 1994, uses quantile regressions and finds that returns to education in the US are higher at the higher quantiles of the conditional distribution of wages. He uses the 0.90-0.10 spread as a measure of within group inequality and finds substantial changes over time.

In the European context, Harmon, Oosterbeek and Walker, 2003, use UK data and find that the returns to schooling are higher for those at the very top of the wage distribution compared to those at the very bottom. Since individuals with higher talent are more likely to be located in the upper part of the (conditional) wage distribution, this result – they argue - points to complementarity between ability and schooling. Martins and Pereira, 2004, use data from 15 European countries and find that individuals who receive higher wages conditional on their characteristics enjoy higher education – related earnings growth. They suggest that over-education, poor school quality and the selection of fields of study with limited ex-post prospects can explain their results.

In contrast with the previous findings, Denny and O'Sullivan, 2007, use UK data and find evidence that unobserved ability and education are substitutes in the earnings function. On a similar line, Schultz and Mwabu, 1996, find complementarity among white South African males, and substitutability among black South Africans.

The idea that individuals located in the upper part of the conditional wage distributions are of higher ability than those located in the bottom part is plausibile but ignores that the allocation of individuals to different points of the distribution could also depend on other factors which are orthogonal to ability itself. Lang, 1993, for instance, distinguishes between cognitive ability, which matters for school and the labour market, and the ability which matters mainly for work<sup>4</sup>. Hornstein, Krusell and Violante, 2006, show that ex-ante identical individuals could end up with different wages because of random matching with available vacancies. Chesher, 2003, calls this labour market fortune. When conditional wages are driven by more than one unobservable factor, the observation that returns to education are higher in the upper quantiles of the conditional distribution of earnings does not suffice to establish complementarity between schooling and ability.

The reviewed studies have in common that they do not address the endogeneity of education. Because of this, their results are best interpreted as interesting associations and correlations, with little to say about causal effects. The econometric literature provides a few approaches for the identification and estimation of causal effects in quantile regressions. One such approach is due to Chesher<sup>5</sup>, who considers non-parametric identification of a structural model with a recursive structure. Chesher, 2001, points out that the continuity of

<sup>&</sup>lt;sup>4</sup> Lang's approach is different from that of Heckman, Stixrud and Urzua, 2006, who distinguish between cognitive ability and social skills. For these authors, both types of ability affect schooling and earnings.

<sup>&</sup>lt;sup>5</sup>Alternative approaches have been developed by Chernozhukov et al, 2005 and Abadie et al, 2002. The assumptions required by the former are too restrictive in our setting, and the latter methodology cannot be applied in our context because it requires that both the instrument and the endogenous regressors are binary variables.

the endogenous regressor is needed for the unambiguous definition of quantiles<sup>6</sup>, and guarantees the point identification of the quantiles of interest. When the continuity assumption fails, Chesher's approach can be extended (see Chesher, 2003, 2005) but does not generally lead to point identification of the function describing the impact without further assumptions. Importantly, the case of an endogenous binary regressor cannot be dealt within this set-up.

The estimation of the exogenous impact functions and inference in the parametric case are discussed by Ma and Koenker, 2006. They assume that the conditional quantile functions are known up to a finite number of parameters and add some technical regularity conditions. In their framework, the conditional quantile functions need not be linear in the parameters and the asymptotic theory is developed for nonlinear quantile regression estimation.

Arias, Hallock and Sosa-Escudero, 2001, use data on US twins to address the issue of the endogeneity of education in quantile wage regressions. They propose a two stages estimator and find that returns to education increase with the quantiles of the conditional distribution of earnings. They interpret this as evidence that ability and education are complements. Their methodology, however, has been recently questioned by Ma and Koenker, 2006, who use Montecarlo simulations to show that the two-stage quantile regression which replaces education with predicted education from the ordinary least square (mean) regression of schooling on the set of instruments performs rather badly in terms of simulated bias when compared with Chesher's WAD estimator and with the control variate approach suggested by Ma and Koenker and implemented in this paper.

#### 3. The Empirical Model

Following Card, 2001 and Ashenfelter and Rouse, 1998, assume that individuals - or their parents - choose years of schooling to maximize

$$U(W_i, S_i) = I n(W_i) - f(S_i)$$
<sup>[1]</sup>

<sup>&</sup>lt;sup>6</sup>In Chesher, 2002, there is no requirement on the scale of the regressors and of the instruments but a completeness condition has to be met.

where W = g(S) is (net) earnings, S is years of schooling and the index *i* is for the individual. At the optimum, individuals select S so as to equate the marginal costs to the (expected) marginal benefits of schooling. Let marginal costs MC be increasing in schooling and decreasing in cognitive ability

$$MC(S_i) = r_i + \theta S_i - \kappa A_i$$
<sup>[2]</sup>

where r captures both the access to funds and the tastes for schooling, and assume the following Mincerian earnings function

$$I n(W_i) = \alpha + \beta S_i + S_i (\lambda A_i + \phi U_i) + A_i + U_i + \gamma_W X_i$$
[3]

where  $A \sim G_A(0, \sigma_A^2)$  is individual cognitive ability, which we take to be known to individuals at the time of their choice, X is a vector of controls and  $U \sim G_U(0, \sigma_U^2)$  is an idiosyncratic error orthogonal to ability. In the language of Chesher, 2003, the latent random variable U is fortune in the labour market, but other interpretations are possible, as discussed in the introduction, and include a zero mean demand shock which affects the relative productivity of jobs and skills (Gosling, Machin and Meghir, 2000, Machin and Van Reenen, 1998).

The specification in Eq. [3] implies that schooling influences both the location and the scale of the earnings distribution. Ability affects earnings both directly and via its interaction with schooling, as in Ashenfelter and Rouse, 1998, who distinguish between the absolute and the comparative advantage of higher talent<sup>7</sup>. Education and ability are complements in the production of human capital when  $\lambda > 0$ , and substitutes when  $\lambda < 0$ . Another feature of the selected specification is the possibility that shocks to the composition of labour demand - or random luck U - have different effects on earnings depending upon the level of accumulated schooling. When  $\varphi > 0$  these shocks are skill –

<sup>&</sup>lt;sup>7</sup>Heterogeneous returns associated to ability are a common assumption in this literature at least since Griliches, 1977.

biased (Katz and Murphy, 1992). With expected marginal benefits of schooling given by

$$\mathsf{MB}(\mathsf{S}_i) = \beta + \lambda \mathsf{A}_i \tag{4}$$

optimal schooling  $S^*$  is equal to

$$\mathsf{S}_{\mathsf{i}}^* = \frac{\beta - \mathsf{r}_i}{\theta} + \frac{\lambda + \kappa}{\theta} \mathsf{A}_i$$
<sup>[5]</sup>

Notice that the relevant variation in ability for the schooling decision involves the slope of the log earnings function, not its intercept<sup>8</sup>. In the private optimum, schooling increases with individual ability A if  $\lambda + \kappa > 0$ . This is the case either if ability and schooling are complements or if the effect of ability on the marginal costs is large enough to offset the substitutability between A and S. Two economic models suggest a positive relationship between ability and schooling: the signalling model and a variant of the human capital model (see Blackburn and Neumark, 1993). In the former model, more able individuals have lower marginal costs of schooling and self-select into higher education. In the latter model, ability increases the marginal benefits of education and reduces the marginal costs of schooling.

We assume that  $1 + \phi S > 0$ , which guarantees that log wages are a monotonic function of the random effect U. A feature of the model is that schooling in [5] is correlated with ability A, which affects log earnings both directly and via its effects on education in [3], but not with the random shock U. Unless we can adequately control for ability, the standard orthogonality condition required for the consistency of ordinary least squares estimation of [3] fails. Consistent estimates can be obtained, however, if there exists a variable Z which is correlated with schooling but not with individual ability conditional on schooling (see Card, 2001, for an extensive discussion)<sup>9</sup>. As discussed in detail in the next section, Z in this

<sup>&</sup>lt;sup>8</sup>As discussed in Card, 1995, "..individuals with higher earnings opportunities at each level of education (i.e. with higher intercepts in their log earnings functions) may well invest less in schooling, since they have a higher opportunity cost of attending school..".

<sup>&</sup>lt;sup>9</sup>When returns to education are heterogeneous traditional IV estimates capture local average treatment effects. See Blundell et al, 2005, for a review of the econometric issues.

paper is the number of years of compulsory education YCOMP.

Omitting subscripts for simplicity, the earnings - cum - education model presented above cab be written in the format of an exactly identified triangular model, as in Chesher's approach

$$I n(W) = \alpha + \beta S + S(\lambda A + \phi U) + A + U + \gamma_W X$$
[6]

$$S = \gamma_S X + \pi Z + \xi A$$
<sup>[7]</sup>

where  $\xi = \frac{\lambda + \kappa}{\theta}$ . Define  $\tau_{A} = G_{A}(A_{\tau_{A}})$  and  $\tau_{U} = G_{U}(U_{\tau_{U}})$ , where  $A_{\tau_{A}}$  and  $U_{\tau_{U}}$  are the  $\tau$  - quantiles of the distributions of A and U, respectively. Furthermore define  $Q_{W}[\tau_{U} | S, X, Z]$  and  $Q_{S}[\tau_{A} | X, Z]$  as the conditional quantile functions corresponding of log wages and years of education. Ma and Koenker, 2007, show that recursive conditioning yields the following model

$$\mathsf{Q}_{\mathsf{W}}[\tau_{\mathsf{U}} \mid \mathsf{Q}_{\mathsf{S}}(\tau_{\mathsf{A}} \mid \mathsf{X}, \mathsf{Z}), \mathsf{X}, \mathsf{Z}] = \mathsf{Q}_{\mathsf{S}}[\tau_{\mathsf{A}} \mid \mathsf{X}, \mathsf{Z}]\pi(\tau_{\mathsf{A}}, \tau_{\mathsf{U}}) + \gamma_{\mathsf{W}}\mathsf{X} + \mathsf{G}_{\mathsf{A}}^{-1}(\tau_{\mathsf{A}}) + \mathsf{G}_{\mathsf{U}}^{-1}(\tau_{\mathsf{U}})$$
[8]

$$O_{S}[\tau_{A} | X, Z] = \gamma_{S} X + \pi Z + \xi G_{A}^{-1}(\tau_{A})$$
[9]

Given the restrictions imposed by equations [6] and [7], the key parameter of interest  $\pi(\tau_A, \tau_U)$  has the following structure

$$\pi(\tau_{\rm A},\tau_{\rm U}) = \beta + \lambda \mathsf{G}_{\rm A}^{-1}(\tau_{\rm A}) + \phi \mathsf{G}_{\rm U}^{-1}(\tau_{\rm U})$$
<sup>[10]</sup>

 $\pi(\tau_A, \tau_U)$  is a function describing the quantile treatment effect of education on earnings. It offers a panoramic view of the stochastic relationship between schooling S and log wages, and describes the effects of a perturbation in the distribution of schooling on the various quantiles of the distribution of earnings. Rather than exogenously altering the value of S, we alter its various quantiles  $Q_S$ , and study how the quantiles  $Q_W$  of the distribution of earnings are affected (Ma and Koenker, 2006).

If we set  $\tau_U$  so that U is fixed at its  $\tau_U$  quantile, changes of  $\tau_A$  in  $\pi(\tau_A, \tau_U)$  reflect how

the distribution of A affects the  $\tau_U$  quantile of the response ln(W). On the other hand, if we fix  $\tau_A$  and allow  $\tau_U$  to vary, we can shed light on how the  $\tau_A$  quantile of S affects the entire distribution of lnW (Ma and Koenker, 2006). By integrating  $\pi(\tau_A, \tau_U)$  with respect to  $\tau_A$ , we obtain the mean quantile treatment effects, which show how returns to education vary along the distribution of labour market luck for individuals of average ability. Further integration yields the mean treatment effect, which corresponds to the two stages least squares estimate. Conditioning on different values of  $\tau_U$  one can investigate whether returns to education increase as we move from lower to higher quantiles of the distribution of ability at different points of the (conditional) wage distribution. In our application, there is evidence of complementarity between education and ability.

Estimation and inference for  $\pi(\tau_A, \tau_U)$  are discussed estensively by Ma and Koenker (2006), who focus on two approaches, namely the "weighted average derivative" (WAD) approach and the "control variate" (CV) approach. They show that the latter approach yields a more efficient estimator when the model is correctly specified and both estimators have a superior performance compared to alternative parametric estimators (see Ma and Koenker, 2006 for details). In this paper, we implement the CV approach, which consists of the following three steps. First, we estimate the conditional quantile functions of schooling S and compute the control variate

$$A(\tau_A) = S - \hat{Q}_S(\tau_A \mid X, Z)$$
<sup>[11]</sup>

where S is observed schooling and  $\hat{Q}_{s}(\tau_{A} | X, Z)$  is the estimated conditional quantile. Second, we augment the conditional quantile functions of  $\ln W$  with the relevant control variate and its interaction with schooling. Finally, we simultaneously estimate the selected quantiles of  $\ln(W)$  conditional on  $Q_{s}[\tau_{A} | X, Z], X, Z]$  as in the hybrid model in equation [8] and obtain the variance - covariance matrix by bootstrapping.

#### 4. Our Empirical Strategy

We identify the causal effect of education on the distribution of wages by using the exogenous variation of schooling induced by compulsory school reforms implemented at different times and with different intensity in 12 European countries after the Second World War. The crucial difference between our study and previous literature using the same instruments (see for instance Oreopoulos, 2006, and the references therein) is that our analysis is not limited to the exploration of the conditional mean impact of schooling on wages. Rather, we permit heterogeneity in the impact of education at different points of the conditional distribution of earnings.

In our empirical application, identification relies on the following assumptions:

- individuals with higher ability stay in school longer (monotonicity with respect to A in Eq. [7]);
- individuals with a luckier draw from the distribution of wage offers have higher wages (monotonicity with respect to U in Eq.[6]);
- when the schooling decision is made, individuals can only form expectations about their future draw from the distribution of wage offers (triangular structure in the unobservables);
- years of compulsory schooling YCOMP have an exogenous impact on the distribution of years of education and/or the educational attainment of individuals: more education (the treatment) is assigned to individuals on the basis of their date of birth and the latter was not chosen by their parents on the basis of future education - related wage gains;
- when the timing of the implementation of school reforms varies across municipalities of the same country, it is uncorrelated with general education levels<sup>10</sup>;
- educational reforms do not affect log wages other than through the individual's education level, in other words they are excluded from the wage equation (triangular structure in the observables).

We use conditional quantile functions as in Eqs. [8] and [9] and allow the conditional quantiles of years of education to differ across countries up to a constant, holding the value

<sup>&</sup>lt;sup>10</sup> Black, Devereux and Salvanes, 2003, test this hypothesis in the case of Norway and find no systematic relationship.

of the other conditioning variables as fixed. We pool data from several countries to increase the number of points on the support of the instrumental variable YCOMP<sup>11</sup>. By pooling countries, we exploit the fact that the timing of compulsory school reforms varies across countries and by so doing we can distinguish school reform from cohort fixed effects.

We select for each country a school reform affecting compulsory education and define  $T \equiv (C - \overline{C}_k)$  as the distance between birth cohort C and the cohort  $\overline{C}_k$ , defined as the first cohort potentially affected by the change in mandatory school leaving age in country k. Since each selected reform occurs at a different point in time, our instrument varies both across countries and over cohorts. For each country, we construct a pre - treatment and post - treatment sample composed of the individuals born within the range defined by 7 years before and 7 years after the year of birth of cohort  $\overline{C}_k$ . The breadth of the window is designed to exclude the occurrence of other compulsory school reforms, which would blur the difference between pre- and post-treatment in our data. Our choice also trades off the increase in sample size with the need to reduce the risk that unaccounted confounders affect our results.

Borrowing from Angrist, Imbens and Rubin, 1996, the individuals with  $T \ge 0$  who have changed their educational attainment as a result of the reforms are defined as "compliers"<sup>12</sup>. Table 1 presents for each country in our sample the selected reform, the year of birth of the first cohort potentially affected by the reform, the change in the minimum school leaving age and in the years of compulsory education induced by the reform, and the expected change in school attainment, expressed in terms of the ISCED classification. Our information is drawn from Eurybase, the Eurydice database on education systems in Europe, from personal communications with national experts and from other country-specific sources. The description of each reform and the explanation of our choice of  $\overline{C}_k$  for each country are relegated to Section 2 of the Appendix.

The selected reforms increased the minimum school leaving age by one year in Austria, Germany, Ireland, Netherlands and Sweden; by two years in Denmark, France and Spain;

<sup>&</sup>lt;sup>11</sup>The support of this variable consists of 7 points.

<sup>&</sup>lt;sup>12</sup>Individuals whose nationality is unknown and/or who are not citizens of the country in which they live at time of the interview are excluded from the analysis. The relative share of compliers is affected by migration flows within Europe. If for instance a German citizen belongs to the first cohort potentially affected but migrated as an adult from Italy, where he received his education, we cannot expect his education to be affected by the change in German schooling laws.

by three years in Finland, Greece, Italy and by four years in Belgium<sup>13</sup>. In some of these countries, the timing of the introduction of the reform varied by region - this is the case of Germany, Finland and Sweden. Since we do not have access to data at the municipality level, in Finland and Sweden we define the year of the reform in each area as the year when the largest share of municipalities in the area experienced the change in the schooling legislation (see Table A.1for Finland in the Appendix).

In our sample, the modal compulsory number of years of education before the reforms is 8 years. The first cohorts potentially affected by the reforms were born between 1941 and 1969, with a relative concentration between the late 1940s and the late 1950s<sup>14</sup>. Furthermore, the most commonly expected change in qualifications is the attainment of ISCED level 3 (upper secondary education).

Tables A.2, A.3 and A.4 in the Appendix summarize the existing empirical evidence on the effects of some of these compulsory school reforms on individual education and earnings as well as the instrumental variable estimates of the average returns to schooling. While the increase in compulsory schooling induced by each reform varies across countries, ranging from 1 additional year of schooling to 3 or 4, the estimated impact on educational attainment (in terms of years of education) is close to 0.3 additional years of schooling, with little cross-country variation. Although the estimates of the effect of compulsory school reforms on educational attainment are broadly similar across European countries, this does not hold when one looks at the effects of longer schooling on wages: while in some countries the evidence suggests zero returns to compulsory schooling<sup>15</sup>, in some other countries returns to longer compulsory schooling are as high as 15%-20%. As discussed by Pischke et al, 2005, for the case of (West) Germany, the following factors may lead to finding no returns to compulsory schooling: (i) measurement errors; (ii) wage rigidity; (iii) the role of apprenticeship; (iv) the heterogeneity of returns, with individuals affected by compulsory schooling being the low-return group; (v) the type of skills learned in school around the time of school leaving age and the relevance of these skills for the labour market.

<sup>&</sup>lt;sup>13</sup>Notice that in Italy, Belgium, Finland, France and in the Netherlands, these reforms were accompanied by a change in school design, typically the postponement of tracking.

<sup>&</sup>lt;sup>14</sup> Close to 83 percent of the individuals belonging to the first cohorts affected by the reforms were born between 1947 and 1959.

<sup>&</sup>lt;sup>15</sup>This is the case for France (see Grenet, 2004) and Germany (see Pischke et al, 2005, and Fertig and Kluwe, 2005).

Another reason might be that returns to education depend on the qualification individuals achieve, regardless of whether the issued certification has legal value, or of the actual time spent in full-time education. As Grenet, 2004, suggests for France, the actual quantity of education attained is far less important than the qualifications held by individuals in determining these returns (p.30).

#### 5. The Data

We pool together data drawn from the 8<sup>th</sup> wave of the European Community Household Panel (ECHP) for the year 2001, the first wave of the Survey on Household Health, Ageing and Retirement in Europe, or SHARE, for the year 2004, and the waves 1993 to 2002 of the International Social Survey Program (ISSP). The countries included in our study are: Austria, Belgium, Denmark, Finland, France, Germany, Greece, Ireland, Italy, the Netherlands, Spain and Sweden. Tables B.1, B.2 and B.3 in the Appendix show for each country in the dataset the sample size for each survey and wave, and the relevant age range<sup>16</sup>.

Our dependent variable is the log of (gross) hourly earnings expressed at 2000 prices and purchasing power parity units. Table B.4 in the Appendix reports the country specific consumer price indices, exchange rates and PPP values. Additional information on earnings, hours worked and the proportion employed is also in the Appendix, see Tables B.5 to B.11. We measure educational attainment with years of education. Since in some countries and datasets the available information is on the highest attained qualification, we convert it into years of education by assuming that each individual requires the customary number of years to complete a degree<sup>17</sup>.

We assume that educational attainment does not change after age 25, and restrict our

<sup>&</sup>lt;sup>16</sup>The European Community Household Panel data used in this paper are from the December 2003 release (contract 14/99 with the Department of Economics, University of Padova). This paper uses data from SHARE 2004. The SHARE data collection has been primarily funded by the European Commission through the 5th framework programme (project QLK6-CT-2001-00360 in the thematic programme Quality of Life). Additional funding came from the US National Institute on Aging (U01 AG09740-13S2, P01 AG005842, P01 AG08291, P30 AG12815, Y1-AG-4553-01 and OGHA 04-064). Data collection in Austria (through the Belgian Science Policy Office) and Switzerland (through BBW/OFES/UFES) was nationally funded. The SHARE data set is introduced in B.Supan et al., 2005); methodological details are contained in B.Supan et al, 2005b.

<sup>&</sup>lt;sup>17</sup>See the Appendix for further details.

sample to include only individuals aged 26 to 65<sup>18</sup>. The final sample consists of 18328 individuals, and its distribution across the 12 countries is shown in the last column of Table 2, which includes also the sample mean by country of log real earnings, years of schooling, years of compulsory schooling, average age and percentage of males. Educational attainment is highest in Finland (15.15 years) and lowest in Spain (11.05 years). Average age is highest in Sweden (50.41) and lowest in Belgium (33.13), which reflects the different timing of the selected reforms.

Since we intend to identify from the data the causal relationship between the distribution of schooling and the distribution of earnings, we need to control as accurately as possible for additional factors affecting the dependent variable. Therefore, we include in the empirical specification both country and survey dummies, individual age and its square. We also control for country - specific macroeconomic effects by using the first lags of the unemployment rate, aggregate productivity, measured by real GDP per head, and the OECD index of the strictness of employment protection. Because of the existing evidence on gender differences in education and earnings, we estimate gender-specific years of education equations, and allowed a gender dummy to interact with a number of covariates in the earnings equation (as explained below).

Trend-like changes in log wages relative to the time of the reform are controlled with a second order polynomial in K=T+7 – where T is the distance between each cohort and the first cohort potentially affected by the reform - and its interactions with country dummies<sup>19</sup>. Empirical research has shown that individual earnings are significantly affected by the conditions prevailing in the labour market at the time of first labour market entry (see for instance Baker, Gibbs and Holmstrom, 1994). To capture these effects, we match to each individual the country and gender specific labour participation rate at the age of estimated labour market entry<sup>20</sup>. The underlying idea is that entry wages are likely to be higher when

<sup>&</sup>lt;sup>18</sup>We also exclude individuals with more than 30 years of schooling. We repeated the analysis by considering only individuals who were aged at least 28 at the time of the interview. Results are robust and are not reported for brevity. We prefer not to exclude individuals aged between 26 and 28 since this procedure would lead to drop from the analysis individuals potentially affected by the reforms in some countries, for instance in Spain and Finland.

<sup>&</sup>lt;sup>19</sup>The relatively low order of the polynomial follows the suggestions by Lee and Card, 2008. Compared to higher order polynomials, the second order specification is the most parsimonious and provides adequate fit of the data.

 $<sup>^{20}</sup>$ We estimate entry to occur after the completion of schooling. We use a three-years moving average of the

the labour market is tight and labour participation rates are high.

Changes in educational attainment after a compulsory school reform could be due to the reform itself or to confounding factors, which may alter the incentives to invest in education at the time of the reform but independently of it. To illustrate, take a reform that increases the minimum school leaving age from 14 to 15 in a certain year. If individuals at age 14 - or their parents - find it more attractive to invest in education because of a reduction in the opportunity costs generated by a contemporaneous increase in the unemployment rate, they might invest more independently of the reform. Similarly, the actual implementation of school leaving laws may vary across countries and over time with changes in economic conditions. Implementation is known to be more difficult in poorer countries, and, ceteris paribus, in households with a higher number of children. To control for these confounders, we construct two variables, the unemployment rate and the real GDP per head, and match these variables to individuals around the age when the school reform is supposed to have taken place. For instance, assume that the critical age is 14 for Austrian citizens born in 1957. For these individuals the relevant values of the three variables described above are those corresponding to 1971.

#### 6. The Findings

We start the presentation of our results with the relationship between education and the conditional distribution of wages when education is treated as exogenous. The estimates are shown in Table 3 separately for males and females, and show that the returns to one year of education increase as we move from the lowest to the highest quantile of the distribution. Moreover, returns are consistently higher for females than for males, in line with previous results in the literature – see Harmon, Walker and Westergaard Nielsen, 2001. Using the 90-10 log wage differential as a measure of conditional wage inequality, we find that this is increased by 2.0% for one additional year of education for males, by 2.4% for females.

Education, however, cannot be treated as exogenous in the presence of unobserved ability. Table 4 presents the results of the first stage regression of years of education against

macro variables to smooth out measurement errors in the date of labour market entry.

all the exogenous controls plus the instrument YCOMP. In most cases the latter variable always attracts a statistically significant and positive coefficient, with the exception of the top decile of the distribution of ability for males. Using the Stock and Staiger rule of thumb, which suggests that when there is a single endogenous variable the selected instrument is weak if the F – test for its inclusion in the auxiliary regression is lower than 10, we find evidence of weakness for the two top deciles in the case of males and for the top decile in the case of females. Furthermore, the size of the effect of compulsory school leaving laws on attained education is much larger for the lowest quantile of the distribution of ability.

These findings suggest that compulsory schooling laws are particularly effective at the lower tail of the distribution of ability: for individuals located below the 10th quantile, a one year increase in compulsory education increases actual attainment by 40 percent of a year in the case of females and by 30 percent of a year in the case of males, which compares to 10 percent of a year for individuals with median ability. These results are in line with expectations, which suggest that the bulk of compliers should be among the less able (and wealthy). While for males the impact of school reforms on educational attainment tends to die out at the upper decile of the conditional distribution of ability, for females the effect remains statistically significant above median ability. One explanation is that better educated individuals, especially females, react to the increase in the minimum school leaving age by upgrading their own education, in an effort to maintain at least in part their relative advantage over the less educated.

We apply the control variate approach due to Ma and Koenker, 2006, and estimate the full matrix of quantile treatment effects, which describe the impact of education on earnings for different quantiles of the distribution of ability and labour market fortune. We focus for brevity on the following quantiles: 0.10, 0.30, 0.50, 0.70 and 0.90 and proceed as follows: first, we run separately for males and females quantile regressions of [9] and compute the control variate for each selected quantile of the conditional distribution of ability. Second, we augment [8] with each estimated control variate and estimate separate quantile regressions by pooling all data and allowing for the full set of interaction of the explanatory variables with the gender dummy. For each regression, we test whether blocs of interactions with gender are statistically significant, and simplify the empirical specification by dropping those blocs which fail to pass the test. Last, we estimate simultaneous quantile

regressions for each quantile of the distribution of ability, using the parsimonius specification, and obtain the estimate of the variance – covariance matrix by bootstrapping<sup>21</sup>. In all cases, we never reject the hypothesis that the control variate and its interaction with schooling are significantly different from zero, and that education is endogenous in the wage regressions.

The estimated percentage increase in log earnings associated to one additional year of education and its standard error are reported separately for males and females in Table 5 for the selected quantiles of the distribution of ability A ( $\tau_A$ ) and labour market fortune U ( $\tau_U$ ). The final row in each section of the table is obtained by integrating the quantile treatment effects with respect to ability, which produces the mean quantile treatment effect. If we compare this effect to the one estimated by treating education as exogenous, we notice that the former is higher than the latter, and that the gap is larger for the lowest decile of the distribution of the random term U.

To interpret our findings, select for instance the bottom decile of the distribution of ability,  $\tau_A = 0.10$ . We find that the estimated returns to education for males are equal to 7.48 percent for the individuals at the bottom decile of the distribution of earnings, to 5.55 percent for the individuals at the median decile and to 5.98 percent for individuals at the top decile. The corresponding returns for females are higher but follow a similar pattern. As illustrated by Figures 1 and 2, the profile of estimated returns is broadly declining, albeit not always monotonically, as we move from the lowest to the highest decile of the distribution of labour market fortune, and independently of the selected decile of the distribution of ability.

Table 6 presents tests for equal coefficients for each selected decile of the distribution of ability. Differences in these parameters can be interpreted as measures of education-induced conditional inequality: one should remember, however, that coefficients on

<sup>&</sup>lt;sup>21</sup> To facilitate the convergence of the estimates in the bootstrapping exercise in the presence of a large number of regressors, we operate as follows: define the relevant regression as  $Q(y) = \Gamma X + \Omega Y$ , where X and Y are suitable vectors of variables. We first estimate each quantile regression separately and compute the index  $\overline{\Gamma}X$ . Next, we redefine the dependent variable as  $Q(y) - \overline{\Gamma}X$  and estimate quantile regressions for different quantiles of the distribution of wages simultaneously using STATA built in functions. In practice, the vector X includes two survey dummies, the polynomial in the second order in K and its inteactions with country dummies.

conditioning variables also vary across deciles, thus conditional inequality will also reflect these differences. With this caveat in mind, we can say that education-related conditional inequality declines with additional education, and significantly so – in a statistical sense – with the exclusion of the index involving the highest decile. When we average returns across the distribution of abilities, we find that the reduction in conditional inequality induced by an additional year of education is close to 1.5 percentage points. With constant marginal returns, this corresponds to about 5 percent for a three – years degree.

In summary, our estimates point to a relationship of substitutability between schooling S and the random effect U: individuals who are less fortunate in the labour market can partially compensate for poor luck with higher returns from investment in education. Therefore, when poor luck depends on circumstances beyond individual control, education policies targeted at the less fortunate which encourage additional schooling contribute to reducing conditional inequality.

Next consider again Table 5 but select a decile in the distribution of the random error U. Independently of gender and of the selected decile, there is evidence that the returns to education fall as we move from the bottom to the top decile of the distribution of ability. Figures 3 and 4 provide a vivid illustration of this pattern in the results. This qualitative finding remains even if we disregard the two top deciles, because of the weakness of our instrument, especially in the case of males. To illustrate, returns to education fall from 7.88 to 5.88 percent when we select the median decile in the distribution of error U, focus on females and on the 10<sup>th</sup> and 70<sup>th</sup> decile of the distribution of ability.

We interpret this as evidence that ability and education are substitutes in the production of human capital. When we increase education by an additional year, individuals located in the upper part of the distribution of ability gain less compared to individuals with less than median ability. Since ability and parental background are closely intertwined - see Cuhna et al, 2005 - our results point to the fact that those better endowed have less to gain from additional education. The size of the gap is again not small, and is equal at most to slightly more than 3 percentage points.

Since earnings in our model depend both on ability A and on the random error U, an alternative reading of Table 5 is along the main diagonal: as we move from the upper left to the bottom right corner, we consider individuals who are increasingly endowed in both

ability and labour market fortune. Independently of gender, our estimates suggest that the better endowed have lower returns to investment in education.

Overall, our findings confirm the results by Ashenfelter and Rouse, 1998, based on a sample of American twins. Using a different methodology – quantile regressions – and a sample of European individuals, we find – as they do - that better endowed individuals have lower returns to education. The fact that better endowed individuals typically have higher attained education but lower marginal benefits of schooling point either to lower costs of funds or to a negative relationship between the marginal costs of schooling and individual ability, as in equation [2].

Table 5 can also be used to calculate the corresponding structural parameters  $\beta$ ,  $\phi$  and  $\lambda$ , according to equation [10]. All we have to do is to compute the empirical c.d.f.'s of the error terms, A and U, and invert them to obtain estimates of  $G_A^{-1}(\tau_A)$  and  $G_U^{-1}(\tau_U)$ . Then from the estimated values of  $\pi(\tau_A, \tau_U)$  one can estimate  $\beta$ ,  $\phi$  and  $\lambda$  by GLS. GLS in this context is an optimal minimum distance estimator, but requires estimating the variance-covariance matrix of the whole  $\pi(\tau_A, \tau_U)$  matrix, something that is beyond the scope of this paper. Card and Krueger, 1992, show that almost identical estimates of both coefficients and standard errors obtain if a weighted least squares estimator (WLS) is used instead of GLS (that is, if observations are weighed inversely to the estimated variance of the dependent variable, and covariance terms are ignored).

We find that the WLS estimated value of  $\beta$  is 0.051 for males and 0.070 for females see Table 7. Furthermore,  $\lambda$  is negative, statistically significant and in the range -0.0021 / -0.0025, depending on gender. Finally, estimated  $\varphi$  is also negative, and in the range -0.0089 / -0.0119<sup>22</sup>. These estimates confirm that both unobserved ability A and labour market luck U are substitutes to education in the production of human capital and earnings.

The policy implications of our results are important. First, suppose that earnings and productivity are closely related, a plausible assumption. Then education policies aimed at raising the educational attainment of the less fortunate and talented are grounded on equity considerations if less lucky individuals are so because of circumstances outside their own control. Equity, however, trades off with efficiency if the marginal returns to schooling are

<sup>&</sup>lt;sup>22</sup> These results do not significantly change when we omit the two top quantiles of the distribution of ability – see columns (2) and (4) in Table 7. The estimates of  $\varphi$  suggest that the constraint 1+  $\varphi$ S is always verified.

higher for the more talented, because this group has lower marginal costs of education. Our findings that labour market returns to schooling are highest among the less fortunate and talented has the important implication that equity and efficiency need not trade-off, because these higher returns can more than compensate the higher schooling costs. Second, the uncovered substitutability between education and ability is relevant for the design of an optimal education policy. Such policy should be elitist and favour brighter children only if education and talent are complements in the production of human capital. When this is not the case, as our results suggest, redistributive education that targets the less able and fortunate can pay off both on equity and efficiency grounds.

To illustrate, consider a lump sum fixed subsidy paid to the less privileged that improves access to schooling, funded by a lump sum education tax on the more privileged. Since ability and parental background are strongly correlated, such policy increases equality of opportunity by raising the educational attainment of the less fortunate and reducing that of the more fortunate. If ability and schooling are substitutes, this policy is efficient, because the gain in marginal benefits for the less talented more than compensates the loss for the more talented.

Last but not least, individual ability is strongly affected both by genetic and by environmental factors (see Cunha and Heckman, 2007). If education and ability are substitutes in the production of human capital and earnings, then additional investment in the former can contribute to reducing the differences produced by the latter.

#### Robustness

Since log hourly wages are only available for employees, our sample is the result of a selection process involving the decision to participate to the labour market and having an employee job. Unless we take this selection process explicitly into account, the error term in equation [6] is unlikely to have zero mean. More important for our purposes is the concern that selection into employment may be affected by the number of years of compulsory education. If this was the case, the validity of our instrument would fail to hold.

We investigate this by defining B as a dummy variable equal to 1 if log earnings are observed and to 0 otherwise. Failure to observe wages could be due to the participation

decision, to the choice between employment, unemployment and self-employment or to the presence of missing wage data. We estimate a probit model for variable B using all the controls described above plus the predicted years of schooling from the first stage regression of years of schooling on compulsory years of education. If the latter affected the selection process, we would expect that predicted years attract a statistically significant coefficient. It turns out that the estimated coefficient is equal to 0.023 for males, with a bootstrapped standard error equal to 0.076 (p-value: 0.300), and to 0.088 for females (standard error: 0.055 and p-value: 0.107). Therefore, there is no evidence in our data supporting the view that the years of compulsory education are significantly associated to the endogenous selection of workers into paid employment, and we conclude that our instrument is not invalidated by failure to explicitly consider such selection.

#### Conclusions

In this paper, we have treated the countries of Europe as regions of a single country, and country specific compulsory school reforms as episodes of a broad European reform, which has taken place in each region at a different point in time. By so doing, we have been able to generate the country and time variation that was absent in previous European research (see Harmon and Walker 1995). Confirming that there are no free lunches, we have made this progress at the price of pooling together data from three fairly heterogeneous surveys, which is likely to lower the precision of our estimates.

There are three main results: first, we find evidence that additional schooling reduces conditional wage dispersion. Second, compulsory school reforms affect mainly the individuals at the lower end of the distribution of educational attainment: for these individuals one additional year of compulsory education is estimated to translate into 30 to 40 percent of a year of additional education or higher. This percentage falls to 10 or below for the rest of the population. Third, there is evidence that education and ability are substitutes in the generation of earnings.

We believe that the policy implications of our results are rather clear. First, compulsory schooling law reforms have had a rather pervasive effect, highest among the less talented but present also – albeit to a substantially lower extent - among the better endowed,

especially females. Second, the pursuit of an elitist education policy because of efficiency considerations is questionable, because the basic assumption of complementarity between education and ability fails to hold. Third, education policies which focus on the equality of opportunity for the less fortunate and less talented can be justified not only on equity but also on efficiency grounds if the additional benefits – which we have shown to be higher than for the more talented – are sufficiently large to offset the additional costs.

#### Appendix

#### A. The Educational Reforms used in this Study

In this section we provide a brief description of the educational reforms considered in the study. Furthermore, we motivate the choice of the first cohort potentially affected. We devote a paragraph to each country considered. Further details on country specific educational systems and reforms are in Fort, 2006.

#### Austria

The 1962 School Amendement Act increased compulsory schooling by one year, from 8 to 9 years. Pupils who were 14 years old or younger at the time the reform was introduced were compelled to attend an additional year of schooling. This suggests that the individuals potentially affected by the reform are those born in 1948 and afterwards. However, individuals born in 1947 who might have already left school when the reform was introduced were required to go back to school and complete the additional year. Therefore, we select the cohort born in 1947 as the first cohort potentially affected by the reform.

#### Belgium

In 1983 (Law of 28 June 1983), the length of compulsory schooling was increased to 18 years (from 8 to 12 compulsory years of education), which could be completed with part time scholling during the final three years. Student potentially affected by the reform were those aged 14 or younger in 1983, i.e. those born after 1969.

#### Denmark

Two major reforms of compulsory schooling took place in Denmark in 1958 and 1971. In 1958 compulsory schooling years were increased by 3 years (from 4 to 7) and in 1971 they were further increased by additional two years (from 7 to 9). Pupils who were 14 years (or younger) in 1971 were potentially affected by the 1971 reform. We only consider this reform in this study.

#### Finland

The relevant reform considered in this study took place during the 1970s. The reform introduced a new curriculum and changed the structure of the educational system, increasing compulsory education from 6 to 9 years. The reform was adopted gradually by Finnish municipalities. Since we do not have access to data at the municipality level, we define the year of the reform in each area as the year when the largest share of municipalities in the area experienced the change in the schooling legislation, as reported in Table finlandeyfeatures below. Following Pekkarinen, 2005, we consider the cohort aged 11 when the reform was implemented as the first cohort potentially affected.

#### France

During the XX century, compulsory schooling age in France was extended twice: from 13 to 14 in 1936 and from 14 to 16 in 1959 (Bethoin Reform). The 1936 reform affected mainly pupils born after 1923, whereas the 1959 reform - which was implemented from 1967 after a long transition period - affected individuals who were born from 1953 onwards (see Grenet, 2004).

#### Germany

The peculiar political situation of the country produced the separate evolution of two distinct education systems between 1949 and 1990. We refer to reader to Pischke et al, 2005, and Pischke, 2003, for a description of the compulsory school reforms and for the selection of the first cohort potentially affected in each German Lander.

#### Greece

In 1975 the Greek Parliament increased compulsory education by three years (from 6 to 9). Individuals potentially affected by this change are those who were 12 in 1975. In particular, those born 1963 and later were compelled to attend 3 additional years of schooling, whereas those born in 1962 were not.

#### Ireland

Compulsory schooling was modified in 1972, when the school leaving age was raised to the age 15. A further raise in compulsory schooling age (to 16 years) announced in 1998, came into effect when the Education (Welfare) Act (2000) became law. Individuals potentially affected by the 1972 reform are the individuals who were 14 in 1972. These individuals where compelled to attend an additional year of schooling, whereas individuals who were 14 in 1971 were not. Therefore we choose 1958 as the first cohort potentially affected.

#### Italy

Junior high school became effectively compulsory in Italy only since 1963. Compliance with the 1963 reform was not instantaneous: only in 1976 the proportion of children attending junior high school approached 100%. According to Brandolini and Cipollone, 2002, the individuals potentially affected by the reform were those born after 1949.

#### The Netherlands

From 1975 onwards, all three-year educational programs in the Netherlands were extended to four years and compulsory schooling leaving age was increased by one year, from 15 to 16. Oosterbeek et al, 2004, highlight that the implementation of the extension of lower vocational education started in 1973. Since August 1 1973 all lower vocational programs had a lenght of four year. Students who were behind the second year could still graduate in a three-year course. Students who started a three-year course of lower vocational education on August 1 1971 could still graduate in 1974. All the following cohorts had to take a four-year course. According to Oosterbeek et al, 2004 ``students born on or after August 1, 1959 had to follow ten years of full time education. This is also the first cohort of students to encounter a complete four-year lower vocational educational education regime".

#### Spain

The compulsory school reform considered in this study was carried out under the 1970 General Act on Education and Financing of Educational Reform (LGE), and increased compulsory years of education from 6 to 8. Individuals potentially affected by the reform were those born in 1957 and after (see Pons and Gonzalo, 2003, p.753 and Table A1 p.767).

#### Sweden

According to Meghir and Palme, 2005, compulsory school reform in Sweden was gradually implemented between 1949 and 1962. The take-up of the experiment varied over the period 1949-1962 across municipalities with the largest number of municipalities involved in the years 1961/1962 (39.4%; 18,665 classes; 436,595 students). It was fully implemented only in 1962. Unfortunately, we do not have access to data at the municipality level but only at the county level. For the purposes of this paper, and based on personal communication with Marten Palme, we considered as potentially affected by the reform all the individuals born after 1950.

#### **B.** The Construction of the Key Variables

Is this section we provide further detail on the costruction of the key variables used in the empirical analysis.

#### Education

The ISSP survey generally includes information on the highest qualification attained at the time of the interview and on the number of years spent at school. When the latter variable is missing, years of education are computed using the information on the highest qualification attained and/or the age at which the individual finished his/her studies.

The SHARE survey collects information on the highest attained qualification and generates a variable corresponding to years of education (see the survey documentation at http://www.share-project.org/ for further details). Finally, the ECHP survey collects information on both years of education, age at which the individual finished his/her studies and the highest attained qualification. We use the variable pt024 in this dataset (see the user manual for more details).

#### Wages

The heterogeneity across surveys increases as one examines earnings, labour force status, and hours worked. Since earnings in the ISSP surveys are recorded on a categorical scale, with the number of categories varying across countries and surveys, we use mid-points of each category. The ECHP data include information both on net and on gross monthly earnings, and we use the latter. Finally, the SHARE survey collects information only on gross yearly earnings<sup>23</sup>. We transform the available information on earnings from the three surveys on a monthly basis, using 2000 consumer prices and PPP units.

Depending on the survey, data on working hours are: total hours worked per week in the main job (SHARE); total hours - including paid overtime - worked per week in the main job or business (ECHP); total number of hours worked per week in the main job (ISSP). In the 1997 ISSP survey hours are reported on a categorical scale, and we take mid-points. Finally, information on employment status is: self reported current employment status (SHARE); self-reported current employment status (ECHP); self -reported economic position (ISSP). Tables desstatalesage1-desstatemalesage3 present detailed information on the earnings, hours and proportion employed by country, year and survey used.

#### **Additional variables**

The aggregate variables used in the estimates are:

- labor force, population and unemployment: ILO Labor Force Statistics, www.laborsta.org.
- completed fertility rate: Eurostat online statistics
- gdp per head in 1990 international dollars: Maddison, 2007.
- We use linear interpolation to replace the few missing values in each of these variables.

 $<sup>^{23}</sup>$ An algorithm to impute missing values has been implemented; for details, see the documentation on the survey (see SHARE, 2005b).

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# Tables and Figures

Table 1.	Selected	compulsorv	school	reforms.	bv	country

		First	-		Expected	Age at
		cohort	Change	Change	change	school
Country	Reform	poten-	in min.	in years	in quali-	entry at
·		tially	sch. leav.	of comp.	fication	the time of
		affected	age	school.	(ISCED)	the reform
Austria	1962	1947	$14 \rightarrow 15$	$8 \rightarrow 9$	↑ ISCED 2	6
Belgium	1983	1969	$14 \rightarrow 18$	$8 \rightarrow 12$	$\uparrow$ ISCED 3	6
Denmark	1971	1957	$14 \rightarrow 16$	$7 \rightarrow 9$	$\uparrow$ ISCED 3	7
Finland (Uusima)	1977	$1966^{+}$	$13 \rightarrow 16$	$6 \rightarrow 9$	$\uparrow$ ISCED 3	7
Finland (Etela-Suomi)	1976	$1965^{+}$	$13 \rightarrow 16$	$6 \rightarrow 9$	$\uparrow$ ISCED 3	7
Finland (Ita-Suomi)	1974	$1963^{+}$	$13 \rightarrow 16$	$6 \rightarrow 9$	$\uparrow$ ISCED 3	7
Finland (Vali-Suomi)	1973	$1962^{+}$	$13 \rightarrow 16$	$6 \rightarrow 9$	$\uparrow$ ISCED 3	7
Finland (Pohjois-Suomi)	1972	$1961^{+}$	$13 \rightarrow 16$	$6 \rightarrow 9$	$\uparrow$ ISCED 3	7
France	$1959^{++}$	1953	$14 \rightarrow 16$	$8 \rightarrow 10$	$\uparrow$ ISCED 3	6
Germany (Schleswig-Holstein)	1956	1941	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Germany (Hamburg)	1949	1934	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Germany (Niedersachsen)	1962	1947	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Germany (Bremen)	1958	1943	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Germany (Nordrhein-Westphalen)	1967	1953	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Germany (Hessen)	1967	1953	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Germany (Rheinland-Pfalz)	1967	1953	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Germany (Baden-Würtemberg)	1967	1953	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Germany (Bayern)	1969	1955	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Germany (Saarland)	1964	1949	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Greece	1975	1963	$12 \rightarrow 15$	$6 \rightarrow 9$	$\uparrow$ ISCED 2	6
Ireland	1972	1958	$14 \rightarrow 15$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6
Italy	1963	1949	$11 \rightarrow 14$	$5 \rightarrow 8$	$\uparrow$ ISCED 2	6
Netherlands	$1975^{\bullet}$	$1959^{\bullet}$	$15 \rightarrow 16$	$9 \rightarrow 10$	$\uparrow$ ISCED 2	6
Spain	1970	$1957^{*}$	$12 \rightarrow 14$	$6 \rightarrow 8$	$\uparrow$ ISCED 2	6
Sweden	1962	1950**	$14/15 \rightarrow 15/16$	$8 \rightarrow 9$	$\uparrow$ ISCED 3	6/7

\* Pons and Gonzalo, 2002, p.753 and Table A1 p.767. + Pekkarinnen, 2005, p.5. ++ Reform

implemented in 1967, see Grenet, 2004.  $\ ^{\bullet}$  Reform implemented in 1973 (Oosterbeek et al.,

2004). \*\* Personal communication with Martin Palme.

Table 2.	Means o	f the l	key <sup>.</sup>	variables.	Sample	size:	18328
----------	---------	---------	------------------	------------	--------	-------	-------

	$\log W$	S	YCOMP	Age	%Males
Austria	2.220	12.181	8.767	50.900	0.492
Belgium	2.470	14.887	9.782	33.125	0.465
Denmark	2.798	13.667	8.030	44.186	0.477
Finland	2.366	15.153	7.511	37.151	0.496
France	2.399	13.410	9.017	47.074	0.525
Germany	2.439	12.127	8.620	45.649	0.590
Greece	2.005	12.929	7.509	38.270	0.562
Ireland	2.265	12.356	8.534	39.331	0.574
Italy	2.367	12.556	7.097	49.066	0.590
Netherlands	2.574	14.166	9.445	37.702	0.592
Spain	2.116	11.049	7.099	43.136	0.626
Sweden	2.328	12.197	8.465	50.410	0.480

Table 3. Quantile effects when education is treated as exogenous. Sample size: 18328. By gender (9936 males and 8392 females)

Coeff. (s.e.)	$\tau = 0.10$	$\tau = 0.30$	$\tau = 0.50$	$\tau = 0.70$	$\tau = 0.90$
Males	.019***(.002)	.026***(.001)	.033***(.001)	.035***(.001)	.039***(.002)
Females	$.027^{***}(.003)$	$.037^{***}(.001)$	$.043^{***}(.001)$	$.050^{***}(.001)$	$.051^{***}(.002)$

Note: each regression includes a constant, country dummies, K,  $K^2$  and their interactions with country dummies, survey dummies, age, age squared, a gender dummy, the lagged country specific unemployment rate and GDP per capita, the country and gender specific labour force participation rate at the estimated time of labour market entry, the country specific GDP per head and unemployment rate at the age affected by the country specific reform. Details on these coefficients are available from the authors upon request. Three stars, two stars and one star for statistically significant coefficients at the 1%, 5%, and 10% confidence level.

Table 4. First stage effect of YCOMP on S. Sample size: 18328.

Males	$\tau_A = 0.10$	$\tau_A = 0.30$	$\tau_A = 0.50$	$\tau_A = 0.70$	$\tau_A = 0.90$
Coeff. (s.e.)	$.354^{***}(.007)$	$.056^{***}(.012)$	$.120^{***}(.006)$	.078** (.035)	.026 (.071)
F-test (p-val.)	2146.6 (.000)	19.1 (.000)	307.6 (.000)	4.86 (.027)	0.13(.714)
Females	$\tau_{A} = 0.10$	$\pi = -0.20$	- 050	- 0.70	- 0.00
remares	$T_A = 0.10$	$\tau_A = 0.30$	$\tau_A = 0.50$	$\tau_A = 0.70$	$\tau_A = 0.90$
Coeff. (s.e.)	$\frac{7_A - 0.10}{.416^{***}(.016)}$	$7_A = 0.30$ .284***(.020)	$7_A = 0.50$ .072***(.007)	$7_A = 0.70$ .219***(.029)	$\frac{\tau_A = 0.90}{.135^{***}(.065)}$

Note: see Table 3.  $\tau_A$  denotes the quantile of the distribution of ability.

Table 5. Heterogeneous returns to schooling. Quantile treatment effects.Sample size: 18328.

Males	$\tau_{U} = 0.10$	$\tau_U = 0.30$	$\tau_U = 0.50$	$\tau_U = 0.70$	$\tau_U = 0.90$
$\tau_A = 0.10$	$.0748^{***}_{.004}$	$.0583^{***}_{.004}$	$.0555^{***}_{.003}$	$.0550^{***}_{.004}$	$.0598^{***}$
$\tau_A = 0.30$	$.0625^{***}$	$.0476^{***}$	$.0462^{***}$	$.0420^{***}$	$.0503^{***}$
$\tau_A=0.50$	$.0665^{***}$	$.0492^{***}$	$.0478^{***}$	$.0432^{***}$	$.0469^{***}$
$\tau_A=0.70$	$.0486^{***}$	$.0396^{***}$	.0448***	.0411***	.0471***
$\tau_A=0.90$	.0468***	$.0329^{***}$	$.0384^{***}$	$.0332^{***}$	.0452***
Mean effect	.0598	0.0456	.0465	.0429	.0499

Females	$\tau_U = 0.10$	$\tau_U = 0.30$	$\tau_U = 0.50$	$\tau_U = 0.70$	$\tau_U = 0.90$
$\tau_A = 0.10$	.0952***	.0780***	.0788***	$.0820^{***}_{.005}$	$.0759^{***}$
$\tau_A=0.30$	$.0838^{***}$	$.0701^{***}$	$.0713^{***}$	$.0730^{***}$	.0702***
$\tau_A = 0.50$	.0847***	.0679***	.0690***	.0707***	.0646***
$\tau_A = 0.70$	.0689***	$.0573^{***}$	$.0588^{***}$	$.0615^{***}$	$.0612^{***}$
$\tau_A=0.90$	$.0631^{***}$	.0502***	$.0527^{***}_{.003}$	$.05555^{***}$	$.0567^{***}_{.006}$
Mean effect	.0792	.0645	.0655	.0655	.0674

Note: see Table 3.  $\tau_U$  denotes the quantile of the distribution of labour market fortune and  $\tau_U$  denotes the quantile of the distribution of ability. Bootstrapped (100 replications) standard errors in small characters. <sup>+</sup>Mean effect: average (over  $\tau_A$ ) quantile treatment effect.

Males	$\delta_{30-10}$	$\delta_{50-10}$	$\delta_{70-10}$	$\delta_{90-10}$
$\tau_A = 0.10$	-0.0165	-0.0193	$-0.0198^{**}$	-0.0150
$\tau_A = 0.30$	$-0.0149^{**}$	$-0.0163^{**}$	$-0.0206^{**}$	-0.0122
$\tau_A = 0.50$	$-0.0172^{**}$	$-0.0187^{***}$	$-0.0233^{***}$	$-0.0196^{**}$
$\tau_A = 0.70$	$-0.0090^{*}$	-0.0038	-0.0075	-0.0015
$\tau_A = 0.90$	$-0.0139^{**}$	-0.0084	$-0.0136^{*}$	-0.0016
Mean effect	-0.0143	-0.0133	-0.0169	-0.0099
Females	$\delta_{30-10}$	$\delta_{50-10}$	$\delta_{70-10}$	$\delta_{90-10}$
$\hline \hline T_A = 0.10$	$\delta_{30-10}$ -0.0172***	$\frac{\delta_{50-10}}{-0.0164^{***}}$	$\delta_{70-10}$ -0.0132*	$\delta_{90-10}$ -0.0193**
$\tau_A = 0.10$	$-0.0172^{***}$	$-0.0164^{***}$	$-0.0132^{*}$	$-0.0193^{**}$
$\tau_A = 0.10$ $\tau_A = 0.30$	$\begin{array}{c} -0.0172^{***} \\ -0.0137^{**} \end{array}$	$-0.0164^{***}$ $-0.0125^{**}$	$-0.0132^{*}$ -0.0108	$-0.0193^{**}$ -0.0136
$ au_A = 0.10 \\  au_A = 0.30 \\  au_A = 0.50  ext{}$	$\begin{array}{r} -0.0172^{***} \\ -0.0137^{**} \\ -0.0168^{***} \end{array}$	$\begin{array}{r} -0.0164^{***} \\ -0.0125^{**} \\ -0.0158^{**} \end{array}$	$-0.0132^{*}$ -0.0108 $-0.0140^{**}$	$-0.0193^{**}$ -0.0136 $-0.0201^{**}$

Table 6. The Effect of a marginal increase in schooling on the conditional log wage differential. Sample size: 18328.

Note: see Tables 3 and 5. Legend:  $\delta_{\tau_2-\tau_1} = \frac{\partial Q_Y(\tau_2,X,S,A)}{\partial S} - \frac{\partial Q_Y(\tau_1,X,S,A)}{\partial S}$ .

Table 7. Estimates of  $\beta$ ,  $\lambda$  and  $\phi$ .

	Males	Males	Females	Females
	(1)	(2)	(3)	(4)
β	$0.051^{***}$	$0.050^{***}$	$0.070^{***}$	$0.072^{***}$
$\lambda$	$-0.0021^{***}$	$-0.0022^{***}$	$-0.0025^{***}$	$-0.0021^{***}$
$\phi$	$-0.004 - 0.0089^{***}$	$-0.013^{***}$	$-0.003 - 0.0091^{***}$	$-0.0119^{***}$
	.003	.0032	.0017	.0016
Observations	25	15	25	15
R Squared	0.680	0.692	0.856	0.836

Note: columns (1) and (3) are estimates based on the 25 estimated returns in Table 5 (by gender). Columns (2) and (4) are based instead on excluding the bottom two rows of Table 5 and retaining 15 estimated returns (by gender). The regressors  $G_A^{-1}(\tau_A)$  and  $G_U^{-1}(\tau_U)$  are computed using the deciles of the empirical distributions.

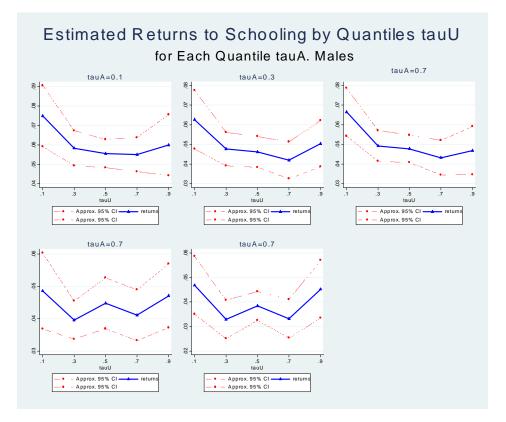


Figure 1: Estimated returns conditional on  $\tau_A.$  Males

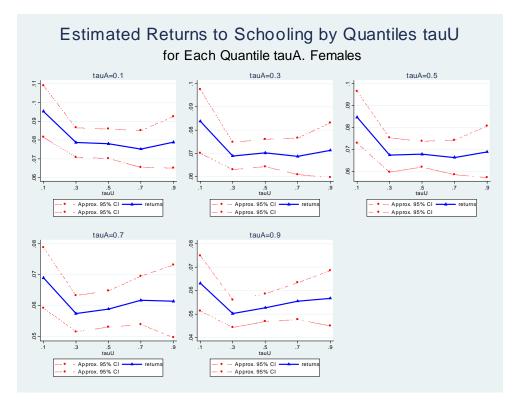


Figure 2: Estimated returns conditional on  $\tau_A.$  Females

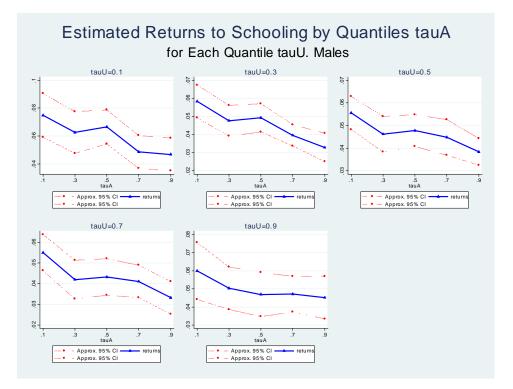


Figure 3: Estimated returns conditional on  $\tau_U$ . Males

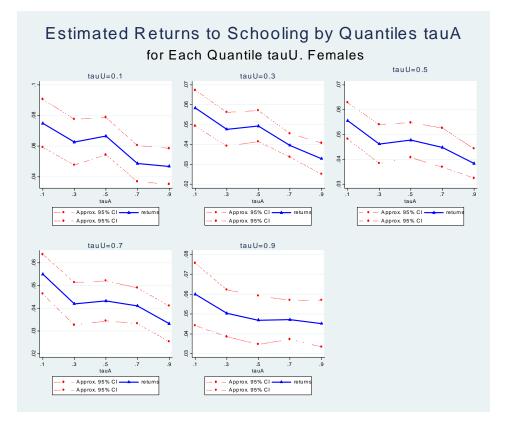


Figure 4: Estimated returns conditional on  $\tau_U$ . Females

	Adop	Adoption of the comprehensive school reform							
Major Regions	1972	1973	1974	1975	1976	1977			
Uusima	0.00	1.15	0.00	6.19	25.16	67.50			
Etela-Suomi	4.26	6.65	12.22	28.36	48.51	0.00			
Ita-Suomi	1.07	31.13	37.69	30.11	0.00	0.00			
Vali-Suomi	4.79	37.61	26.14	31.45	0.00	0.00			
Pohjois-Suomi	52.43	5.40	42.17	0.00	0.00	0.00			

Table A. 1: The distribution of individuals born in 1960-66 across the major regions of Finland and the year of adoption of the comprehensive school reform.

We thank Tuomas Pekkarinnen for kindly providing this table. The table is based on data from the Finnish Longitudinal Census Data Files (FLCD, Statistics Finland, years 1970,1975,1980,1985,1990,1995 and 2000).

Table A. 2: Effect of school reforms on educational attainment across European countries: evidence from the literature. Countries: Denmark, Finland, France, Germany (West), Italy, Ireland, Netherlands, Sweden.

Country	Reform	Effect on Ed. Attainment	Data and References
Denmark	1958	+0.35(women)-+0.4(men)yrs of edu controlling for trend	Danish National Work Environment Cohort ( <i>WECS</i> ) Study, 1990,1995, see Arendt [2005]
Finland	1972-1977	+0.36 in gender gap	Finnish Longitudinal Census Data ( <i>FLCD</i> ) yrs 1970,1975,) 1980,1985,1990,1995,2000) see Pekkarinen [2005]
France	1957	nearly +0.34 yrs of school. using control group (effect varies according to parental backgr.) -20% drop-out rates among farm-workers' sons	French National Labour Force Survey see Grenet [2004]
Germany (West)	1947-1969	+ 0.28% yrs of school. (applies to students in the basic track)	Qualification and Career Survey ( <i>QaC</i> ), MicroCensus, social security records (1% sample) period 1975-1995, see Pischke and Watcher
Italy	1963	<ul> <li>+2.1% enrolment in 8<sup>th</sup> grade</li> <li>+0.21 yrs of schooling</li> <li>+38% (women), +12% men prop. of</li> <li>those achieving high school degree</li> </ul>	Annual Report on Schooling (1948-1979) Labor Force Survey (October 1992-1997) see Brandolini and Cipollone [2002] Survey on Household Income and Wealth, 1991, see Flabbi [1999]
Ireland	1967 (Fees Abolition)	(-0.1)- (-1.8) yrs of schooling	see Denny and Harmon [2000]
Netherlands	1968	+0.71 yrs (males) + 1.33 yrs (females)	OSA- Labour market Survey (1985,1986 ,1988, 1990, 1992 and 1994) see Plug [2001]
Sweden	1950	+10% (males), +8% (females) prop. of those achieving jun. high sch. +0.27(males), +0.22(females) yrs of sch. ( <i>via</i> propensity score match.) effect varies with ability level	Individual Statistics project data merged with administrative data, 1985-1996 see Meghir and Palme [2005]

Table A. 3: Effect of school reforms on earnings across European countries: evidence from the literature. Countries: Denmark, Finland, France, Germany(West)

Country	Reform	Effect on Earnings	Data and References
Finland	1972-1977	-0.029 (men) 0.012 (women) -0.004 (all) non ac. fathers: -0.032 (men) -0.004 (women) -0.013 (all) academic fathers: -0.027 (men) 0.038 (women) 0.005 (all) (log taxable income, euros)	Finnish Longitudinal Census Data ( <i>FLCD</i> ) yrs 1970,1975,) 1980,1985,1990,1995,2000) see Pekkarinen [2005]
Germany (West)	1947-1969	0.004-0.019 (all) -0.013-0.010 (basic track) (log gross monthly wage) 0.003-0.005 (all) 0.001-0.002 (basic track) (log net monthly income) -0.003-0.005 (all) -0.003-0.005 (all) (log earnings)	Qualification and Career Survey ( <i>QaC</i> ) MicroCensus security security records records (1% sample) period 1975-1995, see Pischke and Watcher [2005]

Table A. 4: Returns to education across European countries (identification exploiting instrumental variables, i.e. reforms of the schooling system). Evidence from the literature. Countries: France, Germany (West), Italy, Ireland, Netherlands, Sweden

Country	Reform	Returns to Education	Data and References
France	1957	0.043-0.046 (DD estimate) 0.018-0.027 (DDD estimate)	French National Labour Force Survey see Grenet [2004]
Germany (West)	1947-1969	0.007-0.032 0.005-0.010	Qualification and Career Survey ( <i>QaC</i> ), MicroCensus, social
Italy	1963	Females, ft workers, (various IV-based id. strat.) -0.028-0.024 (1992) 0.051-0.138 (1997) 0.031-0.088 (1992-97) (log real gross weekly earnings)	Annual Report on Schooling (1948-1979) Labor Force Survey (October 1992-1997) see Brandolini and Cipollone [2002]
		-0.022-0.018 (1992) 0.039-0.109 (1997) 0.024-0.072 (1992-97) (log real net weekly earnings)	see Brandolini and Cipollone [2002]
		0.03 (women) 0.05 (men) (log annual earnings less tax plus no monetary integration)	Survey on Household Income and Wealth, 1991, see Flabbi [1999]

	Ref. $\bar{c}$	( <i>c</i> -7, <i>c</i> +7)	ECHP 2001	
Major Regions			ISSP 2001	ISSP 2002
Uusima	1966	1957-1975	28-42	29-43
Etela-Suomi	1965	1956-1974	29-43	30-44
Ita-Suomi	1963	1954-1972	31-45	32-46
Vali-Suomi	1962	1953-1971	32-46	33-47
Pohjois-Suomi	1961	1952-1970	34-47	34-48

Table B. 1: Summary on the age at which individuals are surveyed. Finland.

Note: Sub-sample of individuals born at most 7 years before and 7 years after the year of birth of the first cohort potentially affected by the reform with no missing data on the following variables relevant for the analysis: age, gender, lagged country specific unemployment rate and GDP per capita, country and gender specific labour force participation rate at the estimated time of labour market entry, the country specific fertility rate, GDP per head and unemployment rate at the age affected by the country specific reform, employment status).

Table B. 2: Summary on the age at which individuals are surveyed. Germany.

	Ref. $\bar{c}$ ( $\bar{c}$ -7, $\bar{c}$ +7) International Social Survey					ECHP				
Country			1993	1995	1996	1997	1998	2000	2002	2001
Schleswig-Holstein	1941	1932-1950	45-57	47-59	48-60	49-61	51-60	55-64	59-65	53-65
Hamburg	1934	1925-1943	52-56	54-59	56-60	56-61	57-61	60	64-66	60-65
Niedersachsen	1947	1938-1956	39-52	41-55	42-56	43-57	44-58	46-60	49-62	47-61
Bremen	1943	1934-1952	50-55	n.a.	49-60	51-60	48-62	51-64	61	51-65
NordrWest., Hessen	1953	1944-1962	33-47	35-49	36-50	37-51	38-52	40-54	42-56	41-55
RheinPf.,Baden-W.										
Bayern	1955	1946-1964	31-45	33-47	34-48	35-49	36-50	38-52	40-54	n.a.
Saarland	1949	1940-1958	34-57	39-53	47-49	49	43-56	53-56	48	45-57

Note: see Table B. 1.

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Table B. 3: Descriptive statistics. Sub-sample of individuals born at most 7 years before and 7 years after the year of birth of the first cohort potentially affected by the reform.

Country	Survey Year	Sample Size	Age Range <sup>+</sup>	Country	Survey Year	Sample Size	Age Range
A	1000	170	41.55	X 1 1	1000	200	20, 42
Austria	ISSP, 1994	172	41-55	Ireland	ISSP, 1993	298	28-42
	ISSP, 1995	191	41-55		ISSP, 1994	271	29-43
	ISSP, 1998	192	45-59		ISSP, 1995	315	30-44
	ISSP, 2000	177	47-61		ISSP, 1996	313	31-45
	ISSP, 2001	184	47-61		ISSP, 2000	349	36-50
	ECHP, 2001	1,065	47-61		ECHP, 2001	968	36-50
	SHARE, 2004	781	50-64	Italy	ISSP, 1993	237	37-51
Belgium	ISSP, 2002	329	26-40		ISSP, 1994	235	38-52
	ECHP, 2001	999	26-39		ISSP, 1997	198	41-55
Denmark	ISSP, 1997	297	33-47		ISSP, 1998	197	42-56
	ISSP, 1998	286	34-48		ECHP, 2001	2,447	45-59
	ISSP, 2000	200	38-52		SHARE, 2004	783	48-62
	ISSP, 2001	264	38-52	Netherlands	ISSP, 1993	630	27-41
	ISSP, 2002	343	38-52		ISSP, 1994	630	28-42
	ECHP, 2001	1,034	37-51		ISSP, 1995	711	29-43
	SHARE, 2004	381	41-54		SHARE, 2004	343	39-52
Finland	ISSP, 2001	335	see Table B. 1	Spain	ISSP,1993	343	30-44
	ISSP, 2002	298	see Table B. 1		ISSP,1995	271	31-45
	ECHP, 2001	1,332	see Table B. 1		ISSP,1997	277	33-47
France	ISSP, 1996	342	37-51		ISSP,1998	262	34-48
	ISSP, 1997	280	38-52		ISSP,2000	480	36-50
	ISSP, 1998	243	38-52		ISSP,2002	174	39-53
	ISSP, 2002	477	42-56		ECHP,2001	461	37-51
	SHARE, 2004	424	44-58		SHARE,2004	2,435	41-54
Greece	ECHP, 2001	2,010	31-45	Sweden	ISSP,1994	345	37-51
	SHARE, 2004	113	38-48		ISSP,1995	321	38-52
Germany	ISSP, 1993	245	see Table B. 2		ISSP,1996	352	39-53
5	ISSP, 1995	327	see Table B. 2		ISSP,1998	304	41-56
	ISSP, 1996	578	see Table B. 2		ISSP,2000	270	44-58
	ISSP, 1997	304	see Table B. 2		ISSP,2002	270	45-59
	ISSP, 1998	265	see Table B. 2		SHARE,2004	1,167	47-61
	ISSP, 2000	273	see Table B. 2		, -	/	
	ISSP, 2002	216	see Table B. 2				
	ECHP, 2001	1,047	see Table B. 2				

Note: see Table B. 1.  $^+$  at the time the survey was carried out.

German data from SHARE 2004 have been excluded because there was no available information on the region of residence. Such information is necessary to assign individuals to the pre- or post-reform groups.

Country			Con	sumer Pri	ce Index		
	1993	1994	1995	1996	1997	1998	1999
Austria	96.84	98.18	96.49	96.61	98.10	99.10	99.36
Belgium	90.86	92.70	93.88	95.16	97.31	99.16	99.13
Denmark	85.04	87.53	90.21	91.71	94.91	96.88	97.52
Finland	106.31	106.51	98.09	96.67	97.09	98.97	98.94
France	91.23	91.96	93.07	94.11	95.81	97.43	97.83
Germany	96.41	97.99	98.98	99.56	100.98	101.10	100.71
Greece	67.07	75.95	82.40	88.17	91.82	95.85	98.15
Ireland	82.40	85.13	87.57	89.06	90.42	94.12	96.98
Italy	84.62	87.65	93.04	96.72	96.58	97.58	98.46
Netherlands	92.75	94.49	94.85	94.83	96.37	98.60	99.74
Spain	84.29	88.83	93.16	96.31	95.65	96.71	97.98
Sweden	101.10	102.85	104.32	97.60	97.44	98.54	99.10
Country		Consu	mer Price	Index		Exchange	PPP
	2000	2001	2002	2003	2004	rates	(2000)
Austria	100	103.64	105.08	106.94	108.99	13.76	13.33
Belgium	100	104.57	107.00	109.19	110.52	40.34	40.38
Denmark	100	103.89	106.13	107.66	106.57	7.46	9.04
Finland	100	104.43	107.44	108.08	108.92	5.95	6.90
France	100	105.50	108.39	110.85	111.41	6.56	6.68
Germany	100	104.48	105.33	105.17	104.81	1.96	1.95
Greece	100	105.12	110.71	116.24	116.85	340.75	263.75
Ireland	100	106.52	110.19	111.79	111.48	0.79	0.85
Italy	100	104.11	107.90	111.31	113.72	1936.27	1715.15
Netherlands	100	104.77	108.86	110.96	112.10	2.20	2.20
Spain	100	106.48	108.57	111.41	112.65	166.39	138.07
Sweden	100	105.90	111.16	115.71	120.19	9.18	10.79

Table B. 4: Price indeces, exchange rates and purchainsing power parity indeces used in the application.

Note: consumer price indeces are from OECD, various publications (base year: 2000). Exchange rates are from Table 10.1 in SHARE, 2005 - for countries covered in SHARE rel. 1 - and from the EU Commission *DG II* (Dec. 2000) for the other countries. PPP stands for Purchainsing Power Parities; those reported are the PPP in the year 2000 are from the European Community Houseohold Panel.

Code	ISSP93	ISSP94	ISSP95	ISSP96	ISSP97
Austria	-			-	-
Belgium	-	-	-	-	-
Denmark	-	-	-	-	(kr)/mm <sup>+</sup>
Finland	-	-	-	-	-
France	-	-	-	/mm (Francs)	(Francs)
Germany	net/mm (DEM)	net/mm (DEM)	net/mm (DEM)	net/mm (DEM)	(DEM)
Greece		-	-	-	-
Ireland	net/w	net/w	gross/yy	gross/yy	-
	(Ir. Pounds)	(Ir. Pounds)	(Ir. Pounds)	(Ir. Pounds)	-
Italy	net/mm	net/mm			(1000 Lire)
	(1000 Lire)	(1000 Lire)			(1000 Lire)
Netherlands	net (Gld)	net (Gld)	net (Gld)	-	-
Spain	(Pta)	(Pta)	/mm (Pta)		(Ptas)
Sweden	NA	gross/mm (Sk)	gross/mm (Sk)	gross/mm (Sk)	/mm (Sk)
Austria	net/mm (Sh.)		net/mm (Sh.)		net/mm (EURO)
Belgium	-	-	-	-	net/mm (EURO)
Denmark	gross/yy (Dkr)	-	gross/yy (Dkr)	/yy (Dkr)+	gross/yy
Finland	-	-			
France	/mm (Francs)		-		net/mm (EURO)
Germany	net/mm (DEM)		net/mm (DEM)		net/mm (EURO)
Greece	-	-	-	-	-
Ireland	gross/w	-	gross/w	-	
	(Ir. Pounds)	-	(Ir. Pounds)	-	
Italy	net/mm	-	-		-
	(1000 Lire)	-	-		-
Netherlands		-		-	
Spain	net/mm (Ptas)		/mm (Ptas)		/mm (EURO)
Sweden	gross/mm (SEK)		gross/mm (SEK)	-	gross/mm (SEK)

Table B. 5: Measures of earnings (currency) in ISSP 1993-2002. Differences between countries and surveys

Note: "-" indicates that the survey was not conducted; empty cells indicate that either additional information is not available or there were no departures from the general coding of the question; net and gross stand for net and gross income respectively, and /yy /mm and /w indicate that the reference period is either a year, the month or the week. <sup>+</sup> the reference time period was not clearly stated in the codebook and it has been deducted from the comparison with other surveys.

Table B. 6: Descriptive statistics. Earnings (at 2000 prices and in PPP units), hours worked, and proportion employed. Sub-sample of individuals born at most 7 years before and 7 years after the year of birth of the first cohort potentially affected by the reform. Males.

Country	Survey Year	Sample size	Prop.	Average (st.err) [min	n-max]
		wage-hours-status	Empl.	Wage/month	Hours/week worked
Austria	ISSP,1994	23-28-29	96.6	1,910.2 (1,049.9) [534.8-3,820.3]	45.9 (15.5) [7-85]
	ISSP,1995	31-28-38	87.9	1,527.3 (625.1) [311-2,527.7]	40.3 (12.4) [15-80]
	ISSP,1998	20-24-27	88.9	1,703 (673.2) [300.1-3,150.8]	43.9 (10.7) [20-80]
	ISSP,2000	19-0-26	92.3	1,985.9 (780.1) [675.1-3,750.7]	n.a.
	ISSP,2001	23-0-33	78.8	1,834.6 (731.8) [651.4-3,618.9]	n.a.
	ECHP,2001	289-346-419	68.5	2,454.5 (1,191.6) [150.8-7,838.5]	44.6 (11.1) [17-90]
	SHARE,2004	133-177-292	56.8	1,861 (2,020.2) [19.7-13,4616.6]	44.4 (13.4) [0-100]
Belgium	ISSP,2002	43-156-159	82.4	1,256.8 (399.6) [512.6-2,2427.5]	44.1 (13) [20-90]
	ECHP,2001	361-415-435	83	2,229.8 (844) [457.9-6,276.1]	43.7 (9.5) [16-90]
Denmark	ISSP,1997	123-119-129	84.5	2,743.4 (1,094.9) [485.5-5,826.2]	40.5 (7.7) [10-70]
	ISSP,1998	124-109-133	81.2	2,679.1 (1,267.7) [665.9-6,183.7]	39.9 (8.5) [5-65]
	ISSP,2000	102-89-102	87.3	3,220.7 (1,303.5) [645.1-5,990.6]	40.7 (7.5) [15-70]
	ISSP,2001	128-112-128	84.4	3,153.4 (1,217.9) [443.6-5,322.6]	41.4 (8.0) [24-81]
	ISSP,2002	140-139-142	85.2	3,156.8 (1,291.8) [607.9-5,644.4]	41.8 (8.7) [20-80]
	ECHP,2001	435-487-523	82.8	2,980.8 (1,169.6) [255.5-9,261.4]	41 (8.7) [15-91]
	SHARE,2004	142-149-174	81.6	3,324.3 (5,067.5) [2.4-60,538.6]	41 (10.4) [8-84]
Finland	ISSP,2001	141-125-148	77.0	2,006.9 (2,621.9) [1.4-30,518.5]	40.4 (10.3) [8-90]
	ISSP,2002	130-114-139	71.2	1,805.1 (863.3) [.8-4,249]	40.6 (10.8) [6-85]
	ECHP,2001	513-641-679	75.3	2,033.4 (886.9) [41.6-8,323.2]	44.6 (11.9) [15-96]
France	ISSP,1996	211-202-221	93.2	2,440.2 (1,717.9) [317.9-3,743]	44.0 (9.7) [18-80]
	ISSP,1997	131-136-136	83.8	2,353.5 (1,474.2) [624.6-7,807.4]	40.7 (13.6) [0-75]
	ISSP,1998	115-101-121	86.8	2,437.4 (1,850.9) [307.1-8,445]	43.8 (9.5) [20-80]
	ISSP,2002	165-147-111	78.4	2,232.0 (1,707.8) [276.2-8,971.8]	43.3 (9.4) [30-70]
	SHARE,2004	159-159-196	81.6	3,079.7 (3,267.5)	43 (12.5) [0,105]
Greece	ECHP,2001	554-993-990	55.0	1,350.9 (817.4) [245.3-12,198.8]	45.2 (11.1) [15-90]

Note: Data on wages in the 1999 International Social Survey Program are reported as deciles of the wage distribution. n.a stands for "not available".

Table B. 7: Descriptive statistics. Earnings (at 2000 prices and in PPP units), hours worked, and proportion employed. Sub-sample of individuals born at most 7 years before and 7 years after the year of birth of the first cohort potentially affected by the reform. Males.

Country	Survey Year	Sample size	Prop.	Average (st.err)	
		wage-hours-status	Empl.	Wage/month	Hours worked
Germany	ISSP,1993	87-111-117	95.7	1,978.5 (959.8) [479.2-7,986.2]	42.0 (6.8) [20-61]
	ISSP,1995	183-182-188	97.3	1,823.6 (584.2) [259.3-2,593]	45.2 (11.4) [9-96]
	ISSP,1996	249-0-313	88.8	1,889.3 (980.4) [257.3-6,4441]	n.a.
	ISSP,1997	159-157-162	74.7	1,799.8 (761.9) [254.1-3,049.8]	38.3 (18) [0-80]
	ISSP,1998	78-108-126	85.7	1,740.4 (568.9) [691.9-3,522.6]	41.7 (8.5) [20-80]
	ISSP,2000	104-113-125	88.8	2,132.8 (1,178.8) [256.6-7,699.4]	46.1 (11.3) [8-84]
	ISSP,2002	92-93-105	77.1	2,412.4 (1,703.3) [190.6-14,296.3]	46.3 (11.0) [25-96]
	ECHP,2001	400-460-525	74.5	3,149.0 (1,358.9) [98.3-8,196.0]	45.6 (9.8) [20-96]
	SHARE,2004	+	+	+	+
Ireland	ISSP,1993	126-107-134	80.0	1,116.4 (574.3) [142.9-4,002.6]	45.3 (11.4) [16-80]
	ISSP,1994	103-82-114	77.2	1,309.2 (699.9) [221.4-4,427.7]	44.7 (10.1) [11-70]
	ISSP,1995	142-151-153	88.2	1,782.5 (996.9) [145.7-4,080.1]	46.1 (13.0) [13-96]
	ISSP,1996	141-148-151	86.1	1,767.1 (1,025.1) [143.2-4,012.0]	45.9 (13.1) [13-96]
	ISSP,2000	0-118-153	86.3	n.a.	46.6 (11.7) [6-85]
	ECHP,2001	315-420-475	65.9	2,413.2 (1,293.7) [239.9-10,127.5]	44.7 (12.9) [10-90]
Italy	ISSP,1993	103-105-111	94.6	1,577.4 (701.3) [103.4-3,513.9]	42.3 (8.5) [18-70]
	ISSP,1994	99-103-105	97.1	1,602.7 (798.2) [166.3-4,356.8]	40.1 (10.3) [8-80]
	ISSP,1997	57-82-82	63.4	2,313.1 (1,090.4) [603.7-5,614.2]	41.1 (14.7) [0-89]
	ISSP,1998	60-77-89	89.9	1,667.3 (968.1) [304.7-7,169.9]	42.8 (13) [6-90]
	ECHP,2001	663-903-1,071	61.6	2,033.3 (1,117.1) [336-10,192.5]	41.4 (8.9) [15-80]
	SHARE,2004	125-185-267	66.7	1,917.9 (2,196.4) [.1-19,853.7]	40 (16.2) [0-100]

Table B. 8: Descriptive statistics. Earnings (at 2000 prices and in PPP units), hours worked, and proportion employed. Sub-sample of individuals born at most 7 years before and 7 years after the year of birth of the first cohort potentially affected by the reform. Males.

Country	Survey Year	Sample size	Prop.	Average (st.err)	
		wage-hours-status	Empl.	Wage/month	Hours worked
Netherlands	ISSP,1993	235-232-268	85.1	1,753.1 (724.9) [429-5,025.1]	41.2 (9.7) [10-90]
	ISSP,1994	238-223-259	84.6	1,710.2 (733.8) [280.7-4,451.4]	39.1 (8.6)[8-70]
	ISSP,1995	290-275-319	85.9	1,869.4 (893.2) [179.8-4,913.8]	38.8 (8.8) [8-84]
	SHARE,2004	104-123-137	88.3	3,075.7 (1,607.1) [123.9-8,519.2]	42.7 (12.5) [6-80]
Spain	ISSP,1993	100-117-139	76.3	1,033.8 (623.5) [193.3-3,867]	40.8 (9.7)[8-84]
	ISSP,1995	92-95-133	76.7	952.8 (496.9) [174.9-2,429.5]	39.8 (9.4) [8-70]
	ISSP,1997	98-125-129	57.4	1,008.4 (614.7) [189.3-3,786.2]	30.9 (20.9) [0-80]
	ISSP,1998	173-185-238	77.8	1,231.2 (598.7) [299.6-3,370.1]	43.5 (8.7) [25-80]
	ISSP,2000	50-51-62	90.3	1,303.7 (683.4) [181.1-3,259.3]	43.3 (6.4) [35-60]
	ISSP,2002	167-209-230	65.7	1,552.8 (974.1) [271-7,589.1]	44.3 (11.6)[8-90]
	ECHP,2001	865-1,107-1,220	70.5	1,938 (1,329.2) [119.3-15,442.8]	44.6 (10.2) [15-96]
	SHARE,2004	94-107-129	79.1	1,651.7 (2,321.5) [.5-20,889.5]	41.7 (14.7) [0-80]
Sweden	ISSP,1994	171-173-175	92	1,516 (576.2) [270.3-2,703.3]	41.1 (8.1)[9-84]
	ISSP,1995	144-139-148	85.1	1,573.2 (508.2) [177.7-2,487.5]	43.4 (6.7) [30-70]
	ISSP,1996	177-172-185	86.5	1,694.7 (628.6) [284.9-2,848.5]	43.6 (9.2) [1-75]
	ISSP,1998	125-125-132	87.9	2,009.7 (2,093) [470.3-23,512.7]	42.7 (10.9) [8-90]
	ISSP,2000	137-126-141	93.6	2,350.8 (2,042) [556.1-23,169.6]	43.6 (8.8) [10-80]
	ISSP,2002	139-139-143	75.6	2,146.9 (1,285.3) [698.1-13,001.6]	40.9 (9.1) [8-75]
	SHARE,2004	520-526-638	82.6	2,354.5 (1,881.2) [2.1-21,840.4]	42.6 (10.8) [1-100]

Table B. 9: Descriptive statistics. Earnings (at 2000 prices and in PPP units), hours worked, and proportion employed. Sub-sample of individuals born at most 7 years before and 7 years after the year of birth of the first cohort potentially affected by the reform. Females.

Country	Survey Year	Sample size	Prop.	Average (st.err)	
		wage-hours-status	Empl.	Wage/month	Hours/week worked
Austria	ISSP,1994	84-81-143	58.0	947.8 (490.0) [382.0-2,368.6]	34.9 (14.3) [3-84]
	ISSP,1995	112-87-158	56.3	921.5 (535.3) [310.0-2,526.7]	38.2 (17.8) [4-80]
	ISSP,1998	88-74-165	41.2	1040.9 (593.9) [300.1-3,150.8]	38.1 (16.5) [5-84]
	ISSP,2000	88-0-151	31.1	976.9 (539.8) [225.0-2,850.5]	n.a.
	ISSP,2001	126-0-151	25.2	658.3 (614.4) [0 - 2,750.3]	n.a.
	ECHP,2001	214-277-646	30.8	1535.5 (935.8) [57.0-5-283.6]	40.8 (15.6) [15-96]
	SHARE,2004	122-151-489	23.5	1,270 (2,208.1) [.1-13-416.5]	33.2 (15.6) [0-100]
Belgium	ISSP,2002	27-138-170	75.3	1,675.0 (2,645.7) [266.1-14,751.6]	34.7 (13.1) [6-85]
	ECHP,2001	454-460-564	78.2	1,664.8 (761.4) [106.6-6,272.7]	35.5 (10.0) [15-84]
Denmark	ISSP,1997	156-148-168	79.8	1,936.9 (861.7) [455.5-5,826.2]	34.6 (7.4) [5-50]
	ISSP,1998	147-130-153	77.1	2,085.2 (816.7) [665.9-6,183.7]	35.8 (7.0) [18-75]
	ISSP,2000	96-88-98	89.8	2,273.8 (867.6) [645.1-5,068.9]	35.5 (8.4) [10-60]
	ISSP,2001	134-119-136	86.0	2,212.8 (960.5) [443.6-5,322.6]	35.5 (7.2) [20-80]
	ISSP,2002	197-192-201	80.1	2,209.5 (919.6) [607.9-5,644.4]	36.1 (6.5) [15-60]
	ECHP,2001	434-441-511	83.6	2,230.8 (793.7) [114.6-8,622.7]	35.5 (8.2) [15-90]
	SHARE,2004	179-179-207	83.6	2,134 (886.2) [1.7-6,572.7]	34.6 (9.6) [0-60]
Finland	ISSP,2001	164-132-187	65.2	1,350.1 (727.4) [1.1-4,161.6]	36.8 (8.5) [2-94]
	ISSP,2002	140-118-159	68.6	1,524 (1,323) [28.1-12,827.3]	35.8 (7.8) [7-71]
	ECHP,2001	509-559-653	77.0	1,488.8 (600.0) [166.5-5,548.8]	37.7 (8.3) [10-96]
France	ISSP,1996	105-90-121	80.2	1,579 (1,027.6) [317.9-7,153.4]	35.8 (9.8) [6-65]
	ISSP,1997	129-144-144	84.7	1,450.7 (863.1) [234.2-5,465.2]	28.7 (16.0) [0-77]
	ISSP,1998	105-95-122	81.1	1,621.7 (933.8) [307.1-5,374.1]	37.1 (8.9) [14-67]
	ISSP,2002	259-212-306	64.4	1,379.7 (798.2) [276.2-4,830.4]	34.2 (10.4) [7-70]
	SHARE,2004	159-164-228	72.8	1,672.6 (2,025.5) [.1-22,021.8]	35.7 (14.3) [0-105]
Greece	ECHP,2001	410-591-1,020	39.6	1,103.6 (444.6) [46.3-3,051.9]	38.4 (9.6) [15-75]
	SHARE,2004	39-44-112	36.6	1,283.5 (766.9) [69.1-4,013.9]	29.1 (20.4) [0-66]

Table B. 10: Descriptive statistics. Earnings (at 2000 prices and in PPP units), hours worked, and proportion employed. Sub-sample of individuals born at most 7 years before and 7 years after the year of birth of the first cohort potentially affected by the reform. Females.

Country	Survey Year	Sample size	Prop.	Average (st.err) [min-max]	
		wage-hours-status	Empl.	Wage/month	Hours worked
Germany	ISSP,1993	62-68-128	53.9	1,099.6 (662.8) [106.5-3,726.9]	32.5 (10.3) [12-60]
	ISSP,1995	113-89-139	60.4	878.9 (594.6) [259.3,2,593.0]	28.4 (10.8) [9-50]
	ISSP,1996	158-0-265	59.6	1,051.1 (737.2) [70.6-6,186.3]	n.a.
	ISSP,1997	121-139-142	50.7	1,027.1 (740.5) [254.1-3,049.8]	22.2 (18.2) [0-80]
	ISSP,1998	71-89-139	50.4	999.2 (518.6) [251.6-2,390.3]	30.8 (12.6) [3-80]
	ISSP,2000	98-108-148	65.5	1,078.3 (629.5) [25.7-3,695.7]	29.6 (13.2) [1-80]
	ISSP,2002	81-67-111	54.0	1,187.2 (1,404.5) [190.6-11,437.1]	33.6 (11.6) [15-80]
	ECHP,2001	368-338-522	61.4	1,611.6 (1,130) [24.5-9,334,1]	34.1 (11.4) [2-85]
Ireland	ISSP,1993	135-78-164	46.3	582.4 (485.1) [142.9-2,573.1]	35.6 (9.2) [7-60]
	ISSP,1994	114-72-157	45.9	793 (644.1) [221.4-4,427.7]	36.3 (11.3) [5-72]
	ISSP,1995	150-159-162	53.7	812.1 (755.1) [145.7-4,080.1]	37 (17.1) [5-96]
	ISSP,1996	149-158-162	49.4	766.3 (743.2) [143.2-4,012]	37.9 (19) [5-96]
	ISSP,2000	0-103-196	55.1	n.a.	30.3 (13.8) [4-99]
	ECHP,2001	277-269-493	50.9	1,333.5 (899.5) [47.5-4,926.1]	30 (10.6) [9-90]
Italy	ISSP,1993	68-64-126	49.2	1,062.4 (560.9) [103.4-3,513.9]	33.7 (10.8) [7-60]
	ISSP,1994	80-76-130	56.9	1,164 (752) [166.3-4,356.8]	37 (13.1) [10-90]
	ISSP,1997	28-113-116	33.6	1,808.9 (1,073.7) [664.1-4,829.5]	14.5 (18.5) [0-60]
	ISSP,1998	46-44-108	41.6	1,025.2 (783.7) [119.5-4,899.5]	35.5 (13.7) [6-60]
	ECHP,2001	464-571-1,376	33.2	1,506 (592.2) [224-5,040.2]	34.9 (10.1) [15-80]
	SHARE,2004	126-162-516	28.4	1,595.3 (1,672.2) [.3-12,408.6]	29.4 (15.3) [0-70]

Table B. 11: Descriptive statistics. Earnings (at 2000 prices and in PPP units), hours worked, and proportion employed. Sub-sample of individuals born at most 7 years before and 7 years after the year of birth of the first cohort potentially affected by the reform. Females.

Country	Survey Year	Sample size	Prop.	Average (st.err)	
		wage-hours-status	Empl.	Wage/month	Hours worked
Netherlands	ISSP,1993	298-176-362	38.7	1,672.5 (800.5) [183.8-5,025.1]	24.7 (12.2) [2-60]
	ISSP,1994	325-187-371	38.3	1,673.2 (821.6) [280.7-5,413.8]	25.1 (13.2) [1-60]
	ISSP,1995	330-223-392	45.1	1,824.3 (893.2) [179.8-4,913.8]	23.3 (11.4) [2-56]
	SHARE,2004	141-145-206	66.0	1,609.9 (1,805) [46.5-19,129.4]	25.9 (11.5) [4-80]
Spain	ISSP,1993	67-47-132	31.8	766.7 (530.3) [193.3-2.685.4]	38.2 (14) [6-96]
	ISSP,1995	55-50-144	34.7	836.5 (528.9) [174.9-1,846.4]	33.8 (12.5) [4-66]
	ISSP,1997	82-130-133	30.1	572.5 (489.9) [189.3-2,366.4]	12.5 (17.6) [0-60]
	ISSP,1998	107-92-242	37.6	921.4 (555.9) [299.6-2,621.2]	36.6 (10.1) [9-60]
	ISSP,2000	48-45-112	41.1	848.8 (505.5) [181.1-1,810.7]	36.1 (11.1) [10-70]
	ISSP,2002	104-121-231	46.3	959.6 (524.3) [271-2927.2]	34.4 (10) [6-55]
	ECHP,2001	527-607-1,215	40.9	1,389.1 (888.3) [54.7-6,941.7]	37.7 (10) [14-90]
	SHARE,2004	97-107-225	42.7	1,958.3 (6,651.9) [20.9-62,668.5]	34.3 (13.4) [0-70]
Sweden	ISSP,1994	116-151-170	82.9	1,093.8 (471.4) [270.3-2,703.3]	34.9 (8.7) [10-60]
	ISSP,1995	170-159-173	88.4	1,774.8 (423.6) [177.7-2,487.5]	36.3 (7.6) [8-55]
	ISSP,1996	152-152-167	85.6	1,255.0 (506) [284,9-2,848.7]	35.6 (9.8) [8-60]
	ISSP,1998	160-155-172	83.7	1,429 (660.7) [94.1-4,984.7]	36.2 (9) [8-72]
	ISSP,2000	119-111-129	83.7	1,820.9 (1,552.9) [556.1-16,682.1]	36.9 (6.7) [15-50]
	ISSP,2002	118-117-127	84.3	1,638.7 (755.3) [436.3-7,417]	36.5 (9.8) [5-75]
	SHARE,2004	520-526-638	79.8	1,707.9 (1,142.9) [.7-16,016.3]	36.3 (10.3) [0-80]