

Firm-Level Social Returns to Education[€]

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Abstract: Do workers benefit from the education of their co-workers? This question is examined first by introducing a model of *on-the-job schooling*, in which educated workers may transfer part of their general skills to uneducated workers. After presenting the theoretical results, we then conduct an empirical analysis drawing on a matched panel of Portuguese firms. Schooling endogeneity is tackled by considering firm fixed effects and instruments based on schooling lags and the lagged share of retirement-age workers. We find evidence of large firm-level social returns (ranging between 14% and 23% – and thus larger than standard estimates of private returns) and of significant returns accruing to less educated workers but not to their more educated colleagues.

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1. Introduction

While the labour economics literature has devoted considerable attention to the estimation of *private* returns to education, relatively little is known about the importance of education for society as a whole. However, from many different points of view, *social* returns to education are the key parameter to take into account. For instance, a better understanding of whether education increases total output (and not simply of how much more the educated earn when compared to the less educated) is of paramount importance in a number of policy questions.

Possibly the most important of these policy questions – which frequently comes up in the public debate – is how should education costs be split between the student and the taxpayer (see Gemmell, 1997). In fact, if most of the returns to education are only private, then the case for public subsidies for education probably comes significantly eroded. A related matter concerns the overall importance of education for economic growth. For instance, some of the endogenous growth literature argues that education can sustain indefinitely positive growth rates of income per capita (Lucas, 1988). Again, the empirical support of these views is far from clear.

One explanation for the relatively small number of empirical studies of social returns to education lies on the demanding estimation strategy required. Firstly, as in many other areas of empirical research, one needs appropriate counterfactuals. In the case of private returns to education, this amounts to information about the earnings of very similar individuals but that have different levels of education: one original strategy based on this idea studies differences in earnings and schooling of twins (Ashenfelter and Krueger, 1994).

Secondly, one needs exogenous variation. For instance, in the twins approach, one may wonder why do individuals as similar as twins have different schooling levels. In particular, it is unclear whether some of those factors driving schooling variability across twins will not also be correlated with wages. In this case, methods based on instrumental variables can generate education variability which can then be used for identification purposes. One important example of this approach is Harmon and Walker (1995), who draw on the increases of school leaving age in the UK to estimate private returns to education. However, even this approach is

not free from criticism as the results obtained may be specific to the group affected or be difficult to interpret without a structural model that provides theoretical context.

Finally, the estimation of social returns to education (unlike private returns) may further have to deal with possible general equilibrium effects. For instance, if high- and low-skill workers are imperfect substitutes, then an increased supply of the former will affect the prices of both types of workers even if spillovers do not exist (see Ciccone and Peri, 2002).

On top of the estimation hurdles described above, the few studies available on the topic of social returns to education have not yet reached any stylised fact, not even on whether there are social returns, not to mention their specific magnitude. In particular, two of the most prominent papers, Acemoglu and Angrist (2000) and Moretti (2004a), find conflicting results, even if drawing on similar data sets for the same country (the U.S.), although covering a slightly different time period.

Acemoglu and Angrist (2000) draw on compulsory schooling laws to identify the impact of schooling on average wages in US states, finding insignificant external returns. On the other hand, Moretti (2004a) uses city demographic structures and the geographical presence of some colleges to find significant impacts of graduates on the wages of workers, particularly those with lower levels of schooling. These contrasting results also extend to other studies that look at the U.S. case but that do not address the endogeneity of schooling: Rauch (1993) finds positive, significant effects while Rudd (2000) finds insignificant effects. Using a new methodology, based on the construction of counterfactuals, Ciccone and Peri (2002) also find insignificant results.

It should, however, be mentioned that while the panorama on *wage* social returns to education is mixed, a more consistent and encouraging set of evidence has been found for other social domains where education may also matter. Studies focusing on productivity (Moretti, 2004b), crime (Lochner and Moretti, 2004), citizenship (Milligan et al., 2004), and intergenerational effects (Behrman and Rosenzweig, 2002, and Currie and Moretti, 2003) find positive and significant effects of education.

It is in the context of this emerging literature on social returns to education that the present paper seeks to make its contribution, in particular by focusing on wage effects and conducting its estimations at the firm level, rather than across cities or regions, as most previous work. Focusing on a less aggregated level of analysis, we are able to sidestep the general-equilibrium effects induced by imperfect substitutability. Moreover, it is also likely that most of the education externalities that affect productivity (and, subsequently, pay) occur within firms, as it is probably at the firm level that workers, including those of different skills or education levels, interact the most.

Our reasoning is motivated by a model of *on-the-job schooling*, a new concept introduced here. In this model, we argue that educated workers transmit part of their education skills to their uneducated counterparts and that the strength of this spillover depends on some characteristics of these skills, including their non-excludability and irreversibility. The model then shows that, for some labour market structures, the existence of spillovers will lead to a stronger relationship between wages and education at the firm level.

Furthermore, some of the model's implications are tested, on a panel of almost 5,000 firms and their workers, followed for up to nine years. The longitudinal dimension of the data allows us to implicitly control for unobserved differences across firms, as in the twins literature. Moreover, within-firm variation of education is driven from the vigorous educational expansion experienced in Portugal, the country under study. Additionally, we also use lagged schooling and the lagged share of workers approaching retirement age as sources of exogenous variation. This is particularly important as the model emphasises the endogeneity of firm-level education.

Consistently with the model, our findings indicate what we interpret as social (firm-wide) levels of returns to education above those commonly obtained in studies of private returns. The social returns are particularly high when focusing at specific firm/job-levels cells. Again as predicted by the model, we also find that the less-educated workers are the ones that benefit the most from increases in their firms' average schooling levels.

The remaining of the paper is structured as follows. The next section describes the “on-the-job schooling” model. Section 3 presents the empirical methodology and the data while Section 4 describes the results, the robustness analysis and some extensions. Section 5 concludes.

2. A Model of On-the-Job Schooling

This section presents a simple model that examines the implications of the admission of more skilled workers in a given firm and the reasons why that may increase, more than proportionately, the wages of all workers in that firm. The key assumption is that education may affect the productivity not only of the educated individuals but also of those workers who have not invested in a higher educational level but who interact with colleagues that have.

2.1. General Outline

The model allows the new skills obtained in higher schooling levels by the new workers to be transferred to other workers that were already in the firm. In a sense, new workers act as informal teachers to unskilled workers. To this extent, skilled workers transfer to unskilled workers general knowledge that those (previously) unskilled workers can now use to achieve higher productivity levels in that or other firm.¹ This is thus another sense in which one may argue that “skills beget skills” (Carneiro and Heckman, 2003).

Three aspects of this process of on-the-job schooling should be underlined. The first aspect is about the *non-excludability* of the spillover: we consider that the educated cannot prevent the spillover from occurring, at least to some degree. We argue this assumption is reasonable because if education is valuable for its skills (and not because of signalling issues), then the educated will presumably have to demonstrate or certify their acquisition of those skills by applying them in firms. This application of skills by the educated in their workplace typically takes place in such a way that it allows other workers, including the less educated colleagues,

¹ We will focus on quantity differences in the schooling of different workers. The model could, however, also be adopted to consider quality differences (e.g. degrees in different subjects).

to learn from the educated. Educated workers therefore become workplace teachers of their uneducated colleagues.

A related point is to emphasise the difference between on-the-job schooling and on-the-job training: unlike the latter, on-the-job schooling does not rest on the *deliberate* decision made by firm owners to improve the skills of *some* specific workers. On-the-job schooling is instead an informal process, potentially affecting all (less educated) workers in a firm.

Irreversibility is the second aspect of on-the-job schooling. This is important in that it creates a permanent spillover from the recently-hired skilled workers to the other workers in the firm. It contrasts to the case when productivity is affected only and simply because there are complementarities between different types of workers in the firm's technology. Under the case of complementarities, once unskilled workers are not interacting with their skilled colleagues, the unskilled marginal product would return to its previous, pre-interaction lower level. In the present case, the unskilled marginal product would stay at its previous, higher level.

The third and last aspect in the transmission process is that schooling here is regarded as providing *general* skills. Examples will obviously depend on the specific level and type of schooling considered but will include greater competence in numeracy and literacy. Interpersonal and other “business” skills may also be relevant here. As argued in Acemoglu and Pischke (1999b), “schooling provides *general-purpose knowledge* by teaching conceptual tools and information, useful in a variety of occupations and industries” (page F140, our italics). Since these types of skills are widely used, across many different firms, their possession will increase the workers' outside options.

Another point is that we allow education to present different degrees of these three aspects (non-excludability, irreversibility and generality) in different contexts. This will imply that the strength of the education spillover may vary across occupations and/or industries, for instance. A precise clarification of the practical importance of spillovers is, of course, an empirical matter, which we will examine in Section 4.

It is also important to clarify that the productivity spillovers discussed here do not necessarily imply wage spillovers. Even if such productivity spillovers do indeed materialise, the extent to which wage spillovers will be observed depends, we argue, on the specific type of the labour market. In particular, if the labour market is sufficiently competitive (i.e., subject to few frictions), the outside option of those unskilled workers will improve in correspondence to the importance of the new skills obtained from their educated colleagues. Alternatively, a similar result will apply if the unskilled workers have enough bargaining power, so that the rents created by the spillover are captured by those workers.²

In these two cases, unskilled workers (or, better, workers who were unskilled before the admission of skilled workers) will be able to extract higher wages from their current employers up to the level of their new marginal product. In this sense, productivity spillovers will imply wage spillovers. Moreover, these two labour market types are in this paper observationally equivalent. We will thus use the phrases “perfectly competitive” and “rent-sharing” interchangeably.

However, in the other extreme case, if the labour market is monopsonistic (due to mobility costs, information constraints or other frictions), the outside option of unskilled workers will not change. As discussed in Acemoglu and Pischke (1999a), what are technologically general skills become *de facto* firm-specific skills in a market with frictions. Therefore, employers will capture the entire surplus generated by the inflow of skilled workers and the wages of unskilled workers will not increase.

After having outlined, in a non-technical way, the key features of the model and highlighted the importance of labour market structures, we now describe the model in greater detail and rigour.

² Martins (2004) finds evidence of rent sharing in the Portuguese labour market, using worker/firm fixed effects and exogenous variation in profits derived from interactions between exchange rates and export shares. Rent sharing is also found to be larger for workers that are more educated and more tenured. Other papers, using a variety of methods, also document substantial levels of rent sharing in additional labour markets, including Blanchflower et al (1996) for the U.S., Hildreth and Oswald (1997) for the UK, Arai (2003) for Sweden, and Kramarz (2003) for France. See also Manning (2003) for a recent and thorough treatment of different theoretical and empirical aspects of monopsony.

2.2. Model Description and Results

We consider a two-period, partial equilibrium framework, in which a single, representative firm chooses the quantity of educated workers it will hire in the second period. There are only uneducated and educated workers in this economy, the former only available (to be hired) in the first period and the latter only available in the second period. In the first period, the firm employs only uneducated workers (L_U), whose quantity is normalised at 1; in the second period, the firm chooses how many educated workers to hire (L_S).³

Furthermore, the wages of the educated workers are exogenously defined at W_S – this is because we assume that the firm, our unit of analysis, will never be big enough to be able to affect wages in the whole economy. Unskilled wages are also exogenously defined at W_U ($W_U < W_S$) – this may be the minimum wage, for instance. The product price is normalised at 1.

The firm generates output according to a Cobb-Douglas production function with decreasing returns to scale: $Y = L_U^\alpha L_S^\beta$, $\alpha + \beta < 1$. However, we also allow the interaction between skilled and unskilled workers to lead to on-the-job schooling, whereby the unskilled workers become more productive. The interaction is assumed to be as follows: $L'_U = \gamma L_S, \left(\frac{1}{L_S} < \gamma < 1 \right)$.

This means that L'_U , the effective quantity of educated workers (i.e. after taking into account their higher productivity) in the second period, will depend (positively) on the number of educated workers hired by the firm. Moreover, the bounds on γ (the spillover parameter) ensure that the inflow of educated workers will never decrease the effective size of the uneducated below their previous size in the first period (which was 1) and will also never increase it above the size of the educated workers.⁴

³ These differences in workers' type over the two periods are motivated by the expansion of secondary and tertiary education systems observed throughout the Western world during the second half of the 20th century. For instance, in the case of Portugal, the country examined in the empirical section, compulsory schooling increased from four to nine in only 20 years.

⁴ The bounds on γ also imply that our analysis is implicitly restricted to certain ranges of the other parameters of the model so that $L_S > 1$. For instance, very large gaps between W_U and W_S are ruled out.

We consider three types of labour-market/spillover scenarios: 1) no spillovers (denoted by letter N); 2) spillovers, in a monopsonistic labour market (M); and 3) spillovers, in a perfectly competitive or rent-sharing labour market (P). As mentioned before, the difference between M and P is that in M the unskilled workers will not benefit from any wage increases as their productivity increases due to the spillover, while in P their pay will be raised accordingly. Pay will stay unchanged in N as there are no spillovers of any kind.

We now examine the results of this model under each specific labour-market/spillover case:

2.2.1. Case 1 (no spillovers)

Profits in the second period are affected only by the choice of L_S and the production function becomes $Y_2 = L_U^\alpha L_S^\beta = L_S^\beta$ (since $L_U=1$). The firm's problem will be to find the optimal L_S that maximises profits: $Max_{L_S^N} \Pi_2 = L_S^{N\beta} - W_U - W_S L_S^N$. The first-order condition and its solution (the labour demand function) are:

$$(1) \quad \frac{\partial \Pi_2}{\partial L_S^N} = 0 \Rightarrow L_S^N = \left(\frac{W_S}{\beta} \right)^{\frac{1}{\beta-1}} = L_S^N(W_S; \beta)$$

It can be easily shown that labour demand (for L_S) falls with the educated wage and increases with β . In order to draw some empirically testable predictions of the model, we now derive some expressions for variables that are amenable to empirical examination and that can be related to other studies in the literature and to the other labour market types in this paper. First, the firm-level average wage and its partial derivative with respect to the number of educated workers will be:

$$(2) \quad \bar{W}^N = \frac{W_U + W_S L_S^N}{1 + L_S^N} \Rightarrow \frac{\partial \bar{W}^N}{\partial L_S} = \frac{W_S - W_U}{(1 + L_S^N)^2}$$

The average wage then increases when more educated workers are hired (if their wage is bigger than that of the uneducated workers – which is true by assumption and common sense). Moreover, average education in the firm (measured as the share of educated workers⁵) and its derivative with respect to L_S are given by:

$$(3) \quad \overline{ed}^N = \frac{L_S^N}{1 + L_S^N} \Rightarrow \frac{d\overline{ed}^N}{dL_S^N} = (1 + L_S^N)^{-2}$$

The last two results [(2) and (3)] imply the following positive relationship between log mean wages, $\ln \overline{W}^N$, and average education, \overline{ed}^N :

$$(4) \quad \frac{\partial \ln \overline{W}^N}{\partial \overline{ed}^N} = \frac{\partial \ln \overline{W}^N}{\partial \overline{W}^N} \frac{\partial \overline{W}^N}{\partial L_S^N} \frac{\partial L_S^N}{\partial \overline{ed}^N} = \frac{W_S - W_U}{\overline{W}^N}$$

The increase of log mean wages with respect to mean education is approximately determined by the percentage wage gap between educated and uneducated workers (i.e. by the standard private return to education). This partial derivative is closely comparable to those at the individual-level Mincer private returns to education studies, which will be taken as a benchmark against which we will evaluate our results.

2.2.2. Case 2 (positive spillovers, monopsonistic market)

The firm's problem in this new case will be to find the optimal L_S that maximises profits. However, now L_S has an additional benefit as it increases the productivity of the uneducated, and the production function becomes $Y_2 = (\gamma L_S)^\alpha L_S^\beta = \gamma^\alpha L_S^{\alpha+\beta}$. On the other hand, since in this case the labour market is assumed to be monopsonistic, the increased productivity of the uneducated will not affect their wages, which will stay unchanged. The new problem faced by

⁵ The expression can be adjusted to take into account the difference in years of education between the educated and the less educated groups. This does not, however, bring about any qualitative difference to our findings.

the firm is then $Max_{L_S^M} \Pi_2 = \gamma^\alpha L_S^M{}^{\alpha+\beta} - W_U - W_S L_S^M$ and the first-order condition and its solution are:

$$(5) \quad \frac{\partial \Pi_2}{\partial L_S^M} = 0 \Rightarrow L_S^M = \left[\frac{W_S}{\gamma^\alpha (\alpha + \beta)} \right]^{\frac{1}{\alpha+\beta-1}} = L_S^M(W_S; \alpha, \beta, \gamma)$$

It can be shown that, as before, labour demand falls with the educated wage and increases with β . Moreover, the average wage will also be similar to that of case 1, the difference lying on the new specific hiring level of educated workers (L_S^M instead of L_S^N):

$$(6) \quad \bar{W}^M = \frac{W_U + W_S L_S^M}{1 + L_S^M} \Rightarrow \frac{\partial \bar{W}^M}{\partial L_S^M} = \frac{W_S - W_U}{(1 + L_S^M)^2}$$

Moreover, firm-level average *formal* education, i.e. disregarding on-the-job schooling, and its derivative with respect to the number of educated workers are also given by similar expressions than before:

$$(7) \quad \bar{ed}^M = \frac{L_S^M}{1 + L_S^M} \Rightarrow \frac{d\bar{ed}^M}{dL_S^M} = (1 + L_S^M)^{-2}$$

The two previous results imply again a similar relationship between log mean wages and education:

$$(8) \quad \frac{\partial \ln \bar{W}^M}{\partial \bar{ed}^M} = \frac{\partial \ln \bar{W}^M}{\partial \bar{W}^M} \frac{\partial \bar{W}^M}{\partial L_S^M} \frac{\partial L_S^M}{\partial \bar{ed}^M} = \frac{W_S - W_U}{\bar{W}^M}$$

2.2.3. Case 3 (positive spillovers, competitive or rent-sharing market)

In this last case, the firm's optimal L_S will take into account the spillover benefit for uneducated workers (same production function as in case 2) but also their wage increase

following their increased productivity. Therefore, the new maximisation problem becomes

$$\text{Max}_{L_S^P} \Pi_2 = \gamma^\alpha L_S^{P\alpha+\beta} - W_U (\gamma L_S^P) - W_S L_S^P$$

Notice that the wage bill corresponding to the unskilled workers considered in the profit function takes into account their effective size, which is now more than one because the spillover made them more productive (recall that, by assumption, $\gamma L_S > 1$). Their specific new size depends on the spillover parameter (γ) and the number of educated workers hired (L_S^P).

The first-order condition and its solution are:

$$(9) \quad \frac{\partial \Pi_2}{\partial L_S} = 0 \Rightarrow L_S^P = \left[\frac{W_S + \gamma W_U}{\gamma^\alpha (\alpha + \beta)} \right]^{\frac{1}{\alpha + \beta - 1}} = L_S^M(W_S, W_U; \alpha, \beta, \gamma)$$

It can be shown that, as before, labour demand falls with the educated wage and increases with β . Moreover, now labour demand is also affected (negatively) by the uneducated wage.

Under these new circumstances, the average wage and its partial derivative with respect to the quantity of educated workers will now assume different expressions:

$$(10) \quad \bar{W}^P = \frac{W_U \gamma L_S^P + W_S L_S^P}{1 + L_S^P} \Rightarrow \frac{\partial \bar{W}^P}{\partial L_S^P} = \frac{W_S + \gamma W_U}{(1 + L_S^P)^2},$$

while average (formal) education in the firm and its derivative remain similar to their equivalents in the other labour markets:

$$(11) \quad \bar{ed}^P = \frac{L_S^P}{1 + L_S^P} \Rightarrow \frac{d\bar{ed}^P}{dL_S^P} = (1 + L_S^P)^{-2}$$

The results in (10) and (11) imply the following relationship between log mean wages and education:

$$(12) \quad \frac{\partial \ln \bar{W}^P}{\partial \bar{ed}^P} = \frac{\partial \ln \bar{W}^P}{\partial \bar{W}^P} \frac{\partial \bar{W}^P}{\partial L_s^P} \frac{\partial L_s^P}{\partial \bar{ed}^P} = \frac{W_s + \gamma W_u}{\bar{W}^P} = \frac{1 + L_s^P}{L_s^P}$$

A similar relationship between $\ln \bar{W}_U^P$, the log wages of the uneducated workers (after spillovers and pay rises), and education, \bar{ed}^P , average education, can also be derived:

$$(13) \quad \frac{\partial \ln \bar{W}_U^P}{\partial \bar{ed}^P} = \frac{\partial \ln \bar{W}_U^P}{\partial \bar{W}_U^P} \frac{\partial \bar{W}_U^P}{\partial L_s^P} \frac{\partial L_s^P}{\partial \bar{ed}^P} = \frac{(1 + L_s^P)^2}{L_s^P}$$

We now present the following additional results, obtained from comparing the previous findings across different labour market models:

Result 1: $L_s^M > L_s^P$

This inequality can be immediately derived from comparing equations (1) and (5). Provided that $\gamma > 0$ (which is true by assumption), more educated workers will be hired under monopsony (case 2) than under competition (case 3). This result holds not only in absolute terms (L_s) but also in relative terms (\bar{ed}). (Moreover, the bigger is γ , the bigger the educated hiring difference across the two market structures.)

The intuition behind Result 1 is straightforward but important. While the firm's marginal benefit of hiring educated workers (L_s) is the same under cases 2 and 3, the marginal cost is greater in case 3. This is because in the latter case the employer has to pay the productivity spillover (γW_u) while in case 2 the unskilled wages are ignored as they are simply sunk costs, in that the wage bill of unskilled workers is not affected by how many educated workers are hired.

Result 2: $\frac{\partial \ln \bar{W}_U^P}{\partial \bar{ed}^P} > \frac{\partial \ln \bar{W}^P}{\partial \bar{ed}^P}$

This can be immediately found from a comparison of (13) and (12). The return to firm-level education, as measurable by a firm-level Mincer equation, is bigger for the subset of the uneducated than for the entire workforce. Moreover, under our assumption that the wages of the educated workers are exogenous and not affected by the number of educated workers in the firm, it can be easily shown that the model predicts no impact of changes in firm-level education on the wages of the educated workers.

$$\text{Result 3: } \frac{\partial \ln \bar{W}^P}{\partial \bar{ed}^P} > \frac{\partial \ln \bar{W}^M}{\partial \bar{ed}^M}$$

A proof of this result is presented in the Appendix.⁶ The result says that average wages are more sensitive to average education in a perfect competition setting than in a monopsonistic labour market.

$$\text{Result 4: } \frac{\partial \left(\frac{\partial \ln \bar{W}^P}{\partial \bar{ed}^P} \right)}{\partial \gamma} > 0$$

The proof of this result is also presented in the Appendix. This inequality means that the sensitivity of average wages to the average education increases with the spillover parameter, in the case of a perfect competition market.

$$\text{Result 5: } \frac{\partial^2 \ln \bar{W}_U^P}{\partial \bar{ed}^{P^2}} > 0$$

This result follows immediately from calculating the partial derivative of (13). This can be interpreted as positive and *increasing* social returns to education, i.e. the wage spillovers to the uneducated in a competitive market increases more than proportionately with the inflow of more educated workers to the firm. Notwithstanding this, a caveat to bear in mind is that while

⁶ The comparison of the two partial derivatives is not immediately obvious as, while the numerator in (12) is much bigger than that of (8), the ranking of the denominators is unclear.

the uneducated benefit increasingly more, there are also increasingly fewer uneducated workers in relative terms.

2.3. Other Comments

The different labour market structures will also lead to different policy implications, given the differences in wage spillovers documented above. In the case of monopsonistic labour markets, employers will be residual claimants of the surpluses generated by the admission of skilled workers. Therefore, since employers will be able to fully internalise the externalities from their hiring choices, the socially optimal level of on-the-job schooling will be achieved. (By social optimum level we mean the level that maximises total output.) There is consequently no need for government intervention, at least from the point of view of society achieving the efficient number of high-skilled hires.

However, if labour markets are competitive or if workers have sufficient bargaining power, employers will no longer be able to benefit from the productivity spillovers that may be generated by the admission of educated workers. This will constrain employers to hire socially suboptimal levels of skilled workers, in which the gap between the optimal and the realised levels will increase with the strength of the productivity spillover. Government intervention may then be important in correcting this inefficiency. Such public policies would make skilled workers cheaper from the point of view of employers and/or include subsidies for employers that hire skilled workers or for the skilled workers themselves.⁷

On another level, it is interesting to notice that some of these implications contrast with those that arise in studies of general training. In particular, it is found in this literature that, in competitive markets, firms should never pay for investment in general training (Becker, 1975). If they do, then workers can leave those firms and move to other firms, where those workers would benefit from higher wages without having incurred the investment costs for their higher

⁷ A parenthetical remark is that this approach implies an additional social cost of (voluntary or involuntary) unemployment, as those individuals out of work will be foregoing opportunities to benefit from on-the-job schooling.

levels of human capital. Conversely, in non-competitive markets, firms may pay at least part of the costs of such investments in general training (Acemoglu and Pischke, 1999b). This will occur if firms take an increasing share of the worker's productivity as the worker's skill level increases: in this case, the subsidisation of general skills can be profitable from the firm's point of view.

However, we have found out that, quite on the contrary, in the case of on-the-job schooling, which is also viewed as a type of general human capital, firms will pay for these skills only in competitive markets. In markets with frictions, on the other hand, it will be the workers that implicitly pay for the skills, as they earn the same wages while their productivity increases due to the spillovers. Without frictions, employers will not be able to keep pay unchanged, as that would imply that workers would move elsewhere, to other firms which would pay then their new productivity levels. This is a reversal result that stems from the non-excludability assumed for on-the-job schooling: if less educated workers had to pay directly to have access to this type of schooling, then we would be back to the outcomes described for investments in general training.

Finally, before we present our empirical analysis, we also wish to compare the externality described here with that of Acemoglu's (1996) influential model. This latter model presents an economy with constant returns to scale and complementarities between human and physical capital. This economy is then shown to generate social increasing returns to human capital from market interactions involving matching frictions between workers and firms. These increasing social returns arise as the return to a worker's human capital will depend *positively* on the investment in human capital made by the remaining workers. The intuition is that the interplay between frictions and complementarities will make firms invest more in physical capital when the workers' investment in human capital increases.

Our model also exhibits some elements of pecuniary externalities, as the external effects emerge from (and are in some cases amplified by) market interactions between firms and workers (the pecuniary aspect). Nevertheless, there is also a technological externality, as the spillover mechanism from educated to less educated workers is at the root of the externality in the first place. On the other hand, the externalities in Acemoglu's (1996) paper can be

characterised as of a strictly pecuniary nature – there is no technologically-imbedded spillover in that model.

There are some additional differences between the two models. In our paper, the externalities stem only from the decisions taken by firms – which are subject to *decreasing* returns to scale – as to how many educated workers to hire in a given period. On the other hand, unlike in Acemoglu (1996), here the spillovers arise only from the educated and benefit only the less educated. This is an important prediction that will be tested empirically in Section 4. One last point is that, although both models underline the importance of labour market frictions, in Acemoglu (1996) they are a key driver of the externalities, while in our paper such frictions are not important in the mechanism generating productivity spillovers. They are however crucial in preventing wage spillovers from arising following any productivity spillovers.

3. Empirical Approach

Based on the theoretical model of on-the-job schooling, our empirical work, introduced in this section, is implemented by aggregating individual-level Mincer (1974) equations to the firm level. Indeed, the model was shown to make clear predictions as to how average, firm-level wages of specific groups of workers or the entire workforce vary with average, firm-level education.

As mentioned in the introduction, it is critical to draw on data that present enough variability over time in educational attainment. This concern about variability motivated our use of data for a country and a period that document large upgrades in the schooling of its workforce: during the 1990s, Portugal experienced a substantial educational catching-up of its population. In our data, presented below, the average years of schooling increased by about 17%, from 5.9 in 1991 to 6.9 in 1999.⁸

⁸ These are very low figures for a European country. They correspond to an average school leaving age of the Portuguese workforce around 12 and 13. Compulsory schooling was only four years of schooling (schooling leave age of 12) until the early 1960's.

This considerable increase is partly related to the rise of the minimum level of schooling from six to nine years that occurred since 1986. Additionally, the legal constraints that prevented the expansion of private universities were lifted in the same period, allowing for a large increase in enrolment in such institutions, as until then the demand for university education clearly exceeded its supply by public institutions. (Pereira and Martins (2001) describe these and other developments of the Portuguese education system in greater detail.)

A second concern mentioned in the introduction is that education variability is exogenous with respect to wage determination. This point is further reinforced in the context of our model, in which the average education level of each firm is an endogenous variable. Moreover, average education is shown to be affected by parameters such as the factor shares α and β and the spillover parameter γ that may vary from firm to firm.

While our consideration of firm fixed effects allows the educational attainment of workers in each firm to be correlated with all time-invariant factors (observed or unobserved) that influence wages at that same firm, our estimates would become inconsistent if any of the variables or parameters mentioned above (including the characteristics of the specific labour market in which the firm is active) varied over time. For instance, a technological shock that changes β will lead to different levels of educated workers and average wages and biased estimates of the role of education.

This is a possibility that may affect the study of Barth (2002), who also looks at firm-level returns to education. Drawing on the longitudinal dimension of his Norwegian worker-level data, but assuming that variation to be exogenous, he finds a significant effect of the establishment average level of education on workers' wages.⁹

Moreover, there are additional sources of endogeneity bias, not outlined by the model but of practical relevance, that can emerge from the interplay between workforce adjustment and time-varying shocks that also affect earnings. For instance, firms experiencing an increasing

⁹ See also Battu et al (2003), who find significantly positive effects in a cross-section study of British establishments, proxying firm average education from the distribution of workers across different occupations.

demand for their products may hire younger and more educated workers and, simultaneously, due to rent sharing, increase the earnings of both stayers and/or entrants above the market benchmark. This would lead to positive spurious correlation between education and wages and thus bias upward the education coefficient in a firm-level wage equation.

Alternatively, firms facing negative demand shocks may simultaneously foster the early retirement of their senior workers and demand pay concessions from their remaining employees. In this case, the education coefficient would be downward biased. Moreover, measurement error may also bias downward that coefficient, as it typically attenuates the estimate towards zero, particularly in panel data models (Griliches and Hausman, 1985).

Given these different and conflicting possibilities, we sought to derive consistent estimates using instrumental variables. The first such variable is the firm's lagged education level. As firms keep unchanged a large share of their workforce in each two subsequent periods, we expect there will be a significantly strong correlation between present and lagged education. However, lagged education is unlikely to have a direct role in current wages, as we control for current education and firm fixed effects.

Our second instrument is the lagged share of workers that are of retiring age in each firm-year. The intuition here is as follows: As workers reach that age, they will sooner or later leave the firm (retirement is, in general, not compulsory in Portugal), typically being replaced by younger and, by force of the above-described expansion of the education system, more educated workers. As for the case of lagged education, it is difficult to think why this lagged share of retirement-age workers should affect directly the current level of firm wages.

Moreover, there is additional exogenous variability in this respect over the period covered, as the retirement thresholds varied over the decade in Portugal for different types of cohorts. A law issued in 1991 determined that retirement age would be adjusted gradually over the decade for women, as until then it had stayed at 62, while it was 65 for men. It was decided that the women's retirement age should converge gradually to the men's level, increasing by six months every year since 1993 until reaching men's retirement age of 65 in 1998. Our instrument takes this legislative change into account.

Before concluding this subsection, we wish to highlight a possible concern with the method pursued in the paper: it may be that the new workers of higher education hired by firms have different unobservable characteristics than those of workers hired on previous occasions. In this case, differences in earnings over time could be attributable to such differences in unobservable characteristics among stayers, leavers and entrants and our estimate of the social return could be biased, as it would capture unobservable factors potentially correlated with education.

However, there is some indirect supporting evidence that this case is not likely, particularly for the medium to large firms considered here. Firstly, these firms (unlike smaller ones) typically set up expensive human resource departments that engage in long and meticulous recruitment processes, targeting and assessing worker characteristics that are unobservables for the labour econometrician. Secondly, since only good matches, from the firm and worker points of view, are likely to be stable, new hires should be comparable to their senior colleagues. Moreover, Barth and Dale-Olsen (2003) find corroborative evidence of “assortative matching” in Norwegian firms, in terms of a positive matching along observed and unobserved productivity characteristics between workers of different educational groups in their firms. Finally, since the education system has been expanding, the ability levels across the new educated cohorts are likely to have fallen. This would suggest that our firm-level returns to education would fall, precisely the opposite result of what we document below.

3.1. Data

As to our data, we use a large matched employer-employee panel, “Quadros de Pessoal” [Personnel Records], which covers the universe of Portuguese firms with at least one employee. This data source is based on a compulsory survey administered by Portugal’s Department of Employment. A large set of variables, concerning both firm and worker characteristics, is collected, including identifiers for each firm and each worker. These identifiers allow for both firms and workers to be followed over time. Moreover, as the survey

is also to be used for inspection purposes, checking firm's compliance with different aspects of labour law, particular care is placed on its reliability.

Silva (2003) draws on this same data to study *county-level* social returns to education, adopting an empirical approach based on displaced workers that move to different regions. Unlike in the present paper, he generally finds very small externalities. However, his estimates may be affected by measurement error, as county-level education variables obtainable from the “Quadros de Pessoal” data set cannot include important categories of workers, such as the self-employed or public servants, not to mention individuals outside the labour force. In any case, consistently with the results in this paper, Silva (2003) documents positive and significant results for his controls for average firm education (3rd row, Table 5, page 45), a finding which further strengthens our own results.

In a first step, the analysis in this paper draws on a representative sample of 80% of the manufacturing sector (and all their workers) for each year between 1991 and 1999. Given that we want to focus on firms that are likely to have hiring policies as consistent as possible over time, we use in our analysis only those firms that are “large” enough, defined here a size of at least 50 workers. Moreover, since we need to examine each firm in several periods, we chose to select only those firms that are present in our data in at least four out of the nine years available.

As we also want to minimise measurement error, we dropped firms-year in which more than 20% of workers have missing or incorrect information in the variables required in the wage equation. This procedure leaves us with 4,830 firms and 27,994 firms-year (more than 90% of the original number of firms-year), representing more than 5.9 million workers-year (and, on average, about 213 workers per firm-year).

The descriptive statistics, presented in Table 1, indicate an average schooling attainment across all firms and years of 6.5 years and an average hourly wage of 3.75 euros per hour (1999 prices). Consistently with our assumption about educational expansion, we found in separate calculations that the educational attainment at each firm increases on average by about 2.2% over two contiguous periods.

4. Results

Given the previous discussions, we consider the following wage equation:

$$(14) \quad y_{it} = \beta_1 \text{educ}_{it} + X_{it}'\beta_2 + \alpha_i + \tau_t + \varepsilon_{it}$$

Here y_{it} is the logarithm of average real hourly earnings of firm i in period t . educ_{it} is the average schooling years of the workers of firm i in period t (the empirical counterpart of the variable \overline{ed} of the theoretical model). X_{it} is a set of average characteristics of those workers and their firm in that period: a quadratic of average experience and average tenure, the share of female workers, and size (log number of workers). α_i is the firm fixed effect, τ_t the year dummy, and ε_{it} denotes the error term.

Table 2 presents the results. For the benefit of generality, we also consider pooled OLS and random effects specifications. In these two cases, which assume orthogonality between schooling and the error term, we find large estimates of returns to education, at .18 and .14, respectively.

In the fixed effects specification, the estimated return falls considerably, but is still statistically significant and economically relevant, at .05 (5%). Moreover, the Hausman test strongly rejects the null that the difference in the random and fixed effects coefficients is not systematic (the p -value is less than .0005), thus favouring the fixed effects specification. However, this fixed effect return is also below most of the equivalent estimates obtained in OLS analysis of private returns to education, suggesting that, at best, there are no spillovers.

However, as discussed in Section 4, there are several reasons for the variability in education not to be exogenous. Therefore, we now consider the instruments in the fixed-effects model. The results, presented in Table 3, support their validity. Firstly, both coefficients for the instruments in the auxiliary regression are statistically significant and positive (the sign

expected from our discussion before). Secondly, the tests of instruments quality (see Bound et al, 1995) are also passed.¹⁰ In the main equation, we find that the education coefficient almost triples with respect to the previous results, increasing from .05 to .133 (14.2%), while it is still precisely determined (p-value of .019). Moreover, the over-identification test is not rejected, with a test statistic of 1.3 (p-value of .25). This is a reassuring result although it has to be taken into account bearing in mind that over-identification tests typically have low power.¹¹

These results are also encouraging in that our estimate of a firm-level social return of 14.2% comfortably exceeds most international OLS estimates of private returns, even those for Portugal, a country which typically ranks at the top of such distribution.¹² This gap between firm- and individual-level returns supports the idea that private returns are not irrelevant from the social point of view, as in signalling models, and that there is a considerable additional spillover effect on top of the private return. Moreover, since we have reasons to believe that rent sharing is an important feature of the Portuguese labour market (Martins, 2004), this higher firm-level estimate is precisely the result expected given the model's implications.

On the other hand, and we mentioned in the introduction, evidence of a stronger relationship between average schooling and average earnings than between individual schooling and individual earnings does not necessarily, in general, imply positive spillovers. At least in the context of regions, imperfect substitution between educated and uneducated workers may also induce such result. In this case, of wider units of analysis, educational expansion may increase

¹⁰ The coefficient for lagged schooling in the auxiliary regression is .08 (with a p-value less than .0005) and the coefficient for the lagged share of workers of retirement age is 1.29 (p-value less than .0005). Moreover, the partial R^2 is reasonably large, at 0.013, and the F-statistic strongly rejects the null that the instruments are jointly equal to zero.

¹¹ We have also considered different retirement-age thresholds (more than 60 or 63 years old) since early retirement applies in some cases. Our results remained largely unchanged. However, as expected, the strength of the instrument becomes weaker as we move farther from the 65 level. These findings are available upon request.

¹² Pereira and Martins (2001) estimate an OLS private return of between 8% and 11% over the 1991-98 period. Other studies include Vieira (1999), who follows the strategy of Harmon and Walker (1995) and documents IV estimates lower than the OLS ones, at around 5%, and Modesto (2003), who examines the self-selection involved in progressing or not from compulsory education and finds marginal returns at that stage not greater than 10%. See also Martins and Pereira (2004), who present OLS and quantile regression results for comparable datasets covering sixteen developed countries.

the earnings of uneducated workers not because they become more productive but just because they become scarcer.

Nonetheless, as we explained before, we believe it is unlikely that this general-equilibrium effect will be relevant in firms, unlike in cities or regions, for instance. Indeed, the model assumes that wages for workers of different skills will be constant from the firms' point of view, as firms as small units cannot affect prices – wages may only change for the less skilled to the extent they benefit from on-the-job schooling spillovers.

We now test these assumptions by contrasting how does pay for the more and less educated evolve as a function of average education in their firm. Our empirical approach is as follows: First, in order to make our estimates of the firm-level return to education less affected by the impact of entrants, we consider only workers that have been in the firm for at least 36 months.¹³ We then consider two different thresholds between “educated” and “uneducated” workers, which are, for each firm-year, the mean and median levels of schooling. After that, we select workers whose education exceeds or is below each threshold and aggregate their characteristics (schooling, experience, etc) for each firm-year, after which we run similar regressions as before. This approach should result in cleaner estimates of the spillovers, as we focus on the impact of average education (determined by both stayers and entrants) on educated and less educated stayers separately.¹⁴

Our empirical model is an extended version of (1), now including the characteristics of each subset of workers (educated and less educated) plus the previously-used control for the average schooling across all workers:

$$(15) \quad y_{ijt} = \beta_1 educ_{it} + \beta_2 educ_{ijt} + X_{ijt}'\beta_3 + \alpha_i + \tau_t + \varepsilon_{it}$$

¹³ 36 months is the general time threshold at which employment contracts have either to become permanent or terminated. Workers with levels of tenure of 36 months or more will thus have a stronger degree of bargaining power, allowing them to benefit from any spillover that may occur, unlike those workers with temporary contracts. Using a subsample of the present data set and the same tenure threshold, Martins (2004) finds in his study of rent sharing that high-tenure workers benefit almost twice from rents their low-tenure colleagues.

¹⁴ It should be recalled that the subgroup of educated stayers has no immediate theoretical counterpart in the model of Section 2.

y_{ijt} is the logarithm of the average earnings of workers of type j (educated or uneducated) in firm i in period t (that are also present in the firm in $t-1$). As before, $educ_{it}$ is the average level of schooling years of the workers of firm i in period t , regardless of whether they were also present in the firm in the previous period. X_{ijt} refer to the same set of average characteristics of the workers of type j in firm i in period t . The remaining variables have the same interpretation as before.

The results are presented in Table 4. With respect to the first-stage equations, we find little differences in the role of the instruments across the two sub-groups (educated and uneducated workers) and across the two education thresholds (mean and median).¹⁵ More interestingly, we find for the main equation that the impact of firm average schooling is much greater for the uneducated stayers than for their educated counterparts. For instance, taking the mean-education threshold, an increase in firm average education of one year significantly increases uneducated workers wages by 0.024. The equivalent increase for educated workers is only 0.008 and not significant. Moreover, the same pattern is obtained for the median threshold, with a wage increases for the uneducated workers of 0.033 and an insignificant wage increase for the educated workers.

On the other hand, this pattern is reversed if we look at the impact of each groups' own schooling. While this impact is not significant for the uneducated workers, the return is significant and ranges between 0.075 and 0.077 for educated stayers.¹⁶ Overall, these results are consistent with the model and, in particular, with the existence of spillovers for the less educated, as these uneducated workers benefit from the schooling of their co-workers, while the educated workers do not.¹⁷

¹⁵ One exception is that the average schooling of the uneducated workers plays a greater role in explaining total average schooling than the average schooling of the educated workers. This is due to the large skewness of the distribution of schooling within firms.

¹⁶ It should be mentioned that instruments for retirement shares are not significant (and in some cases have negative signs). The over-identification test is, however, passed in all specifications.

¹⁷ As a further proof of robustness, a previous version of the paper, focusing on a smaller subset of firms with at least 100 workers, found the same qualitative results, both for all workers jointly and for different education groups separately. The quantitative results were, however, stronger. This may suggest that the "spilloverability" of education is positively affected by firm size.

4.1. Extensions

As a first extension, we replicate our previous analysis of equation (14) but considering now information aggregated at different job levels within each firm, rather than at the firm level, as before. Our motivation is that the model would predict a stronger spillover effect in this case, as there is greater scope for spillovers between educated and less educated workers within a job level, rather than across all job levels: Education is likely to be more excludable across than within job levels.

The “Quadros de Pessoal” data includes information for eight standard job levels, ranging between apprentices and low-skilled blue-collar occupations to high- and top-level management. This range of job levels has to be considered by all firms that submit their information to the Department of Employment. The descriptive statistics for the new resulting data are presented in Table 5. (We considered only the seven job levels above apprenticeships, a level that presents considerable measurement error and/or coding errors.) Notice the large increase in the number of observations, from 27,994 firms-year (Table 1) to 177,662 job-levels-firm-year. Notice also the increase in (unweighted) average education, as the thinner job levels (with fewer workers) typically include more educated individuals.

We then regress log average wages in each job-level/firm/year cell on the mean characteristics of that cell, considering also cell fixed effects and instrumenting education in a similar way than before. The results are presented in Table 6 and indicate a significant and precisely estimated return to average education of 0.209 (23%). This finding is consistent with our expectations under the framework of the on-the-job schooling model since it is considerably larger than our estimate for the firm-level analysis.

Other reassuring results are that the over-identification is passed (p-value of 0.45) and the F-statistic of the instruments is very large. However, the indicators of instrument quality are not

as good as before: the coefficient of retirement shares is not significant and the partial R^2 statistic is relatively low.¹⁸

Finally, as a last extension, we sought to provide a more directly interpretable measure of the economic impact of on-the-job schooling and its spillovers, as measured in the paper, by computing some simple, back-of-the-envelope estimates of how much that type of schooling affects wages. For this exercise we consider a spillover effect of 7%, conservatively halfway between the 14% derived in our first estimation and the 23% obtained for job-level cells, minus 10%, which is approximately the modal estimate for OLS private returns to education for Portugal (see references in a footnote above).

We then borrow from the Lester ranges methodology, as discussed in the rent-sharing literature, and work out the percentage wage gain of an hypothetical worker that moved from a firm at the 10th percentile of the distribution of average firm-level education (4.5 years of schooling) to a different firm at the 90th percentile of the same distribution (9.3 years of schooling). The two firms would have precisely the same average characteristics, except for education. The resulting wage gain would be 34%, a figure that can be regarded as considerably large.

What implications can a figure of this size have? It may have some importance for research that seeks to understand the increasing levels of within wage inequality observed in the U.S. and in the U.K. over the 1980's and, at least, the early 1990's (Katz and Autor, 1999). For instance, a process of increasing education dispersion within firms (which may or not correspond to the case of those countries) would, according to the model, increase within wage inequality, as workers in firms that bring in more educated workers would see their wages increase unlike workers in firms whose workforce's average educational attainment stays stable.

¹⁸ One explanation for these latter findings is that average education at each job-level cell is subject to job upgrading processes which are little affected by retirement-related forces. Moreover, measurement error is likely to be more acute within job-levels taken separately than together in firms, for instance because firms may change their coding practices (as to how to allocate each given worker to a job level). Promotions may also affect the strength of the retiring instrument.

On-the-job schooling can also be interpreted in the context of the emerging literature on education and wage risk (see Pereira and Martins, 2002, Carneiro et al., 2003, and Hogan and Walker, 2003, for some recent contributions). Our model and findings suggest that education is related to wage uncertainty not only in the sense that investments in one's education can be more or less profitable but also because one's wages will be affected by the education of the colleagues one is paired with.

5. Conclusions

We contribute to the literature on social returns to education by putting forward a model of on-the-job schooling and some empirical tests. The model's hypotheses are that the schooling of the educated exhibits some positive degree of non-excludability, that the consequent schooling spillovers that occur inside firms present some irreversibility and, finally, that such schooling is sufficiently general to be of use in other firms. This model is shown to lead to a stronger relationship between wages and education in non-monopsonistic labour markets, the strength of this relationship depending on the size of the education spillovers.

The empirical results were then based on the estimation of Mincer firm-level wage equations applied to a Portuguese matched employer-employee panel. Endogeneity concerns, clarified by the model, led us to control for firm fixed effects and instrument firm average education with its lagged values and the lagged share of retirement-age workers.

Consistently with the predictions of the model, we found firm-level returns to education much above their individual-level counterparts. We also find evidence of significant wage spillovers to less-educated workers: their pay increases by 2% to 3% per extra year of education of workers in their firm. However, the subset of educated workers does not seem to benefit from such spillovers, a result which is again suggested by the model, which assumes that only the less educated benefit from the higher levels of education of the new hires in a firm. The education spillover is also found to be stronger when examining job levels within firms, a more disaggregate level of analysis, which allows for stronger interactions among workers.

Taken as a whole, our evidence indicates that education has a significant external effect on wages, at least within firms. This implies that education has social returns to education that exceed private returns. In other words, the results suggest that there is a multiplier effect in education, as its benefits fall not only on the individuals that invest in their human capital at school but also on the workers that have not made such an investment but then go on to interact with educated colleagues at firms.

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Appendix

A. Proof of Result 3

It was derived before that $\overline{ed}_P < \overline{ed}_M$. This implies that $\overline{ed}_P W_S + (1 - \overline{ed}_P) W_U < \overline{ed}_M W_S + (1 - \overline{ed}_M) W_U$ and that

$W_1 = \frac{W_S + W_U}{\overline{ed}_P W_S + (1 - \overline{ed}_P) W_U} > \frac{W_S + W_U}{\overline{ed}_M W_S + (1 - \overline{ed}_M) W_U} = W_0$. Moreover, in an expression of the

type $Y = \frac{a + bX}{c + dX}$, one has that $\frac{dY}{dX} = \frac{bc - da}{(c + dX)^2}$. Taking $W_2 = \frac{W_S + W_U X}{\overline{ed}_P W_S + (1 - \overline{ed}_P) W_U X}$, the

derivative becomes $\frac{dW_2}{dX} = \frac{W_U W_S (2\overline{ed}_P - 1)}{(\overline{ed}_P W_S + (1 - \overline{ed}_P) W_U X)^2}$. This derivative is positive if $(2\overline{ed}_P -$

$1) > 0$, which is the case if $\overline{ed}_P > 0.5$. The last inequality will hold only if $L_S > 1$, which is true by assumption, implying that $W_2 > W_1$. Since $W_1 > W_0$ (8) and since $W_2 = (12)$ when $X = \gamma L_S$, then it must be the case that $(12) > (8)$. Q.E.D.

B. Proof of Result 4

The sign of the partial derivative of (12), $\frac{\partial \ln \bar{W}^P}{\partial e d^P}$, with respect to γ is

$$\begin{aligned} \text{sign} \left[\frac{\frac{\partial L^P}{\partial \gamma} L^P - \frac{\partial L^P}{\partial \gamma} (1 + L^P)}{L^{P^2}} \right] &= -\text{sign} \left(\frac{\partial L^P}{\partial \gamma} \right). \quad \text{This is then equal to} \\ & -\text{sign} \left[\frac{W_U (W_S + \gamma W_U)^{\frac{1}{\alpha+\beta-1}-1} \frac{1}{\alpha+\beta-1} [\lambda^\alpha (\alpha+\beta)]^{\frac{1}{1-(\alpha+\beta)}}}{[\gamma^\alpha (\alpha+\beta)]^{\frac{2}{\alpha+\beta-1}}} \right] + \\ & + \text{sign} \left[\frac{-\frac{\alpha}{1-(\alpha+\beta)} \gamma^{\frac{\alpha}{1-(\alpha+\beta)}-1} (W_S + \gamma W_U)^{\frac{1}{\alpha+\beta-1}} (\alpha+\beta)^{\frac{1}{1-(\alpha+\beta)}}}{[\gamma^\alpha (\alpha+\beta)]^{\frac{2}{\alpha+\beta-1}}} \right] > 0 \quad \text{Q.E.D.} \end{aligned}$$

Tables

Table 1 - Descriptive Statistics

Variable	Obs	Mean	Std. Dev.	Min	Max
Hourly earnings	27,994	3.75	2.47	0.92	87.16
Log Hourly Earnings	27,994	1.08	0.47	-0.11	4.41
Education	27,994	6.45	2.00	0.00	16.65
Experience	27,994	23.38	6.05	3.72	43.65
Experience²	27,994	712.95	310.96	23.25	1942.44
Tenure	27,994	108.18	63.65	0.00	357.16
Tenure²	27,994	231.36	218.91	0.00	1402.50
Female	27,994	0.42	0.30	0.00	1.00
Firm Size	27,994	213.41	615.13	50	29433
Age	27,994	35.83	5.72	18.51	53.70
Share Retirement Age	27,994	0.01	0.02	0.00	0.21
1991	27,994	0.10	0.29		
1992	27,994	0.10	0.30		
1993	27,994	0.11	0.31		
1994	27,994	0.11	0.32		
1995	27,994	0.12	0.33		
1996	27,994	0.12	0.32		
1997	27,994	0.12	0.33		
1998	27,994	0.12	0.32		
1999	27,994	0.10	0.31		
Lagged Education	23,164	6.38	1.98	0.00	16.65
Lagged Share Retirement	23,164	0.01	0.02	0.00	0.21

Table 2 - Results

	Pooled OLS		Random Effects		Fixed Effects	
	Coeff.	St. Error	Coeff.	St. Error	Coeff.	St. Error
Schooling	0.178**	0.002	0.135**	0.003	0.050**	0.003
Experience	0,072**	0.004	0.066**	0.002	0.035**	0.003
Experience²	-0.001**	0.000	-0.001**	0.000	-0.001**	0.000
Female	-0.348**	0.009	-0.318**	0.009	-0.159**	0.021
Log Size	0.054**	0.004	0.047**	0.003	-0.003	0.005
Adj. R ²	0.7836					
Firms-year	27994		27994		27994	

Notes:

All regressions include a quadratic on tenure and year dummies.

The Hausman test about the difference between the random and fixed effects models is strongly rejected.

Notes:

* - significant at the 5% level

** - significant at the 1% level

Table 3 - Results, Fixed Effects and Instruments

	Coeff.	St. Error
<i>First Stage</i>		
Lagged Schooling	0.082**	0.005
Share of 65 and over	1.289**	0.246
Adjusted R ²	0.5186	
Partial R ²	0.0131	
F-statistic	153.14	(P-value=0,000)
<i>Main Equation</i>		
Schooling	0.133**	0.019
Experience	0.066**	0.008
Experience ²	-0.001**	0.000
Female	-0.128**	0.017
Log Size	0.021**	0.007
Within R ²	0.5302	
Between R ²	0.2502	
Overall R ²	0.2529	
Overid. Test Statistic	1.307	(P-value=0.253)
Observations	23164	

Notes:

* - significant at the 5% level

** - significant at the 1% level

Table 4 - Results, Different Sub-Samples

	Mean Education		Median Education	
	Coeff.	St. Error	Coeff.	St. Error
Uneducated workers (stayers)				
<i>First Stage</i>				
Group Average Schooling	0.544**	0.011	0.596**	0.010
Lagged Total Average Schooling	0.168**	0.006	0.152**	0.006
Share of 65 and over	-0.170	0.279	-0.144	0.270
<i>Main Equation</i>				
Total Average Schooling	0.024**	0.011	0.033**	0.012
Group Average Schooling	-0.008	0.007	0.006	0.008
Overall R ²	0.3928		0.5062	
Observations	22841		22883	
Groups	4824		4828	
Educated workers (stayers)				
<i>First Stage</i>				
Group Average Schooling	0.021**	0.005	0.013**	0.005
Lagged Total Average Schooling	0.179**	0.006	0.188**	0.006
Share of 65 and over	-0.020	0.296	0.004	0.294
<i>Main Equation</i>				
Total Average Schooling	0.008	0.012	-0.003	0.012
Group Average Schooling	0.075**	0.002	0.077**	0.002
Overall R ²	0.7315		0.6504	
Observations	22771		22578	
Groups	4820		4804	

Notes:

Both equations consider the same additional variables as in Table 2, although now they refer to each specific subset of workers (educated and uneducated), and not to the entire firm.

For each period, only workers present in the firm in the current and previous period ("stayers") are considered, except in the total average schooling variable.

* - significant at the 5% level

** - significant at the 1% level

Table 5 - Descriptive Statistics (Firm/Job-levels)

Variable	Cells	Mean	Std. Dev.	Min	Max
Hourly earnings	177,662	4.49	3.86	0.51	198.08
Log Hourly Earnings	177,662	1.24	-0.61	0.67	5.29
Education	177,662	7.43	3.36	0.00	17.00
Experience	177,662	23.82	10.30	0.00	76.00
Experience2	177,662	751.84	542.17	0.00	5776.00
Tenure	177,662	112.62	89.58	0.00	758.00
Tenure2	177,662	254.57	340.69	0.00	5745.64
Female	177,662	0.37	0.37	0.00	1.00
Firm Size	177,662	231.76	647.31	50	29433
Age	177,662	37.25	9.59	14.00	87.00
Share Retirement Age	177,662	0.02	0.08	0.00	1.00
1991	177,662	0.09	0.29		
1992	177,662	0.10	0.30		
1993	177,662	0.10	0.31		
1994	177,662	0.11	0.31		
1995	177,662	0.12	0.33		
1996	177,662	0.12	0.33		
1997	177,662	0.12	0.33		
1998	177,662	0.12	0.32		
1999	177,662	0.11	0.31		
Lagged Education	142,176	7.30	3.31	0.00	17.00
Lagged Share Retirement	142,176	0.02	0.08	0.00	1.00

Table 6 - Results, Fixed Effects and Instruments (Job Levels)

	Coeff.	St. Error
<i>First Stage</i>		
Lagged Schooling	0.015**	0.002
Share of 65 and over	-0.062	0.052
Adjusted R ²	0.3503	
Partial R ²	0.0003	
F-statistic	23.94	(P-value= 0,000)
<i>Main Equation</i>		
Schooling	0.209**	0.037
Experience	0.069**	0.010
Experience ²	-0.001**	0.000
Female	-0.180**	0.010
Log Size	0.041**	0.005
Within R ²		
Between R ²	0.6853	
Overall R ²	0.6515	
Overid. Test Statistic	0.58	(P-value= 0.446)
Observations	142,176	

Notes:

Both equations consider the same additional variables, as in Table 2.

* - significant at the 5% level

** - significant at the 1% level