Labor Economics, 14.661. Lectures 1 and 2: Labor Market Externalities

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Introduction

• General interest over the recent two decades or so on the modeling and estimation of various aspects of labor market externalities.

• Two different aspects of externalities:
  1. Externalities in (local) labor markets due to production, matching or other market interactions.
  2. Externalities in social environments, including schools, friendships, neighborhoods, and networks as well as their implications for social mobility, inequality.

• Both types of externalities may be important in practice and have major welfare consequences.

• Both types of externalities present a range of challenges in estimation.

• These two lectures on labor market externalities, and then we will turn to social mobility and peer effects.
What happens to my wages/earnings if other workers in the same labor market/economy increase their schooling by one year?

The Becker-Mincer-Ben Porath framework silent on this (it answers the question of how much my earnings increased when I increase my schooling by one year).

Useful to distinguish between non-pecuniary externalities (where the effect of others’ education on me as “technological”) vs. pecuniary externalities (where the effect of others’ education on me works through changes in equilibrium prices).
What Is an Externality?

- Imagine your coworkers’ human capital makes you more productive
  - e.g., academics would like to be together with other high-quality academics
- Imagine your production function is
  \[ y = f(h, \bar{H}) \]
  where \( \bar{H} \) is the average human capital of your coworkers.
- This is a *technological* spillover of productivity.
- Is it an externality?
What Is an Externality? (continued)

- Not necessarily.
- If all of the spillover is within the firm, the firm will internalize it in its hiring decisions and in its compensation of different workers with different amounts of human capital.
- In that case, there is a technological spillover, but no labor market externality.
- Externalities require
  1. either that productivity spillovers are beyond firm boundaries
  2. or that firms are unable to compensate workers appropriately for their contribution to their coworkers’ productivity (why would this be the case?)
- Let us now focus on externalities.
Nonpecuniary Externalities

- Nonpecuniary externalities ≈ technological spillovers of productivity that are not internalized by prices.
- Canonical example due to Jane Jacobs *The Economy of Cities*: managers from different companies exchange ideas.
- Very popular in economics (e.g., Lucas’s famous 1986 endogenous growth model)
- What other contexts would this be important in?
Nonpecuniary Positive Externalities

- A simple model of nonpecuniary externalities:
- Suppose that the output (or marginal product) of a worker, $i$, is
  \[ y_i = Ah_i^\nu, \]
  where $h_i$ is the human capital (schooling) of the worker, and $A$ is aggregate productivity.
- Assume that labor markets are competitive. So individual earnings are $W_i = Ah_i^\nu$.
- Key idea: the exchange of ideas among workers raises productivity.
- This can be modeled by allowing $A$ to depend on aggregate human capital. In particular, suppose that
  \[ A = BH^\delta \equiv E[h_i]^\delta, \]  
  where $H$ is a measure of aggregate human capital, $E$ is the expectation operator, $B$ is a constant, and $\delta > 0$ (the case where $\delta < 0$ would have negative externalities).
Individual earnings can then be written as

\[ W_i = A h_i^\nu = BH^\delta h_i^\nu. \]

Taking logs:

\[ \ln W_i = \ln B + \delta \ln H + \nu \ln h_i. \]  \hfill (2)

If external effects are stronger within a geographical area, as seems likely in a world where human interaction and the exchange of ideas are the main forces behind the externalities, then equation (2) should be estimated using measures of \( H \) at the local level.

This is a theory of non-pecuniary externalities, since the external returns arise from the technological nature of equation (1).

Nonpecuniary externalities unattractive for a number of reasons:

1. Very reduced form.
2. Do we really expect workers in chemical factories to have a direct productivity effect on retail workers?
Nonpecuniary Positive Externalities (continued)

- Instead, more compelling (at least from one perspective) sources of spillovers:
  - Interactions in the labor market mediated by prices, but externalities might still be at present; **pecuniary externalities**.
  - Interactions in the product market; when computer users become more productive, they can supply cheaper computers to retail companies, again pecuniary externalities.
  - Interactions via R&D and innovation; the semiconductor or the combustion engine have increased the productivity of many workers in many different sectors of the economy.

- The last one may or may not be a pecuniary externality.

- However, except those working in the labor market, the remaining externalities would be economy-wide (sometimes even world-wide), thus difficult to estimate with cross-sectional or panel data variation.

- Thus, let us focus on labor market interactions.
Pecuniary Externalities

- Do pecuniary externalities matter?
- Not in Arrow-Debreu.
- Why not?
- Could they matter in other environments?
- The answer is “perpaps yes”—if we are away from the complete markets benchmark.
Pecuniary Positive Externalities

- First suggested in Alfred Marshall’s *Principles of Economics* in the context of benefits of geographic concentration of industry.

- A complementary story with labor market imperfections, innovation investment by firms and training by workers developed in Acemoglu (1997).
  - Firms find it profitable to invest in new technologies only when there is a sufficient supply of trained workers to replace employees who quit.

- This is a *nonpecuniary* externality, since it is not built in in the form of technological spillovers, but works through market interactions and results from the fact that prices at which labor is transacted is not equal to its marginal product.

- A related model developed in Acemoglu (1996). Here is simplified version of this model.
A Search Model of Pecuniary Externalities

- Consider an economy lasting two periods, with production only in the second period, and a continuum of workers normalized to 1.
- Take human capital of each worker $i$, $h_i$, as given.
- A continuum of risk-neutral firms.
- In period 1, firms make an irreversible investment decision, $k$, at cost $Rk$.
- Workers and firms come together in the second period.
- The labor market is not competitive; instead, firms and workers are matched randomly, and each firm meets a worker.
- The only decision workers and firms make after matching is whether to produce together or not to produce at all (since there are no further periods).
If firm $f$ and worker $i$ produce together, their output is

$$k_i^\alpha h_i^\nu,$$

where $\alpha < 1$, $\nu \leq 1 - \alpha$.

Since it is costly for the worker-firm pair to separate and find new partners in this economy, employment relationships generate quasi-rents.

Wages will therefore be determined by rent-sharing. Here, simply assume that the worker receives a share $\beta$ of this output as a result of bargaining, while the firm receives the remaining $1 - \beta$ share (a simplified version of Nash bargaining).

An equilibrium in this economy is a set of schooling choices for workers and a set of physical capital investments for firms.
Firm $f$ maximizes the following expected profit function:

$$(1 - \beta) k_f^\alpha \mathbb{E}[h_i^\nu] - Rk_f,$$  \hspace{1cm} (4)

with respect to $k_f$.

Since firms do not know which worker they will be matched with, their expected profit is an average of profits from different skill levels.

The function (4) is strictly concave, so all firms choose the same level of capital investment, $k_f = k$, given by

$$k = \left( \frac{(1 - \beta)\alpha H}{R} \right)^{1/(1-\alpha)},$$ \hspace{1cm} (5)

where

$$H \equiv \mathbb{E}[h_i^\nu]$$

is the measure of aggregate human capital.
Now the equilibrium is straightforward to characterize.

Substituting (5) into (3), and using the fact that wages are equal to a fraction $\beta$ of output, the wage income of individual $i$ is given by

$$W_i = \beta \left((1 - \beta)^\alpha H\right)^{\alpha/(1 - \alpha)} R^{-\alpha/(1 - \alpha)} (h_i)^\nu.$$ 

Taking logs, this is:

$$\ln W_i = c + \frac{\alpha}{1 - \alpha} \ln H + \nu \ln h_i,$$

(6)

where $c$ is a constant and $\alpha/(1 - \alpha)$ and $\nu$ are positive coefficients.

The presence of $\ln H$ on the right hand side corresponds to positive pecuniary externalities (in the local labor market).

What about education choices? This actually is a trickier issue and I will return to it later.
Pecuniary Externalities (continued)

- For now the important thing is that human capital externalities arise here because firms choose their physical capital in anticipation of the average human capital of the workers they will employ in the future.
- Since physical and human capital are complements in this setup, a more educated labor force encourages greater investment in physical capital and to higher wages.
- In the absence of the need for search and matching, firms would immediately hire workers with skills appropriate to their investments, and there would be no human capital externalities.
- Nonpecuniary and pecuniary theories of human capital externalities lead to similar empirical relationships since equation (6) is identical to equation (2), with \( c = \ln B \) and \( \delta = \alpha / (1 - \alpha) \).
- Again presuming that these interactions exist in local labor markets, we can estimate a version of (2) using differences in schooling across labor markets (cities, states, or even countries).
The above models focused on positive externalities to education. In contrast, in a world where education plays a “ranking” role, we might also expect negative externalities.

Suppose that there are different types of jobs with different quality/attributes and higher-quality jobs pay higher wages (this may be due to bargaining reasons).

Suppose also that higher education workers get the higher quality jobs (a sort of assignment rule).

What will happen if all of the other workers in my labor market become more educated?
Consider now signaling. Take the most extreme world in which education is only a signal—it does not have any productive role.

What happens if all of the other workers in my labor market become more educated?

Perhaps the same thing as ranking?

To answer this question, let us consider a simple signaling model (and the evidence for signaling in the labor market.
Baseline Signaling Model

- Consider the following simple model to illustrate the issues.
- There are two types of workers, high ability and low ability.
- The fraction of high ability workers in the population is $\lambda$.
- Workers know their own ability, but employers do not observe this directly. They only observe schooling.
- High ability workers always produce $y_H$, low ability workers produce $y_L$. (No productive role of schooling for simplicity now).
Baseline Signaling Model (continued)

- Workers can invest in education, $e \in \{0, 1\}$.
- The cost of obtaining education is $c_H$ for high ability workers and $c_L$ for low ability workers.
- **Crucial assumption** ("single crossing")
  \[ c_L > c_H \]
- That is, education is more costly for low ability workers. This is often referred to as the "single-crossing" assumption, since it makes sure that in the space of education and wages, the indifference curves of high and low types intersect only once. For future reference, I denote the decision to obtain education by $e = 1$.
- To start with, suppose that education does not increase the productivity of either type of worker.
- Once workers obtain their education, there is competition among a large number of risk-neutral firms, so workers will be paid their *expected productivity*.
Game of incomplete information $\rightarrow$ Perfect Bayesian Equilibrium

Two (extreme) types of equilibria in this game.

1. *Separating*, where high and low ability workers choose different levels of schooling.

2. *Pooling*, where high and low ability workers choose the same level of education.
Separating Equilibrium

- Suppose that we have

\[ y_H - c_H > y_L > y_H - c_L \]  

(7)

- This is clearly possible since \( c_H < c_L \).

- Then the following is an equilibrium: all high ability workers obtain education, and all low ability workers choose no education.

- Wages (conditional on education) are:

\[ w(e = 1) = y_H \] and \( w(e = 0) = y_L \)

- Notice that these wages are conditioned on education, and not directly on ability, since ability is not observed by employers.
Separating Equilibrium (continued)

- Let us now check that all parties are playing best responses.
- Given the strategies of workers, a worker with education has productivity $y_H$ while a worker with no education has productivity $y_L$. So no firm can change its behavior and increase its profits.
- What about workers?
- If a high ability worker deviates to no education, he will obtain $w(e = 0) = y_L$, but
  $$w(e = 1) - c_H = y_H - c_H > y_L.$$
If a low ability worker deviates to obtaining education, the market will perceive him as a high ability worker, and pay him the higher wage $w(e = 1) = y_H$. But from (7), we have that

$$y_H - c_L < y_L.$$

Therefore, we have indeed an equilibrium.

In this equilibrium, education is valued simply because it is a signal about ability.

Is “single crossing important”? 
Pooling Equilibrium

- The separating equilibrium is not the only one.
- Consider the following allocation: both low and high ability workers do not obtain education, and the wage structure is

\[ w(e = 1) = (1 - \lambda) y_L + \lambda y_H \]  
and  
\[ w(e = 0) = (1 - \lambda) y_L + \lambda y_H \]

(Does this wage structure make sense?)
- Again no incentive to deviate by either workers or firms.
- Is this Perfect Bayesian Equilibrium reasonable?
Provided that
\[ y_H - c_H > (1 - \lambda) y_L + \lambda y_H, \] (8)

The answer is no.

- This equilibrium is being supported by the belief that the worker who gets education is no better than a worker who doesn’t.
- But education is more costly for low ability workers, so they should be less likely to deviate to obtaining education.
- This can be ruled out by various different refinements of equilibria.
Pooling Equilibrium (continued)

- The underlying idea: if there exists a type who will never benefit from taking a particular deviation, then the uninformed parties (here the firms) should deduce that this deviation is very unlikely to come from this type.
- This falls within the category of “forward induction” where rather than solving the game simply backwards, we think about what type of inferences will others derive from a deviation.
Pooling Equilibrium (continued)

- Take the pooling equilibrium above. Consider a deviation to $e = 1$.
- There is no circumstance under which the low type would benefit from this deviation, since
  
  $$y_L > y_H - c_L,$$

  and the low ability worker is now getting
  
  $$(1 - \lambda) y_L + \lambda y_H.$$

- Therefore, firms can deduce that the deviation to $e = 1$ must be coming from the high type, and offer him a wage of $y_H$.
- Then (8) ensures that this deviation is profitable for the high types, breaking the pooling equilibrium.
Pooling Equilibrium (continued)

- The reason why this refinement is called The Intuitive Criterion is that it can be supported by a relatively intuitive “speech” by the deviator along the following lines:

  you have to deduce that I must be the high type deviating to $e = 1$, since low types would never ever consider such a deviation, whereas I would find it profitable if I could convince you that I am indeed the high type). Of course, this is only very loose, since such speeches are not part of the game, but it gives the basic idea.

- The overall conclusion: separating equilibria, where education is a valuable signal, may be more likely than pooling equilibria.
- When would this not be the case?
Generalizations

- Suppose education is continuous \( e \in [0, \infty) \).
- Cost functions for the high and low types are \( c_H(e) \) and \( c_L(e) \), which are both strictly increasing and convex, with \( c_H(0) = c_L(0) = 0 \).
- The single crossing property is that
  \[
  c'_H(e) < c'_L(e) \quad \text{for all } e \in [0, \infty),
  \]
  that is, the marginal cost of investing in a given unit of education is always higher for the low type (why is this the right condition?).
- Suppose that the output of the two types as a function of their educations are \( y_H(e) \) and \( y_L(e) \), with
  \[
  y_H(e) > y_L(e) \quad \text{for all } e.
  \]
Generalizations (continued)

- The single crossing property:

\[ C_L(e), C_H(e) \]

\[ C_L(e) \]

\[ C_H(e) \]
Generalizations (continued)

- Again there are many Perfect Bayesian Equilibria, some separating, some pooling and some semi-separating.
- But applying a stronger form of the Intuitive Criterion reasoning, we will pick the *Riley equilibrium* of this game, which is a particular separating equilibrium.

*Riley equilibrium*: first find the most preferred (*first-best*) education level for the low type in the perfect information case

\[
y'_L(e^*_i) = c'_L(e^*_i). 
\]
Generalizations (continued)

- First best diagrammatically:

\[
\begin{align*}
U_H &= w(e) - c_H(e) \\
U_L &= w(e) - c_L(e) \\
U_H^* &= y_H e - C_H(e) \\
U_L^* &= y_L e - C_L(e)
\end{align*}
\]
Then we can write the incentive compatibility constraint for the low type, such that when the market expects low types to obtain education $e_l$, the low type does not try to mimic the high type.

$$y_L(e_l^*) - c_L(e_l^*) \geq w(e) - c_L(e) \text{ for all } e,$$ (9)

Let $e_h$ be the level of education for high type such that this constraint holds as an equality:

$$y_L(e_l^*) - c_L(e_l^*) = y_H(e_h) - c_L(e_h).$$

Question: why did we write $w(e_h) = y_H(e_h)$?

Then the Riley equilibrium is such that low types choose $e_l$ and obtain the wage

$$w(e_l^*) = y_L(e_l^*),$$

and high types choose $e_h$ and obtain the wage

$$w(e_h) = y_H(e_h).$$
Generalizations (continued)

- *Riley equilibrium* diagrammatically:

\[ U_H = w(e) - c_H(e) \]
\[ U_L = w(e) - c_L(e) \]
Generalizations (continued)

Why are high types happy to do this? From the single-crossing property:

\[ y_H(e_h) - c_H(e_h) = y_H(e_h) - c_L(e_h) - (c_H(e_h) - c_L(e_h)) \]
\[ > y_H(e_h) - c_L(e_h) - (c_H(e_l^*) - c_L(e_l^*)) \]
\[ = y_L(e_l^*) - c_L(e_l^*) - (c_H(e_l^*) - c_L(e_l^*)) \]
\[ = y_L(e_l^*) - c_H(e_l^*), \]

High ability workers investing in schooling more than they would have done in the perfect information case, in the sense that \( e_h \) characterized here is greater than the education level that high ability individuals chosen with perfect information, given by

\[ y'_H(e_h^*) = c'_H(e_h^*). \]
Evidence on Labor Market Signaling

For different types of evidence:

1. Do degrees matter?
2. Do compulsory schooling laws affect schooling levels for higher grades?
3. Returns to GED?
4. Investigation of negative externalities

Why are these informative about signaling?

Which ones are more convincing?
Third approach: Tyler, Murnane and Willett.

Passing grades in the Graduate Equivalent Degree (GED) differ by state.

So an individual with the same grade in the GED exam will get a GED in one state, but not in another.

If the score in the exam is an unbiased measure of human capital, and there is no signaling, these two individuals should get the same wages.

If the GED is a signal, and employers do not know where the individual took the GED exam, these two individuals should get different wages.

Using this methodology, the authors estimate that there is a 10-19 percent return to a GED signal.
### Evidence on Labor Market Signaling (continued)

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<th>Experiment 4</th>
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<td>1531**</td>
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<td><strong>Panel B: Minorities</strong></td>
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<td><strong>Difference-in-differences</strong> for minorities</td>
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An interesting result we can see from the above table is that there are no GED returns to minorities. This is also consistent with the signaling view, since it turns out that many minorities prepare for and take the GED exam in prison. Therefore, GED would be not only a positive signal, but also likely a signal that the individual was at some point incarcerated. Hence not a good signal at all.
Now suppose there is signaling, and suppose there are no direct productivity benefits from education.

Contrast two situations: in the first, all individuals have 12 years of schooling and in the second all individuals have 16 years of schooling.

Since education has no productive role, and all individuals have the same level of schooling, in both allocations they will earn exactly the same wage (equal to average productivity).

Therefore, here the increase in aggregate schooling does not translate into aggregate increases in wages.

But in the same world, if one individual obtains more education than the rest, there will be a private return to him, because he would signal that he is of higher ability.

Is this a pecuniary or a nonpecuniary externality?
Now returning to regression models, in a world where signaling or ranking is important, we might also want to estimate an equation of the form (2), but when signaling issues are important, we would expect $\delta$ to be negative.

When others invest more in their education, a given individual’s rank in the distribution declines, hence others are creating a negative externality on this individual via their human capital investment.
Evidence on Labor Market Externalities

- Ordinary Least Squares (OLS) estimation of equations like (2) using city or state-level data yield very significant and positive estimates of $\delta$, indicating substantial positive human capital externalities; e.g., Jim Rauch’s paper in the *Journal of Urban Economics* 1993.
- Similar results later found in many other papers.
- The next table shows it for US states.
### Table 2: OLS Estimates of Private and External Returns to Schooling

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<td><strong>(a) Private Returns</strong></td>
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<tr>
<td>Private return to schooling</td>
<td>0.073</td>
<td>0.068</td>
<td>0.075</td>
<td>0.055</td>
<td>0.069</td>
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<td>(State of residence main effects?)</td>
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<td>Yes</td>
<td>Yes</td>
<td>No</td>
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<td>No</td>
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<td><em>(b) Private and External Returns</em></td>
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<td>0.073</td>
<td>0.068</td>
<td>0.074</td>
<td>0.055</td>
<td>0.068</td>
<td>0.075</td>
<td>0.074</td>
</tr>
<tr>
<td>(State of residence main effects?)</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>No</td>
<td>No</td>
<td>No</td>
<td>No</td>
</tr>
<tr>
<td>External return to schooling</td>
<td>0.073</td>
<td>0.061</td>
<td>0.072</td>
<td>0.136</td>
<td>0.136</td>
<td>0.128</td>
<td>0.160</td>
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<tr>
<td>(State of residence main effects?)</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>No</td>
<td>No</td>
<td>No</td>
<td>No</td>
</tr>
<tr>
<td>N</td>
<td>609,852</td>
<td>626,511</td>
<td>729,695</td>
<td>16,659</td>
<td>72,344</td>
<td>161,029</td>
<td>376,479</td>
</tr>
</tbody>
</table>

Notes: Standard errors corrected for state–year clustering are shown in parentheses. The data are from the Census IPUMS for 1950 through 1990, with 1950–80 and 1980–90.
There are at least two problems with this type OLS estimates.

First problem: high-wage cities or states may attract a large number of high education workers or give strong support to education.

- Rauch uses a cross-section of cities.
- Including city or state fixed affects ameliorates this problem, but does not solve it, since states’ attitudes towards education and the demand for labor may comove. The ideal approach would be to find a source of quasi-exogenous variation in average schooling across labor markets.

Acemoglu and Angrist (2000): exploit exogenous sources of variation due to cross-state differences in compulsory schooling laws. The advantage is that these laws not only affect individual schooling but average schooling in a given area.
Second problem: even if we have an instrument for average schooling in the aggregate, estimates of labor market externalities might be spurious.

In particular, if individual schooling is measured with error (or for some other reason OLS returns to individual schooling are not the causal effect), some of this discrepancy between the OLS returns and the causal return may load on average schooling, even when average schooling is instrumented.

This suggests that we may need to instrument for individual schooling as well (so as to get to the correct return to individual schooling).
A Little Bit of Econometrics

- To elaborate on the second problem, let $Y_{ijt}$ be the log weekly wage, than the estimating equation is

$$Y_{ijt} = X_i' \mu + \delta_j + \delta_t + \gamma_1 \bar{S}_{jt} + \gamma_2 s_i + u_{jt} + \varepsilon_i,$$  \hspace{1cm} (10)

- To illustrate the main issues, ignore time dependence, and consider the population regression of $Y_i$ on $s_i$:

$$Y_{ij} = \mu_0 + \rho_0 s_i + \varepsilon_0 i; \text{ where } \mathbb{E}[\varepsilon_0 i s_i] \equiv 0.$$  \hspace{1cm} (11)

- Next consider the IV population regression using a full set of state dummies. This is equivalent to

$$Y_{ij} = \mu_1 + \rho_1 \bar{S}_j + \varepsilon_1 i; \text{ where } \mathbb{E}[\varepsilon_1 i \bar{S}_j] \equiv 0,$$  \hspace{1cm} (12)

since the projection of individual schooling on a set of state dummies is simply average schooling in each state.
Now consider the estimation of the empirical analogue of equation (2):

\[ Y_{ij} = \mu^* + \pi_0 s_i + \pi_1 \bar{S}_j + \zeta_i; \quad \text{where } \mathbb{E}[\zeta_i s_i] = \mathbb{E}[\zeta_i \bar{S}_j] \equiv 0. \quad (13) \]

Then, we have

\[
\begin{align*}
\pi_0 &= \rho_1 + \phi(\rho_0 - \rho_1) \\
\pi_1 &= \phi(\rho_1 - \rho_0)
\end{align*}
\]

where

\[ \phi = 1 / 1 - R^2 > 1, \]

and \( R^2 \) is the first-stage R-squared for the 2SLS estimates in (12).

Therefore, when \( \rho_1 > \rho_0 \), for example because there is measurement error in individual schooling, we may find positive external returns even when there are none.
A Little Bit of Econometrics (continued)

- What can be done?
- Instrument for both individual and average schooling, we would solve this problem.
- But what type of instrument?
- Consider the relationship of interest:

\[ Y_{ij} = \mu + \gamma_1 S_j + \gamma_2 s_i + u_j + \varepsilon_i, \]  
(15)

which could be estimated by OLS or instrumental variables, to obtain an estimate of \( \gamma_1 \) as well as an average estimate of \( \gamma_{2i} \), say \( \gamma_2^* \).

- An alternative way of expressing this relationship is to adjust for the effect of individual schooling by directly rewriting (15):

\[ Y_{ij} - \gamma_2^* s_i \equiv \tilde{Y}_{ij} \]

\[ = \mu + \gamma_1 S_j + [u_j + \varepsilon_i + (\gamma_{2i} - \gamma_2^*) s_i]. \]  
(16)
In this case, instrumental variables estimate of external returns is equivalent to the Wald formula

\[
\gamma_{1}^{IV} = \frac{\mathbb{E}[\tilde{Y}_{ij}|z_i = 1] - \mathbb{E}[\tilde{Y}_{ij}|z_i = 0]}{\mathbb{E}[\bar{S}_j|z_i = 1] - \mathbb{E}[\bar{S}_j|z_i = 0]}
\]

\[
= \gamma_1 + \left[ \frac{\mathbb{E}[\gamma_{2i}s_i|z_i = 1] - \mathbb{E}[\gamma_{2i}s_i|z_i = 0]}{\mathbb{E}[s_i|z_i = 1] - \mathbb{E}[s_i|z_i = 0]} - \gamma_{2}^* \right]
\]

\[
\times \left[ \frac{\mathbb{E}[s_i|z_i = 1] - \mathbb{E}[s_i|z_i = 0]}{\mathbb{E}[\bar{S}_j|z_i = 1] - \mathbb{E}[\bar{S}_j|z_i = 0]} \right].
\]
This shows that to obtain consistent estimates of external returns to schooling we should set

$$\gamma_2^* = \frac{\mathbb{E}[\gamma_2 i s_i | z_i = 1] - \mathbb{E}[\gamma_2 i s_i | z_i = 0]}{\mathbb{E}[s_i | z_i = 1] - \mathbb{E}[s_i | z_i = 0]}$$

$$= \frac{\mathbb{E}[(Y_{ij} - \gamma_1 S_j) | z_i = 1] - \mathbb{E}[(Y_{ij} - \gamma_1 S_j) | z_i = 0]}{\mathbb{E}[s_i | z_i = 1] - \mathbb{E}[s_i | z_i = 0]}$$ (17)

This is typically not the OLS estimator of the private return, and we should be using some instrument to simultaneously estimate the private return to schooling. The ideal instrument would be one affecting exactly the same people as the compulsory schooling laws.
A Feasible Strategy?

- Quarter of birth instruments might come close to this.
- Since quarter of birth instruments are likely to affect the same people as compulsory schooling laws, adjusting with the quarter of birth estimate, or using quarter of birth dummies as instrument for individual schooling, is the right strategy.
- So the strategy is to estimate an equation similar to (2) or (13) using compulsory schooling laws for average schooling and quarter of birth dummies for individual schooling.
Evidence on Child Labor Laws

Figure 2 CDF DIFFERENCE BY SEVERITY OF CHILD LABOR LAWS

(1 - CDF) difference

0.07
0.06
0.05
0.04
0.03
0.02
0.01
0.00
-0.01
-0.02
-0.03

Highest grade completed

Schooling required to work

7 years
8 years
9 or more
Evidence on Compulsory Schooling Laws

Figure 3 CDF DIFFERENCE BY SEVERITY OF COMPULSORY ATTENDANCE LAWS

(1-CDF) difference

Highest grade completed

Required years of attendance

- 9 years
- 10 years
- 11 or more
### Table 7  2SLS Estimates of Private and External Returns

**Instruments, 1960–1980 and Men Aged 40–49**

<table>
<thead>
<tr>
<th>Instrument Set</th>
<th>Individual Schooling Endogenous</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
</tr>
<tr>
<td><strong>Second-Stage Estimates</strong></td>
<td></td>
</tr>
<tr>
<td><strong>Private return to schooling</strong></td>
<td>0.074</td>
</tr>
<tr>
<td></td>
<td>(0.012)</td>
</tr>
<tr>
<td><strong>External return to schooling</strong></td>
<td>0.003</td>
</tr>
<tr>
<td></td>
<td>(0.040)</td>
</tr>
</tbody>
</table>
The estimation results from using this strategy in Acemoglu and Angrist (2000) suggest that there are no significant external returns. The estimates are typically around 1 percent or less, and statistically not different from zero.

They also suggest that in the aggregate signaling considerations are unlikely to be very important (at the very least, they do not dominate positive externalities).

But caveat: the fweaker significantly weaker is a full set of state times linear trends are included.