

MIT Graduate Labor Economics 14.662 Spring 2015

Instructor Notes

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1 MUNDLAK DECOMPOSITION¹

Consider a series of cross sectional regressions of earnings on educational attainment S_i with establishment fixed effects

$$y_{it} = c_t + r_t S_i + \mu_{j(i,t)t} + v_{it}.$$

The $\mu_{j(i,t)t}$ captures any general effect of working in establishing j in time t that is shared by all workers, regardless of education. Estimating instead a population regression that ignores the establishment fixed effect:

$$y_{it} = c'_t + r_t^M S_i + v'_{it},$$

will yield a population regression coefficient

$$r_t^M \equiv \frac{Cov(y_{it}, S_i)}{Var(S_i)} = r_t + \lambda_t b_t,$$

where

$$b_t \equiv \frac{Cov(\bar{S}_{j(i,t)t}, S_i)}{Var(S_i)},$$

is an educational sorting index, and

$$\lambda_t \equiv \frac{Cov(u_{j(i,t)t}, \bar{S}_{j(i,t)t})}{Var(\bar{S}_{j(i,t)t})},$$

which is the coefficient of a regression of the establishment effect on mean schooling at the establishment.

Thus, the standard Mincerian return to schooling in each year r_t^M is composed of three components:

1. the return to schooling conditional on job quality r_t ;
2. the return to mean establishment schooling λ_t ;
3. and the association between worker and establishment schooling b_t .

Note that we can derive this decomposition from the (hopefully familiar) omitted-variables

¹Due to Mundlak, Yair. 1978. "On the Pooling of Time Series and Cross Section Data." *Econometrica*, 46, 69–85.

bias formula (see, e.g., *Mostly Harmless Econometrics*, section 3.2.2):

$$\begin{aligned}
\underbrace{r_t^M}_{\text{"short"}} &= \underbrace{r_t^l}_{\text{"long"}} + \underbrace{1}_{\text{"effect of omitted"}} \cdot \underbrace{\frac{Cov(\mu_{j(i,t)t}, S_i)}{Var(S_i)}}_{\text{"regression of omitted on included"}} \\
&= r_t^l + \frac{Cov(\mu_{j(i,t)t}, \bar{S}_{j(i,t)t} + \tilde{S}_i)}{Var(\bar{S}_{j(i,t)t} + \tilde{S}_i)} \\
&= r_t^l + \frac{Cov(\mu_{j(i,t)t}, \bar{S}_{j(i,t)t} + \tilde{S}_i)}{Var(\bar{S}_{j(i,t)t})} \frac{Var(\bar{S}_{j(i,t)t})}{Var(\bar{S}_{j(i,t)t} + \tilde{S}_i)} \\
&= r_t^l + \frac{Cov(\mu_{j(i,t)t}, \bar{S}_{j(i,t)t})}{Var(\bar{S}_{j(i,t)t})} \frac{Cov(\bar{S}_{j(i,t)t}, S_i)}{Var(S_i)}
\end{aligned}$$

Here we've defined $\tilde{S}_i \equiv S_i - \bar{S}_{j(i,t)t}$ as the "within-establishment" variation in S_i (i.e. residual from regressing S_i on establishment effects). By construction $Cov(\mu_{j(i,t)t}, \tilde{S}_i) = 0$ and

$$\begin{aligned}
Cov(\bar{S}_{j(i,t)t}, S_i) &= Cov(\bar{S}_{j(i,t)t}, \bar{S}_{j(i,t)t} + \tilde{S}_i) \\
&= Var(\bar{S}_{j(i,t)t})
\end{aligned}$$

We can use the same OVB logic to note that

$$\begin{aligned}
r_t^M &= r_t + \lambda_t b_t \\
&= r_t + \lambda_t \frac{Cov(\bar{S}_{j(i,t)t}, S_i)}{Var(S_i)}
\end{aligned}$$

relates the "short" regression coefficient to the "long" regression of

$$y_{it} = c_t + r_t S_i + \lambda_t \bar{S}_{j(i,t)t} + \omega_{it}.$$

We can thus estimate λ_t by running this regression. Estimating the short and long regression coefficients r_t^M and r_t^l would then allow us to solve for b .

Figure 13 from CHK (NBER Working Paper #18522, November 2012) shows that a large chunk of the apparent rise in the return to schooling is due to the rising 'return' to mean establishment schooling, which effectively triples from 0.04 to 0.11 between 1985 to 2009.

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