

The Price of Experience

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Motivation

Observation: Market return to experience highly correlated with average labor market experience.

Question: To what extent do demographic trends (supply-side changes) drive changes in the market price of experience?

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Correlations:

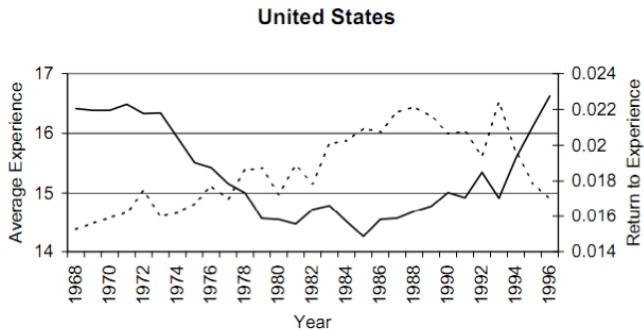
- ▶ U.S. PSID, 1968-1996: $\rho = -0.75$
- ▶ Swedish SAF, 1975-1995: $\rho = -0.97$
- ▶ Danish data, 1980-2003: $\rho = -0.80$

Market return to experience measured as β_t from Mincer wage regression in repeated cross-sections:

$$\log(w_{it}) = \beta_t \text{Experience}_{it} + \gamma_t C_{it} + \varepsilon,$$

C_{it} a vector of control variables.

Motivation, cont.



Approach

- ▶ Assume workers are endowed with two distinct and complimentary production inputs: labor and experience.
- ▶ Efficiency units of labor and experience are only inputs to production.
- ▶ Competitive market pricing: factor prices determined by marginal products.
- ▶ Relative price of experience determined by relative stock of experience to labor.
- ▶ CRS production technology: output, effective labor, and effective experience across workers consistently aggregate.
- ▶ Estimate parameters of production technology, efficiency experience/labor schedules from individual wage equation.

Evaluation

- ▶ Do model estimated parameters support hypothesis of complementarity between experience and labor?
- ▶ If so, how does “experience premium” derived from the aggregate production technology compare with that estimated from the data?

Workers

- ▶ Time is discrete
- ▶ Worker is endowed with one unit of labor each year.
- ▶ Worker of age j accumulates experience e for each year she works more than threshold \underline{h} :

$$e = \sum_{k=0}^j I(h_k > \underline{h})$$

- ▶ Effective factor inputs of agent determined by “life-efficiency schedules” that depend on age (j) and education (e)
- ▶ Labor input: $\lambda_L(j, s)$
- ▶ Experience input: $\lambda_E(j, s) \cdot e$

Workers, cont.

- ▶ Let R_{E_t} and R_{L_t} denote the market prices of experience and labor.
- ▶ Individual productivity term z_{it}
- ▶ Earnings:

$$\begin{aligned}y_{it} &= [R_{L_t} \lambda_L(j_{it}, s_{it}) + R_{E_t} \lambda_E(j_{it}, s_{it}) e_{it}] h_{it} z_{it} \\ &= R_{L_t}(j_{it}, s_{it}) \left[1 + \frac{R_{E_t}}{R_{L_t}} \frac{\lambda_E(j_{it}, s_{it})}{\lambda_L(j_{it}, s_{it})} e_{it} \right] h_{it} z_{it}\end{aligned}$$

- ▶ Experience premium: $\Pi_{E_t} \equiv \frac{R_{E_t}}{R_{L_t}}$.
- ▶ Relative life-cycle efficiency schedule of experience:

$$\lambda_{E/L}(j_{it}, s_{it}) \equiv \frac{\lambda_E(j_{it}, s_{it})}{\lambda_L(j_{it}, s_{it})}$$

Technology

- ▶ CRS production technology mapping aggregate stock of labor L_t and aggregate stock of experience E_t into aggregate earnings Y_t :

$$Y_t = A_t G(L_t, E_t)$$

where A_t represents the aggregate productivity of the composite labor/experience input.

- ▶ Euler's theorem:

$$Y_t = A_t (G_{L_t} L_t + G_{E_t} E_t),$$

where $G_{E_t} = \frac{\partial G}{\partial E_t}$ and $G_{L_t} = \frac{\partial G}{\partial L_t}$

- ▶ Competitive factor prices:

$$R_{L_t} = A_t G_{L_t}$$

$$R_{E_t} = A_t G_{E_t}$$

Aggregation

Sum individual earnings equation over individuals i for a given t :

$$\begin{aligned}\sum_i y_{it} &= R_{L_t} \sum_i \lambda_L(j_{it}, s_{it}) z_{it} h_{it} + R_E \sum_i \lambda_E(j_{it}, s_{it}) z_{it} h_{it} e_{it} \\ &= A_t G_{L_t} L_t + A_t G_{E_t} E_t \\ &= Y_t\end{aligned}$$

where

$$\begin{aligned}L_t &= \sum_i \lambda_L(j_{it}, s_{it}) z_{it} h_{it} \\ E_t &= \sum_i \lambda_E(j_{it}, s_{it}) e_{it} z_{it} h_{it}\end{aligned}$$

Complementarity

- ▶ Assume G is CES:

$$Y_t = A_t (L_t^\mu + \delta E_t^\mu)^{\frac{1}{\mu}}$$

- ▶ $\mu \leq 1 \Rightarrow$: necessary condition for relative factor supply to determine relative factor prices.

Log wage equation

Combine aggregate production function with individual earnings equation:

$$\ln w_{it} = \ln A_t + \ln G_{L_t} + \ln \lambda_L(j_{it}, s_{it}) + \ln [1 + \Pi_{E_t} \lambda_{E/L}(j_{it}, s_{it}) e_{it}] + \ln z_{it}$$

where

$$w_t = \frac{y_{it}}{h_{it}}$$
$$G_{L_t} = \left(1 + \delta \left(\frac{E_t}{L_t}\right)^\mu\right)^{\frac{1}{\mu} - 1}$$
$$\Pi_{E_t} = \delta \left(\frac{E_t}{L_t}\right)^{\mu - 1}$$

Life-cycle efficiency schedules

- ▶ Life-cycle efficiency schedules for labor and experience:

$$\begin{aligned}\lambda_L(j, s) &= \exp(\lambda_{L,1}(s)j + \lambda_{L,2}(s)j^2) \\ \lambda_E(j, s) &= \exp(\lambda_{E,1}(s)j + \lambda_{E,2}(s)j^2)\end{aligned}$$

- ▶ Relative life-cycle efficiency schedule of experience:

$$\lambda_{E/L}(j, s) = \exp(\lambda_{E/L,1}(s)j - \lambda_{E/L,2}(s)j^2)$$

where

$$\begin{aligned}\lambda_{E/L,1}(s) &= \lambda_{E,1}(s) - \lambda_{L,1}(s) \\ \lambda_{E/L,2}(s) &= \lambda_{E,2}(s) - \lambda_{L,2}(s)\end{aligned}$$

- ▶ Individual productivity:

$$\ln z_{it} = \alpha_t \chi_{it}$$

where χ_{it} is a vector of observables.

Fully parametrized log wage equation

Substitute efficiency schedules into log wage equation:

$$\begin{aligned} \ln w_{it} = & \ln A_t + \ln G_{L_t} + (\lambda_{L,1}(s_{it})j_{it} + \lambda_{L,2}(s_{it})j_{it}^2) \\ & + \ln \left[1 + \delta \left(\frac{E_t}{L_t} \right)^{\mu-1} \exp(\lambda_{E/L,1}(s_{it})j_{it} + \lambda_{E/L,2}(s_{it})j_{it}^2) e_{it} \right] \\ & + \alpha_t \chi_{it} + \varepsilon_{it} \end{aligned}$$

Identification of aggregate parameters

- ▶ Given the measurement of E_t and L_t , variation of experience premium Π_{E_t} in relation to relative factor endowment $\frac{E_t}{L_t}$.
- ▶ Correlation between Π_{E_t} and $\frac{E_t}{L_t}$ identifies μ .
- ▶ Average magnitude of Π_{E_t} relative to $\frac{E_t}{L_t}$ identifies δ , subject to scale factor.
- ▶ Normalize $\lambda_E(0, s) = \lambda_L(0, s) = 1$.

Two-step estimation

First stage:

Estimate life-cycle efficiency parameters and time-varying coefficients ($\hat{\alpha}_t$, \hat{D}_t , and $\hat{\Pi}_{E_t}$) from the following equation:

$$\begin{aligned}\ln w_{it} &= D_t + (\lambda_{L,1}(s_{it})j_{it} + \lambda_{L,2}(s_{it})j_{it}^2) \\ &\quad + \ln [1 + \Pi_E \exp(\lambda_{E/L,1}(s_{it})j_{it} + \lambda_{E/L,2}(s_{it})j_{it}^2) e_{it}] \\ &\quad + \alpha_t \chi_{it} + \varepsilon_{it}\end{aligned}$$

where $D_t \equiv \ln A_t + \ln G_{L_t}$ and $\Pi_{E_t} \equiv \delta \left(\frac{E_t}{L_t}\right)^{\mu-1}$

From first-stage estimates:

- ▶ Obtain $\hat{\lambda}_E(j, s)$ from

$$\hat{\lambda}_E(j, s) = \hat{\lambda}_{E/L}(j, s) \hat{\lambda}_L(j, s)$$

- ▶ Then

$$\hat{L}_t = \sum_i \hat{\lambda}_L(j_{it}, s_{it}) \hat{z}_{it} h_{it}, \quad \hat{E}_t = \sum_i \hat{\lambda}_E(j_{it}, s_{it}) \hat{z}_{it} h_{it} e_{it}$$

Two-stage estimation, cont.

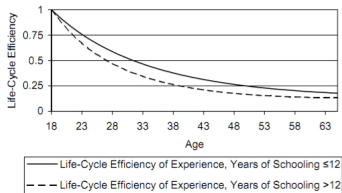
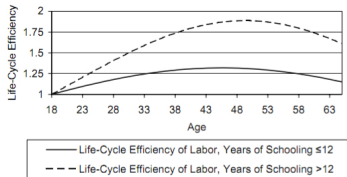
Second stage: Use first-stage estimates to estimate μ and δ from the following equation:

$$\begin{aligned} \ln w_{it} = & \hat{D}_t + \left(\hat{\lambda}_{L,1}(s_{it})j_{it} + \hat{\lambda}_{L,2}(s_{it})j_{it}^2 \right) \\ & + \ln \left[1 + \delta \left(\frac{\hat{E}_t}{\hat{L}_t} \right)^{\mu-1} \exp \left(\hat{\lambda}_{E/L,1}(s_{it})j_{it} + \hat{\lambda}_{E/L,2}(s_{it})j_{it}^2 \right) e_{it} \right] \\ & + \hat{\alpha}_t \chi_{it} + \varepsilon_{it} \end{aligned}$$

Data

- ▶ Benchmark: PSID over 1968-1996 sample period.
- ▶ Use actual experience rather than potential experience: important for separately identifying the returns to experience.
- ▶ Variation in experience premium and experience-labor ratio comes from across time.
- ▶ Robustness: U.S. census
- ▶ Impute actual experience from PSID estimates.
- ▶ Variation in experience premium and experience-labor ratio comes from across states.

Life Cycle Efficiency Schedules



Technology parameters

Table 1: PSID estimates of technology parameters.

| Parameter | Second Stage | | Full Estimation | |
|-----------|--------------|-----------|-----------------|-----------|
| | Estimate | St. Error | Estimate | St. Error |
| μ | -2.352 | 0.061 | -2.367 | 0.175 |
| δ | 6.931 | 0.447 | 7.028 | 1.354 |
| R^2 | 0.926 | | 0.926 | |
| $RMSE$ | 0.596 | | 0.596 | |

Correlation between experience premium $\hat{\Pi}_{E_t}$ and experience-labor ratio \hat{E}_t/\hat{L}_t estimated from first stage: -0.96

Actual vs. Predicted Experience Premium

- ▶ Actual experience premium: $\hat{\Pi}_{E_t}$ estimated in the first stage.
- ▶ Predicted experience premium: Analytic expression for Π_{E_t} evaluated at estimated technology parameters $\hat{\mu}$ and $\hat{\delta}$, and estimated aggregate experience/labor ratio, \hat{E}_t/\hat{L}_t .
- ▶ Correlation is 0.97

