

Educational Rat Race, Positional Externalities, and Intergenerational Mobility

Ernest Liu
Princeton University

Shaoda Wang
Chicago Harris

Hanzhe Zhang
Michigan State University

Washington University in St. Louis
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Educational rat race

- ▶ Competition for high-quality jobs or opportunities can induce parents and students to engage in excessive education.
 - ▶ analogous to a *rat race* in which everyone runs faster just to stay in place.
- ▶ In many societies, a fixed number of top positions (elite college slots, prestigious jobs) drives families to invest heavily in education to improve their children's chances of winning these opportunities.
- ▶ This paper develops a tractable overlapping-generations model to investigate this phenomenon
 - ▶ as well as income inequality and intergenerational mobility.

Positional externalities

- ▶ When skills and jobs are positively matched, the equilibrium features over-investment in education relative to the social optimum.
- ▶ This over-investment is driven by a *positional externality*:
 - ▶ individual families do not account for the fact that improving their child's rank in the skill distribution comes partly at the expense of others.
- ▶ Empirical evidence: We show that when income declines more steeply with college rank in a country, the share of private spending on education tends to increase in that country.

Income inequality and intergenerational mobility

- ▶ Given the optimal policy of parents in each generation, we derive a linear law of motion for (log) skill across generations.
- ▶ The process for log human capital is mean-reverting: children's log skill is an affine function of their parents' log skill.
- ▶ Under mild conditions, the economy converges to a stationary distribution of skills, which is lognormal.
- ▶ We also derive the intergenerational elasticity of income (IGE) in the stationary state.
 - ▶ The IGE is increasing in the strength of human capital transmission (education technology and any direct skill inheritance) but is mitigated by the randomness in job allocations.

Literature: Rat race and positional externalities

- ▶ The paper is related to premarital/pre-matching investments, e.g., Bhaskar and Hopkins (2016, JPE), Zhang (2021, JPE), Bhaskar, Li and Yi (2023, JPE).
 - ▶ recent surveys of investment and matching literature: Hopkins (2022) and Nöldeke and Samuelson (2024).
- ▶ Positional externalities have appeared in the context of status competition, e.g., Hopkins and Kornienko (2004, AER), Becker, Murphy and Werning (2005, JPE), Ray and Robson (2012, Ecma)
 - ▶ empirical analysis: Kim, Tertilt and Yum (2024, AER) quantify educational overinvestment due to status competition in South Korea

Literature: Intergenerational mobility

- ▶ We provide a general-equilibrium model of altruistic parents deciding on educational investments for children who compete in the labor market to derive a stationary law of motion for skill formation that involves aspects of nature and nurture, as well as intergenerational mobility.
 - ▶ Becker and Tomes (1979, JPE), Becker and Tomes (1986, JOLE) and Becker, Kominers, Murphy and Spenkuch (2018, JPE)
 - ▶ Relatedly, Becker and Barro (1988, QJE) and Barro and Becker (1989, Ecma) study fertility and intergenerational mobility in overlapping-generations models.
 - ▶ a stylized version of the skill formation of Cunha and Heckman (2007, AERPP), Cunha, Heckman and Schennach (2010, Ecma), and Heckman and Mosso (2014, ARE)

Roadmap

- ▶ Key economic insights of positional externalities in the educational rat race in the two-period model
- ▶ Empirical evidence
- ▶ Inequality and intergenerational mobility in the tractable overlapping-generations model

Parent's education investment choice

- ▶ Parent i in generation t has income $y_{i,t}$.
- ▶ Parent i chooses to spend $e_{i,t}$ to invest in a child's education.
- ▶ Education translates into a child's skill:

$$\ln s_{i,t+1} = \eta_e \ln e_{i,t} + \varepsilon_{i,t+1}^s,$$

where $\varepsilon_{i,t}^s \sim N(0, \sigma_{\varepsilon^s}^2)$ and $0 < \eta_e < 1$.

- ▶ Later on we will generalize to a fully dynamic problem in which the parent's skill could also influence the child's (and subsequently the future generations') skill:

$$\ln s_{i,t+1} = \eta_e \ln e_{i,t} + \eta_s \ln s_{i,t} + \varepsilon_{i,t+1}^s,$$

Child's labor market matching and income

- Skills are matched noisily with job qualities via the Gaussian copula:

$$\begin{pmatrix} \ln s_{i,t+1} \\ \ln q_{i,t+1} \end{pmatrix} \sim N \left(\begin{pmatrix} \mu_{s,t+1} \\ \mu_q \end{pmatrix}, \begin{pmatrix} \sigma_{s,t+1}^2 & \rho \sigma_{s,t+1} \sigma_q \\ \rho \sigma_{s,t+1} \sigma_q & \sigma_q^2 \end{pmatrix} \right),$$

where $0 \leq \rho \leq 1$.

- The pool of opportunities is fixed, $\ln q_{i,t} \sim N(\mu_q, \sigma_q^2)$, but for any $\rho > 0$, higher skills are more likely to draw higher opportunities.
 - $\rho = 1$: perfectly positive-assortative matching.
 - $\rho = 0$: uniformly random matching.
- A child's income on the job is

$$\ln y_{i,t+1} = (1 - \alpha) \ln s_{i,t+1} + \alpha \ln q_{i,t+1} + k.$$

Microfoundation when $\rho = 1$

- ▶ Cobb-Douglas matching output $O(i,j) = As_i^{1-\alpha}q_j^\alpha$.
- ▶ Perfectly transferable utility.
- ▶ Stable outcome is a pair of payoff function (x,y) such that
 - ▶ $y_i + x_j = O(i,j)$ for matched pairs (i,j) , and
 - ▶ $y_i + x_j \geq O(i,j)$ for all feasible (i,j) .
- ▶ Under lognormal distributions of skills and job qualities,

$$\ln s \sim N(\mu_s, \sigma_s) \text{ and } \ln q \sim N(\mu_q, \sigma_q),$$

- ▶ Matching is perfectly positive assortative: $\Phi_s(s) = \Phi_q(q)$
- ▶ Workers' payoffs are $\ln y_i = (1 - \alpha) \ln s_i + \alpha \ln q_i + \text{constant}$.

Microfoundation when $\rho < 1$

- ▶ Cobb-Douglas matching output $O(i,j) = As_i^{1-\alpha}q_j^\alpha$.
- ▶ Matching is positive assortative based on a latent index

$$\rho \cdot \frac{\ln s_i - \mu_s}{\sigma_s} + \sqrt{1 - \rho^2} \cdot \varepsilon_i, \quad \varepsilon_i \sim N(0, 1).$$

- ▶ Division of surplus is through Nash bargaining

$$y_i = \theta As_i^{1-\alpha}q_j^\alpha,$$

so

$$\ln y_i = \ln(\theta A) + (1 - \alpha) \ln s_i + \alpha \ln q_i.$$

Parent's utility function

- We first analyze a simple problem in which the child consumes the entire income, generating utility $\ln y_{i,t+1}$.
- Each parent decides how much to spend on education, solving

$$u(y_{i,t}) = \max_{e_{i,t}} \{ \ln(y_{i,t} - e_{i,t}) + \delta \mathbb{E} [\ln y_{i,t+1} | e_{i,t}] \}.$$

- In the dynamic model, the child also decides on education investment for the grandchild, so altruistic parents have a dynastic utility function

$$u_t(y_{i,t}, s_{i,t}) = \max_{e_{i,t}} \{ \ln(y_{i,t} - e_{i,t}) + \delta \mathbb{E} [u_{t+1}(y_{i,t+1}, s_{i,t+1}) | e_{i,t}, s_{i,t}] \}.$$

Expected job quality

- Define the percentile ranks of $s_{i,t+1}$ and $q_{i,t+1}$:

$$\Phi_s(s_{i,t+1}) \equiv \frac{\ln s_{i,t+1} - \mu_{s,t+1}}{\sigma_{s,t+1}} \text{ and } \Phi_q(q_{i,t+1}) \equiv \frac{\ln q_{i,t+1} - \mu_q}{\sigma_q}.$$

- The Gaussian copula job matching process implies

$$\Phi_q(q_{i,t+1}) = \rho \cdot \Phi_s(s_{i,t+1}) + \sqrt{1 - \rho^2} \cdot \varepsilon_{i,t+1}^q.$$

- Given $\varepsilon_{i,t+1}^q \sim N(0, 1)$,

$$\mathbb{E} \left[\frac{\ln q_{i,t+1} - \mu_q}{\sigma_q} \middle| s_{i,t+1} \right] = \rho \frac{\ln s_{i,t+1} - \mu_{s,t+1}}{\sigma_{s,t+1}}.$$

- Rearranged,

$$\mathbb{E} [\ln q_{i,t+1} | s_{i,t+1}] = \mu_q + \rho \frac{\sigma_q}{\sigma_{s,t+1}} (\ln s_{i,t+1} - \mu_{s,t+1}).$$

Expected income given skill

- Recall income for a job

$$\ln y_{i,t+1} = (1 - \alpha) \ln s_{i,t+1} + \alpha \ln q_{i,t+1} + k.$$

- Hence, we can write income generation given skill as

$$\begin{aligned} & \mathbb{E}[\ln y_{i,t+1} | s_{i,t+1}] \\ = & (1 - \alpha) \ln s_{i,t+1} + \alpha \left[\mu_q + \rho \frac{\sigma_q}{\sigma_{s,t+1}} (\ln s_{i,t+1} - \mu_{s,t+1}) \right] + k \\ = & \left(1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} \right) \ln s_{i,t+1} + \text{constants} \end{aligned}$$

Expected income given education

- Recall skill given education

$$\ln s_{i,t+1} = \eta_e \ln e_{i,t} + \varepsilon_{i,t+1}^s.$$

- Child's expected income given education $e_{i,t}$ is

$$\mathbb{E}[\ln y_{i,t+1} | e_{i,t}] = \underbrace{\left(1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}}\right)}_{\equiv S_{t+1}} \eta_e \ln e_{i,t} + \text{constants}$$

Parent's education investment decision

- ▶ Parent's investment decision solves

$$\max_{e_{i,t}} \{ \ln(y_{i,t} - e_{i,t}) + \delta \mathbb{E} [\ln y_{i,t+1} | e_{i,t}] \}.$$

- ▶ Plugging in the expected income, we translate the problem to

$$\max_{e_{i,t}} \{ \ln(y_{i,t} - e_{i,t}) + \delta S_{t+1} \eta_e \ln e_{i,t} + \text{constants} \}.$$

- ▶ First-order condition:

$$\frac{1}{y_{i,t} - e_{i,t}} = \delta S_{t+1} \eta_e \frac{1}{e_{i,t}}.$$

- ▶ We get a constant rate of income spent on education:

$$R_t \equiv \frac{e_{i,t}}{y_{i,t}} = \frac{\delta S_{t+1} \eta_e}{1 + \delta S_{t+1} \eta_e}.$$

Investment incentives

- Let's better understand

$$S_{t+1} = 1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}}; S = 1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_s}$$

in

$$R_t \equiv \frac{e_{i,t}}{y_{i,t}} = \frac{\delta S_{t+1} \eta_e}{1 + \delta S_{t+1} \eta_e} = \frac{\delta \left(1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} \right) \eta_e}{1 + \delta \left(1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} \right) \eta_e}.$$

- Note that we can rewrite the maximization problem as

$$\max_{e_{i,t}} \left\{ \underbrace{\ln(y_{i,t} - e_{i,t})}_{\text{consumption}} + \underbrace{\delta(1 - \alpha)\eta_e \ln e_{i,t}}_{\text{child income due to skill}} + \underbrace{\delta \alpha \mathbb{E}[\ln q_{i,t+1} | e_{i,t}]}_{\text{child income due to job quality}} \right\}.$$

Positional incentives

- The sensitivity of expected log job quality with respect to education

$$\frac{d\mathbb{E}[\ln q_{i,t+1}|e_{i,t}]}{d\ln e_{i,t}} = \alpha\rho\frac{\sigma_q}{\sigma_{s,t+1}},$$

can be decomposed into

$$\underbrace{\frac{d\ln s_{i,t+1}}{d\ln e_{i,t}}}_{\eta_e} \underbrace{\frac{d\Phi_s(\ln s_{i,t+1})}{d\ln s_{i,t+1}}}_{\frac{1}{\sigma_{s,t+1}}} \underbrace{\frac{d\Phi_q(\mathbb{E}[\ln q_{i,t+1}|s_{i,t+1}])}{d\Phi_s(\ln s_{i,t+1})}}_{\rho} \underbrace{\frac{d\mathbb{E}[\ln q_{i,t+1}|s_{i,t+1}]}{d\Phi_q(\mathbb{E}[\ln q_{i,t+1}|s_{i,t+1}])}}_{\sigma_q}$$

Investment rate R_t

Higher investment rate

$$R_t \equiv \frac{e_{i,t}}{y_{i,t}} = \frac{\delta S_{t+1} \eta_e}{1 + \delta S_{t+1} \eta_e} = \frac{\delta \left(1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} \right) \eta_e}{1 + \delta \left(1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} \right) \eta_e}$$

when

- ▶ the agent is more patient/altruistic (i.e., higher δ);
- ▶ education translates into more skills (i.e., higher η_e);
- ▶ correlation between skill and opportunity is higher (i.e., higher ρ)
- ▶ more investment translates into bigger gains in the skill quantiles (i.e., lower $\sigma_{s,t+1}$), and/or
- ▶ a higher skill quantile translates into bigger gains in expected job quality (i.e., higher σ_q).
 - ▶ parents invest in education to improve skill because higher skill (i) raises output and (ii) raises the expected opportunity (given a fixed job pool).

Equilibrium over-investment

- Socially optimal education investment is

$$R_t^{\text{opt}} \equiv \frac{e_{i,t}^{\text{opt}}}{y_{i,t}} = \frac{\delta (1 - \alpha) \eta_e}{1 + \delta (1 - \alpha) \eta_e}.$$

- In contrast, privately optimal education investment is

$$R_t \equiv \frac{e_{i,t}}{y_{i,t}} = \frac{\delta \left(1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} \right) \eta_e}{1 + \delta \left(1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} \right) \eta_e}.$$

- For any $\rho > 0$, the wedge is positive, implying an over-investment by parents, due to positive-assortative matching, a positional externality.

- To formally see this, note that a utilitarian planner solves

$$\max_{e_{i,t}} \left\{ \int_i [\ln(y_{i,t} - e_{i,t}) + \delta \mathbb{E}[\ln y_{i,t+1} | e_{i,t}]] \, di \right\}$$

subject to

$$\ln y_{i,t+1} = (1 - \alpha) \ln s_{i,t+1} + \alpha \ln q_{i,t+1},$$

$$\ln s_{i,t+1} = \eta_e \ln e_{i,t} + \varepsilon_{i,t+1}^s,$$

and

$$\ln q_{i,t+1} = \mu_q + \rho \frac{\sigma_q}{\sigma_{s,t+1}} \left(\ln s_{i,t+1} - \int_j \ln s_{j,t+1} \, dj \right) + \sigma_q \sqrt{1 - \rho^2} \varepsilon_{i,t+1}^q,$$

where $\varepsilon_{i,t+1}^q \sim N(0, 1)$. We can combine the three constraints to write

$$\mathbb{E}[\ln y_{i,t+1} | e_{i,t}] = \alpha \mu_q + (1 - \alpha) \eta_e \ln e_{i,t} + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} (\ln s_{i,t+1} - \int_j \ln s_{j,t+1} \, dj)$$

► Hence, the social planner's problem is

$$\max_{e_{i,t}} \int_i [\ln(y_{i,t} - e_{i,t}) + \delta(1 - \alpha)\eta_e \ln e_{i,t}] di + \underbrace{\alpha\rho \frac{\sigma_q}{\sigma_{s,t+1}} \left(\int_i \ln s_{i,t+1} di - \int_j \ln s_{j,t+1} dj \right)}_{=0},$$

which implies a simplified problem of

$$\max_{e_{i,t}} \int_i [\ln(y_{i,t} - e_{i,t}) + \delta(1 - \alpha)\eta_e \ln e_{i,t}] di$$

and a solution of

$$R_t^{\text{opt}} = \frac{e_{i,t}^{\text{opt}}}{y_{i,t}} = \frac{\delta(1 - \alpha)\eta_e}{1 + \delta(1 - \alpha)\eta_e}.$$

Proposition 1 (Over-Investment Rat Race)

The decentralized equilibrium features over-investment in education relative to the social optimum as long as job assignment is positively assortative ($\rho > 0$). In particular, the equilibrium education rate R exceeds the efficient rate R^{opt} , with the wedge

$$R - R^{opt} = \frac{\delta \eta_e \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} (1 - \alpha)}{\left[1 + \delta (1 - \alpha) \eta_e \right] \left[1 + \delta \left(1 - \alpha + \alpha \rho \frac{\sigma_q}{\sigma_{s,t+1}} \right) \eta_e \right]} > 0 \quad \text{for } \rho > 0.$$

Equivalently, parents invest too much in children's education, driven by the private incentive to boost their child's relative standing (the "rat race"), which the social planner would avoid.

Key testable implication

- The education investment rate

$$R \equiv \frac{e_i}{y_i} = \frac{\delta S \eta_e}{1 + \delta S \eta_e}$$

is increasing in skill-income elasticity

$$S = \frac{d\mathbb{E}[\ln y_i | s_i]}{d \ln s_i} = \frac{d\mathbb{E}[\ln y_i | s_i]}{d\Phi_s(\ln s_i)} \underbrace{\frac{d\Phi_s(\ln s_i)}{d \ln s_i}}_{>0}$$

Empirical education investment rate R : Measure 1

► Household education expenditure rate

$$\hat{R} = \frac{\text{Total Private Household Expenditure on Education}}{\text{Gross Domestic Product (GDP)}} \times 100.$$

- in harmonized national household survey data provided by the UNESCO Institute for Statistics (UIS)
- It reflects the macroeconomic weight of education financing borne by households.
- Private household education expenditure includes tuition payments, fees, textbooks, and other direct education-related expenditures reported in national household income and expenditure surveys.

Empirical education investment rate R : Measure 2

► Household secondary education expenditure rate

$$\hat{R} = \frac{\left(\frac{\text{Total Household Expenditure on Secondary Education}}{\text{Total Secondary Enrollment}} \right)}{\text{GDP per capita}} \times 100.$$

- constructed by UNESCO UIS using household survey data aligned with macroeconomic indicators from the World Bank's World Development Indicators (WDI) measures per-student household expenditure in secondary education (ISCED levels 2 and 3).
- allows cross-country comparisons of the relative financial effort required for secondary schooling.
- can be interpreted as the proportion of the average annual income required to finance one student's secondary education, serving as a proxy for the intensity of private investment in human capital.

Revelio Lab LinkedIn Data

- ▶ 12+ million college graduates from 2000 to 2015 worldwide with complete information on education, work, and country
 - ▶ 624+ million LinkedIn profiles in total
 - ▶ 129+ million reported bachelor degree information
- ▶ Revelio Lab estimated a salary based on companies, titles, countries, etc.

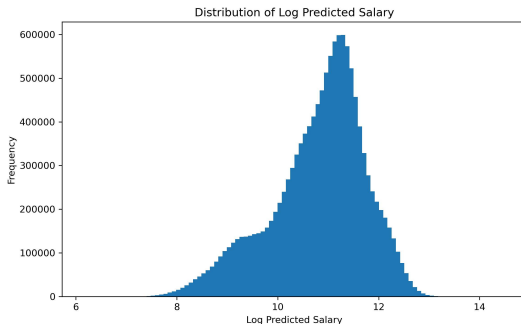
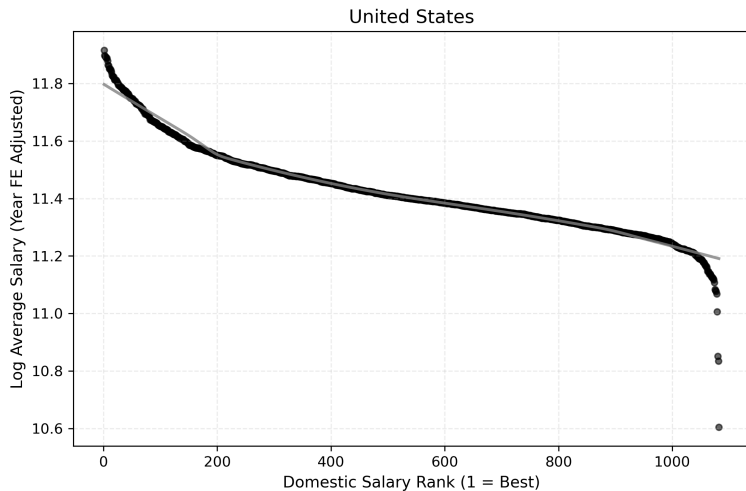


Table: Top 10 Universities by Earnings

Rank	US	China	Korea
1	Univ of Pennsylvania	Tsinghua University	KAIST
2	Duke University	Peking University	Seoul National University
3	Princeton University	University of Science and Technology of China	Yonsei University
4	Stanford University	Central University of Finance and Economics	Korea University
5	Yeshiva University	Beijing University of Posts and Telecommunications	Hongik University
6	Caltech	Zhejiang University	Ajou University
7	Yale University	Beihang University	Sogang University
8	MIT	Xiamen University	Chung-Ang University
9	George Washington Univ	Shanghai Jiao Tong University	Hanyang University
10	Bentley University	Nanjing University	Ewha Womans University

Average log salary by college rank



University rank gradient

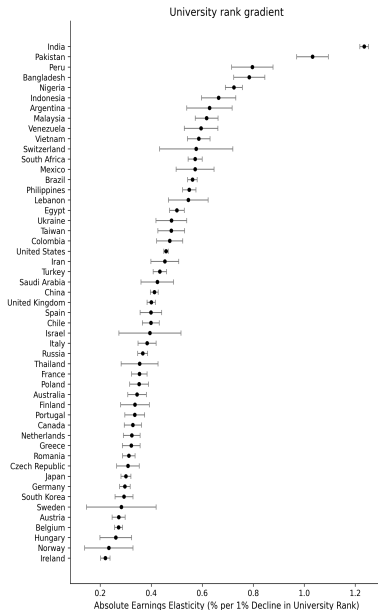
- ▶ Run regression:

$$\overline{\ln y}_{uc} = c + \beta_c \ln R_{uc} + \epsilon_{uc}$$

- ▶ $\overline{\ln y}_{uc}$: average log salary of university u in country c
 - ▶ $\ln R_{uc}$: log rank of the university
 - ▶ $\hat{\beta}_c$: percentile increase in income by percentile increase in rank
- ▶ University rank gradient:

$$|\hat{\beta}_c| = \frac{d\mathbb{E}[\ln y_i | s_i]}{d\Phi_s(\ln s_i)}$$

3. Empirical evidence

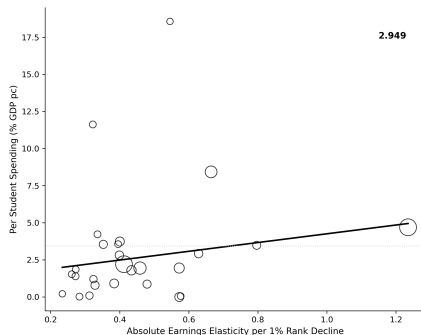
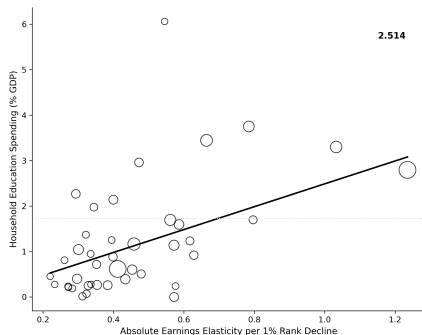


Positive relationships between university rank gradient and education investment rate

	(1)	(2)
	HH edu spend (% GDP)	Per-student spend (% GDP pc)
β_c	2.514*** (0.324)	2.949*** (0.881)
Constant	-0.025 (0.253)	1.310* (0.718)
Observations	41	27
R^2	0.607	0.310

Notes. Country-level weighted least squares (WLS) regressions. The regressor is the country-specific ranking gradient β_c , defined as the absolute earnings elasticity (in percentage points) per 1% decline in university rank. Weights are proportional to population (normalized by mean). Standard errors in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.10$.

Positive relationships between university rank gradient and education investment rate



Robustness checks

The positive relationships are robust to alternative specifications:

- ▶ Alternative years after graduation: 1-year, 5-year, 10-year
- ▶ Alternative subsamples
 - ▶ Postgraduate degrees: including or excluding PhDs
 - ▶ education and work in the same country
- ▶ Alternative measures of β_c
 - ▶ Explanatory power R^2 of a university fixed effects regression.
 - ▶ Sorting power ROC-AUC of university: how well university identity predicts the likelihood of entering high-income jobs

Overlapping generations model

- ▶ Time is discrete, indexed by generation $t = 0, 1, 2, \dots$
- ▶ Each generation t , there is a unit mass of families
 - ▶ indexed by $i \in [0, 1]$.
- ▶ Each family consists of one parent and one child.
- ▶ Each agent lives as a parent for one period, during which they work, earn income, consume, and decide how much to invest in their child's education.
- ▶ The child then becomes the parent of generation $t + 1$.

Parent's education investment choice

- ▶ Parent i has income $y_{i,t}$.
- ▶ Parent i chooses to spend $e_{i,t}$ to invest in a child's education.
- ▶ ~~Education translates into skill:~~

~~$$\ln s_{i,t+1} = \eta_e \ln e_{i,t} + \varepsilon_{i,t+1}^s,$$~~

- ▶ Education and parent's skill translate into child's skill:

$$\ln s_{i,t+1} = \eta_e \ln e_{i,t} + \eta_s \ln s_{i,t} + \varepsilon_{i,t+1}^s,$$

where $\varepsilon_{i,t}^s \sim N(0, \sigma_{\varepsilon^s}^2)$ and $0 < \eta_e < 1$.

Child's labor market matching

- Skills are matched noisily with job qualities via the Gaussian copula:

$$\begin{pmatrix} \ln s_{i,t+1} \\ \ln q_{i,t+1} \end{pmatrix} \sim N \left(\begin{pmatrix} \mu_{s,t+1} \\ \mu_q \end{pmatrix}, \begin{pmatrix} \sigma_{s,t+1}^2 & \rho \sigma_{s,t+1} \sigma_q \\ \rho \sigma_{s,t+1} \sigma_q & \sigma_q^2 \end{pmatrix} \right),$$

where $0 \leq \rho \leq 1$.

- The pool of opportunities is fixed, $\ln q_{i,t} \sim N(\mu_q, \sigma_q^2)$, but for any $\rho > 0$, higher skills are more likely to draw higher opportunities.
- Once a child is matched with a job, income is generated:

$$\ln y_{i,t+1} = (1 - \alpha) \cdot \ln s_{i,t+1} + \alpha \cdot \ln q_{i,t+1}.$$

Parent's utility function

- ▶ We first analyze a simple problem in which the child consumes the entire income, generating utility $\ln y_{i,t+1}$.
- ▶ Each parent decides on how much to spend on education, solving

$$u(y_{i,t}) = \max_{e_{i,t}} \{ \ln(y_{i,t} - e_{i,t}) + \delta \mathbb{E} [\ln y_{i,t+1} | e_{i,t}] \}.$$

- ▶ Altruistic parents have a dynastic utility function

$$u_t(y_{i,t}, s_{i,t}) = \max_{e_{i,t}} \{ \ln(y_{i,t} - e_{i,t}) + \delta \mathbb{E} [u_{t+1}(y_{i,t+1}, s_{i,t+1}) | e_{i,t}, s_{i,t}] \}.$$

Equilibrium utility

- In the two-period model, it is solved that

$$u(y_{i,t}) = \ln((1 - R_t)y_{i,t}) + S_{t+1}\eta_e \ln(R_t y_{i,t}) + \text{constants}.$$

- In the overlapping-generations model, guess and verify:

$$u_t(y_{i,t}, s_{i,t}) = A_t + B_t \ln(y_{i,t}) + C_t \ln(s_{i,t}).$$

Solving for equilibrium investment

- Plug $u_{t+1}(y_{i,t+1}, s_{i,t+1})$ in $u_t(y_{i,t}, s_{i,t})$:

$$\ln(y_{i,t} - e_{i,t}) + \delta \mathbb{E}[A_{t+1} + B_{t+1} \ln(y_{i,t+1}) + C_{t+1} \ln(s_{i,t+1}) | e_{i,t}, s_{i,t}]$$

- Note

$$\mathbb{E}[\ln s_{i,t+1} | e_{i,t}, s_{i,t}] = \eta_e \ln e_{i,t} + \eta_s \ln s_{i,t};$$

$$\mathbb{E}[\ln y_{i,t+1} | e_{i,t}, s_{i,t}] = S_{t+1} \mathbb{E}[\ln s_{i,t+1} | e_{i,t}, s_{i,t}] + \text{constants},$$

- The utility expression becomes

$$\ln(y_{i,t} - e_{i,t}) + \delta[B_{t+1}S_{t+1} + C_{t+1}][\eta_e \ln e_{i,t} + \eta_s \ln s_{i,t}] + \text{constants}$$

Equilibrium investment

- Equilibrium constant education investment rate:

$$R_t \equiv \frac{e_{i,t}}{y_{i,t}} = \frac{\delta [B_{t+1}S_{t+1} + C_{t+1}] \eta_e}{1 + \delta [B_{t+1}S_{t+1} + C_{t+1}] \eta_e}.$$

- We solve for nested expressions of

$$\begin{aligned} A_t &= \ln(1 - R_t) + \delta \left\{ A_{t+1} + \alpha \left[\mu_q - \rho \frac{\sigma_q}{\sigma_{s,t+1}} \mu_{s,t+1} \right] \right\} \\ &\quad + \delta \eta_e [B_{t+1}S_{t+1} + C_{t+1}] \ln R_t \\ B_t &= 1 + \delta \eta_e [B_{t+1}S_{t+1} + C_{t+1}] \\ C_t &= \delta \eta_s [B_{t+1}S_{t+1} + C_{t+1}] \end{aligned}$$

Law of motion for skill and income

► Income dynamics

$$\ln y_{i,t} = S_t \ln s_{i,t} + \alpha \left[\mu_q - \rho \frac{\sigma_q}{\sigma_{s,t}} \mu_{s,t} + \sigma_q \sqrt{1 - \rho^2} \varepsilon_{i,t}^q \right],$$

where $\varepsilon_{i,t}^q \sim N(0, 1)$.

► Skill dynamics

$$\begin{aligned} \ln s_{i,t+1} = & \eta_e \ln R_t + (\eta_e S_t + \eta_s) \ln s_{i,t} \\ & + \alpha \eta_e \left[\mu_q - \rho \frac{\sigma_q}{\sigma_{s,t}} \mu_{s,t} + \sigma_q \sqrt{1 - \rho^2} \varepsilon_{i,t}^q \right] + \varepsilon_{i,t+1}^s. \end{aligned}$$

Stationary skill and income distributions

- Time-invariant skill variance σ_s^2 :

$$\sigma_s^2 = (\eta_e S + \eta_s)^2 \sigma_s^2 + \alpha^2 \eta_e^2 (1 - \rho^2) + \sigma_\epsilon^2.$$

- The dynamic path of skill inequality is

$$\sigma_{s,t+1}^2 = (\eta_e S_t + \eta_s)^2 \sigma_{s,t}^2 + \alpha^2 \eta_e^2 (1 - \rho^2) \sigma_q^2 + \sigma_{\epsilon^s}^2.$$

- Time-invariant income variance:

$$\sigma_y^2 = S^2 \sigma_s^2 + \alpha^2 (1 - \rho^2) \sigma_q^2.$$

- The dynamic path of income inequality is

$$\sigma_{y,t}^2 = S_t^2 \sigma_{s,t}^2 + \alpha^2 (1 - \rho^2) \sigma_q^2.$$

Income variance

$$\sigma_y^2 = \left[\eta_e \left((1 - \alpha) + \alpha \rho \frac{\sigma_q}{\frac{XY + \sqrt{X^2 Y^2 + (1 - X^2)(Y^2 + Z)}}{1 - X^2}} \right) + \eta_s \right]^2 \\ \times \left(\frac{XY + \sqrt{X^2 Y^2 + (1 - X^2)(Y^2 + Z)}}{1 - X^2} \right)^2 + \alpha^2 (1 - \rho^2) \sigma_q^2,$$

where

$$X = \eta_e^2 (1 - \alpha) + \eta_e \eta_s + \eta_s,$$

$$Y = \alpha \eta_e^2 \rho \sigma_q,$$

$$Z = \alpha^2 \eta_e^2 (1 - \rho^2) \sigma_q^2 + \eta_\varepsilon^2 \sigma_\varepsilon^2.$$

Comparative statics of income variance

- ▶ $\alpha \rightarrow \uparrow$
- ▶ $\sigma_q \rightarrow \uparrow$
- ▶ $\eta_e \rightarrow \uparrow$
- ▶ $\eta_s \rightarrow \uparrow$
- ▶ $\eta_\varepsilon, \sigma_\varepsilon \rightarrow \uparrow$
- ▶ $\rho \rightarrow \text{ambiguous}$
 - ▶ Direct effect: Higher ρ decreases $(1 - \rho^2) \rightarrow$ tends to reduce the “match noise” component \rightarrow tends to lower σ_y^2 .
 - ▶ Indirect effect: Higher ρ increases S (through $\alpha \rho \frac{\sigma_q}{\sigma_s}$) \rightarrow increases persistence \rightarrow tends to raise σ_y^2 .

Intergenerational mobility

- Intergenerational elasticity of income (IGE)

$$\iota = \frac{d \ln y_{i,t+1}}{d \ln y_{i,t}}.$$

- For any t ,

$$\iota_t = \frac{S_{t+1}}{S_t} \cdot (\eta_e S_t + \eta_s).$$

- In the stationary equilibrium,

$$\iota = \eta_e S + \eta_s.$$

Conclusion

- ▶ Educational over-investment due to positional externalities
 - ▶ Empirical evidence consistent with key comparative statics
- ▶ A tractable overlapping-generations model with closed-form expressions of skill and income dynamics
 - ▶ Implications for intergenerational mobility

THANK YOU!

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