Educational Investment in Spatial Equilibrium: Evidence from Indonesia

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How does migration shape large-scale educational investment?

- Governments invest \$3 trillion in education annually (World Bank 2022)
 - In Indonesia, 61,807 new primary schools (Sekolah Dasar INPRES program, 1973-1978)
- Educational investment targets students locally
 - But graduates migrate and seek employment nationally

This paper

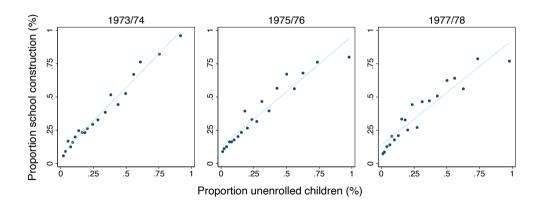
- Aggregate and distributional effects of the INPRES program
 - Difference-in-difference with long-run outcomes (Duflo 2001)
 - Spatial heterogeneity in returns to education + implications for program design
- Spatial equilibrium model of complementary education + migration decisions
 - Returns to education: rural incentives depend on urban wages (if mobile)
 - Regional convergence: rural schools increase urban output (if mobile)
- **Results:** aggregate output \uparrow (8%), inequality \updownarrow (people \downarrow 5%, places \uparrow 12%)
 - Mobility magnifies each effect

Literature

- Education and migration at scale in general equilibrium
 - Education: Khanna 2021, Dinerstein et al. 2022 (no migration)
 - Migration: Dahl 2002, Bryan et al. 2014, Bryan & Morten 2019 (no education)
 - Both: Eckert & Kleineberg 2021, Agostinelli et al. 2022 (no school construction)
- INPRES program evaluation with aggregate effects and counterfactuals
 - Duflo 2001/2004, Martinez-Bravo 2017, Ashraf et al. 2020, Bazzi et al. 2021
- Place-based policy with portable human capital benefits
 - Glaeser & Gottlieb 2008, Kline & Moretti 2014, Busso et al. 2013, Austin et al. 2018

Data and Stylized Facts

The INPRES program built 62,000 new primary school (1973-1978)





Data

- Treatment at district level
 - INPRES school construction (1973-1978)
 - Pre-program primary schools, child populations, enrollment rates
- Long-run outcomes at individual level
 - SUSENAS household surveys (2011-2014)
 - Districts of residence and birth, years of schooling, monthly wages

Difference-in-difference variation in school construction (Duflo 2001)

$$Y_{ijk} = \delta_j + \delta_k + \beta S_j T_k + C_j T_k \phi + \varepsilon_{ijk}$$

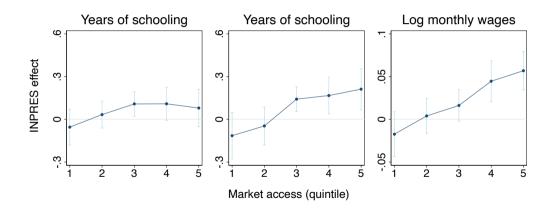
$$Y_{ijk} = \delta_j + \delta_k + X_j S_j T_k \beta + C_j T_k \phi + \varepsilon_{ijk}$$

- Young vs. old students in age cohorts k
 - Young exposed to new schools, but old not
- Many vs. few new schools in origin districts j
 - ullet More schools o bigger difference between young and old cohorts

Long-term education and wage effects

Outcomes	Estimate	SE	Obs
Years of schooling	0.103**	(0.0424)	233,517
 For wage earners 	0.121**	(0.0495)	89,404
Log monthly wages	0.0195**	(0.00916)	89,404

Driven by labor market access





Spatial equilibrium model

- Government constructs schools
 - Build human capital that is portable (aggregate output)
- 2 Individuals invest in education
 - In a district, more schools \rightarrow better access \rightarrow lower costs of education
- Individuals migrate for work
 - Mobility gives rural students access to high urban wages (person-based inequality)
 - But rural students leave after graduation (place-based inequality)

Education and migration frictions

$$U(e,\epsilon) = \alpha_{\ell} \varepsilon_{jk\ell}^{\alpha} [\underbrace{(1-\overbrace{\tau_{j\ell}^{m}}) w_{\ell} h_{jk} \varepsilon_{jk\ell}^{h} e^{\eta} \epsilon}_{\text{net labor income}} - \underbrace{(1+\overbrace{\tau_{jk}^{e}}) c \varepsilon_{jk\ell}^{c} e}_{\text{cost of education}}]$$

- Individual i, origin j, age cohort k, destinations ℓ
 - For ijk, each destination has $e_\ell^*(\epsilon_\ell) = \max_e U_\ell(e,\epsilon_\ell)$ and $U_\ell(\epsilon_\ell) = U_\ell(e_\ell^*,\epsilon_\ell)$
- Compare destinations, then pick best to get choice probabilities

$$\pi_{jk\ell} = rac{ ilde{w}_{jk\ell}^{m{\sigma}}}{\sum_{\hat{\ell}} ilde{w}_{jk\hat{\ell}}^{m{ heta}}} \quad ext{for} \quad ilde{w}_{jk\ell} \equiv lpha_{\ell}^{1-\eta} (1 - au_{j\ell}^{m{m}}) w_{\ell} ilde{arepsilon}_{jk\ell}$$

Choice probabilities, education, and wages

LHS variables observed in data

$$\begin{split} \pi_{jk\ell} &= \tilde{w}^{\theta}_{jk\ell} / \sum_{\hat{\ell}} \tilde{w}^{\theta}_{jk\hat{\ell}} \\ \overline{\text{educ}}_{jk\ell} &= \mathbb{E}\left[e^* \mid \text{individuals choose } \ell\right] \\ \overline{\text{wage}}_{jk\ell} &= \mathbb{E}[w_{\ell}h_{jk}\varepsilon^h_{jk\ell}e^{\eta}\varepsilon \mid \text{individuals choose } \ell, \, e = e^*] \end{split}$$

Education and wages are increasing in labor market access

$$\overline{\mathsf{educ}}_{jk\ell}$$
, $\overline{\mathsf{wage}}_{jk\ell} \propto \left(\sum_{\hat{\ell}} \tilde{w}_{jk\hat{\ell}}^{\theta}\right)^{\frac{1}{\theta(1-\eta)}} \equiv \mathsf{MA}_{jk\ell}$



Human capital function (INPRES as IV)

$$\begin{aligned} \mathsf{wage}_i & \propto \mathsf{hcap}_i = (\mathsf{educ}_i)^{\eta} \\ & \qquad \qquad \Downarrow \\ \\ \mathsf{log}\, \mathsf{wage}_{ijk} & = \delta_j + \delta_k + \eta \, \mathsf{log}\, \mathsf{educ}_{ijk} + \pmb{C}_j T_k \pmb{\phi} + \varepsilon_{ijk} \\ \\ \mathsf{log}\, \mathsf{educ}_{ijk} & = \delta_j + \delta_k + \beta S_j T_k + \pmb{C}_j T_k \pmb{\phi} + \varepsilon_{ijk} \\ & \qquad \qquad \Downarrow \\ \\ \widehat{\eta} & = 0.688^{**} (0.311) \end{aligned}$$

Education and migration costs (INPRES as DD)

Other parameters (INPRES as moments)

$$\sum_{i=1}^{n} [y_i - \exp(x_i \hat{\beta})] x_i = 0$$

- Poisson pseudo-maximum likelihood (Santos Silva & Tenreyro 2006)
 - Common in spatial models to accommodate zeros in choice probabilities

$$\frac{\log \mathsf{educ}_{jk\ell} - \log \mathsf{\overline{wage}}_{jk\ell}}{\Delta_{\ell} \log \mathsf{\overline{educ}}_{jk\ell}, \quad \Delta_{\ell} \log \mathsf{\overline{wage}}_{jk\ell}, \quad \Delta_{\ell} \log \pi_{jk\ell}}$$
 INPRES treatment effects



Quantifying aggregate and distributional effects

- **Evaluate** relative to zero-construction counterfactual
- **Decompose** effects of mobility by mechanism
 - And separate each from the general equilibrium effects
 - Diff-in-diff avoids model but only captures net effects
- Study program design
 - By simulating alternative allocations of school construction

The program increased aggregate output by 8%

	Aggregate output
Zero construction	1.00
+ Direct effect of construction	1.02
+ Migration	1.03
+ Migration-induced schooling	1.07
+ New equilibrium wages	1.08

- Small gains without migration (direct effect) or without education (sorting)
 - Complementarity between education and migration
 - Gains from sorting are already large (Bryan et al. 2014)

With especially large benefits for rural students

	Inequality (people)
Zero construction	1.00
+ Direct effect of construction	0.99
+ Migration	0.98
+ Migration-induced schooling	0.96
+ New equilibrium wages	0.95

- Expanded opportunity for rural students with high marginal returns
 - \bullet Decreased inequality between rural and urban students by 5%

But also increased inequality across places by 12%

	Inequality (places)
Zero construction	1.00
+ Direct effect of construction	0.99
+ Migration	1.02
+ Migration-induced schooling	1.11
+ New equilibrium wages	1.12

- The program explicitly aimed to encourage regional convergence
 - But mobility places convergence in tension with output gains
 - Rural regions still enjoy net gains, but urban regions gain more

Equity-efficiency tradeoff under mobility

	Aggregate	Inequality	Inequality
	output	(people)	(places)
Actual INPRES allocation	1.08	0.95	1.12
Prioritizing rural regions + Halving migration costs	1.09	0.93	1.14
	1.13	0.90	1.18
Prioritizing urban regions + Halving migration costs	1.04	0.97	1.06
	1.08	0.93	1.12

- Rural construction generates large returns but widens rural-urban gap (still Pareto)
- Alternative: schools + roads, although rural out-migration will rise (not Pareto)



Summary

- Evaluating large-scale educational investment in spatial equilibrium
 - Indonesia's INPRES program built 62,000 primary schools in 1970s
 - Aggregate output \uparrow (8%), person-based inequality \downarrow (5%), place-based \uparrow (12%)
- Education and migration are complementary
 - Big gains for rural students who leave rural regions behind