

# Teacher Labor Market and the Dynamics of Inequality

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# Motivation

- Teachers account for less than 5% of the labor force but play a disproportionate role in the production of human capital
- Reward structure in teacher labor market affects:
  - Selection of teachers
  - Achievements of students (e.g., test scores, earnings)
- This paper studies the **dynamic spillover effects** of teacher labor market on income inequalities in the aggregate labor market
- Putting the teacher labor market in a dynamic GE context

# This paper

- An OLG model of occupation choice & child investments:
  1. Teacher quality affected by (1) the relative returns to human capital across occupations and (2) the **endogenous** human capital distribution
  2. Teacher quality has differential impacts across the income distribution
- **New mechanism**: endogenous human capital formation amplifies the occupation selection channel
- Key parameter: substitutability b/w private and public inputs in education
- Analytical solutions + empirical evidence from duty-to-bargain laws  $\implies$  **closed-form identification**
- Counterfactual + model-based decompositions

# Preview of Findings

1. Performance-based compensation in the teacher labor market:
    - Increases inequality among teachers
    - Reduces inequalities elsewhere
    - Raise intergenerational mobility
  2. With skill-biased technical change (SBTC) in non-teaching occupations:
    - Teacher labor market greatly propagates the effects on inequalities
    - To offset the direct effects of SBTC, the returns to human capital among teachers need to rise relatively more than the SBTC itself
- ⇒ existing one-generation estimates understate long-run effects on teacher quality, child outcomes, and inequalities

# Literature

- Education and inequality: Benabou (2002), Durlauf & Seshadri (2018), Caucutt & Lochner (2020), Fogli & Guerrieri (2019)  
**Contribution**: role of the teacher labor market (supply side)
- Teacher market: Hoxby (1996), Bacolod (2007), Lovenheim & Willèn (2019), Tincani (2021), Biasi & Sarsons (2022), Cohodes et al. (2023)  
**Contribution**: dynamic spillover effects in GE
- Aggregate impacts of occupational reward structure: Murphy, Shleifer, & Vishny (1991), King and Levine (1993), Acemoglu (1995)  
**Contribution**: new quantification strategy applied to teachers

# Roadmap

Model

Solution, Dynamics, and Mechanism

Identification and Calibration

Counterfactual Results

# Model Overview

- Two-period OLG: children and adults
- Two occupations: teachers and non-teachers (workers)
- Human capital production w/ parental investments & teacher quality
- In each period: occupation selection, then make child investments

# Occupation Choice

- In period  $t$ , heterogeneous human capital  $h \sim F_t(h)$
- Individuals make occupation choice after observing idiosyncratic preference shock  $\nu$  (Gumbel w/ parameter  $\theta$ ):

$$\max_{j \in \{1,2\}} \mathbb{1}_{j=1} \underbrace{(\log(w_t \cdot h^{\psi_1} \cdot \kappa) + \nu)}_{\text{teachers}} + \mathbb{1}_{j=2} \underbrace{\log(h^{\psi_2})}_{\text{workers}}$$

- $w_t$  is the relative wage across occupations
- $\psi_j$  is the occupation-specific returns to human capital
- $\kappa$  is the non-pecuniary benefits for being teachers



# Child Investments

- Workers have children and solve

$$\max_{c, e \geq 0} \log(c) + \beta \log(\mathbb{E}_\epsilon h')$$

$$c + e = y = h^{\psi_2}$$

- Human capital production function  $H(h, e, \xi, \epsilon)$ :

$$\log(h') = A + \underbrace{\log(\epsilon)}_{\text{normal dist.}} + \underbrace{\delta_0 \xi_t + \delta_1 \log(e) + \delta_2 \cdot \xi_t \cdot \log(e)}_{\text{translog}} + \underbrace{\rho \log(h)}_{\text{residual persistence}}$$

- $\xi_t$  is the  $z$ -score of average teacher human capital in the population

$$\xi_t = \frac{\overline{\log(h_1)} - \mu_t}{\sigma_t}$$

# Discussions of Assumptions

- A reduced-form approach w/o micro-founding the assignment problem between children and teachers (Sattinger 1975, Seshadri 2003)
  - $\delta_1 \log(e)$  captures both purchasing more books and getting matched with more effective teachers
  - $\delta_2 < 0 \implies$  higher  $\xi_t$  reduces the role of parental investments
- $\xi_t$  being  $z$ -score of teacher human capital in the population
  - “Relative” learning: parallel shifts in  $\mu_t$  does not affect  $\xi_t$
  - Results will be stronger if the level of human capital matters
- One-dimensional human capital  $h$ 
  - A Roy model generates the same intuition, but requires estimating multi-dimensional human capital production function
- Homogeneous preferences regarding occupations

more

robustness

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# Teacher Labor Market

- Assume  $F_h(h) \sim \text{lognormal}(\mu_t, \sigma_t^2)$  - verified in equilibrium
- Share of teachers by  $h$  is

$$l(h) = \frac{(w_t \kappa h^{\psi_1 - \psi_2})^\theta}{1 + (w_t \kappa h^{\psi_1 - \psi_2})^\theta} \approx (w_t \kappa h^{\psi_1 - \psi_2})^\theta$$

- Aggregate share of teacher  $\pi_t$  is

$$\pi_t = \int l(h) dF_t(h) = (w_t \kappa)^\theta \cdot \exp \left( \mu_t (\psi_1 - \psi_2) \theta + \frac{((\psi_1 - \psi_2) \sigma_t \theta)^2}{2} \right)$$

- If  $\pi_t$  is fixed in real life,  $\{w_t, \kappa, \psi_1\}$  need to adjust to clear the market

# Teacher Quality

- Teacher quality depends on both relative skill bias and  $\sigma_t$

$$\xi_t = \frac{\overline{\log(h_1)} - \mu_t}{\sigma_t} = (\psi_1 - \psi_2) \cdot \sigma_t \cdot \theta$$

- Change in teacher quality can be **decomposed** as

$$\underbrace{d \log(\xi)}_{\text{change in teacher quality}} = \underbrace{d \log(\psi_1 - \psi_2)}_{\text{change in selection}} + \underbrace{d \log(\sigma)}_{\text{change in h.c. dispersion}}$$

- Endogeneity of  $\sigma_t$  is the key to dynamic effects

# Optimal Investment

- Optimal private education investment (interior solution)

$$\frac{e}{y} = \frac{\beta(\delta_1 + \delta_2 \xi_t)}{1 + \beta(\delta_1 + \delta_2 \xi_t)} \approx \beta(\delta_1 + \delta_2 \xi_t)$$

- Substitute back to the human capital production function

$$\log(h') = A + \log(\epsilon) + f(\xi_t; \vec{\delta}) + (\rho + \psi_2(\delta_1 + \delta_2 \xi_t)) \log(h)$$

where  $\vec{\delta} = \{\delta_0, \delta_1, \delta_2\}$  and

$$f(\xi_t; \vec{\delta}) = \delta_0 \xi_t + (\delta_1 + \delta_2 \xi_t) \cdot \log(\beta(\delta_1 + \delta_2 \xi_t))$$

# Human Capital Dynamics

- H.c. dist. follows an AR(1) process that preserves lognormality:

$$\log(h') = A + \log(\epsilon) + f(\xi_t; \vec{\delta}) + (\rho + \psi_2(\delta_1 + \delta_2 \xi_t)) \log(h)$$

- The transition path is analytically characterized (c.f. Benabou 2002)

$$\begin{cases} \mu_{t+1} = f(\xi_t; \vec{\delta}) + (\rho + \psi_2(\delta_1 + \delta_2 \xi_t)) \cdot \mu_t - \sigma_\epsilon^2/2 \\ (\sigma_{t+1})^2 = (\rho + \psi_2(\delta_1 + \delta_2 \xi_t))^2 \cdot \sigma_t^2 + \sigma_\epsilon^2 \\ \xi_t = (\psi_1 - \psi_2) \cdot \sigma_t \theta \end{cases}$$

# Mechanism

- Suppose  $\psi_1 < \psi_2$  so that  $\xi_t < 0$ , a reduction in  $\psi_1$ 

evidence

  - Reduces teacher quality  $\xi_t$
  - If  $\delta_2 < 0$ , raises  $\text{IGE}_t = \rho + \psi_2(\delta_1 + \delta_2\xi_t)$
  - Raises  $\sigma_{t+1}$  because  $\sigma_{t+1}^2 = \text{IGE}_t^2 \cdot \sigma_t^2 + \sigma_\epsilon^2$
  - Reduces teacher quality  $\xi_{t+1}$  even further as  $\xi_{t+1} = (\psi_1 - \psi_2) \cdot \sigma_{t+1}\theta$
  - ...
- Spillover to non-teacher markets as  $F_{t+1}(h)$  changes
- Effects depend on other parameters because parental investment  $e$  endogenously respond to teacher quality  $\xi$



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# The Model Summarized

- 15 unknowns:  $\underbrace{\delta_0, \delta_1, \delta_2, A, \rho, \sigma_\epsilon^2}_{\text{h.c. technologies}}, \underbrace{\kappa, \theta, \beta}_{\text{preferences}}, \underbrace{\psi_1, \psi_2}_{\text{labor market}}, \underbrace{\xi, w, \mu, \sigma^2}_{\text{equilibrium objects}}$

Steady-state relationships in the model:

$$\frac{\mathbb{E}(y_1)}{\mathbb{E}(y_2)} = w \cdot \exp \left( \mu(\psi_1 - \psi_2) + \frac{\sigma^2}{2}(\psi_1 - \psi_2)(\psi_1 + \psi_2 + 2\psi_1\theta) \right) \quad (1)$$

$$\text{CV}(y_1) = \sigma\psi_1 \quad (2)$$

$$\text{CV}(y_2) = \sigma\psi_2 \quad (3)$$

$$\pi = (w\kappa)^\theta \cdot \exp \left( \frac{\mu\xi}{\sigma} + \frac{\xi^2}{2} \right) \quad (4)$$

$$\xi = (\psi_1 - \psi_2) \cdot \theta\sigma \quad (5)$$

$$\text{IGE} = \rho + \psi_2(\delta_1 + \delta_2\xi) \quad (6)$$

$$\sigma^2 = \frac{\sigma_\epsilon^2}{1 - \text{IGE}^2} \quad (7)$$

$$\mu = \frac{A + \delta_0\xi + (\delta_1 + \delta_2\xi) \cdot \log\left(\frac{e}{y}\right) - \sigma_\epsilon^2/2}{1 - \text{IGE}} \quad (8)$$

$$\frac{e}{y} = \beta(\delta_1 + \delta_2\xi) \quad (9)$$

- 15 unknowns with 9 equations
- Can make one normalization to pin down the scale of h.c. (set  $\psi_2 = 1$ )
- Still need additional information/moments

# Effects of Changes in Teacher Pay Rigidities - $\psi_1$

$$\frac{\partial \overline{\log(y')}}{\partial \psi_1} = \psi_2 \cdot \underbrace{\sigma \theta}_{\frac{\partial \xi}{\partial \psi_1}} \left[ \underbrace{\delta_0 + \delta_2 \log(e/y) + \psi_2 \delta_2 \mu}_{\text{direct effect through } \xi} + \underbrace{\beta \delta_2 \cdot \frac{\delta_1 + \delta_2 \xi}{e/y}}_{\text{indirect through } e} \right] \quad (10)$$

$$\frac{\partial^2 \log y'}{\partial \psi_1 \partial \log y} = \sigma \theta \cdot \psi_2 \cdot \delta_2 \quad (11)$$

$$\frac{\partial \mathbb{E}(y_1)}{\partial \psi_1} = \mathbb{E}(y_1) \cdot (\mu + \sigma^2(\psi_1 + \theta(2\psi_1 - \psi_2))) \quad (12)$$

$$\frac{\partial \pi}{\partial \psi_1} = \pi \cdot (\theta \mu + \sigma \theta \xi) \quad (13)$$

# Identification

**Proposition:** The model is identified up to the calibration of  $\rho$  if we observe  $\mathbb{E}(y_1)/\mathbb{E}(y_2)$ ,  $CV(y_1)$ ,  $CV(y_2)$ ,  $\pi$ , IGE,  $e/y$  and measure the left-hand-sides of Equations (10)-(13).

**Proof:** Given that  $\psi_2 = 1$ , Equation (3) identifies  $\sigma$ ; then Equation (2) identifies  $\psi_1$ ; Equation (7) identifies  $\sigma_\epsilon$ . Combining Equations (12) and (13) by substituting out  $\mu$  identifies  $\theta$ . Then, Equation (5) identifies  $\xi$ ; Equation (13) identifies  $\mu$ ; Equation (1) identifies  $w$ ; Equation (4) identifies  $\kappa$ ; Equation (11) identifies  $\delta_2$ ; Equation (6) identifies  $\delta_1$  given that I calibrated  $\rho$ ; Equation (9) identifies  $\beta$ ; and lastly Equation (8) identifies  $A$ .

- Choose conservative value  $\rho = 0.6 \cdot \text{IGE}$  following Lefgren et al. (2012)

# Empirical Setting

- Passage of duty-to-bargain (DTB) laws across states
- Mandated that districts have to negotiate in good faith with a union that has been elected for the purposes of collective bargaining
- NBER collective bargaining law dataset (1960-1996)
- In recent years some states discontinued collective bargaining requirements  $\Rightarrow$  variations from expiration of preexisting collective bargaining agreements (Biasi 2021)

# Effects on Next Generation

- Lovenheim & Willèn (2019) on long-run outcomes (ACS)

$$Y_i = \beta \cdot \text{DTB exposure}_i + \zeta X_i + \epsilon$$

- 10-year exposure reduces next gen's earnings by 2.36% on average
- Effects are 4.74 pp smaller for White and Asian households (who have 40% higher income than Black and Hispanic parents)  $\implies \delta_2 < 0$

# Effects on Teacher Labor Market

- Effects on teachers' employment and earnings (CPS-ASEC)

$$Y_{\text{state,year}} = \beta \cdot \text{DTB}_{\text{state,year}} + \text{State FE} + \text{Year FE} + \epsilon$$

	(1)	(2)	(3)
	Teacher share	CV(teacher)	Average teacher earnings
DTB	-0.351**	-0.0292*	-591.3
	(0.110)	(0.0138)	(425.4)
# Observations	1378	1364	1378

*Notes:* This table displays the results of regression (21). Standard errors in parentheses. \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$ . Average teacher earnings is measured in year 2000 dollars.

- Results are qualitatively consistent with Biasi (2021)



# Link Back to Model Predictions

- Interpreting DTB laws as reductions in  $\psi_1$ , measured using

robustness

$$\Delta\psi_1 = \psi_1 \cdot \frac{\Delta\text{CV}(y_1)}{\text{CV}(y_1)}$$

- Revisit Equations (10)-(13) and rewrite them as:

$$\Delta\overline{\log(y')} = \text{CV}(y_2) \cdot \theta \left[ \delta_0 + \delta_2 \log(e/y) + \psi_2 \delta_2 \mu + \beta \delta_2 \cdot \frac{\delta_1 + \delta_2 \xi}{e/y} \right] \cdot \Delta\psi_1 \quad (10')$$

$$\frac{\Delta^2 \log y'}{\Delta \log y} = \theta \delta_2 \cdot \text{CV}(y_2) \cdot \Delta\psi_1 \quad (11')$$

$$\Delta \log(y_1) = [\mu + \sigma^2(\psi_1 + \theta(2\psi_1 - \psi_2))] \cdot \Delta\psi_1 \quad (12')$$

$$\frac{\Delta\pi}{\pi} = (\theta\mu + \sigma\theta\xi) \cdot \Delta\psi_1 \quad (13')$$

# Calibration Result

Human capital formation parameters			Value	Preference parameters			Value
$\delta_0$	teacher effect (level)		-0.48	$\kappa$	teacher cost		0.21
$\delta_1$	investment effect		-0.42	$\theta$	taste shock dispersion		3.07
$\delta_2$	teacher effect (gradient)		-0.88	$\beta$	weight on children's h.c.		0.35
$A$	TFP of h.c. production		1.12	Equilibrium objects			Value
$\rho$	residual persistence		0.22	$\xi$	teacher quality		-0.64
$\sigma_\epsilon$	ability shock dispersion		0.72	$w$	relative wage		2.18
Labor market parameters			Value	$\mu$	average h.c.		1.15
$\psi_1$	skill bias (teachers)		0.73	$\sigma$	h.c. dispersion		0.77
$\psi_2$	skill bias (non-teachers)		1				

# Non-Targeted Moments

- One standard deviation change of  $\xi$  raises human capital by 1.6 percent
  - Chetty et al. (2014): 1.3 percent
- Using Kmenta (1967) approximation, the implied elasticity of substitution between  $e$  and  $\xi$  is around 2
  - Blankenau (2015): 1.92
  - Kotera and Seshadri (2017): 2.43

# Roadmap

Model

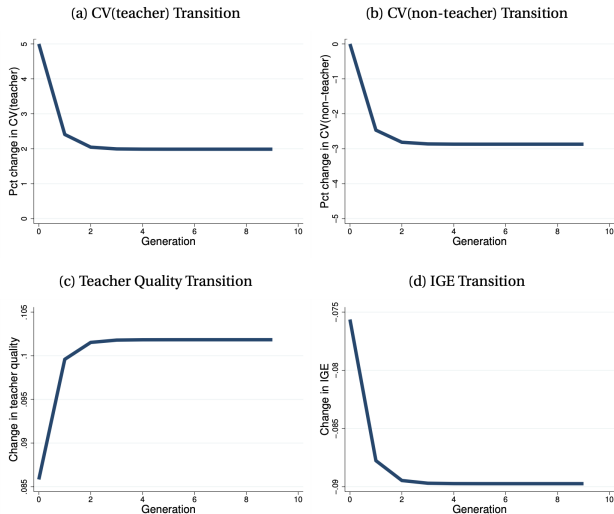
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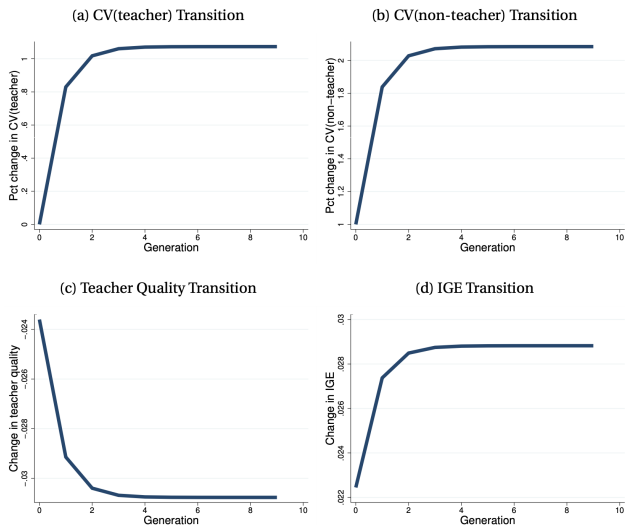
# Counterfactual 1: Performance-Based Compensation

- Increase  $\psi_1$  by 5% (similar magnitude as DTB)



# Counterfactual 2: SBTC

- Increase  $\psi_2$  rises by 1% - “convexification” (Autor, Goldin & Katz 2020)



# Decompose the Source of Changing Inequality

- First, because  $CV(y_2) \equiv \sigma\psi_2$ , I decompose  $CV(y_2)$ :

$$\underbrace{CV(y_2)' - CV(y_2)}_{\text{changes in inequality}} = \underbrace{\sigma'(\psi_2' - \psi_2)}_{\text{direct effect}=48\%} + \underbrace{(\sigma' - \sigma)\psi_2}_{\text{indirect h.c. dist.}=52\%}$$

- To understand what drives changes in  $\sigma$ , I conduct a decomposition of IGE because in the steady-state  $\sigma$  is proportional to  $\sqrt{1 - IGE^2}$

$$\underbrace{\psi_2'(\delta_1 + \delta_2\xi') - \psi_2(\delta_1 + \delta_2\xi)}_{\text{changes in IGE}} = \underbrace{\psi_2\delta_2(\xi' - \xi)}_{\text{teacher quality}=94\%} + \underbrace{(\delta_1 + \delta_2\xi')(\psi_2' - \psi_2)}_{\text{parental income}=6\%}$$

- $\psi_1$  needs to rise **relatively more** than  $\psi_2$  to neutralize the effects of the SBTC on inequalities in non-teaching markets

# Conclusion

- Dynamic spillover effects of teacher labor market on aggregate inequality
- Identify and calibrate the model using empirical evidence from DTB laws
- Key takeaways:
  1. Performance-based compensation trades off income inequality among teachers versus non-teachers
  2. Teacher market greatly amplifies the effects of SBTC
- Same mechanism applies to other occupations such as doctors and elected officials – avenue for future research



# Skill Requirement by Occupation

Table 1: Skill Importance by Occupation

	Teacher	Non-Teachers	
	Value	Mean	Standard Deviation
Complex Problem Solving Skills	3.53	3.19	0.50
Resource Management Skills	2.40	2.39	0.66
Social Skills	3.27	2.89	0.52
System Skills	3.17	2.85	0.60
Technical Skills	1.45	1.93	0.79

Notes: This table displays the importance of each cross-functional skills by teachers and non-teachers in the O\*NET dataset.

# Declining Teacher Quality Over Time

TABLE 3.—DECLINE IN TEACHER QUALITY AS EVIDENCED BY TEST SCORES:  
FRACTION SCORING LOWER OR UPPER 20% ON IQ, AFQT AMONG  
FEMALE TEACHERS

<i>Birth Cohort</i>	<i>Above 80th Percentile</i>	<i>Below 20th Percentile</i>
<b>1941–45</b>	0.41	0.08
<b>1946–49</b>	0.40	0.05
<b>1951–53</b>	0.34	0.06
<b>1957–59</b>	0.44	0.06
<b>1960–62</b>	0.20	0.12
<b>1963–64</b>	0.19	0.19

Source: NLS-YW,YM,Y79. Sample for this table includes black and white female respondents with at least two years of college who ever taught when they were aged 21 to 30. Sample selection is further described in the text and appendix A.

Source: Bacolod (2007)

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# Interpretation of DTB Laws

What if DTB laws not only affect  $\psi_1$ , but also  $\kappa$  and  $w$ ?

$$\begin{cases} \Delta \log(y_1) = \Delta \psi_1 [\mu + \sigma^2(\psi_1 + \theta(2\psi_1 - \psi_2))] + \Delta w \\ \frac{\Delta \pi}{\pi} = \Delta \psi_1 \cdot (\theta \mu + \sigma \theta \xi) + \theta(\Delta w + \Delta \kappa) \end{cases} \quad (14)$$

They threat the identification of  $\theta$  and  $\mu$

- **Insight:** alternative values of  $\theta$  does not affect counterfactual results
- Results remain unchanged as long as either  $\Delta w$  or  $\Delta \kappa$  equals to zero because we can choose  $\theta$  exogenously
- If both  $\Delta w$  and  $\Delta \kappa$  are non-zero, then need additional moment (e.g., teacher value-added)  $\Rightarrow$  model performs well in that dimension

# Heterogeneous Assumptions

- Suppose instead there is a systematic relationship between taste and human capital, consider

$$\nu = \psi_3 \log(h) + \tilde{\nu}$$

- Individual's choice problem becomes

$$\max_{j \in \{1,2\}} \mathbb{1}_{j=1} (\log(wh^{\psi_1} \cdot \kappa \cdot h^{\psi_3}) + \tilde{\nu}) + \mathbb{1}_{j=2} \log(h^{\psi_2})$$

- Model is largely unchanged, but need another moment to identify  $\psi_3$
- Can use teacher value-added

back