Teacher Quality Policy When Supply Matters

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We want to raise teacher productivity. How?

Objectives

Policies

	Existing teachers work harder	Existing teachers work better	Higher-ability teacher workforce
Performance pay	Repeated failure (in US)		??
Performance- based non- retention	No evidence either way		??
Coaching and mentoring		++ (Taylor & Tyler 2011)	

Jumping off points

Safelite performance pay (Lazear 2000)

- Big effects of pay-for-performance on effort & selection.
- Lazear (2003) suggests same for education.

A long literature on performance measurement (AKA value added modeling).

- Ongoing debates about statistical properties in low-stakes settings.
- Little formal consideration of how the measures will be used.
- <u>Tennessee POINT (Springer et al. 2010)</u>. Performance bonuses up to \$15,000 per year had no effect.
 - Three-year experiment with volunteers.
 - Gets at effort margin; selection impossible to study using RCTs
- <u>Staiger & Rockoff (JEP 2010)</u>. Model selection effects of performance-based firing rules.
 - No model of the labor market -- only tradeoff is ability vs. experience.
 - Optimal policy: Fire 80% of teachers after year 2.

This paper

- Goal: Understand potential selection effects of performance pay and performance-based retention, taking account of:
 - (Some of the) known imperfections of performance measures
 - Labor market effects (aka self selection contraints)
- Strategy: Develop dynamic model of occupation choice (teaching vs. other) & calibrate with plausible parameters.
 - Focus on role of limited information
 - Set aside influence activities / goal distortion / manipulation treat performance measure as noisy but unbiased.
- Policy counterfactuals: New teacher contracts
 - Implemented by entire education sector
 - Established and permanent

The logic of the model

- Let τ be a teacher's true ability, and μ_t her perceived ability (posterior mean) after t years.
- Alternative contracts change the expected compensation as a function of τ . The more reliable the performance measure, the steeper the slope of E[w | τ].
- Incentives for selection depend on E[E[w | τ] | μ_t]. This flattens the slope if teachers do not have information about their own ability.
- But we care about selection on τ , not on μ . This creates more flattening if teachers' information is limited.
- In the model, teachers start with a small amount of private information and learn more from subsequent performance measures.

Contracts

1. Baseline: No firing, salaries rise with experience.

2. Bonuses

- 20% bonuses if 0.5($y_1 + y_2$) > y^{PP}
- 1st year teachers ineligible.
- y^{PP} calibrated so 25% of current teachers qualify.
- Not a tournament more could qualify if ability distribution rose.
- Base salaries reduced to yield same total number of teachers.

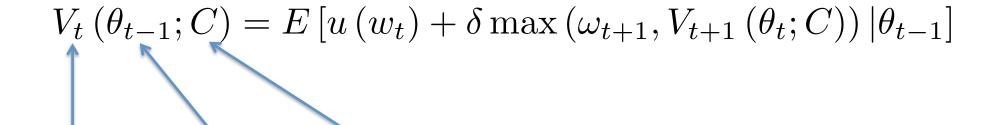
3. Firing

- Teacher fired if district's posterior mean falls below a threshold y^F.
- Posterior mean:

$$\overline{y} \frac{\sigma_{\tau}^{2}}{\sigma_{\tau}^{2} + \frac{\sigma_{\varepsilon}^{2}}{t}}$$

- y^F calibrated so 10% of current teachers would be fired immediately.
- Firing reduces future earnings distribution by 10%.
- Salaries increased to yield same total number of teachers.

Dynamic optimization problem

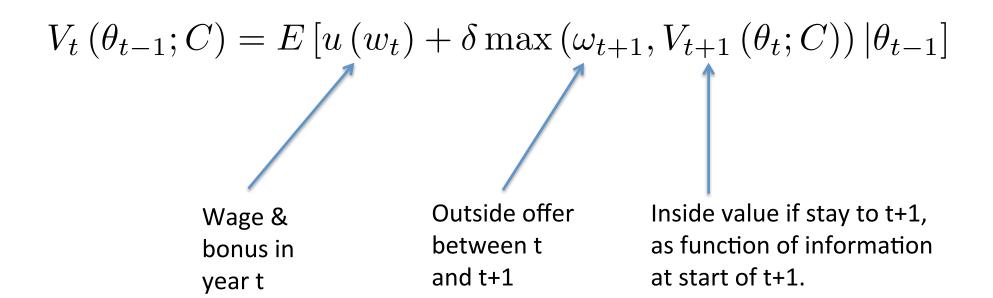


Value if remain in teaching in year t

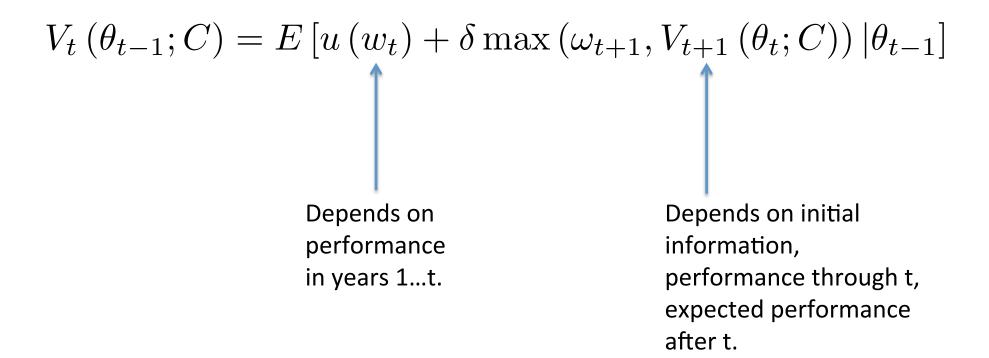
State Contract variables:
Initial information, performance in years 1

through t-1

Dynamic optimization problem

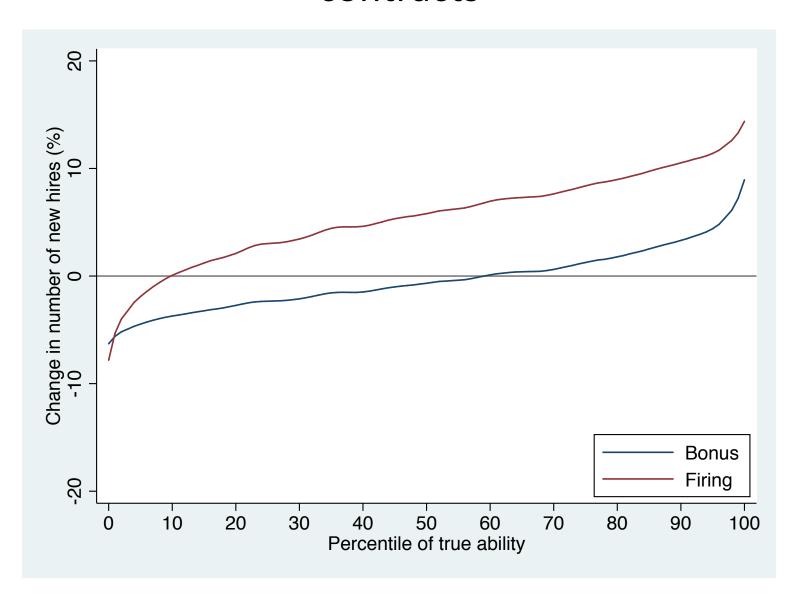


Dynamic optimization problem

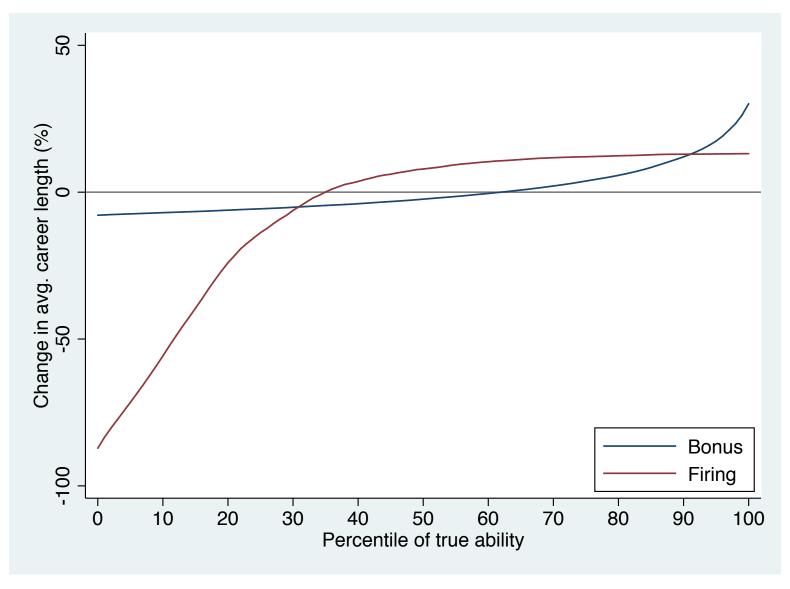


Expectation is over posterior distribution of own ability, distribution of performance in t, t+1, ..., T given ability.

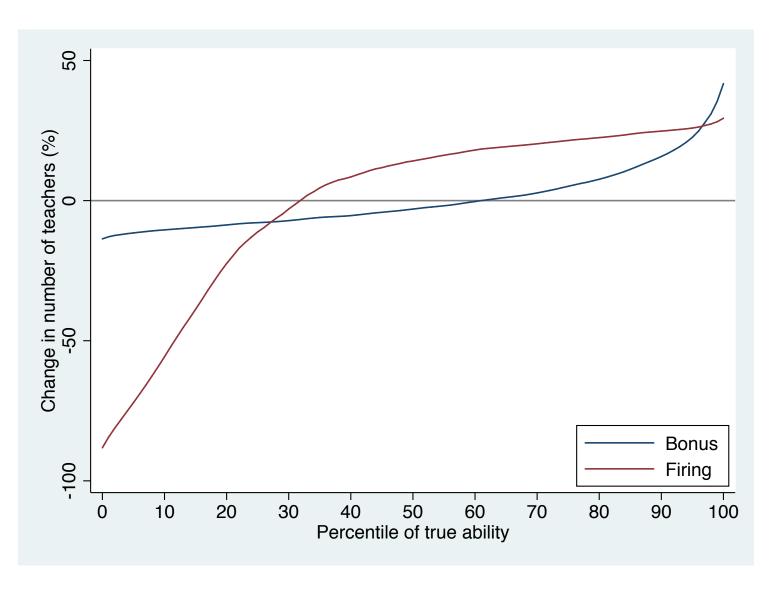
Entrants by ability under bonus & firing contracts



Career length by ability



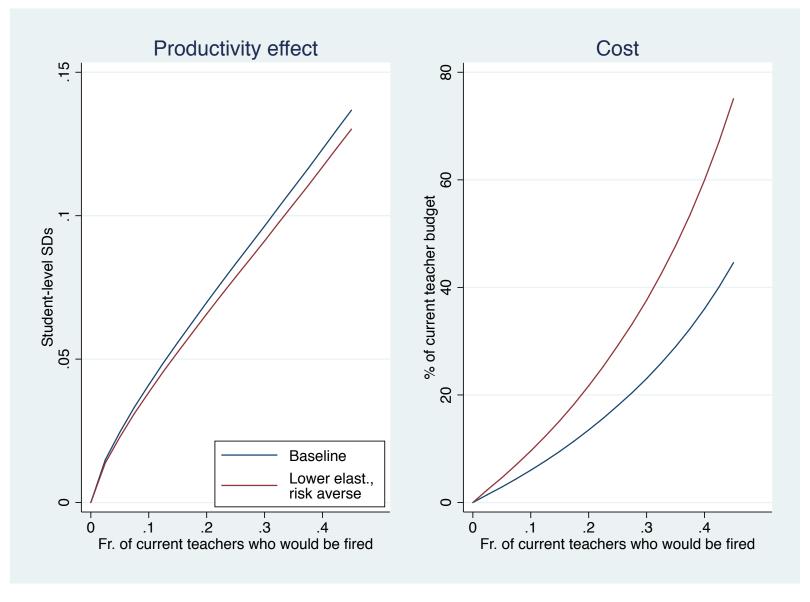
Total # of teachers by ability



Impact of alternative contracts

	Base	Bonuses	Firing
Ability: Mean (SD)	0.000 [0.150]	0.015 [0.153]	0.040 [0.130]
Fr. 1 st year	8.0%	8.0%	8.1%
Fr. 1 st three years	30.9%	30.8%	31.0%
Avg. experience	8.8	8.9	9.1
Impacts: Mean (SD)	-0.011 [0.151]	0.004 [0.155]	0.029 [0.134]
Base salary (rel. to baseline)		-3.6%	+5.4%
Total wage bill (rel. to baseline)		+1.8%	+5.9%
Net impact on effectiveness		+0.015	+0.041

Varying the firing rate



Multiple tasks & influence activities

Multiple tasks

- Suppose two dimensions of output, A & B, with $corr(\tau_A, \tau_B) = 0.4$.
- Reward A but want B.

Influence activities

- Can raise measured performance by E, at personal cost $c(E) = kE^2$.
- Calibrate k: c(0.137) = 0.2. [SD($\tau_A \mid \tau_B$)=0.137].
- Option A: E doesn't affect dimension-B output.
- Option B: E comes at expense of dimension B.

Impact of firing policy with multiple tasks & influence activities

	Measured effect	True effect on measured output	True effect on unmeasured output	Cost
Baseline	+0.042	+0.042	n/a	+6.1%
Multiple outputs	+0.042	+0.042	+0.020	+6.1%
Influence activity – nondistorting	+0.037	+0.030	+0.013	+3.9%
Influence activity – distorting	+0.037	+0.030	+0.005	+3.9%

Conclusions

- Can't predict effect of changing the teaching contract without accounting for the teacher labor market.
- When labor market responses are incorporated:
 - Both bonuses & firing policies have positive effects.
 - Both are expensive (but still pass cost benefit test).
 - Plausible effects are not enormous.
 - Effects evaporate if allow for multiple outputs & influence activities.
- Caveat 1: Model is cooked to make the policies look good.
 - Highly elastic labor supply
 - Lots of private information
 - Little risk aversion
 - Unbiased performance measure
- Caveat 2: Many key parameters are made up.
 - Traditional program evaluations / social experiments will be uninformative.
 - Keys: Labor supply, private information, potential to screen on entry, outside labor market return to teaching experience, impact of firing, potential for goal distortion.