# Pay by Design <br> Teacher Performance Pay Design and the Distribution of Student Achievement 

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## Teacher Performance Pay



- Teachers among most important inputs to student achievement (Aaronson, Barrow, and Sander, 2003; Rockoff, 2004; Rivkin, Hanushek, and Kain, 2005; Hanushek and Rivkin, 2010; Rivkin, 2006; Chetty, Friedman and Rockoff , 2013)
- But often work in settings where they face incentives that are weak or misaligned with improving student outcomes (Lazear, 2003)


## Teacher Performance Pay

- Widespread policy interest in motivating teachers by linking pay to performance metrics - commonly student exam scores:
- US, Australia, UK, Israel; Mexico, Chile, Kenya, India, Pakistan, China
- But, mixed evidence on effectiveness

Effects of Teacher Performance Pay Programs on Achievement (After 1 year)


## Performance Pay Design

- One reason for mixed evidence on teacher performance pay may be because the design of performance pay varies across studies (Neal, 2011)
- Two design features which vary across studies:
- Design feature 1: the way in which student achievement scores are used to measure teacher performance \& mapping to rewards
- Design feature 2: the size of the rewards
- Despite the theoretical importance of these design features, there is little empirical evidence about how varying them affect:
- Student achievement on average
- The achievement of different types of students
- Theoretically compelling designs may not outperform simple/ transparent schemes in practice


## This Study

Randomized trial across 216 primary schools in rural western China to study

1. How different ways of using student achievement to measure and reward teacher performance affect teacher effort and student achievement
2. Whether the size of potential rewards matters
3. How different performance pay designs affect achievement among low, medium, and high achieving students within the classroom? (i.e. do teachers "triage" students in response to incentives)

## Rest of the Presentation

- Background/Context
- Study in Rural China
- Teacher Performance Pay Policy in 2009
- Experimental Design and Interventions
- Sampling/Data
- Results + Discussion


## Rural China: low levels of learning

- Rural-urban achievement gap grows as children progress through the education system ( 0.8 SD gap in Math by grade 6).

grade 3
grade 6
Low levels of learning despite large, large increases in government expenditures on rural, compulsory education (NBS, 2011)


## Teacher Performance Pay Policy in China

- 2009 Teacher Performance Pay Policy
- Increased teacher salaries to the level of other local civil servants
- Stipulated that $30 \%$ of increase be awarded based on performance
- How was the policy actually implemented?
- Teacher performance based mainly on inputs (e.g. class hours) and subjective measures
- Little variation in actual rewards: $\mathbf{3 0 0}$ yuan difference per semester between top and bottom teacher on average
- Teachers rankings done WITHIN schools (potentially problematic)
- When evaluated on student scores, rankings based on levels




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## Experimental Design

| Teacher incentive treatment (outcome-based <br> design feature x payout size design feature) | X. Large <br> incentive payout | Y. Small incentive <br> payout |
| :--- | :--- | :--- |
| A. Control | A. 52 schools |  |
| B. Levels incentive | BX. 26 schools | BY. 28 schools |
| C. Gains incentive | CX. 26 schools | CY. 30 schools |
| D. Pay for percentile incentive | DX. 26 schools | DY. 28 schools |

Math teachers in 216 schools
Approximately 8,000 grade 6 students

## Underlying Structure <br> (Common to all treatment groups)

- Incentives tied to student achievement as measured by scores on standardized math exams
- Teachers compete in rank-order tournament with teachers in other schools
- No explicit penalty for missing students, but potential disqualification


# Design Feature 1: Different ways of using student 

 achievement to measure and reward teacher performance (Teacher Performance Indices)Levels Incentive: Rewards teachers based on student achievement on an end-of-the-year exam

Gains Incentive: Rewards teachers based on gains in achievement from the start to the end of the year

Pay for percentile incentive: Reward teachers based on pay for percentile index: Within similar comparison sets (among students with similar baseline scores), rank students by scores on endline exam and give them a percentile rank. Averaged them to create pay for percentile index (Neal, 2011).

- Explicitly accounts for multiple students (Barlevy \& Neal, 2012)


## Design Feature 2: Large vs. Small Rewards



Top reward in small group $\boldsymbol{\sim} 1$ month pay

## Incentive＂Agreement＂

## 老师编码

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＂提高农村小学数学教学质量试点项目＂激励协议

甲方：西北大学西北社会经济发展研究中心
乙方： $\qquad$老师

一，甲方在前期调研的基础上，经与当地教育主管部门协商，特邀请乙方参加＂提高农村小学数学教学质量＂试点项目。该项目主要是通过将数学老师的潡励奖金与老师的教学厉量挂钧，给老师提供一个＂优绩优䂨＂的机会。

二，本项目主要分为四个阶段：
－第一阶段：培训会。在2013年9月初，甲方将邀请乙方参加为期两天的堷训会，详细介绍项目的实施方案，请乙方填写调查表，双方签署本协议。
－第二阶段：统一的标准化数学考试。在2014年春季学期末，甲方将对乙方任数学老师的某个班的学生进行统一的标准化数学考试。
－第三阶段：衡量教学质量并计算潡梮奖金。基于乙方的班的学生在标准化数学考试中的表现，街量乙方的教学质量。乙方能从甲方㨍到多少奖金，取决于乙方的教学质量，以及的表现，教学质量相对于其余参与项目老师的百分位。计算方法如下：
－教学质量——数学成殸的增加值百分位，
1，以你班的小明同学为例，基于小明在 2013 年 5 月和 2014 年春季学期末标准化数学考试中的成绩，计算小明两次考试成绩的增加值。
2，从全部参加项目的老师所教的全部学生中，找出在2013年5月那次标准化数学考试中跟小明得分完全一样的同学，我们称为跟小明＂起点相同的同学＂

3，按品考试成殸的增加值，将小明与跟他所有＂起点相同的同学＂进行比较，计算小明在这帮同学中的增加值百分位。

## Incentive Agreement Guide



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## Sampling and Data

## Sample

- 16 Counties in Tianshui (Gansu) and Yulin (Shaanxi)
- 216 Schools (243 Math Teachers)
- All $6^{\text {th }}$ grade students, about 8,000 Students Total


## Data

- 2 waves of pre-program math scores
- Teacher Survey at Baseline (Sept. 2013)
- Detailed information on teacher characteristics, existing incentives, perceptions, social preferences
- Endline Math Exam (constructed test w/ good properties)
- Detailed Student, Teacher, School Surveys (May 2014)


## Sampling: Study Locations



## Estimation Strategy

- Main specification (for child $i$ in school $j$ ):

$$
Y_{i j c}=\alpha+T^{\prime}{ }_{j c} \beta+x_{i j c}^{\prime} \gamma+\lambda_{c}+\varepsilon_{i j c}
$$

- $Y_{i j c}$ Outcome of interest at the endline (e.g. math scores)
- $T_{j c}$ Vector of treatment dummies
- $x_{i j c}$ Baseline student, teacher, school characteristics
- $\lambda_{c}$ Block/strata (county) fixed effects
- Standard errors account for clustering at the school level
- Significance based on p-values adjusted for multiple hypotheses (Romano and Wolf)
- Pre-analysis plan filed in AEA registry before data available
- Balance across treatment arms on baseline characteristics


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## Average Impacts on Math Scores

 (Design Feature 1: Teacher Performance Indices)|  | (1) | (2) | (3) | (4) |
| :---: | :---: | :---: | :---: | :---: |
| Levels | 0.056 | 0.084 |  |  |
|  | (0.048) | (0.052) |  |  |
| Gains | 0.012 | 0.001 |  |  |
|  | (0.051) | (0.050) |  |  |
| Pay-for-percentile | 0.128* | 0.148** |  |  |
|  | (0.064) | (0.064) |  |  |
| Small |  |  | 0.063 | 0.081 |
|  |  |  | (0.053) | (0.055) |
| Large |  |  | 0.064 | 0.067 |
|  |  |  | (0.045) | (0.046) |
| Baseline Scores | Yes | Yes | Yes | Yes |
| Strata FE | Yes | Yes | Yes | Yes |
| Other Controls |  | Yes |  | Yes |
| P-value: Gains - Levels | 0.390 | 0.114 |  |  |
| P-value: P4Pct - Levels | 0.236 | 0.292 |  |  |
| P-value: P4Pct - Gains | 0.078 | 0.023** |  |  |
| P-value: Large - Small |  |  | 0.989 | 0.778 |
| Observations | 7,454 | 7,373 | 7,454 | 7,373 |

* p<0.1 after adjustment.

Average Impacts on Math Scores (Design Feature 2: Large vs Small Rewards)

|  | (1) | (2) | (3) | (4) |
| :---: | :---: | :---: | :---: | :---: |
| Levels | 0.056 | 0.084 |  |  |
|  | (0.048) | (0.052) |  |  |
| Gains | 0.012 | 0.001 |  |  |
|  | (0.051) | (0.050) |  |  |
| Pay-for-percentile | 0.128* | 0.148** |  |  |
|  | (0.064) | (0.064) |  |  |
| Small |  |  | 0.063 | 0.081 |
|  |  |  | (0.053) | (0.055) |
| Large |  |  | 0.064 | 0.067 |
|  |  |  | (0.045) | (0.046) |
| Baseline Scores | Yes | Yes | Yes | Yes |
| Strata FE <br> Other Controls | Yes | Yes | Yes | Yes |
|  |  | Yes |  | Yes |
| P-value: Gains - Levels | 0.390 | 0.114 |  |  |
| P-value: P4Pct - Levels | 0.236 | 0.292 |  |  |
| P-value: P4Pct - Gains | 0.078 | 0.023** |  |  |
| P-value: Large - Small |  |  | 0.989 | 0.778 |
| $\frac{\text { Observations }}{\text { Robust standard errors }}$ | 7,454 | 7,373 | 7,454 | 7,373 |
| * $\mathrm{p}<0.1$ after adjustment. |  |  | $1 \text { Pe }$ |  |

## Average Impacts on Math Scores

(Teacher Performance Index by Reward Size)

## Note: Not Pre-specified

|  | Small |  |  | Large |  |
| :--- | :---: | :---: | :---: | :---: | :---: |
|  | $(1)$ | $(2)$ | $(3)$ | $(4)$ |  |
| Levels | 0.046 | 0.080 |  | 0.064 | 0.081 |
|  | $(0.059)$ | $(0.067)$ |  | $(0.059)$ | $(0.061)$ |
| Gains | 0.049 | 0.037 |  | -0.033 | -0.033 |
|  | $(0.064)$ | $(0.063)$ |  | $(0.060)$ | $(0.061)$ |
| Pay-for-percentile | 0.089 | 0.131 |  | $0.163^{* *}$ | $0.165^{* *}$ |
|  | $(0.094)$ | $(0.100)$ |  | $(0.059)$ | $(0.060)$ |
|  |  |  |  |  |  |
| Baseline Scores | Yes | Yes |  | Yes | Yes |
| Strata FE | Yes | Yes |  | Yes | Yes |
| Other Controls |  | Yes |  |  | Yes |
|  |  |  |  |  |  |
| P-value: Gains - Levels | 0.974 | 0.730 |  | 0.153 | 0.100 |
| P-value: P4Pct - Levels | 0.648 | 0.667 |  | 0.157 | 0.237 |
| P-value: P4Pct - Gains | 0.690 | 0.546 |  | $0.005^{* *}$ | $0.004^{* *}$ |
|  |  |  |  |  |  |
| Observations | 4655 | 4609 |  | 4678 | 4628 |

Robust standard errors accounting for clustering at the school level in parentheses. *** p<0.01, ** $\mathrm{p}<0.05, * \mathrm{p}<0.1$. Other controls include student gender, age, parent's education, a household wealth index, class size, teacher experience and teacher base salary.

## Mechanisms



## Impacts on teacher behavior

- Like Glewwe et al. (2010) and Muralidharan and Sundararaman (2011), we find little effect on many types of teacher behavior in the classroom (reported by students):
- Classroom engagement
- Care
- Classroom management
- Communication with students
- We find no significant effect on self-reported teacher effort.
- While we do find impacts of all types of incentives on studentreported times being tutored outside of class, these do not explain the significantly larger impact of pay-for-percentile.


## Effect on Amount and Type of Curricula Taught



## More curricula coverage - was it at the expense of intensity of instruction (teachers just went faster)?

Treatment effects (large rewards group) on easy, hard, medium test items:

|  |  | Easy <br> (1) | Medium (2) | Hard <br> (3) |
| :---: | :---: | :---: | :---: | :---: |
| (1) | Levels Incentive | 0.029 | 0.094 | 0.075 |
|  |  | (0.044) | (0.50) | (0.052) |
| (2) | Gains Incentive | -0.006 | -0.010 | 0.019 |
|  |  | (0.036) | (0.050) | (0.053) |
| (3) | Pay-for-Percentile Incentive | 0.105** | 0.092 | 0.16** |
|  |  |  | (0.062) | (0.067) |

Pay-for-percentile led to gains in easy \& hard items- suggesting it increases both the coverage \& intensity of instruction

## Distributional Effects of the Incentive Designs: Within Classes



## Distributional Effects of the Incentive Designs: Within Classes



## Summary of Results

- Of the different teacher performance indices used to incentivize teachers, only pay for percentile has significant effects on average. It is accompanied by meaningful changes in curricular coverage
- Doubling size of reward does not have statistically significant effect
- limited power to test within incentive designs
- Only large increase in point estimate for pay-for-percentile
- Pay for percentile produced broad-based gains across distribution of student achievement
- For levels, gains teachers focus on students for whom perceived returns to effort highest


## Limitations/Contributions

Limitations: Only examined impacts after year 1.

## Contributions:

1) Head-to-head experimental test of alternative approaches of mapping student achievement into rewards for individual teachers (including "pay-forpercentile")

- Adds to previous work testing individual vs group incentives
(Muralidharan and Sundararaman 2011; Behrman et al. 2012), standard vs
loss aversion-based incentives (Fryer et al. 2012)

2) First experimental study of reward size in teacher performance pay
3) Closely examine how different performance pay designs affect distribution of student achievement within the class (and how teachers likely multitask across students)

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