







# Using Big Data to Solve Economic and Social Problems

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## **K-12 Education**

### K-12 Education: Background

• U.S. spends nearly \$1 trillion per year on K-12 education

- Decentralized system with substantial variation across schools
  - Public schools funded by local property taxes → sharp differences in funding across areas
  - Private schools and growing presence of charter schools

### **K-12 Education: Overview**

- Main question: how can we maximize the effectiveness of this system to produce the best outcomes for students?
  - Traditional approach to study this question: qualitative work in schools
  - More recent approach: analyzing big data to evaluate impacts
- References:

Chetty, Friedman, Hilger, Saez, Schanzenbach, Yagan. "How Does Your Kindergarten Classroom Affect Your Earnings? Evidence from Project STAR" *QJE* 2011.

Reardon, Kalogrides, Fahle, Shores. "The Geography of Racial/Ethnic Test Score Gaps." Stanford CEPA Working Paper 2016

Fredriksson, Ockert, Oosterbeek. "Long-Term Effects of Class Size." QJE 2012

Chetty, Friedman, Rockoff. "Measuring the Impacts of Teachers I and II" AER 2014

### Using Test Score Data to Study K-12 Education

- Primary source of big data on education: standardized test scores obtained from school districts
  - Quantitative outcome recorded in existing administrative databases for virtually all students
  - Observed much more quickly than long-term outcomes like college attendance and earnings

### **Using Test Score Data to Evaluate Primary Education**

- Common concern: are test scores a good measure of learning?
  - Do improvements in test scores reflect better test-taking ability or acquisition of skills that have value later in life?

- Chetty et al. (2011) examine this issue using data on 12,000 children who were in Kindergarten in Tennessee in 1985
  - Link school district and test score data to tax records
  - Ask whether KG test score performance predicts later outcomes

### **A Kindergarten Test**

- I'll say a word to you. Listen for the *ending* sound.
- You circle the picture that *starts* with the same sound













#### **College Attendance Rates vs. KG Test Score**



#### Marriage by Age 27 vs. KG Test Score



### **Studying Differences in Test Score Outcomes**

 Test scores can provide a powerful data source to compare performance across schools and subgroups (e.g., poor vs. rich)

Problem: tests are not the same across school districts and grades
→ makes comparisons very difficult

- Reardon et al. (2016) solve this problem and create a standardized measure of test score performance for all schools in America
  - Use 215 million test scores for students from 11,000 school districts across the U.S. from 2009-13 in grades 3-8

### Making Test Score Scales Comparable Across the U.S.

- Convert test scores to a single national scale in three steps:
  - 1. Rank each school district's average scores in the statewide distribution (for a given grade-year-subject)
  - 2. Use data from a national test administered to a sample of students by Dept. of Education to convert state-specific rankings to national scale
    - Ex: suppose CA students score 5 percentiles below national average
    - Then a CA school whose mean score is 10 percentiles below CA mean is 15 percentiles below national mean
  - 3. Convert mean test scores to "grade level" equivalents

# Nationwide District Achievement Variation, 2009-2013



Average Test Scores, by School District, Grades 3-8, 2009-2013



© (2016) sean f. reardon, Demetra Kalogrides, Erin Fahle, Kenneth Shores, and Benjamin Shear. Stanford Education Data Archive: seda.stanford.edu



### **Achievement Gaps in Test Scores by Socioeconomic Status**

 Next, use these data to examine how test scores vary across socioeconomic groups

 Define an index of socioeconomic status (SES) using Census data on income, fraction of college graduates, single parent rates, etc.

# Academic Achievement and Socioeconomic Status US School Districts, 2009-2013



# Academic Achievement and Socioeconomic Status California and Massachusetts School Districts, 2009-2013







### How Can We Improve Poorly Performing Schools?

 There are many school districts in America where students are two grade levels behind national average, controlling for SES

- How can we improve performance in these schools?
  - Simply spending more money on schools is not necessarily the solution...

#### **Test Scores vs. Expenditures on Primary Education Across Countries**



Spending per pupil in primary education (World Bank 2009)

### **Two Policy Paradigms to Improve Schools**

- Two distinct policy paradigms to improve schools
  - 1. Government-based solutions: improve public schools by reducing class size, increasing teacher quality, etc.
  - 2. Market-based solutions: charter schools or vouchers for private schools

- Contentious policy debate between these two approaches
  - We will consider each approach in turn

### **Government-Based Solutions: Improving Schools**

### **Improving Schools: The Education Production Function**

 Improving public schools requires understanding the education production function

How should we change schools to produce better outcomes?

Better Teachers?



Smaller Classes?



Better Technology?



### **Effects of Class Size**

- Begin by analyzing effects of class size
- Cannot simply compare outcomes across students who are in small vs. large classes
  - Students in schools with small classes will generally be from higherincome backgrounds and have other advantages
  - Therefore, this simple comparison in observational data will overstate causal effect of class size
- Need to use experimental/quasi-experimental methods instead

### **Effects of Class Size: Tennessee STAR Experiment**

- Student/Teacher Achievement Ratio (STAR) experiment
  - Conducted from 1985 to 1989 in Tennessee
  - About 12,000 children in grades K-3 at 79 schools

- Students and teachers randomized into classrooms within schools
  - Class size differs: small (~15 students) or large (~22 students)
  - Classes also differ in teachers and peers

### **Effects of Class Size: Tennessee STAR Experiment**

 Evaluate impacts of STAR experiment by comparing mean outcomes of students in small vs. large classes

 Report impacts using regressions of outcomes on an indicator (0-1 variable) for being in a small class [Krueger 1999, Chetty et al. 2011]

	Dep Var:	Test Score	College Attendance	Earnings
	Outcome	(1)	(2)	(3)
Small Class		4.81 (1.05)	2.02% (1.10%)	-\$4 (\$327)
Observations		9,939	10,992	10,992
Mean of Dep. \	/ar.	48.67	26.4%	\$15,912

	Dep Var:	Test Score	College Attendance	Earnings
		(1)	(2)	(3)
Small Class		<del>,</del> 4.81	2.02%	-\$4
	Estimated Impact	(1.05)	(1.10%)	(\$327)
Observations	•	9,939	10,992	10,992
Mean of Dep. Var.		48.67	26.4%	\$15,912

Estimated impact of being in a small KG class: 4.81 percentile gain in end-of-KG test score

	Dep Var:	Test Score (1)	College Attendance (2)	Earnings (3)
Small Class		4.81	2.02%	-\$4
Observations	Standard Error	9,939	(1.10%)	(\$327)
Mean of Dep. \	/ar.	48.67	26.4%	\$15,912

95% confidence interval = estimate +/-1.96 times standard error  $\rightarrow$  95% CI for test score impact = 2.71 to 6.91 percentiles

*If we repeat experiment 100 times, 95 of the 100 estimates will lie between 2.71 and 6.91 percentiles* 

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95% confidence interval: (-\$645,\$637)

→ Earnings impact could be as large as **\$637** (a 4% increase)

### **Effects of Class Size: Quasi-Experimental Evidence**

 Limitation of STAR experiment: sample is too small to estimate impacts of class size on earnings precisely

- Fredriksson et al. (2013) use administrative data from Sweden to obtain more precise estimates
  - No experiment here; instead use a quasi-experimental method: regression discontinuity

### **Regression Discontinuity Using Class Size Cutoffs**

- Sweden imposes a maximum class size of 25 students
  - School that has 26 students in a given grade will therefore have two classes of 13 students each
  - School that has 25 students may have one class of 25 students
- School that have 26 students in a grade are likely to be comparable to those that have 25 students

Can identify causal effects of class size by comparing outcomes in schools with 26 vs. 25 students in a given grade

#### **Discontinuities in Class Size Created by Maximum Class Size Rule**



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#### Test Score Achievement: Regression Discontinuity Estimates



#### **Test Score Achievement: Regression Discontinuity Estimates**



Test scores jump by 0.2 standard deviations (8 percentiles) at cutoff → Reducing class size by 5 students causes 8 percentile increase in scores

#### **Earnings Impacts: Regression Discontinuity Estimates**



### **Lessons on Class Size**

- Reducing class sizes in primary school by hiring more teachers can have large returns
  - Present value of lifetime earnings of a child growing up in a family at 25<sup>th</sup> percentile is about \$500,000 on average
  - -4% earnings gain from smaller class = \$20,000
  - Dividing a class of 30 students into two would increase total earnings of students by more than \$600,000
  - Costs (hiring another teacher and an additional room) likely to be well below \$600,000