What Are We Learning about Learning from Experiments? (It Depends.)

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Cataloging what works

Lexicographic evidence rankings Highly clustered evaluations Meta-analysis Global policy prescriptions

What have we learned?

Contract teachers Class size Private schools

Where are we going?

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Cataloging what works

Lexicographic evidence rankings

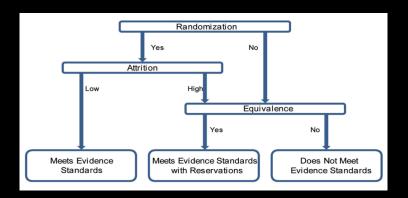
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1 of 4: Lexicographic evidence rankings



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2 of 4: Highly clustered evaluations

Caroline Fiennes, *SSIR* 2013, "Most Charities Shouldn't Evaluate Their Work"

 $impact = idea \times implementation$

"the ideas used by charities don't need to be evaluated again, because they've been amply evaluated already. ... All the charity then needs to do is run the programs well."

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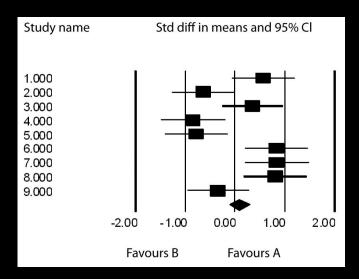
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3 of 4: Meta-analysis



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4 of 4: Global policy prescriptions

Table 2.4 Program costs						
Program (recurring annual expenditure)	Source and method of calculation	Randomized?	Cost (US\$2000 millions)			
Education		Tundonace.	minonsy			
Remedial teaching based on the Balsakhii model developed by Pratham	Source: Bunerjee et al. 2003. Cost per child-year), BalkhushiCost, from the Prathan Casto, Savahable at www.pratham.org/reports. We use the mean of costs Pune regions. Country cost = BalkashiPop * BalkhushiCost * GDPCorrection*	Yes 1 2001-2002 Rep reported for the	644 orts (cited on June 28, Dehli, Mumbai, and			
Universal education based on a 40:1 pupil-teacher ratio	Source: Angrist and Lavy 1999 ¹ Calculation: Country cost = NotInSchool × SchoolCost	No	1.544			
School inputs (uniforms and textbooks)	Source: Kremer, Moulin, and Namunyu 2003. Calculation: Unit linguasCost (p. 44). Transportation Country cost = linguasCost * SchoolAge * PPPCorrection	Yes costs ignored.	2,268			
Schooling vouchers	Source: Angrist et al. 2002. Chiculation: Assumes that everyone is sufficiently motivated to achieve satisfactory performance, hence qualifying for the vouchers, and ignoring general equilibrium effects due to the resultant increase in private school fees. Unit Pencher Cog used mor arcalculations is the increase in partial contactional expenditure per rottery winner, given in Country cost: et 4(15.5). Enthick Cog with the Country cost: et 4(15.5). Enthick Cog with Country cost: et 4(15.5). Enthick Cog with Cog with Country cost: et 4(15.5). Enthick Cog with Cog wi					
Monetary rewards to parents for sending children to school	Source: Behman, Sepupta, and Todd 2001. Calculation: Assumes that if the subsidy is large enough, everyone will we therefore everyone will get the subsidy. Unit Subsidif Cost is calculated fire 1). This is multiplied by seven because at any time there are four cohorts Country cost = (7/15) * Subsidif Cost * Child Pop * GDP Corrects*.					
Nutrition supplementation	Country Cost = (1715) * Substay Cost * ChitaPop * GDP Correction					
Iron	Source: Bobonis, Miguel, and Sharma 2004. Calculation: Unit <u>IrrorCost</u> data from Miguel and Bobonis, private commonlyst age between uges two and six. Country cost = (5/15) * <u>IrrorCost</u> * ChildPop * PPPCorrection	Yes sunication. The p	346 rogram covers five			

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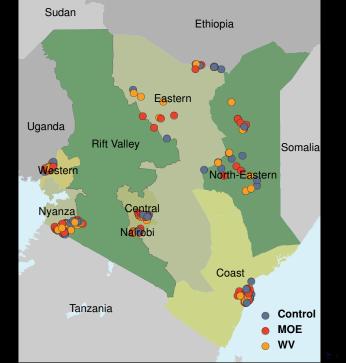
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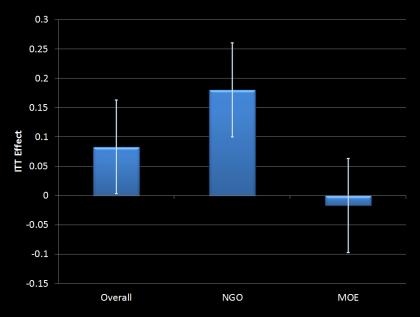
Conclusions

Contract teachers

- ► Muralidharan & Sundararaman (2008) Andhra Pradesh Contract teachers ⇒ +0.15 std. dev.
- ▶ Duflo, Dupas, & Kremer (2009) Western Kenya Contract teachers $\Rightarrow +0.21$ std. dev. Class size reduction \Rightarrow no effect on scores



Treatment Effect of Contract Teachers on Test Scores



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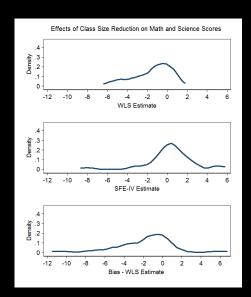
Private schools

Where are we going?

Class-size effects

		Point estimate & std. error		
			Controlling	Controlling
		No	for	for
	Country	Controls	observables	unobservables
RDD				
Angrist & Lavy (1999)	Israel	0.322	0.019	-0.261
		(.039)	(.044)	(.113)***
Urquiola (2006)	Bolivia	0.07	0.01	-0.21
		(0.03)**	-0.03	(0.07)**
Asadulla (2005)	Bangladesh		0.25	3.5
			(0.115)***	(1.03)***
RCT				
Krueger (1999)	USA			-0.271
				(.072)***
Banerjee et al (2007)	India	0.027		0.064
		(.0125)**		(.118)
Duflo et al (2012)	Kenya			-0.06 4
				(.024)**

Class-size effects: Woessman & West (2006) TIMSS data



Class-size effects: Lessons

- 1. Most striking feature is high variance, not high mean.
- 2. Heterogeneity is real (statistically significant) and affects both treatment and selection parameters.
- No clear encompassing theory to explain widely variant OLS and (quasi-) experimental results – as required for any external validity claims.

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Return to Private Schooling

		Point estimate & std. error		
			Controlling	Controlling
		No	for	for
	Country	Controls	observables	unobservables
None				
Cox & Jimenez (1991)	Colombia	0.22	0.55	
Cox & Jimenez (1991)	Tanzania	-0.14	0.97	
Aggregation Hsieh & Urquiola (2006)	Chile	-0.714 (1.188)		-0.51 (1.390)
Tabarrok (2013)	India			0.224
Bold et al (2012)	Kenya	0.79 (0.046)***		(0.036)*** 0.98 (0.41)**
RCT				
Angrist et al (2002)	Colombia		0.379 (0.111)***	0.291 (0.153)**

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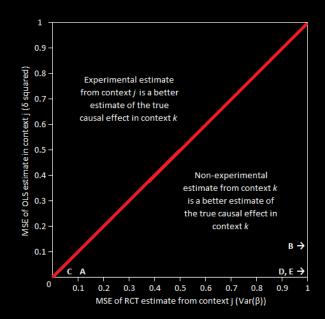
OLS 'here' vs an RCT 'there'

For non-experimental estimates:

$$\mathsf{MSE}(\tilde{\beta}_k) = \underbrace{\mathsf{Var}(\tilde{\beta}_k)}_{\substack{\mathsf{Sampling} \\ \mathsf{error}}} + \underbrace{(\tilde{\beta}_k - \beta_k)^2}_{\substack{\mathsf{Omitted} \\ \mathsf{var. \ bias}}}$$

For experimental estimates:

$$\mathsf{MSE}(\hat{\beta}_j) = \underbrace{\mathsf{Var}(\hat{\beta}_j)}_{\substack{\mathsf{Sampling error} \\ \mathsf{in context } j}} + \underbrace{\mathsf{Var}(\beta)}_{\substack{\mathsf{Variance of true effect} \\ \mathsf{across contexts}}}$$



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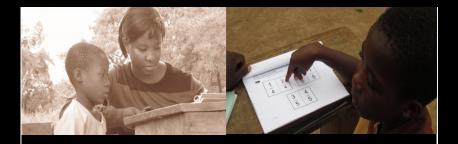
Alternative evaluation models

Straw man model?

Define an intervention defined at school level; evaluate in a controlled context; extrapolate.

Alternative models:

- Evaluate scale-up within large programs (Duflo & Kiessel 2013, Ghana TCAI).
- Mechanism experiments (Ludgwig, Kling, and Mullainathan) applied within organizations to production of outputs.
- ► Move up the bureaucratic supply chain (Rasul & Roger 2013, Nigerian Civil Service).



Teacher Community Assistants Initiative

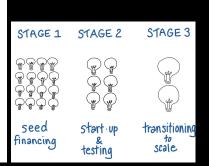
A project of the Ghana Education Service in collaboration with:

Ghana National Association of Teachers (GNAT) National Youth Employment Program (NYEP) Innovations for Poverty Action (IPA)



Development Innovation Ventures (DIV) @ USAID





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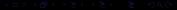
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Conclusions

For evaluators:

- Specify context: Intervention may be both bigger and narrower than assumed.
- ► Encompass rather than trump: OLS estimates are facts demanding explanation.

For evaluation users:

- Avoid lexicographic preference for internal over external validity.
- Commission more, faster, cheaper RCTs with fewer grand ambitions.