Outsourcing Service Delivery in a Fragile State: Experimental Evidence from Liberia

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How to improve service delivery in fragile states?

Give money

▶ Bottleneck imposed by state capacity → Standard development aid is usually least effective in these places (Burnside & Dollar, 2000; Collier & Dollar, 2002)

Build state capacity

- Hard and slow. Efforts to build stronger institutions often fail (Pritchett & Woolcock, 2004)
- Outsourcing provision to sidestep "poor governance"
 - Private management better than public (Bloom & Van Reenen, 2010; Bloom, Sadun, & Van Reenen, 2015)

Theoretical and empirical analyses of outsourcing suggest caution

 Contractors have incentives to cut quality on non-contracted/non-monitored processes/outcomes (Hart, Shleifer, & Vishny, 1997)

Empirically

 Better outcomes in some cases (e.g., water services in Argentina (Galiani, Gertler, & Schargrodsky, 2005))

Failed in others (e.g., prisons in the U.S. (Useem & Goldstone, 2002))

PPP aims to overcome efficiency/equity trade-off

 Efficiency: Private schools are on average better managed than public schools (Bloom, Lemos, Sadun, & Van Reenen, 2015; Muralidharan & Sundararaman, 2015)

 Equity: Fee-charging private schools may increase inequality and sorting (Hsieh & Urquiola, 2006; Lucas & Mbiti, 2012; Zhang, 2014)

 "Solution": Public funding with private management (and restrictions on selection)

This paper: RCT across 185 schools in Liberia

- Outsource management of public schools to private providers
 - ▶ 7% of public primary schools and 12% of students
 - Randomly assign treatment at the school level (matched-pairs)
 - Sample students from enrollment records prior to treatment

This paper: RCT across 185 schools in Liberia

- Outsource management of public schools to private providers
 - 7% of public primary schools and 12% of students
 - Randomly assign treatment at the school level (matched-pairs)
 - Sample students from enrollment records prior to treatment
- Treatment increases test scores by .19 σ after 1 year
- Heterogeneity by provider: Highest= 0.26σ , lowest=0
- Contracts matter
 - Removal of students from over-subscribed schools
 - Dismissal of under-performing teachers
- Results reflect additional inputs and different management
 Non-experimental mediation analysis suggest both mattered

1. Contracting out public services

- ► Theory: Hart et al. (1997)
- Empirically: Water (Galiani et al., 2005), health (Loevinsohn & Harding, 2005), education (Aslam, Rawal, & Saeed, 2017)
- \Rightarrow Contracting out management of public schools

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- 2. Charter schools
 - ▶ Lotteries to overcome endogeneity: Chabrier et. al. (16)
 - ▶ Truncated distribution of estimated effects: Tuttle et. al. (12)
 - ⇒ Treatment effects: Across all schools and providers

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4. Service delivery

- State capacity: (Ladner & Persson, 2009; Besley & Persson, 2010; Muralidharan et al., 2016)
- \Rightarrow Outsourcing in the absence of state capacity

Introduction

Context: Low learning & a weak state

The experiment: Private management of public schools

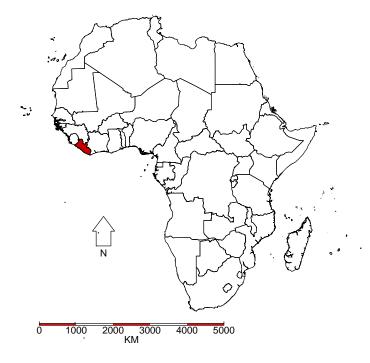
Results

Introduction

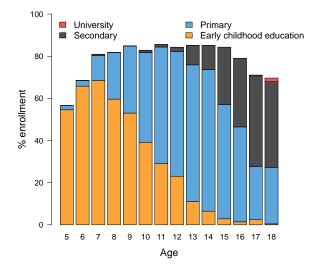
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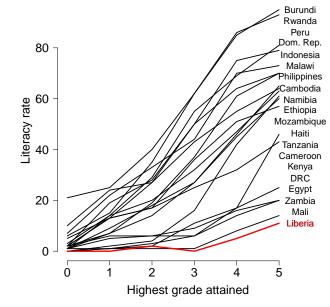


Low enrollment and backlog of overage children



Note: Based on 2014 Household Income and Expenditures Survey.

Schooling \neq learning



Source: Oye, Pritchett, and Sandefur (2016)

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Liberia is outsourcing education. Can it work?



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Liberia is outsourcing education. Can it work?

The Economist

Ashes to classes Liberia's bold experiment in school reform

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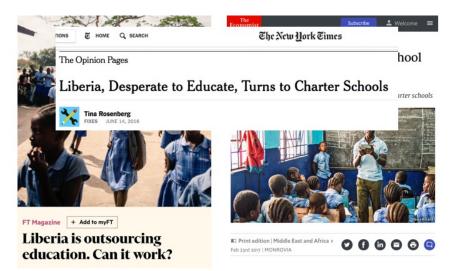
A war-scorched state where almost nothing works tries charter schools



C Print edition | Middle East and Africa > Feb 23rd 2017 | MONROVIA

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🚨 Welcome 🗉



What are "Partnership Schools for Liberia"?





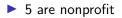
non-selective

- staffed by teachers on government payroll
- and managed by 8 private contractors

with a \$50 per pupil subsidy (+ fundraising)

More details What do contractors do? How does this compare to other PPPs?

8 Private providers





▶ 6 were contracted through competitive tender

• More details • What do contractors do? • How does this compare to other PPPs?



Uncover implementer-specific treatment effects



Experimental details

Randomize at the school level

Uncover implementer-specific treatment effects





► Endogenous participation in lotteries → population TE

Uncover implementer-specific treatment effects

Endogenous sorting of implementers into regions

Randomize within the pairs that each implementer accepted



Uncover implementer-specific treatment effects





► Endogenous participation in lotteries → population TE

Uncover implementer-specific treatment effects

Endogenous sorting of students (supply or demand driven)

Sample students from enrollment records prior to treatment

Track them and assign to original schools for analysis (ITT)

School: Facilities, expenditure, management.

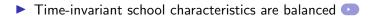
 Teachers: Socio-demographic, qualifications/experience, knowledge test.

Classroom observations: Stallings

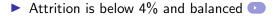
Students: 20 students across all grades (2015/2016 log)

Households: Survey 7-10 HH (only follow ups)

Time-invariant characteristics are balanced and attrition is low



Time-invariant student characteristics are balanced



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Empirical specification

$$Y_{isp} = \alpha_p + \beta T_s + \varepsilon_{isp}$$
(1)
$$Y_{isp} = \alpha_p + \beta T_s + \gamma X_i + \delta Z_s + \varepsilon_{isp}$$
(2)

- Y_{isp}: Outcome for student i in school s and pair p
- α_p : Matched-pair fixed effects (stratification dummies)
- ► *T_s*: Treatment dummy
- ► X_i: Student level controls
- ► Z_s : School level controls show me
- Clustered s.e. at the school-level

Introduction

Context: Low learning & a weak state

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Results

Test scores

Learning gains varied by provider Contracting details matter What explains learning gains?

Test design

- One-on-one tests
 - Literacy cannot be assumed at any grade level
 - Prevents cheating
- Single adaptive test for all students
 - Capture a wide range of abilities
 - Comparability across grades
- Item response theory (IRT) to estimate student's ability
 - Widely used and "best practice" (GRE, SAT, PISA, TIMSS)
 - Simultaneously estimate the test taker's ability and the difficulty of the questions
 - Comparability across students (if overlap is imperfect)
- Normalize the scores with respect to the control group

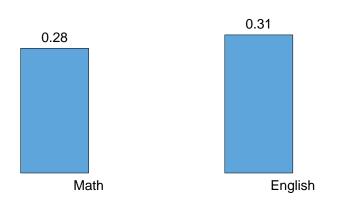
Test scores increased by $.19\sigma$

	One year follow-up		
	Difference	Difference	Difference
	<i>.</i>	(F.E.)	(F.E. + Controls)
	(1)	(2)	(3)
English	0.17**	0.17***	0.18***
	(0.08)	(0.04)	(0.03)
Math	0.17***	0.19***	0.18***
	(0.07)	(0.04)	(0.03)
Composite	0.17**	0.19***	0.19***
	(0.07)	(0.04)	(0.03)
Observations	3,492	3,492	3,492

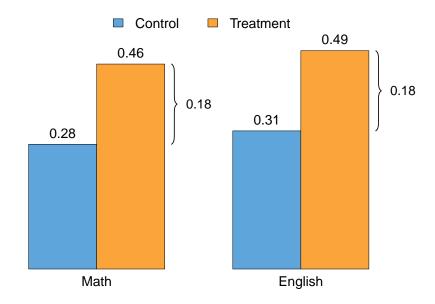
▶ Teaching to the test? → First wave → Timing → Without Bridge

"Business as usual" learning is \sim 0.3 σ per academic year





Treatment is roughly ~ 0.62 extra years of schooling



Other outcomes

No heterogeneity by school characteristics

No heterogeneity by student characteristics

No evidence of student selection •

No effect on enrollment (more on this soon)

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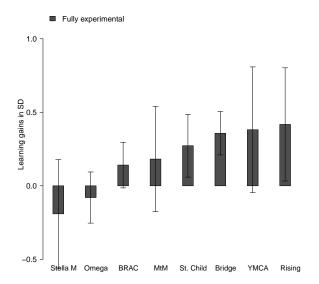
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Test scores

Learning gains varied by provider

Contracting details matter What explains learning gains?

Final thoughts



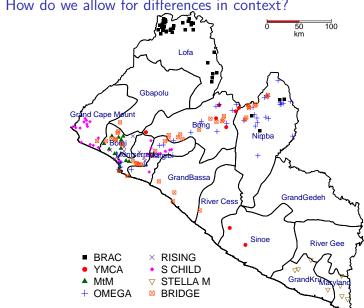
Two problems when comparing providers

1. They work in different contexts

Raw estimates for each provider are correct (internal validity)

But they aren't immediately comparable (external validity)

2. Sample sizes for most providers are small



1. How do we allow for differences in context?

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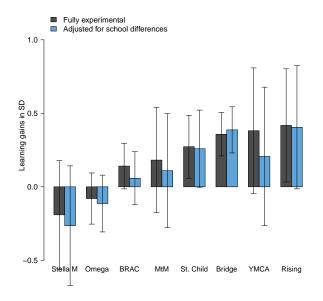
Control for school characteristics

Some schools will score better for reasons unrelated to treatment

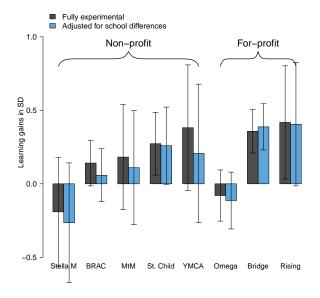
Control for the interactions of characteristics with treatment

 Raising scores through treatment will be easier in some contexts

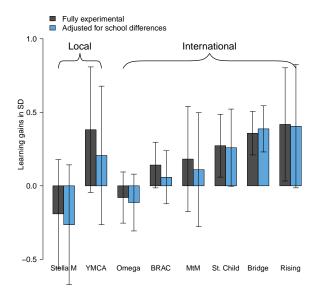
▶ Controls













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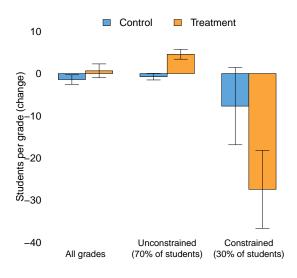
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All contractors allowed to cap class sizes

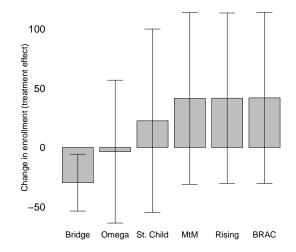
 Largest provider bypassed the competitive procurement and negotiated a bilateral agreement

Lump-sum grants (as opposed to per-pupil funding)

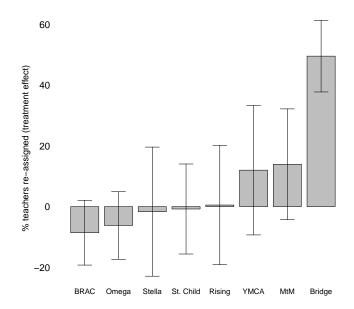
 Limitations on removing government teachers verbally stipulated (as opposed to written in the contract) No effect on total enrollment, but in constrained schools enrollment went down



Removing students from schools where class sizes were large



Removing incumbent teachers



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What explains learning gains?

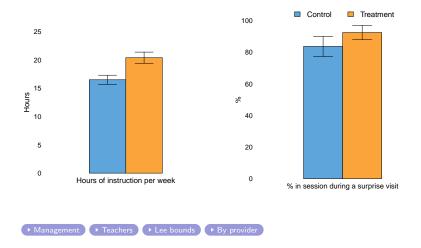


 Which changes mattered for learning outcomes? (Non-experimental) What explains learning gains?

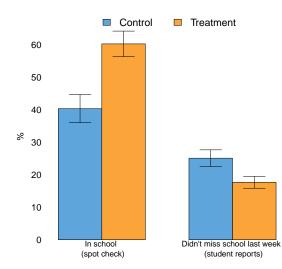
What changed? (Experimental)

 Which changes mattered for learning outcomes? (Non-experimental)

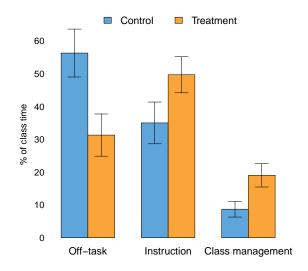
Treatment schools have more instructional time



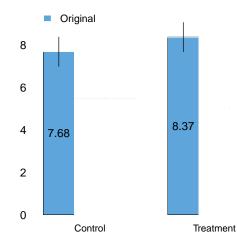
Teachers are more likely to be in school...



...and quality of instruction is higher

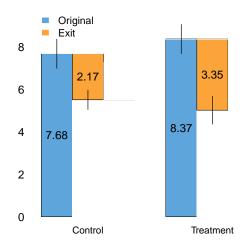


Teachers per school: baseline, entry, and exit

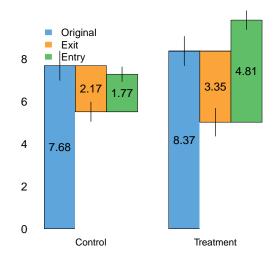


→ Table) (→ By provider

Teachers per school: baseline, entry, and exit



Teachers per school: baseline, entry, and exit



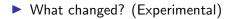
→ Table → By provider

Treatment schools get new teaching graduates

	(1) Treatment	(2) Control	(3) Difference	(4) Difference (F.E)
Age in years	39.09	46.37	-7.28***	-7.10***
	(11.77)	(11.67)	(1.02)	(0.68)
Experience in years	10.59	15.79	-5.20***	-5.26***
	(9.20)	(10.77)	(0.76)	(0.51)
% has worked at a private school	47.12	37.50	9.62**	10.20***
	(49.95)	(48.46)	(3.76)	(2.42)
Test score in standard deviations	0.13	-0.01	0.14*	0.14**
	(1.02)	(0.99)	(0.07)	(0.06)



What explains learning gains?



 Which changes mattered for learning outcomes? (Non-experimental)

How to choose relevant mediators?

- "Normally impossible to measure all possible mediators. Indeed, it may be impossible to merely *think* of all possible mediators" Bullock and Ha (2011)
- Adding an exhaustive list reduces degrees of freedom
- "Double Lasso" to selects relevant controls (Belloni, Chernozhukov, & Hansen, 2014; Urminsky, Hansen, & Chernozhukov, 2016)

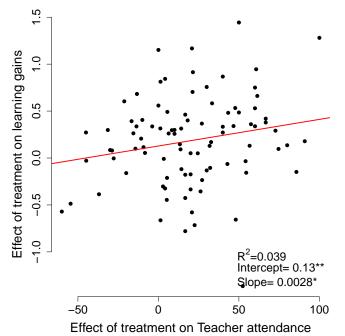
Lasso #1: What is highly correlated with learning?

Lasso #2: What did the experiment change?

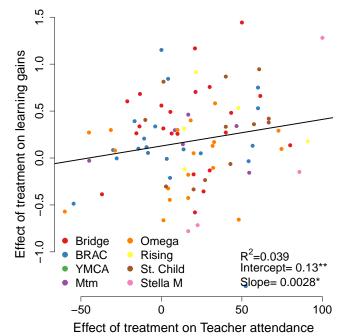
Mediator

Teachers' age Teacher attendance Hrs/week Teachers' Experience % time management

Where teacher attendance increases, so do test scores



Where teacher attendance increases, so do test scores



Correlation between treatment effects at the match-pair level

Variable	Learning
Teachers' age	-0.37***
Teacher attendance	0.20*
Teachers' experience	-0.16
Hours/Week	0.15
% time management	0.057

Decomposition of the treatment effect: Half is management

Mediator	% of total treatment effect
Teachers' age	60.77%
Teacher attendance	15.43%
Hrs/week	14.70%
Teachers' Experience	-13.51%
% time management	3.59%
Direct	19.02%



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Can outsourcing public education raise learning levels in fragile states?

- .19 σ ~0.62 extra years of schooling
- Highest performing= 0.26σ , lowest=0
- Largest provider unenrolled pupils from schools with large class sizes and removed 74% of incumbent teachers
- Questions regarding contracts/procurement
 - Broad statements about PPP may be simplistic
 - Managing/contracting providers requires some state capacity
 - Dynamic contracting where apriori quality is unknown
 - Contracts are incomplete and subject to regulatory capture
 - Mission alignment (Besley & Ghatak, 2005)
 - Competition requires active encouragement

Thank you



Asante Sana







PSL and traditional public schools

	Control schools	PSL treatment schools
Management		
Who owns school building?	Government	Government
Who employs and pays teachers?	Government	Government
Who manages the school and teachers?	Government	Provider
Who sets curriculum?	Government	${\sf Government} + {\sf provider \ supplement}$
Funding		
Primary user fees (annual USD)	Zero	Zero
ECE user fees (annual USD)	\$38	Zero
Extra funding per pupil (annual USD)	NA	50 + independent fund-raising
Staffing		
Pupil-teacher ratios	NA	Promised one teacher per grade,
		allowed to cap class sizes at 45-65 pupils
New teacher hiring	NA	First pick of new teacher-training graduates

Back

	Liberia PSL	South Africa	UK Academy	USA Charters	Punjab PSSP	Punjab vouchers	Philippines vouchers	India RTE	Uganda Secondary
Year started	2016	2016	2001	1991	2016	2006	2005	2012	2007
# Schools	93	7	5,000	7,000	500	1,700	c. 6,000	91,000	800
# Students	27,000	6,000	2million+	2.7million	c. 50,000	500,000	c. 1million	c. 1.7mill	440,000
Туре	Contract Mgmt	Contract Mgmt	Contract Mgmt	Contract Mgmt	Contract Mgmt	Voucher	Voucher	Subsidy	Subsidy
No fee?	~	\checkmark	\checkmark	\checkmark	>	×	×	>	×
Non-profit?	×	\checkmark	\checkmark	-	>	×	×	~	×
Non-selective?	~	\checkmark	\checkmark	\checkmark	>	 ✓ 	x	×	×
Govt teacher contracts	~	-	-	×	×	×	×	×	×
Teachers in unions	✓	\checkmark	~	×	×	×	x	×	×
Accountable for outcomes	~	\checkmark	~	\checkmark	~	~	\checkmark	×	×
National curriculum	~	\checkmark	×	-	~	~	\checkmark	~	~
Govt buildings	✓	\checkmark	~	-	>	×	×	×	×
	More public					•			More private



What do providers do? Depends on the provider

Textbooks/Paper/Notebook: YMCA/BRAC/MtM

Technology (e.g., scripted lessons in tablets): Bridge/Omega

Community engagement: MtM/Rising/St Child

Teacher training: Rising

Teacher guides: Rising/MtM/Bridge

Back
 More details

What do providers do? Depends on the provider

		Provider							
t		Stella M	YMCA	Omega	BRAC	Bridge	Rising	St. Child	MtM
bdd	Provider staff visits at least once a week(%)	0	54	13	93	76	94	91	96
-Su	Heard of PSL(%)	42	85	61	42	87	90	68	85
der	Heard of provider(%)	46	96	100	95	100	100	100	100
Provider Support	Has anyone from (provider) been to this school?(%)	42	88	100	94	100	100	99	100
۵.	T	10	00	70	0.4		74	0.1	0.0
	Textbooks(%)	12	96	73	94	99	71	94	96
	Teacher training(%)	0	77	62	85	87	97	93	96
70	Teacher received training since Aug 2016(%)	23	46	58	45	50	81	58	37
Ever provided	Teacher guides (or teacher manuals)(%)	0	69	75	54	97	94	68	98
20	School repairs(%)	0	12	25	24	53	52	13	93
声	Paper(%)	0	92	30	86	70	97	88	98
Š	Organization of community meetings(%)	0	54	27	69	73	87	83	91
	Food programs(%)	0	8	2	1	1	10	0	17
	Copybooks(%)	4	65	30	92	18	97	94	91
	Computers, tablets, electronics(%)	0	0	94	0	99	3	3	2
		-							
	Provide/deliver educational materials(%)	0	4	45	17	18	26	29	50
	Observe teaching practices and give suggestions(%)	0	19	45	81	65	45	74	85
sit	Monitor/observe PSL program(%)	0	12	23	11	13	13	35	65
Ę	Monitor other school-based government programs(%)	0	0	7	5	10	6	18	9
cen	Monitor health/sanitation issues(%)	0	8	9	2	5	0	10	28
Most recent visit	Meet with PTA committee(%)	0	12	8	10	7	0	21	41
los	Meet with principal(%)	0	12	54	36	38	6	51	63
2	Deliver information(%)	0	12	36	16	8	6	16	35
	Check attendance and collect records(%)	42	23	43	56	39	19	66	70
	Ask students questions to test learning(%)	4	4	24	33	18	58	44	43

Schools in the RCT are better than the average public school in the country $% \left({{{\rm{CT}}_{\rm{sch}}} \right)$

	(1) RCT (Treatment and control)	(2) Other public schools	(3) Difference
Students: ECE	142.68	112.71	29.97***
	(73.68)	(66.46)	(5.77)
Students: Primary	151.55	132.38	19.16*
	(130.78)	(143.57)	(10.18)
Students	291.91	236.24	55.67***
	(154.45)	(170.34)	(12.15)
Classrooms per 100 students	1.17	0.80	0.37***
	(1.63)	(1.80)	(0.13)
Teachers per 100 students	3.04	3.62	-0.58**
	(1.40)	(12.79)	(0.28)
Textbooks per 100 students	99.21	102.33	-3.12
	(96.34)	(168.91)	(7.88)
Chairs per 100 students	20.71	14.13	6.58***
	(28.32)	(51.09)	(2.38)
Food from Gov or NGO	0.36	0.30	0.06
	(0.48)	(0.46)	(0.04)
Solid building	0.36	0.28	0.08*
	(0.48)	(0.45)	(0.04)
Water pump	0.62	0.45	0.17***
	(0.49)	(0.50)	(0.04)
Latrine/toilet	0.85	0.71	0.14***
	(0.33)	(0.45)	(0.03)
Observations	185	2,420	2,605



Time-invariant characteristics are balanced and attrition is low

Tr	(1) reatment	(2) Control	(3) Difference	(4) Difference (F.E)
	-0.003	-0.080	-0.077	-0.070
((0.169)	(0.156)	(0.230)	(0.232)
	14.130	13.978	-0.152	0.000
((3.652)	(3.615)	(5.138)	(5.094)
;	80.435	79.570	-0.865	-0.361
((4.159)	(4.204)	(5.913)	(4.705)
(68.043	75.129	7.086	7.079
((6.308)	(7.165)	(9.547)	(8.774)
_	(6.308) MIS	`	,	, (,

Time-invariant characteristics are balanced and attrition is low

	(1) Treatment	(2) Control	(3) Difference	(4) Difference (F.E)
Age in years	12.292	12.390	0.098	0.052
	(0.070)	(0.068)	(0.169)	(0.112)
% male	56.253	54.825	-1.427	-1.720
	(1.185)	(1.192)	(2.048)	(1.269)
Wealth index	0.025	-0.006	-0.031	0.010
	(0.037)	(0.037)	(0.140)	(0.060)
% in top wealth quartile	0.219	0.199	-0.020	-0.017
	(0.010)	(0.010)	(0.026)	(0.014)
% in bottom wealth quartile	0.284	0.266	-0.018	-0.012
	(0.011)	(0.011)	(0.039)	(0.019)
ECE before grade 1	0.820	0.834	0.014	0.013
-	(0.009)	(0.009)	(0.025)	(0.017)

Time-invariant characteristics are balanced and attrition is low

	(1)	(2)	(3)	(4)		
	Treatment	Control	Difference	Difference (F.E)		
% interviewed	96.01	95.98	-0.03	-0.23		
	(0.47)	(0.47)	(0.63)	(0.44)		
 First wave Back 						

Balance using EMIS data

	(1) Treatment	(2) Control	(3) Difference	(4) Difference (F.E)
Students: ECE	148.51	136.72	11.79	11.03
	(76.83)	(70.24)	(10.91)	(9.74)
Students: Primary	159.05	143.96	15.10	15.68
	(163.34)	(86.57)	(19.19)	(16.12)
Students	305.97	277.71	28.26	27.56
	(178.49)	(124.98)	(22.64)	(19.46)
Classrooms per 100 students	1.21	1.13	0.09	0.08
	(1.62)	(1.65)	(0.24)	(0.23)
Teachers per 100 students	3.08	2.99	0.09	0.09
	(1.49)	(1.30)	(0.21)	(0.18)
Textbooks per 100 students	102.69	95.69	7.00	7.45
	(97.66)	(95.40)	(14.19)	(13.74)
Chairs per 100 students	18.74	22.70	-3.96	-4.12
	(23.06)	(32.81)	(4.17)	(3.82)
Food from Gov or NGO	0.36	0.36	-0.01	-0.01
	(0.48)	(0.48)	(0.08)	(0.05)
Solid building	0.39	0.33	0.06	0.06
	(0.49)	(0.47)	(0.07)	(0.06)
Water pump	0.56	0.67	-0.11	-0.12*
	(0.50)	(0.47)	(0.07)	(0.06)
Latrine/toilet	0.85	0.86	-0.01	-0.01
	(0.35)	(0.32)	(0.05)	(0.05)
Observations	92	93	185	185



First wave sampling

	(1) Treatment	(2) Control	(3) Difference	(4) Difference (F.E
Number of students sampled	24.8	24.6	0.13	0.035
	(5.74)	(5.10)	(0.81)	(0.81)
Found at the school	18.2	16.7	1.49***	1.555***
	(2.30)	(4.70)	(0.55)	(0.54)
Found at home	1.73	2.91	-1.18**	-1.223**
	(2.12)	(3.97)	(0.48)	(0.47)
Interviewed	19.8	19.5	0.30	0.320
	(0.83)	(2.18)	(0.25)	(0.26)
Observations	88	90	178	171

▶ Back

Control variables

Student controls

Wealth index Age Gender Grade (2015/2016)

School controls

Enrollment (2015/2016) Infrastructure quality (2015/2016) Travel time to nearest bank Rurality

▶ Back

Test scores increased by .19 σ

	One year follow-up					
	Difference	Difference (F.E.)	Difference (F.E. + Controls)			
	(1)	(2)	(3)			
English	0.17**	0.17***	0.18***			
	(0.08)	(0.04)	(0.03)			
Math	0.17***	0.19***	0.18***			
	(0.07)	(0.04)	(0.03)			
Composite	0.17**	0.19***	0.19***			
	(0.07)	(0.04)	(0.03)			
New modules	0.17**	0.20***	0.19***			
	(0.07)	(0.04)	(0.04)			
Conceptual	0.12**	0.13***	0.12***			
	(0.05)	(0.04)	(0.04)			
Observations	3,492	3,492	3,492			

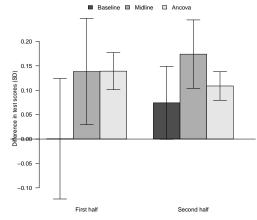
Without Bridge: Test scores increased by 0.1σ

	One year follow-up						
	Difference	Difference (F.E.)	Difference (F.E. + Controls)				
	(1)	(2)	(3)				
English	0.13*	0.13***	0.13***				
	(0.07)	(0.04)	(0.03)				
Math	0.09	0.09**	0.10***				
	(0.06)	(0.04)	(0.03)				
Abstract	0.03	0.03	0.04				
	(0.05)	(0.03)	(0.04)				
Composite	0.10	0.10***	0.11***				
	(0.07)	(0.04)	(0.03)				
New modules	0.17**	0.20***	0.19***				
	(0.07)	(0.04)	(0.04)				
Conceptual	0.12**	0.13***	0.12***				
	(0.05)	(0.04)	(0.04)				
Observations	2,643	2,643	2,643				



First round of data is "contaminated" by short-run treatment effects

Test scores (all questions)







PPP increased test scores by .19 σ

	Bas	seline	One year follow-up					
	Difference	Difference (F.E.)	Difference	Difference (F.E.)	Difference (F.E. + Controls)	Difference (ANCOVA)		
	(1)	(2)	(3)	(4)	(5)	(6)		
English	0.05	0.09*	0.17**	0.17***	0.18***	0.13***		
	(0.08)	(0.05)	(0.08)	(0.04)	(0.03)	(0.02)		
Math	0.08	0.08*	0.17***	0.19***	0.18***	0.14***		
	(0.07)	(0.04)	(0.07)	(0.04)	(0.03)	(0.02)		
Composite	0.07	0.08*	0.17**	0.19***	0.19***	0.14***		
	(0.07)	(0.05)	(0.07)	(0.04)	(0.03)	(0.02)		
New modules			0.17**	0.20***	0.19***	0.16***		
			(0.07)	(0.04)	(0.04)	(0.03)		
Conceptual			0.12**	0.13***	0.12***	0.10***		
			(0.05)	(0.04)	(0.04)	(0.04)		
Observations	3,496	3,496	3,492	3,492	3,492	3,492		

▶ Back ▶ Without Bridge

Without Bridge: PPP increased test scores by .19 σ

	Bas	seline	One year follow-up				
	Difference	Difference (F.E.)	Difference	Difference (F.E.)	Difference (F.E. + Controls)	Difference (ANCOVA)	
	(1)	(2)	(3)	(4)	(5)	(6)	
English	-0.00	0.03	0.13*	0.13***	0.13***	0.11***	
	(0.09)	(0.04)	(0.07)	(0.04)	(0.03)	(0.03)	
Math	0.00	-0.00	0.09	0.09**	0.10***	0.09***	
	(0.07)	(0.04)	(0.06)	(0.04)	(0.03)	(0.02)	
Abstract	0.01	0.02	0.03	0.03	0.04	0.04	
	(0.07)	(0.05)	(0.05)	(0.03)	(0.04)	(0.03)	
Composite	-0.00	0.01	0.10	0.10***	0.11***	0.10***	
	(0.08)	(0.04)	(0.07)	(0.04)	(0.03)	(0.02)	
New modules			0.17**	0.20***	0.19***	0.16***	
			(0.07)	(0.04)	(0.04)	(0.03)	
Conceptual			0.12* [*]	0.13***	0.12***	0.10***	
			(0.05)	(0.04)	(0.04)	(0.04)	
Observations	2,654	2,654	2,643	2,643	2,643	2,643	



No heterogeneity by student characteristics

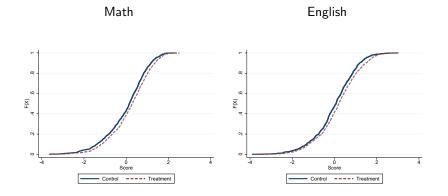
(

	Male (1)	Top wealth quartile (2)	Bottom wealth quartile (3)	Grade (4)
Treatment	0.20***	0.18***	0.17***	0.16
	(0.047)	(0.035)	(0.035)	(0.10)
Treatment \times covariate	-0.021	0.030	0.061	0.0050
	(0.068)	(0.066)	(0.050)	(0.020)
No. of obs.	3,492	3,492	3,492	3,492
• Distribution plots 🔶 Quai	ntile treatme	nt effects 🚺 Without Bi	ridge 🕩 Back	

Quantile treatment effects

Quantile	0.1	0.25	0.5	0.75	0.9			
	(1)	(2)	(3)	(4)	(5)			
Panel A: Math								
Treatment	0.24***	0.22***	0.16***	0.13***	0.24***			
	(0.037)	(0.026)	(0.027)	(0.024)	(0.022)			
No. of obs.	3,492	3,492	3,492	3,492	3,492			
Panel B: Er	Panel B: English							
Treatment	0.14***	0.14***	0.16***	0.18***	0.20***			
	(0.018)	(0.024)	(0.019)	(0.021)	(0.020)			
No. of obs.	3,492	3,492	3,492	3,492	3,492			
Panel C: Co	omposite							
Treatment	0.22***	0.21***	0.16***	0.18***	0.22***			
	(0.031)	(0.026)	(0.023)	(0.022)	(0.023)			
No. of obs.	3,492	3,492	3,492	3,492	3,492			

Distribution plots



→ Quantile → Back

Without Bridge: No heterogeneity by student characteristics

	Male (1)	Top wealth quartile (2)	Bottom wealth quartile (3)	Grade (4)
Treatment	0.12**	0.10**	0.11***	0.15
	(0.050)	(0.040)	(0.040)	(0.11)
Treatment \times covariate	0.0028	0.071	0.039	-0.0079
	(0.071)	(0.069)	(0.060)	(0.020)
No. of obs.	2,643	2,643	2,643	2,643

No heterogeneity by school characteristics

	Enrollment(2015/2016) (1)	Facilities (PCA) (2)	Rural (3)	Time to nearest bank (4)
Treatment	0.16*	0.19***	0.24***	0.18***
	(0.093)	(0.032)	(0.050)	(0.050)
$Treatment \times covariate$	0.000069	-0.020	-0.068	0.000061
	(0.00029)	(0.028)	(0.077)	(0.00058)
No. of obs.	3,492	3,492	3,492	3,492
▶ Without Bridge → Bag	ck			

No heterogeneity by school characteristics

31*** 0.11)	0.11***	0.24***	0.043
0 11)	(0,020)	(0.050)	(0.0-0)
0.11)	(0.038)	(0.050)	(0.056)
00069**	-0.031	-0.14*	0.00093
00035)	(0.034)	(0.078)	(0.00061)
2,643	2,643	2,643	2,643
(00035)	00035) (0.034)	00035) (0.034) (0.078)

No evidence of student selection

(1)	(2)	(3)
Same school	Same school	Same schoo
0.061	0.012	0.021
(0.082)	(0.026)	(0.019)
-0.0042		
(0.0064)		
	-0.011	
	(0.028)	
		-0.0061
		(0.011)
3,487	3,487	3,428
	Same school 0.061 (0.082) -0.0042 (0.0064)	Same school Same school 0.061 0.012 (0.082) (0.026) -0.0042 -0.011 (0.0064) -0.011

Without Bridge: No evidence of student selection

	(1)	(2)	(3)
	Same school	Same school	Same school
Treatment	0.14*	0.029	0.030*
	(0.071)	(0.021)	(0.016)
Treatment imes Age	-0.0089*		
	(0.0054)		
Treatment \times Male		-0.011	
		(0.025)	
Treatment \times Asset Index (PCA)			0.0013
			(0.010)
No. of obs.	2,638	2,638	2,632

Controls

 School characteristics (per PAP): Enrollment (2015/2016), Facilities (PCA), Rural, Time to nearest bank



Robust to including different controls

Bayesian hierarchical model

$$Y_{isgc} = \alpha_g + \beta_c treat_s + \varepsilon_{isgc}$$
(3)

$$\widehat{\beta}_c \sim N(\beta_c, \sigma_c^2)$$
 (4)

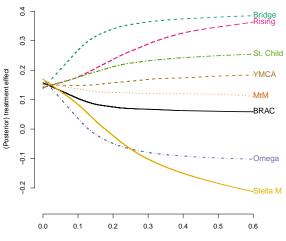
The Bayesian hierarchal model further assumes that

$$\beta_c \sim N(\mu, \tau^2)$$
 (5)

1. Place a prior distribution over μ and τ^2

2. Estimate the posterior distribution of β_c

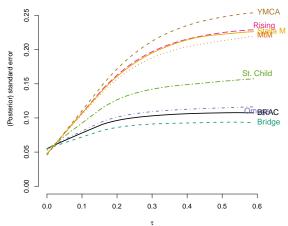
Prior choice
 Posteriors
 Back



τ

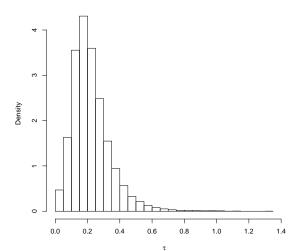
Back

Posterior standard errors by provider for different values of $\boldsymbol{\tau}$





Posterior distribution of $\boldsymbol{\tau}$



Posterior treatment effects and standard errors for different prior

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	BRAC	Bridge	YMCA	MtM	Omega	Rising	St. Child	Stella M
Flat prior	0.080	0.329***	0.126	-0.037	0.242	0.210	-0.026	0.159
	(0.098)	(0.097)	(0.162)	(0.114)	(0.176)	(0.130)	(0.187)	(0.180)
Cauchy (0,25)	0.080	0.329***	0.127	-0.037	0.241	0.209	-0.025	0.160
	(0.098)	(0.097)	(0.162)	(0.114)	(0.176)	(0.130)	(0.186)	(0.180)
Half-normal	0.081	0.327***	0.127	-0.035	0.241	0.208	-0.023	0.160
	(0.097)	(0.097)	(0.161)	(0.114)	(0.175)	(0.128)	(0.186)	(0.178)
Half-t(4)	0.080	0.327***	0.127	-0.035	0.239	0.208	-0.022	0.160
	(0.098)	(0.097)	(0.160)	(0.114)	(0.175)	(0.128)	(0.184)	(0.178)

▶ Back

No effect on total enrollment, but attendance increases

	(1)	(2)	(3)	(4)
	Treatment	Control	Difference	Difference (F.E)
Panel A: School level data (N =	175)			
Enrollment 2015/2016	298.45	264.11	34.34	34.18*
	(169.74)	(109.91)	(21.00)	(20.28)
Enrollment 2016/2017	309.71	252.75	56.96***	56.89***
	(118.96)	(123.41)	(18.07)	(16.29)
15/16 to $16/17$ enrollment change	11.55	-6.06	17.61	24.60*
	(141.30)	(82.25)	(17.19)	(14.35)
Attendance % (spot check)	48.02	32.84	15.18***	15.56***
	(24.52)	(26.54)	(3.81)	(3.13)
% of students with disabilities	0.59	0.39	0.20	0.21
	(1.16)	(0.67)	(0.14)	(0.15)
Panel B: Student level data (N =	= 3,627)			
% enrolled in the same school	80.74	83.34	-2.61	0.79
	(39.45)	(37.27)	(3.67)	(2.07)
% enrolled in school	94.14	94.00	0.14	1.22
	(23.49)	(23.76)	(1.33)	(0.87)
Days missed, previous week	0.85	0.85	-0.00	-0.06
	(1.42)	(1.40)	(0.10)	(0.07)

Without Bridge: No effect on total enrollment, but attendance increases

	(1)	(2)	(3)	(4)
	Treatment	Control	Difference	Difference (F.E)
Panel A: School level data (N =	136)			
Enrollment 2015/2016	263.25	247.94	15.31	15.31
	(97.38)	(92.57)	(16.06)	(14.83)
Enrollment 2016/2017	293.26	246.17	47.09**	46.68**
	(121.13)	(112.15)	(19.87)	(18.55)
15/16 to $16/17$ enrollment change	29.45	-0.67	30.12**	30.19**
	(82.40)	(67.50)	(12.82)	(12.09)
Attendance % (spot check)	48.11	29.80	18.31***	18.58***
	(26.30)	(26.01)	(4.45)	(3.49)
% of students with disabilities	0.66	0.41	0.25	0.26
	(1.27)	(0.73)	(0.18)	(0.18)
Panel B: Student level data (N =	= 2,775)			
% enrolled in the same school	85.26	83.74	1.52	2.97**
	(35.47)	(36.91)	(2.34)	(1.46)
% enrolled in school	94.97	95.06	-0.09	0.71
	(21.86)	(21.67)	(1.11)	(0.69)
Days missed, previous week	0.68	0.86	-0.18*	-0.19***
	(1.25)	(1.43)	(0.09)	(0.07)



No effect on total enrollment, but in constrained schools, enrollment went down

	(1) Δ enrollment	(2) % same school	(3) % in school	(4) Test scores
Constrained=0 \times Treatment	5.30***	4.04***	1.64**	0.15***
	(1.11)	(1.39)	(0.73)	(0.034)
$Constrained{=}1 imes Treatment$	-11.7*	-12.8	0.070	0.35***
	(6.47)	(7.74)	(4.11)	(0.11)
No. of obs.	1,635	3,625	3,485	3,490
Mean control (Unconstrained)	-0.75	82.09	93.38	0.13
Mean control (Constrained)	-7.73	84.38	94.81	-0.08
$\alpha_0 = \text{Constrained} - \text{Unconstrained}$	-17.05	-16.79	-1.57	0.20
p-value ($H_0: lpha_0 = 0$)	0.01	0.03	0.71	0.07

→ Without Bridge → Back

Without Bridge: Enrollment goes up

	(1) Δ enrollment	(2) % same school	(3) % in school	(4) Test scores
Constrained=0 \times Treatment	3.09***	3.94***	1.42*	0.12***
	(1.04)	(1.48)	(0.74)	(0.038)
$Constrained{=}1 imes Treatment$	18.9**	-11.4	-10.1*	0.0076
	(8.55)	(7.03)	(5.32)	(0.12)
No. of obs.	1,256	2,773	2,636	2,641
Mean control (Unconstrained)	-0.54	82.63	93.93	0.09
Mean control (Constrained)	-9.96	79.49	100.00	-0.26
$\alpha_0 = \text{Constrained} - \text{Unconstrained}$	15.84	-15.31	-11.55	-0.11
p-value ($H_0: lpha_0 = 0$)	0.07	0.04	0.04	0.36

▶ Back

Lasso

- 1. Standardize variables
- 2. What mediators are correlated with learning the most (i.e., what explains learning differences?)
- 3. What mediators are correlated with the experiment the most (i.e., what did the experiment change?)

Each Lasso is estimated via:

$$\widehat{\beta}_t = \underset{\beta}{\operatorname{argmin}} ||\mathbb{Y} - \mathbb{X}\beta||_2^2 \text{ subject to } ||\beta||_1 \le t$$

or alternatively:

$$\widehat{\beta}_{\lambda} = \operatorname*{argmin}_{\beta} ||\mathbb{Y} - \mathbb{X}\beta||_{2}^{2} + \lambda ||\beta||_{1}$$



By provider

	(1) BRAC	(2) Bridge	(3) YMCA	(4) MtM	(5) Omega	(6) Rising	(7) St. Child	(8) Stella M	(9) p-value
% teachers dismissed	-8.66	49.55***	13.92	-6.17	0.69	-0.77	-1.53	11.91	< 0.001
	(6.46)	(7.18)	(11.13)	(6.74)	(11.88)	(8.97)	(12.90)	(13.01)	
% new teachers	38.12***	70.73***	47.11**	22.59*	20.57	35.91**	-9.52	35.78*	0.0060
	(11.15)	(13.10)	(18.67)	(11.92)	(20.17)	(15.21)	(26.30)	(21.03)	
School in session	0.16*	0.08	0.17	-0.13	0.22	0.05	0.27	0.27	0.026
	(0.08)	(0.09)	(0.13)	(0.10)	(0.15)	(0.11)	(0.16)	(0.16)	
Hours per week	0.80	7.60***	6.45***	3.93***	1.65	3.94**	-3.44	2.74	0.0020
	(1.42)	(1.55)	(2.46)	(1.41)	(2.52)	(1.91)	(3.29)	(2.72)	
Teacher attendance	13.02*	19.92***	17.76**	17.22***	24.81**	24.58***	28.27**	19.97**	0.26
	(6.97)	(6.27)	(8.76)	(6.24)	(10.07)	(8.48)	(12.27)	(9.46)	
Off-task (Classroom obs)	-14.46*	-13.64	-11.47	-13.07	-12.56	-11.27	-10.53	-12.07	0.77
	(8.36)	(8.43)	(10.40)	(8.32)	(10.46)	(9.61)	(11.12)	(10.74)	
Observations	40	45	8	12	38	10	24	8	

→ Back (instructional time) → Back (attendance) → Back (time on task) → Back (teachers)

More inputs and more and better teachers

	()	(-)	(-)	(.)
	(1)	(2)	(3)	(4)
	Treatment	Control	Difference	Difference (F.E)
Panel A: School-level outcomes (N	V = 185)			
Number of teachers	9.62	7.02	2.60***	2.61***
	(2.82)	(3.12)	(0.44)	(0.37)
Pupil-teacher ratio (PTR)	32.20	39.95	-7.74***	-7.82***
	(12.29)	(18.27)	(2.31)	(2.12)
New teachers	4.81	1.77	3.03***	3.01***
	(2.56)	(2.03)	(0.34)	(0.35)
Teachers dismissed	3.35	2.17	1.18**	1.16**
	(3.82)	(2.64)	(0.48)	(0.47)
Panel B: Teacher-level outcomes (N = 1,167			
Age in years	39.09	46.37	-7.28***	-7.10***
	(11.77)	(11.67)	(1.02)	(0.68)
Experience in years	10.59	15.79	-5.20***	-5.26***
	(9.20)	(10.77)	(0.76)	(0.51)
% has worked at a private school	47.12	37.50	9.62**	10.20***
	(49.95)	(48.46)	(3.76)	(2.42)
Test score in standard deviations	0.13	-0.01	0.14*	0.14**
	(1.02)	(0.99)	(0.07)	(0.06)
Panel C: Classroom observation (V = 185)			
Number of seats	20.64	20.58	0.06	0.58
	(13.33)	(13.57)	(2.21)	(1.90)
% with students sitting on the floor	2.41	4.23	-1.82	-1.51
	(15.43)	(20.26)	(2.94)	(2.61)
% with chalk	96.39	78.87	17.51***	16.58***
	(18.78)	(41.11)	(5.29)	(5.50)
% of students with textbooks	37.08	17.60	19.48***	22.60***
	(43.22)	(35.25)	(6.33)	(6.32)
% of students with pens/pencils	88.55	79.67	8.88**	8.16**
	(19.84)	(30.13)	(4.19)	(4.10)

	(1)	(2)	(3)	(4)
	Treatment	Control	Difference	Difference (F.E)
Panel A: School-level outcomes (N = 140)			
Number of teachers	9.76	6.74	3.01***	3.01***
	(3.16)	(2.98)	(0.52)	(0.41)
Pupil-teacher ratio (PTR)	30.66	39.63	-8.97***	-8.97***
	(12.23)	(17.32)	(2.53)	(2.32)
New teachers	4.24	1.83	2.41***	2.41***
	(2.46)	(2.19)	(0.39)	(0.39)
Teachers dismissed	1.94	2.29	-0.34	-0.34
	(1.61)	(2.78)	(0.38)	(0.31)
Panel B: Teacher-level outcomes	(N = 902)			
Age in years	40.18	46.21	-6.03***	-5.86***
	(12.07)	(11.83)	(1.21)	(0.78)
Experience in years	11.36	15.38	-4.01***	-4.05***
	(9.74)	(10.82)	(0.91)	(0.57)
% has worked at a private school	41.54	34.36	7.18*	7.48***
	(49.32)	(47.55)	(4.21)	(2.67)
Test score in standard deviations	0.07	-0.03	0.10	0.10
	(1.00)	(1.03)	(0.08)	(0.07)
Panel C: Classroom observation (N = 140)			
Number of seats	17.20	18.51	-1.31	-2.00
	(10.65)	(11.82)	(2.18)	(1.74)
% with students sitting on the floor	3.28	4.00	-0.72	0.00
-	(17.96)	(19.79)	(3.62)	(3.11)
% with chalk	95.08	80.00	15.08**	13.04*
	(21.80)	(40.41)	(6.36)	(6.67)
% of students with textbooks	38.06	20.33	17.73**	21.04***
	(44.30)	(37.97)	(7.77)	(7.34)
% of students with pens/pencils	87.63	78.06	9.56*	7.71
	(20.99)	(31.60)	(5.17)	(5.11)

Without Bridge: More inputs and more and better teachers

Management improves

	(1)	(2)	(3)	(4) D:ff
	Treatment	Control	Difference	Difference (F.E)
% school in session	92.47	83.70	8.78*	8.66*
	(26.53)	(37.14)	(4.75)	(4.52)
Instruction time (hrs/week)	20.40	16.50	3.90***	3.93***
	(5.76)	(4.67)	(0.77)	(0.73)
Intuitive score (out of 12)	4.08	4.03	0.04	0.02
	(1.35)	(1.38)	(0.20)	(0.19)
Time management score (out of 12)	5.60	5.69	-0.09	-0.10
	(1.21)	(1.35)	(0.19)	(0.19)
Principal's working time (hrs/week)	21.43	20.60	0.83	0.84
	(11.83)	(14.45)	(1.94)	(1.88)
% of time spent on management	74.06	53.64	20.42***	20.09***
	(27.18)	(27.74)	(4.12)	(3.75)
Index of good practices (PCA)	0.41	-0.00	0.41***	0.40***
	(0.64)	(1.00)	(0.12)	(0.12)
Observations	92	93	185	185



Without Bridge: Management improves

	(1)	(2)	(3)	(4)
	Treatment	Control	Difference	Difference (F.E)
% school in session	91.43	80.00	11.43*	11.43**
	(28.20)	(40.29)	(5.88)	(5.59)
Instruction time (hrs/week)	19.06	16.40	2.66***	2.66***
	(5.44)	(4.59)	(0.85)	(0.79)
Intuitive score (out of 12)	4.03	3.92	0.11	0.09
	(1.44)	(1.37)	(0.24)	(0.23)
Time management score (out of 12)	5.60	5.68	-0.08	-0.09
	(1.22)	(1.44)	(0.23)	(0.23)
Principal's working time (hrs/week)	21.45	20.98	0.48	0.48
	(10.73)	(14.37)	(2.14)	(2.00)
% of time spent on management	69.41	52.70	16.72***	17.03***
	(25.89)	(27.76)	(4.62)	(4.16)
Index of good practices (PCA)	0.36	0.04	0.32**	0.32**
	(0.66)	(0.97)	(0.14)	(0.12)
Observations	70	70	140	140



Teachers attendance and time on-task increases

	(1)	(2)	(3)	(4)
	Treatment	Control	Difference	Difference (F.E)
Panel A: Spot checks ($N = 185$)				
% on schools campus	60.32	40.38	19.94***	19.79***
	(23.10)	(25.20)	(3.56)	(3.48)
% in classroom	47.02	31.42	15.60***	15.37***
	(26.65)	(25.04)	(3.80)	(3.62)
Panel B: Student reports ($N = 185$)				
Teacher missed school previous week (%)	17.69	25.12	-7.43***	-7.55***
	(10.75)	(14.92)	(1.91)	(1.94)
Teacher never hits students (%)	54.71	48.21	6.50**	6.56***
	(18.74)	(17.06)	(2.63)	(2.52)
Teacher helps outside the classroom $(\%)$	50.00	46.59	3.41	3.55
	(18.22)	(18.05)	(2.67)	(2.29)
Panel C: Classroom observations ($N = 185$)				
Instruction (active + passive) (% of class time)	49.68	35.00	14.68***	14.51***
	(32.22)	(37.08)	(5.11)	(4.70)
Classroom management (% class time)	19.03	8.70	10.34***	10.25***
	(20.96)	(14.00)	(2.62)	(2.73)
Teacher off-task (% class time)	31.29	56.30	-25.01***	-24.77***
	(37.71)	(42.55)	(5.91)	(5.48)
Student off-task (% class time)	50.41	47.14	3.27	2.94
	(33.51)	(38.43)	(5.30)	(4.59)

Without Bridge

Teachers attendance and time on-task increases

	(1)	(2)	(3)	(4)
	Treatment	Control	Difference	Difference (F.E)
Panel A: Spot checks ($N = 140$)				
% on schools campus	59.42	38.68	20.74***	20.74***
	(24.29)	(25.92)	(4.25)	(4.20)
% in classroom	44.52	29.06	15.46***	15.46***
	(26.70)	(25.59)	(4.42)	(4.20)
Panel B: Student reports ($N = 140$)				
Teacher missed school previous week (%)	16.25	24.64	-8.39***	-8.39***
	(10.95)	(14.65)	(2.19)	(2.26)
Teacher never hits students (%)	52.59	45.07	7.52**	7.52**
	(17.96)	(16.48)	(2.91)	(2.94)
Teacher helps outside the classroom (%)	50.83	47.61	3.22	3.22
	(18.87)	(18.76)	(3.18)	(2.76)
Panel C: Classroom observations ($N = 140$)				
Instruction (active + passive) (% of class time)	48.57	30.71	17.86***	17.86***
	(33.11)	(35.85)	(5.83)	(5.45)
Classroom management (% class time)	18.57	9.86	8.71***	8.71***
	(21.35)	(15.37)	(3.14)	(3.24)
Teacher off-task (% class time)	32.86	59.43	-26.57***	-26.57***
	(38.57)	(42.63)	(6.87)	(6.35)
Student off-task (% class time)	47.96	45.80	2.16	2.16
	(33.99)	(38.39)	(6.13)	(5.41)

Lee bounds

	(1) Treatment	(2) Control	(3) Difference	(4) Difference (F.E)	(5) 90% CI Lee bounds
Panel A: Spot check ($N = 929$)					
% on schools campus	68.15	52.40	15.75***	14.17***	2.51
	(46.64)	(50.00)	(4.45)	(3.75)	28.11
% in classroom	50.96	41.05	9.91**	9.96**	-1.34
	(50.04)	(49.25)	(4.78)	(3.86)	24.44
B: Classroom observation ($N = 143$))				
Active instruction (% class time)	38.12	30.13	7.98	7.62	-4.75
	(28.93)	(32.11)	(4.86)	(4.75)	19.92
Passive instruction (% class time)	16.24	12.80	3.44	4.72	-4.93
	(17.18)	(19.83)	(2.95)	(3.23)	9.62
Classroom management (% class time)	20.82	10.67	10.16***	10.33***	0.77
	(21.06)	(14.83)	(2.85)	(3.32)	16.99
Teacher off-task (% class time)	24.82	46.40	-21.58***	-22.66***	-40.24
	(32.65)	(41.09)	(5.92)	(6.26)	-10.32
Student off-task (% class time)	55.06	57.60	-2.54	-5.19	-16.05
	(31.23)	(34.87)	(5.26)	(4.88)	12.63
Panel C: Inputs ($N = 143$)					
Number of seats	20.64	20.58	0.06	0.58	-7.22
	(13.33)	(13.57)	(2.21)	(1.90)	5.36
% with students sitting on the floor	2.41	4.23	-1.82	-1.51	-7.48
	(15.43)	(20.26)	(2.94)	(2.61)	2.76
% with chalk	96.39	78.87	17.51***	16.58***	9.47
	(18.78)	(41.11)	(5.29)	(5.50)	27.85
% of students with textbooks	37.08	17.60	19.48***	22.60***	-1.21
	(43.22)	(35.25)	(6.33)	(6.32)	34.87
% of students with pens/pencils	88.55	79.67	8.88**	8.16**	1.36
	(19.84)	(30.13)	(4.19)	(4.10)	20.98

Without Bridge: Lee bounds

	(1)	(2)	(3)	(4)	(5)
	Treatment	Control	Difference	Difference (F.E)	90% CI
					Lee bounds
Panel A: Spot check (N = 740)					
% on schools campus	67.88	50.76	17.12***	14.41***	7.83
	(46.75)	(50.07)	(5.25)	(4.15)	32.16
% in classroom	49.39	38.60	10.79*	9.20**	-1.47
	(50.06)	(48.76)	(5.67)	(4.41)	22.86
B: Classroom observation ($N = 102$)					
Active instruction (% class time)	39.05	27.22	11.83**	10.98*	-4.30
	(30.31)	(31.47)	(5.74)	(5.69)	27.41
Passive instruction (% class time)	14.92	12.59	2.33	3.14	-7.89
	(16.55)	(21.03)	(3.54)	(4.10)	9.92
Classroom management (% class time)	20.63	12.78	7.86**	7.65*	-3.95
	(21.54)	(16.42)	(3.52)	(4.11)	16.77
Teacher off-task (% class time)	25.40	47.41	-22.01***	-21.76***	-44.55
	(33.01)	(41.49)	(7.01)	(7.67)	-7.81
Student off-task (% class time)	53.17	59.07	-5.90	-8.24	-22.73
	(31.77)	(33.66)	(6.08)	(6.01)	12.32
Panel C: Inputs ($N = 102$)					
Number of seats	17.20	18.51	-1.31	-2.00	-8.10
	(10.65)	(11.82)	(2.18)	(1.74)	4.02
% with students sitting on the floor	3.28	4.00	-0.72	0.00	-7.74
	(17.96)	(19.79)	(3.62)	(3.11)	5.30
% with chalk	95.08	80.00	15.08**	13.04*	4.99
	(21.80)	(40.41)	(6.36)	(6.67)	27.73
% of students with textbooks	38.06	20.33	17.73**	21.04***	-8.17
	(44.30)	(37.97)	(7.77)	(7.34)	37.33
% of students with pens/pencils	87.63	78.06	9.56*	7.71	-0.07
	(20.99)	(31.60)	(5.17)	(5.11)	24.12

Students and parents like PPP schools more

	(1) Treatment	(2) Control	(3) Difference	(4) Difference (F.E)
Panel A: Household behavior (N = $1,115$)				
% satisfied with school	74.87	67.46	7.42**	7.44**
	(19.25)	(23.95)	(3.20)	(3.23)
% paying any fees	48.11	73.56	-25.45***	-25.69***
	(50.01)	(44.14)	(4.73)	(3.26)
Fees (USD/year)	5.72	8.04	-2.32**	-2.89***
	(10.22)	(9.73)	(0.96)	(0.61)
Expenditure (USD/year)	65.52	73.61	-8.09	-6.74
	(74.78)	(79.53)	(6.96)	(4.13)
Engagement index (PCA)	-0.11	-0.09	-0.02	-0.03
	(0.84)	(0.91)	(0.07)	(0.06)
Panel B: Student attitudes ($N = 3,492$)				
School is fun	0.58	0.53	0.05**	0.05**
	(0.49)	(0.50)	(0.02)	(0.02)
I use what I'm learning outside of school	0.52	0.49	0.04	0.04***
	(0.50)	(0.50)	(0.02)	(0.02)
If I work hard, I will succeed.	0.60	0.55	0.05*	0.04***
	(0.49)	(0.50)	(0.03)	(0.02)
Elections are the best way to choose a president	0.90	0.88	0.03*	0.03***
	(0.30)	(0.33)	(0.01)	(0.01)
Boys are smarter than girls	0.69	0.69	-0.00	0.01
	(0.46)	(0.46)	(0.02)	(0.01)
Some tribes in Liberia are bad	0.76	0.79	-0.03	-0.03**
	(0.43)	(0.41)	(0.02)	(0.01)

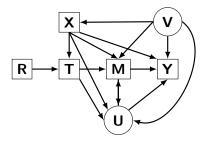
Without Bridge: Students and parents like PPP schools more

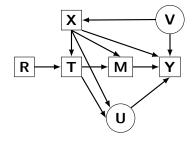
	(1) Treatment	(2) Control	(3) Difference	(4) Difference (F.E)
Danal A. Hawashald bahavian (N. 917)				
Panel A: Household behavior (N = 817) % satisfied with school	75.04	60.50	F F1	F F1
% satisfied with school	75.04	69.53	5.51	5.51
0/ · · · ·	(18.78)	(23.96)	(3.64)	(3.73)
% paying any fees	57.41	77.46	-20.04***	-20.36***
	(49.50)	(41.84)	(5.50)	(3.67)
Fees (USD/year)	5.69	8.01	-2.32**	-2.34***
	(9.61)	(9.27)	(1.00)	(0.68)
Expenditure (USD/year)	65.42	68.04	-2.62	-0.37
	(75.46)	(70.54)	(6.40)	(4.78)
Engagement index (PCA)	-0.12	-0.08	-0.04	-0.06
	(0.88)	(0.92)	(0.08)	(0.06)
Panel B: Student attitudes ($N = 2,643$)				
School is fun	0.57	0.53	0.04	0.04**
	(0.50)	(0.50)	(0.03)	(0.02)
I use what I'm learning outside of school	0.52	0.49	0.04	0.05***
Ũ	(0.50)	(0.50)	(0.03)	(0.02)
If I work hard, I will succeed.	0.57	0.55	0.02	0.03
	(0.50)	(0.50)	(0.03)	(0.02)
Elections are the best way to choose a president	0.90	0.88	0.02	0.02*
	(0.30)	(0.33)	(0.02)	(0.01)
Boys are smarter than girls	0.71	0.70	0.01	0.02
,	(0.45)	(0.46)	(0.02)	(0.02)
Some tribes in Liberia are bad	0.73	0.77	-0.05*	-0.03*
	(0.45)	(0.42)	(0.02)	(0.02)



Decompose the treatment effect - Mediation analysis

Causal relationships under different models





Under assumption sequential ignorability

Note: Based on Figure 1 in Heckman and Pinto (2015).



Decompose the treatment effect - Mediation analysis

$$M_{isg} = \alpha_g + \beta_6 treat_g + \gamma_6 X_i + \delta_6 Z_s + u_i$$
(6)

$$Y_{isg} = \alpha_g + \beta_7 treat_g + \gamma_7 X_i + \delta_7 Z_s + \theta_7 M_{is} + \varepsilon_i$$
(7)

Back

Key assumption

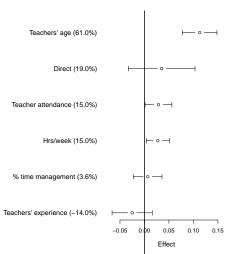
Sequential ignorability (Imai, Keele, & Yamamoto, 2010)]

$$Y_i(t',m), M_i(t) \perp T_i | X_i = x$$
(8)

$$Y_i(t',m) \perp M_i(t) | X_i = x, T_i = t$$
(9)



Material inputs don't matter, teachers do (and so does teacher attendance)



Direct and mediation effects



1. How do we allow for differences in context? Adjust for baseline differences

	(1) BRAC	(2) Bridge	(3) LIYONET	(4) MtM	(5) Omega	(6) Rising	(7) St. Child	(8) Stella M	(9) p-value equality
Students	31.94	156.19***	-23.03	35.49	-0.83	31.09	-19.16	-22.53	.00092
	(27.00)	(25.48)	(49.01)	(27.69)	(53.66)	(34.74)	(59.97)	(59.97)	
Teachers	1.23*	2.72***	1.42	1.70**	1.16	0.59	1.13	0.76	.66
	(0.70)	(0.66)	(1.28)	(0.72)	(1.40)	(0.90)	(1.56)	(1.56)	
PTR	-4.57	5.77*	-8.47	-5.45	-6.02	2.34	-10.62	-7.29	.079
	(3.27)	(3.09)	(5.94)	(3.36)	(6.50)	(4.21)	(7.27)	(7.27)	
Latrine/Toilet	0.18**	0.28***	0.26*	0.25***	0.23	0.22**	0.06	0.18	.96
	(0.08)	(0.07)	(0.14)	(0.08)	(0.16)	(0.10)	(0.17)	(0.17)	
Solid classrooms	0.63	2.81***	2.64*	-0.11	1.85	1.59*	-1.95	1.30	.055
	(0.75)	(0.71)	(1.36)	(0.77)	(1.49)	(0.97)	(1.67)	(1.67)	
Solid building	0.28***	0.22***	0.19	0.09	0.26*	0.19*	0.23	0.23	.84
-	(0.08)	(0.07)	(0.14)	(0.08)	(0.15)	(0.10)	(0.17)	(0.17)	
Nearest paved road (KM)	-9.25***	-10.86***	-7.13*	-8.22***	-4.47	-7.13***	-4.56	-7.79*	.78
	(2.03)	(1.91)	(3.67)	(2.08)	(4.01)	(2.60)	(4.48)	(4.48)	

▶ Back ▲ Back2

2. How do we overcome small samples?

Different providers: Same program

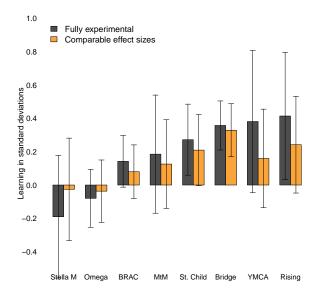
Bayesian hierarchical model ala Rubin (1981)

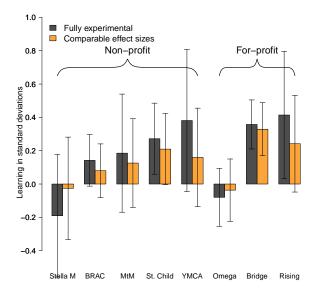
"Pool power" across providers

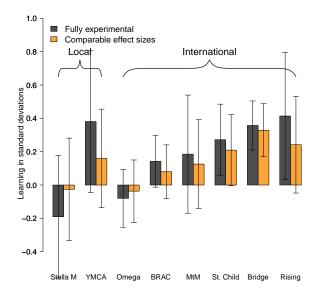
 Final estimate: Weighted average of providers' ITT and average ITT

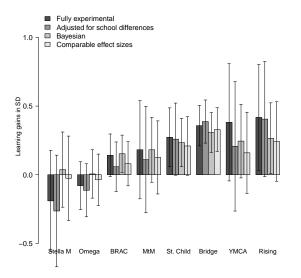
Proportions depend on provider's sample size





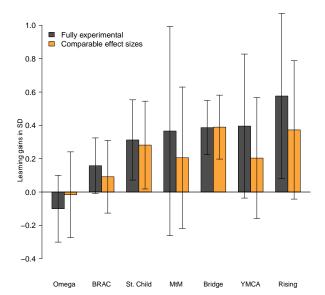






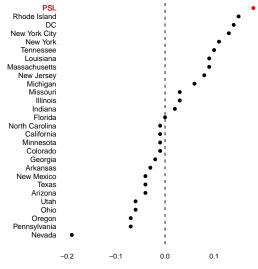
▶ Back 1 ▶ Back 2 ▶ Back 3

Treatment on the treated



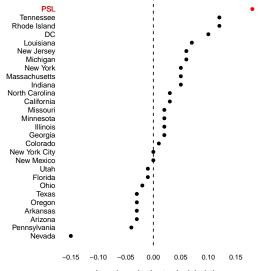


Math



Learning gains in standard deviations

English



Learning gains in standard deviations

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