

UNOBSERVED HETEROGENEITY, SCHOOL CHOICE AND THE EFFECT OF VOUCHER SCHOOL: EVIDENCE FROM CHILE

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Introduction

Does competition improve the efficiency and quality of publicly financed education?: **Open Question**

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- We assess the relative performance of private-voucher schools
- We use panel data of 8th and 10th graders
- We model a structural model of sequential schooling decisions with endogenous outcomes and unobserved heterogeneity
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- We model a structural model of sequential schooling decisions with endogenous outcomes and unobserved heterogeneity
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We find a positive effect of private-voucher education, and an important role of unobserved ability in decisions and outcomes

The Chilean School System

After 1981, three types of schools were established:

- **Public schools.** Financed by the per-student voucher subsidy and administered by local municipalities. They are not allowed to select students, and have dismissal restrictions.
- **Private-voucher schools.** Financed by the per-student voucher subsidy and run by privates. Selection processes are allowed, and they can freely hire and dismiss teachers.
- **Private-fee-paying schools.** Financed by the fees charged to parents and run by privates. They behave as profit-maximizing firms.

Literature Review

- Factors underlying school choice: Gallego and Hernando (2009); and Makovec, Mizala and Barrera (2010)
- Link between socioeconomic status and academic achievement: Mizala, Romaguera and Ostoic (2005); and Mizala, Romaguera and Urquiola (2006)
- Private-voucher schools' relative performance: Mizala and Romaguera (2000); Tokman (2002); Sapelli and Vial (2005); Contreras and Santos (2009); and Lara, Mizala and Repetto (2009)

Model for School Choice and Test Scores under Unobserved Heterogeneity

- 1 Schooling decisions are modeled in two periods of time: 8th and 10th grades
- 2 Two dimensions of decisions are taken into account: voucher vs. public; **and** whether or not students are forced to switch schools
- 3 The decision process depends on observed and unobserved characteristics
- 4 Test scores are observed conditional on the optimal decision, and are also determined by observed and unobserved characteristics
- 5 The unobserved heterogeneity is interpreted as a combination of different inherent abilities (cognitive and non-cognitive)

Scheme of Decisions

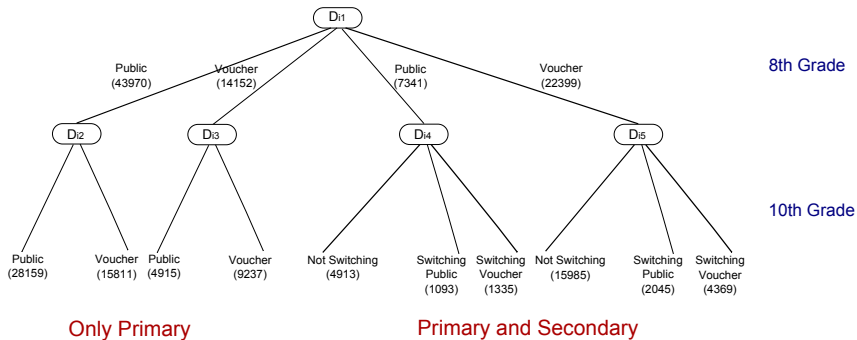
◀ Back to Estimates

◀ Back to Analysis

◀ Back to TE 1

◀ Back to TE 2

Figure 1: Scheme of Decisions



The System of Test Scores and School Choices in 8th grade

- Parents decide s^* , the optimal school-type:

$$s^* = \operatorname{argmax}_{\{s\}} \{I(s)\}$$

where

$$I(s) = \gamma_s Z(s) + \eta(s) \quad \text{for } s = 1, \dots, \bar{S}.$$

- We also observe tests scores given s^* :

$$T(s) = \beta_s X(s) + \varepsilon(s) \quad \text{for } s = 1, \dots, \bar{S},$$

- We allow $\eta(s)$, $\eta(s')$, $\varepsilon(s)$, $\varepsilon(s')$ to be correlated, for any $s \neq s'$

- Following Hansen, Heckman and Mullen (2004) we impose a factor structure to our model

$$\begin{aligned}\eta(s) &= \alpha_s f + \nu(s) \\ \varepsilon(s) &= \lambda_s f + \tau(s) \quad \text{for } s = 1, \dots, \bar{S},\end{aligned}$$

where we assume $\nu(s) \perp\!\!\!\perp \tau(s) \perp\!\!\!\perp f$

- This model is non-parametrically identified

The System of Test Scores and School Choices in 10th grade

- Conditional on the school-type in primary school (s^p), parents decide the type of secondary school, s^* :

$$s^* = \operatorname{argmax}_{\{s|s^p\}} \{I(s|s^p)\},$$

where

$$I(s|s^p) = \gamma_{s|s^p} Z(s|s^p) + \alpha_{s|s^p} f + \nu(s|s^p) \quad \text{for } s = 1, \dots, \bar{S}$$

- Test scores are modeled as

$$T(s|s^p) = \beta_{s|s^p} X(s|s^p) + \lambda_{s|s^p} f + \tau(s|s^p) \quad \text{for } s = 1, \dots, \bar{S}$$

- And, as before, we assume $\nu(s|s^p) \perp\!\!\!\perp \tau(s|s^p) \perp\!\!\!\perp f$

Empirical Implementation

The key insight is that, conditional on unobserved abilities, all error terms are mutually independent. Thus, likelihood function can be written as:

$$\begin{aligned} \Gamma(\mathbf{T}_i, \mathbf{D}_i | X_i, Z_i) &= \prod_{i=1}^N \int f(\mathbf{T}_i, \mathbf{D}_i | X_i, Z_i, f) dF(f) \\ &= \prod_{i=1}^N \int \left(\prod_{j=1}^J [f(\mathbf{T}_{ij} | D_{ij} = 1, X_i, f) \times \Pr(D_{ij} = 1 | Z_{ij}, f)]^{D_{ij}} \right) dF(f) \end{aligned}$$

- We also assume:

$$f \sim p_1 N(\mu_1, \sigma_1^2) + p_2 N(\mu_2, \sigma_2^2) + p_3 N(\mu_3, \sigma_3^2)$$

- The model is estimated using Markov Chain Monte Carlo (MCMC) methods

Data

- We construct a two-period panel using SIMCE 2004 and 2006
- We examine test scores in Language, Mathematics, Social Sciences and Natural Sciences
- We control for family background and environment characteristics (gender, grade repetition, parents' education, attendance to preschool, HH income, HH members, geographic location, access to computer, books, fees, etc.)
- The number of students of our final sample is 87,862

Estimates

▶ Go to Scheme of Decisions

Findings:

- Females perform better on Language tests, but do worse on every other subject
- Particular pattern for the attendance to preschool variable: positive for private-vouchers and negative for public
- The unobserved component of our model is a strong predictor of schooling decisions and test scores:
 - Sorting on school-type in 8th grade
 - Sorting on switching school decisions for secondary education
 - Positive effect on test scores, at all schooling levels

Simulations and Goodness of Fit

- We simulate our model using the estimated coefficients and the sample data
- We draw samples of the unobserved factor and make predictions of schooling choices and outcomes

Table 1: Goodness of Fit - Schooling Decisions

Total	Only Primary				Primary and Secondary			
Actual	0.66				0.34			
Model	0.66				0.34			
p-value ^a	0.78				0.78			
School Type 2004	Public		Voucher		Public		Voucher	
Actual	0.76		0.24		0.25		0.75	
Model	0.75		0.25		0.25		0.75	
p-value ^a	0.66		0.66		0.65		0.65	
Switching Decision					No	Yes	No	Yes
Actual					0.67	0.33	0.71	0.29
Model					0.67	0.33	0.70	0.30
p-value ^a					0.94	0.94	0.03	0.03
School Type 2006	Public	Voucher	Public	Voucher	Public	Voucher	Public	Voucher
Actual	0.64	0.36	0.35	0.65	0.45	0.55	0.32	0.68
Model	0.64	0.36	0.35	0.65	0.45	0.55	0.33	0.67
p-value ^a	0.84	0.84	0.87	0.87	0.92	0.92	0.28	0.28

Table 2: Goodness of Fit - Test Scores

Language					
Total School Type 2004	Only Primary				
		Public		Voucher	
	Actual	-0.24 (0.93)		0.01 (0.96)	
	Model	-0.24 (0.94)		0.01 (0.95)	
p-value ^a	0.04		0.00		
School Type 2006		Public	Voucher	Public	Voucher
	Actual	-0.30 (0.93)	-0.18 (0.91)	-0.06 (0.98)	0.01 (0.93)
	Model	-0.31 (0.93)	-0.17 (0.90)	-0.07 (0.99)	-0.01 (0.93)
	p-value ^a	0.00	0.00	0.00	0.00
Mathematics					
Total School Type 2004	Only Primary				
		Public		Voucher	
	Actual	-0.27 (0.90)		0.00 (0.96)	
	Model	-0.27 (0.92)		0.00 (0.93)	
p-value ^a	0.00		0.00		
School Type 2006		Public	Voucher	Public	Voucher
	Actual	-0.32 (0.91)	-0.17 (0.91)	-0.12 (0.94)	0.01 (0.93)
	Model	-0.32 (0.91)	-0.16 (0.90)	-0.11 (0.95)	-0.03 (0.93)
	p-value ^a	0.00	0.00	0.00	0.00

Table 3: Goodness of Fit - Test Scores

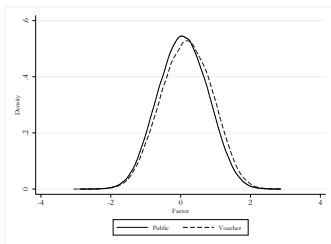
Language							
Total	Primary and Secondary						
School Type 2004	Public			Voucher			
Actual	0.00 (1.06)			0.25 (0.95)			
Model	0.03 (1.00)			0.23 (0.94)			
p-value ^a	0.00			0.00			
Switching Decision	No	Yes	No		Yes		
School Type 2006	Public		Voucher		Public		Voucher
Actual	0.02 (1.09)	-0.05 (0.95)	-0.11 (0.93)		0.31 (0.96)		0.11 (0.99)
Model	0.05 (0.99)	-0.05 (1.02)	-0.06 (0.95)		0.29 (0.94)		0.20 (1.00)
p-value ^a	0.00	0.00	0.00		0.00		0.00
Mathematics							
Total	Primary and Secondary						
School Type 2004	Public			Voucher			
Actual	0.02 (1.07)			0.24 (0.96)			
Model	-0.05 (0.98)			0.22 (0.95)			
p-value ^a	0.00			0.00			
Switching Decision	No	Yes	No		Yes		
School Type 2006	Public		Voucher		Public		Voucher
Actual	0.05 (1.11)	-0.06 (0.93)	-0.11 (0.93)		0.35 (0.95)		0.10 (0.99)
Model	0.09 (1.01)	-0.06 (0.98)	-0.05 (0.94)		0.33 (0.93)		0.20 (1.99)
p-value ^a	0.00	0.00	0.00		0.00		0.00

Sorting on Ability

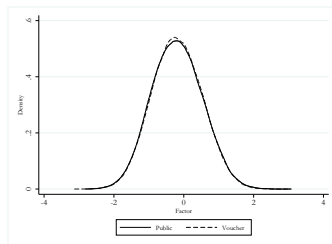
▶ [Go to Scheme of Decisions](#)

Figure 2: Distribution of Factor by School Type in 8th grade

Only Primary



Primary and Secondary

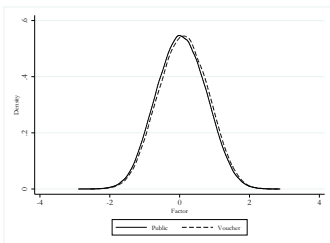


▶ Go to Scheme of Decisions

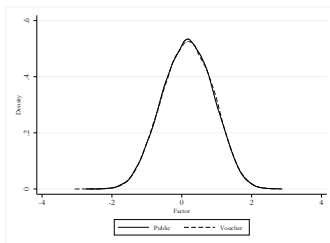
Figure 3: Distribution of Factor by School Type in 10th grade

(A) Only Primary

i. Public 8th grade



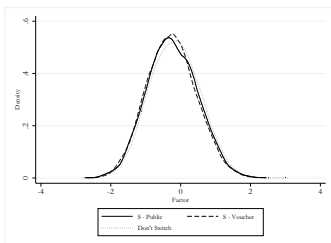
ii. Voucher 8th grade



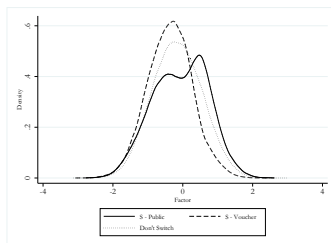
▶ Go to Scheme of Decisions

Figure 4: Distribution of Factor by School Type in 10th grade
(B) Primary and Secondary

i. Public 8th grade



ii. Voucher 8th grade



Treatment Effects

- We assess the return of attending a voucher school in 10th grade using ATE, TT, TUT treatment parameters
- Binary set-up: Let Y_1 denote test scores for students in voucher schools, and Y_0 denote test scores for students in public schools. Let also D denote the observed treatment participation. We define

$$ATE = \int \int E(Y_1 - Y_0 \mid X = x, f = \theta) dF_{X,f}(x, \theta)$$

$$TT = \int \int E(Y_1 - Y_0 \mid X = x, f = \theta, D = 1) dF_{X,f}(x, \theta)$$

$$TUT = \int \int E(Y_1 - Y_0 \mid X = x, f = \theta, D = 0) dF_{X,f}(x, \theta)$$

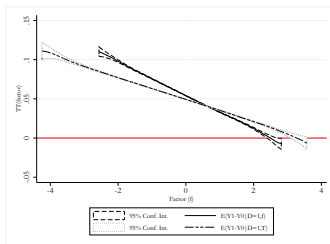
▶ Go to Scheme of Decisions

Table 5: People with positive potential gains if treated

School Decision Path		Language		Mathematics	
		%	\bar{f}	%	\bar{f}
Only Primary, Public 8th grade	$Y_1 - Y_0 > 0$	0.83	-0.03	0.87	0.02
	$Y_1 - Y_0 < 0$	0.17	0.55	0.13	0.46
Only Primary, Voucher 8th grade	$Y_1 - Y_0 > 0$	0.62	0.00	0.67	0.01
	$Y_1 - Y_0 < 0$	0.38	0.47	0.33	0.51
Primary and Secondary, Public 8th grade, Switching	$Y_1 - Y_0 > 0$	0.54	-0.39	0.48	-0.41
	$Y_1 - Y_0 < 0$	0.46	-0.08	0.52	-0.10
Primary and Secondary, Voucher 8th grade, Switching	$Y_1 - Y_0 > 0$	0.64	-0.30	0.61	-0.42
	$Y_1 - Y_0 < 0$	0.36	-0.20	0.39	-0.02

Figure 5: TT Parameter as a Function of Unobserved Ability and Initial Test Scores (Only Primary)

i. Public 8th grade



ii. Voucher 8th grade

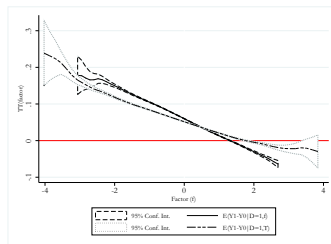
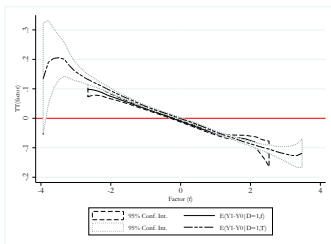
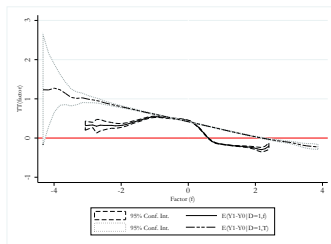


Figure 6: TT Parameter as a Function of Unobserved Ability and Initial Test Scores (Primary and Secondary, Switched School)

i. Public 8th grade



ii. Voucher 8th grade



Final Comments

- Our results suggest that the effect of attending a private-voucher school is positive and statistically significant
- The unobserved component of our model is a strong predictor of the model, and some interesting patterns are found on the sorting on ability:
 - Students in private-voucher schools are on average more able than students in public schools, in 8th grade
 - When facing the switching decision at the end of 8th grade, students switching from a voucher school to a public one are the most able ones
- The benefits of attending a private-voucher school are larger for students with low ability
- From the point of view of public policies, our results suggest to encourage the attendance to private-voucher schools