### Hacking Gender Stereotypes

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### **Motivation**

- Employment opportunities and wage growth are rising more rapidly among occupations that require high level of both social and math skills (Deming, 2017; Fayer et al., 2017)
- The supply of those skills is insufficient also due to the limited participating of women in STEM and coding (Kahn and Ginther, 2017; Adams and Kirchmaier, 2016)
- Several initiatives around the word are trying to promote STEM education among female students or adult women  $\rightarrow$  limited evidence on the self-selecting and impact of these interventions

### **Research questions**

- 1. Who are the girls that self-select into coding clubs?
- 2. Which is the impact of coding clubs on girls?
- **3.** Which is the impact of coding clubs on the formation of stereotypes in schools?

In this presentation, I will show evidence on the first question and some preliminary results on the second from the pilot data.



#### **Data and Experimental Design**

**Program Sign-Up** 

Preliminary Results (from pilot)

Conclusions

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Parents and Peers

## **Girls Code It Better**

- The idea: from a private employment agency in Italy  $\rightarrow$  firms wants to hire more women in STEM and coding
- Participation in the coding clubs:
  - 20 middle-school girls (11-14 years old), 1 trained teacher, 1 coach maker
  - voluntary, free of charge for girls
  - at school in the afternoon, 45 hours per school year

#### • Project-based methodology:

- Coding and new technologies (automation, web and app design, 3D printing)
- Team work and communication

### **Girls Code It Better**



# **Data Collection**

- **Survey Data from Students.** We collect survey data at the end of the school year from all students in the schools included in the randomization
  - Response rate and matching: 85%
- Administrative Data and Standardized Test Score: track choice, teachers' track recommendation, grades (to be obtained soon)

# **Experimental Design**

**Randomization:** individual level randomization within each school, conditional on receiving more than 20 application.

• 16 schools with rationing

	(1)	(2)	(3)	(4)
	Control	Treated	Diff	Norm. Diff.
Immigrant	0.061	0.165	0.099**	0.237
	(0.240)	(0.372)	(0.035)	
Mum less than high-school	0.545	0.523	-0.075	-0.032
	(0.500)	(0.500)	(0.069)	
Mum has a university degree	0.455	0.477	0.075	0.032
	(0.500)	(0.500)	(0.069)	
Mum works in STEM	0.148	0.183	0.044	0.066
	(0.357)	(0.387)	(0.055)	
Mum has a high wage	0.717	0.665	-0.016	-0.079
	(0.453)	(0.473)	(0.054)	
Dad less than high-school	0.626	0.605	-0.051	-0.030
	(0.486)	(0.490)	(0.064)	
Dad has a university degree	0.374	0.395	0.051	0.030
	(0.486)	(0.490)	(0.064)	
Dad works in STEM	0.337	0.303	-0.002	-0.051
	(0.475)	(0.461)	(0.054)	
Dad has a high wage	0.576	0.583	0.018	0.010
	(0.497)	(0.494)	(0.061)	
Observations	99	266	365	

# **Gender Gaps**

Variable	Boys	Girls	P-value		
Panel A: Academic Interests					
Plans: University	0.543	0.700	0.000		
	(0.498)	(0.458)			
Like Math	0.479	0.384	0.000		
	( 0.500)	(0.486)			
Like Italian	0.302	0.440	0.000		
	(0.459)	(0.496)			
STEM High-School	0.415	0.372	0.004		
	(0.493)	(0.484)			
Classic High-School	0.494	0.707	0.000		
	(0.500)	(0.455)			
STEM Occupations	0.547	0.334	0.000		
	(0.498)	(0.472)			
Non-STEM Occupations	0.436	0.460	0.091		
	(0.496)	(0.499)			
Panel B: Barriers to achieve Educational Goals					
Barrier: Gender Unfit	0.410	0.535	0.000		
	(0.492)	(0.499)			
Barrier: Ability Math	0.349	0.437	0.000		
	(0.477)	(0.496)			
Observations	2244	2250			



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### How do Girls Self-Select into Coding Clubs?

We may expect that girls who self-select into coding clubs are (Ertl et al., 2017):

- less prone to stereotypic influences;
- have stronger STEM interest;
- highly educated parents working in STEM (especially mothers).

Investigating the characteristics associated with take-up of these types of programs is of crucial importance for designing effective policies to address gender gaps in STEM.

### How do Girls Self-Select into Coding Clubs?

Variable	Not Apply	Apply	P-value
Panel A: Family Background			
Immigrant	0.190	0.061	0.006
	(0.392)	(0.240)	
Mum less than high-school	0.582	0.545	0.396
	(0.493)	(0.500)	
Mum has a university degree	0.418	0.455	0.391
	(0.493)	(0.500)	
Mum works in STEM	0.133	0.148	0.756
	(0.340)	(0.357)	
Mum has a medium-high wage	0.648	0.717	0.143
	(0.478)	(0.453)	
Dad less than high-school	0.614	0.626	0.913
-	(0.487)	(0.486)	
Dad has a university degree	0.386	0.374	0.919
	(0.487)	(0.486)	
Dad works in STEM	0.282	<b>0.337</b>	0.246
	(0.450)	(0.475)	
Dad has a medium-high wage	0.582	0.576	0.945
	( 0.493)	(0.497)	
Dad works in STEM Dad has a medium-high wage	0.282 ( 0.450) 0.582 ( 0.493)	0.337 ( 0.475) 0.576 ( 0.497)	0.246

#### How do Girls Self-Select into Coding Clubs?

Panel B: Academic Interests   Plans: University 0.691 0.707 0.454   (0.462) (0.457) 0.454 0.365 0.434 0.128   Like Math 0.365 0.432 (0.482) 0.408) 0.128   Like Italian 0.439 0.404 0.293 0.406 (0.493)	Variable	Not Apply	Apply	P-value
Plans: University 0.691 0.707 0.454   (0.462) (0.457) 0.454 0.365 0.434 0.128   Like Math 0.365 0.434 0.128 0.482) 0.404 0.293   Like Italian 0.439 0.404 0.293 0.404 0.293	Panel B: Academic Interests			
(0.462) (0.457) Like Math 0.365 0.434 0.128 (0.482) (0.498) Like Italian 0.439 0.404 0.293 (0.496) (0.493) CTEMARY 4.04 0.293	Plans: University	0.691	0.707	0.454
Like Math 0.365 0.434 0.128 (0.482) (0.498) Like Italian 0.439 0.404 0.293 (0.496) (0.493)		(0.462)	(0.457)	
(0.482) (0.498) Like Italian 0.439 0.404 0.293 (0.496) (0.493)	Like Math	0.365	0.434	0.128
Like Italian 0.439 0.404 0.293 ( 0.496) ( 0.493)		(0.482)	(0.498)	
(0.496) (0.493)	Like Italian	0.439	0.404	0.293
CTENTIC 1 1 0.000		(0.496)	(0.493)	
STEM High-School 0.354 0.444 0.042	STEM High-School	0.354	0.444	0.042
(0.478) (0.499)		(0.478)	(0.499)	
Classic High-School 0.708 0.758 0.366	Classic High-School	0.708	0.758	0.366
( 0.455) ( 0.431)		( 0.455)	( 0.431)	
STEM Occupations 0.308 0.374 0.079	STEM Occupations	0.308	0.374	0.079
(0.462) (0.486)		(0.462)	(0.486)	
Non-STEM Occupations 0.468 0.404 0.261	Non-STEM Occupations	0.468	0.404	0.261
(0.499) (0.493)		(0.499)	(0.493)	
Panel C: Barriers to achieve Educational Goals				
Barrier: Gender Unfit 0.533 0.657 0.018	Barrier: Gender Unfit	0.533	0.657	0.018
(0.499) (0.477)		(0.499)	(0.477)	
Barrier: Ability Math 0.444 0.434 0.687	Barrier: Ability Math	0.444	0.434	0.687
(0.497) (0.498)		(0.497)	(0.498)	
Explicit gender stereotypes 0.351 0.354 0.725	Explicit gender stereotypes	0.351	0.354	0.725
( 0.477) ( 0.480)		(0.477)	(0.480)	
Observations 1885 99	Observations	1885	99	



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# The Impact of Coding Clubs Share of students interested in STEM Occupations



# The Impact of Coding Clubs Placebo



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# Mechanisms: Barrier to Achieve Educational Goals





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## **Summary and Conclusions**

- We show that girls who self-select to participate in coding clubs have:
  - slightly more educated mothers (but not statistically significant);
  - high interest in STEM occupations;
  - BUT they perceive their gender unfit to achieve her goals.
- Participation in coding clubs is a promising venue to increase interest in STEM occupations for women and mitigate the perception that own gender is unfit to achieve their goal

#### THANK YOU!

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