## **Online Appendix**

"Liquidity Constraints, Spillovers, and Entrepreneurship: Direct and Indirect Effects of a Cash Transfer Program'

May 4, 2018

## A IV with a Cluster-Level Instrument

**Proposition A.1.** Let  $z_{ivt}$  be an instrumental variable. If the period-cluster conditional variance of  $z_{ivt}$  is zero,  $Var(z_{ivt}|v,t) = 0$ , then the IV estimator for  $\tau$  in equation (4.2) is equivalent to the IV estimator for  $\tau$  in the following equation:

$$y_{ivt} = \beta_0 + \tau d_{ivt} + \mu_v + \mu_t + u_{ivt}.$$
 (A.1)

*Proof.* Let  $y_{ivt}^*$ ,  $d_{ivt}^*$ , and  $\overline{d}_{vt}^*$  be cluster-period mean-centered versions of  $y_{ivt}$ ,  $d_{ivt}$ , and  $\overline{d}_{vt}$ , respectively.

Suppose equation (4.1) is the true equation, but we instead estimate the following model:

$$y_{ivt} = \beta_0 + \beta_1 d_{ivt} + \mu_v + \mu_t + u_{ivt}, \tag{A.2}$$

in which  $\overline{d}_{vt}$  is omitted.

Let  $z_{vt}$  be an instrumental variable such that  $Var(z_{vt}|v,t) = 0$ . Then the (within-group) IV estimator for  $\beta_1$  in equation (A.2) is:

$$\begin{aligned} \widehat{\beta}_{1}^{IV} &= \frac{\sum_{ivt} z_{vt}^{*} y_{ivt}^{*}}{\sum_{ivt} z_{vt}^{*} d_{ivt}^{*}} \\ &= \frac{\sum_{ivt} z_{vt}^{*} y_{ivt}^{*}}{\sum_{vt} z_{vt}^{*} \sum_{i} d_{ivt}^{*}} \\ &= \frac{\sum_{ivt} z_{vt}^{*} y_{ivt}^{*}}{\sum_{ivt} z_{vt}^{*} d_{vt}^{*}} = \widehat{\tau}^{IV} \end{aligned}$$

Thus the formula is exactly the same as if we estimate equation (4.2) using  $z_{vt}$  as an instrumental variable. Using similar steps as in Proposition 4.1, we can show that  $\hat{\tau}^{IV}$ , as well as  $\hat{\beta}_1^{IV}$ , is a consistent estimator for the overall effect,  $(\beta_1 + \beta_2)$ .

		2001			2004			2006	
	Total	Urban	Rural	Total	Urban	Rural	Total	Urban	Rural
Poverty headcount	0.301	0.250	0.579	0.285	0.241	0.534	0.225	0.183	0.466
<sup>9</sup> rogram coverage	0.064	0.044	0.174	0.178	0.146	0.360	0.227	0.188	0.450
Share of benefits		0.599	0.401		0.686	0.314		0.708	0.292
Number of obs.	368,605	316,793 $51,812$	51,812	378,658	378,658 $326,322$	52,336	389,807	389,807 336,502	53,305

**Table B1:** Poverty Headcount and Program Coverage

Estimates are obtained using PNAD. 'Poverty headcount' is measured by the proportion of people with household per capita income below the poverty line (half of the 2001 minimum wage). 'Program coverage' is measured by the proportion of people participating in the program. 'Share of benefits' is the ratio between the total amount of transfers going to either urban or rural areas and the total amount of transfers distributed by CCT programs in the country.

	Std.				Number of
	Mean	Dev.	Min.	Max.	municipalities
2001					
Number of households	128.1	290.4	19	3,505	796
Sample size	52.4	128.1	5	$1,\!571$	796
2004					
Number of households	136.8	305.1	23	$3,\!575$	796
Sample size	54.3	131.8	5	1,751	796
2006					
Number of households	143.8	322.7	28	$3,\!884$	796
Sample size	56.4	136.1	5	1,753	796

 Table B2:
 Number of Observations per Municipality

The sample comprises men aged between 25 and 45 years old, with no college degree, and living in urban areas. This sample also excludes public servants and employers with more than five employees.

	Decision of being a small entrepreneur in					
	Ser	vices	Sa	les	Manufa	cturing
	FE	IV	FE	IV	FE	IV
	(1)	(2)	(3)	(4)	(5)	(6)
program coverage, $\overline{d}$	0.040***	0.056***	0.007	0.015	0.010	0.006
	(0.012)	(0.017)	(0.009)	(0.013)	(0.008)	(0.011)
age (x10)	0.031***	$0.031^{**}$	0.023**	0.023**	0.001	0.001
	(0.012)	(0.012)	(0.012)	(0.012)	(0.010)	(0.010)
squared age $(x100)$	-0.002	-0.002	-0.001	-0.001	0.002	0.002
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)
white	$0.016^{***}$	$0.015^{***}$	$0.015^{***}$	$0.015^{***}$	$0.006^{***}$	$0.006^{***}$
	(0.002)	(0.002)	(0.001)	(0.001)	(0.001)	(0.001)
black	-0.006***	-0.006***	-0.005***	-0.005***	-0.005***	-0.005***
	(0.002)	(0.002)	(0.002)	(0.002)	(0.001)	(0.001)
married	0.000	0.000	$0.012^{***}$	$0.012^{***}$	$0.006^{***}$	$0.006^{***}$
	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)
elementary education	$0.011^{***}$	$0.011^{***}$	$0.011^{***}$	$0.011^{***}$	$0.008^{***}$	$0.008^{***}$
	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)
primary education	$0.012^{***}$	$0.012^{***}$	$0.015^{***}$	$0.015^{***}$	$0.003^{*}$	$0.003^{*}$
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)
high school	$0.022^{***}$	$0.022^{***}$	$0.013^{***}$	$0.013^{***}$	-0.002	-0.002
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)
log of population	-0.010	-0.011	-0.015	-0.015	0.002	0.002
	(0.011)	(0.011)	(0.010)	(0.009)	(0.008)	(0.008)
year = 2001	0.020***	$0.023^{***}$	-0.009***	-0.008***	-0.002	-0.003
	(0.003)	(0.004)	(0.002)	(0.003)	(0.002)	(0.002)
year = 2004	0.001	0.001	0.000	0.001	-0.001	-0.001
	(0.001)	(0.001)	(0.002)	(0.002)	(0.001)	(0.001)
Municipality Fixed-Effects	Yes	Yes	Yes	Yes	Yes	Yes
Number of observations	$112,\!117$	$112,\!117$	$112,\!117$	$112,\!117$	$112,\!117$	$112,\!117$

Table B3: Overall Effect of Cash Transfers on Different Types of Business

\*\*\*, \*\*, \* represent statistical significant at the 1%, 5% and 10% levels, respectively. Standard errors in parentheses are clustered by municipality. Sample includes only men with high school diploma or less. FE columns present the fixed-effect regressions obtained using the within-group method. IV columns present the fixed-effect, Instrumental-Variable regressions with 'program coverage' instrumented by the interactions between municipal quotas and year dummies.

	Decision of being a small entrepreneur				
	Without	children	With c	hildren	
	FE	IV	FE	IV	
	(1)	(2)	(3)	(4)	
program coverage, $\overline{d}$	0.028	0.044	0.090***	0.112***	
	(0.021)	(0.029)	(0.025)	(0.030)	
individual benefit, $d$	-0.013***	-0.013***	-0.050***	-0.058***	
	(0.005)	(0.005)	(0.009)	(0.012)	
age (x10)	$0.071^{**}$	$0.071^{**}$	$0.062^{**}$	$0.062^{**}$	
	(0.029)	(0.029)	(0.028)	(0.028)	
squared age $(x100)$	-0.004	-0.004	-0.002	-0.002	
	(0.004)	(0.004)	(0.004)	(0.004)	
white	$0.031^{***}$	$0.031^{***}$	$0.037^{***}$	$0.037^{***}$	
	(0.002)	(0.002)	(0.003)	(0.003)	
black	-0.010**	-0.010**	-0.021***	-0.021***	
	(0.004)	(0.004)	(0.004)	(0.004)	
married	$0.027^{***}$	$0.027^{***}$	0.028***	0.028***	
	(0.002)	(0.002)	(0.003)	(0.003)	
elementary education	$0.027^{***}$	$0.027^{***}$	0.029***	$0.028^{***}$	
	(0.003)	(0.003)	(0.003)	(0.003)	
primary education	$0.027^{***}$	$0.027^{***}$	0.029***	0.029***	
	(0.003)	(0.003)	(0.004)	(0.004)	
high school	$0.029^{***}$	$0.029^{***}$	$0.031^{***}$	$0.031^{***}$	
	(0.003)	(0.003)	(0.004)	(0.004)	
log of population	0.002	0.001	-0.064***	-0.065***	
	(0.022)	(0.022)	(0.023)	(0.023)	
year = 2001	0.004	0.006	0.001	0.005	
	(0.005)	(0.006)	(0.006)	(0.007)	
year = 2004	-0.002	-0.001	-0.002	-0.001	
	(0.003)	(0.003)	(0.004)	(0.004)	
Municipality Fixed-Effects	Yes	Yes	Yes	Yes	
N. of obs all sample	$63,\!348$	$63,\!348$	65698	65698	
N. of obs $d = 0$	$60,\!630$	60,630	$52,\!458$	$52,\!458$	

Table B4: Indirect and Direct Effects on Entrepreneurship, With and Without Children

\*\*\*, \*\*, \* represent statistical significant at the 1%, 5% and 10% levels, respectively. Sample includes only men with high school diploma or less. Standard errors in parentheses are clustered by municipality. All coefficients are estimated using Seemingly Unrelated Regressions (SUR). The indirect effect (program coverage) is estimated using the sample of non-participants, whereas the direct effect (individual benefit) is estimated using all sample and bias corrected according to Lemma 4.1. Columns (1) and (2) present the estimates of effects on individuals without children in their household. Columns (3) and (4) present the estimates of effects on individuals living with children under 15 years old. The FE column shows the fixed-effect regression obtained using the within-group method. The IV column shows fixed-effect, Instrumental-Variable regression with 'program coverage' instrumented by the interactions between municipal quotas and year dummies.

Panel A: Individuals with	Fixed-Effect Model							
			Formal	Informal	Informa			
	Entrep.	Jobless	employee	employee	self-emp			
program coverage, $\overline{d}$	0.05***	0.023	0.04	-0.087**	-0.025			
program coverage, a	(0.016)	(0.026)	(0.035)	(0.034)	(0.035)			
individual benefit, $d$	-0.038***	0.035***	-0.044***	0.026*	0.021			
individual benenit, a	(0.006)	(0.011)	(0.014)	(0.014)	(0.015)			
Municipality Fixed-Effects	Yes	Yes	Yes	Yes	Yes			
Year dummies	Yes	Yes	Yes	Yes	Yes			
Individual covariates	Yes	Yes	Yes	Yes	Yes			
N. of obs all sample	90,648	90,648	90,648	90,648	90,648			
N. of obs $d = 0$	76,566	76,566	76,566	76,566	76,566			
		Instrum	ental-Variabl	e Model				
			Formal	Informal	Informa			
	Entrep.	Jobless	employee	employee	self-emp			
program coverage, $\overline{d}$	0.064***	0.055	-0.007	-0.109***	-0.004			
	(0.022)	(0.034)	(0.047)	(0.041)	(0.045)			
individual benefit, $d$	-0.043***	0.043***	-0.051***	0.01	0.042**			
	(0.007)	(0.016)	(0.018)	(0.019)	(0.021)			
Municipality Fixed-Effects	Yes	Yes	Yes	Yes	Yes			
Year dummies	Yes	Yes	Yes	Yes	Yes			
Individual covariates	Yes	Yes	Yes	Yes	Yes			
N. of obs all sample	90,648	90,648	90,648	90,648	90,648			
N. of obs $d = 0$	76,566	76,566	76,566	76,566	76,566			
Panel B: Individuals with	High-Scho	ol Diploma		·				
	i ingn bono	-	ed-Effect Mo	Effect Model				
			Formal	Informal	Informa			
	Entrep.	Jobless	employee	employee	self-emp			
program coverage, $\overline{d}$	0.086*	0.017	-0.037	-0.014	-0.052			
r88-, -	(0.047)	(0.048)	(0.073)	(0.035)	(0.039)			
individual benefit, $d$	-0.039***	0.03**	-0.053**	0.029**	0.034**			
· · · · · · · · · · · · · · · · · · ·	(0.011)	(0.013)	(0.021)	(0.014)	(0.015)			
	()			()				
Municipality Fixed-Effects	Yes	Yes	Yes	Yes	Yes			
Municipality Fixed-Effects Year dummies	Yes Yes	Yes Yes	Yes Yes	Yes Yes	Yes Yes			
Year dummies								
Year dummies Individual covariates	Yes Yes	Yes Yes	Yes Yes	Yes Yes	Yes Yes			
Year dummies Individual covariates N. of obs all sample	Yes	Yes	Yes	Yes	Yes			
Year dummies Individual covariates N. of obs all sample	Yes Yes 38,398	Yes Yes 38,398 36,522	Yes Yes 38,398	Yes Yes 38,398 36,522	Yes Yes 38,398			
Year dummies Individual covariates N. of obs all sample	Yes Yes 38,398	Yes Yes 38,398 36,522	Yes Yes 38,398 36,522	Yes Yes 38,398 36,522	Yes Yes 38,398 36,522			
Year dummies Individual covariates N. of obs all sample	Yes Yes 38,398	Yes Yes 38,398 36,522	Yes Yes 38,398 36,522 ental-Variabl	Yes Yes 38,398 36,522 e Model	Yes Yes 38,398 36,522 Informa			
Year dummies Individual covariates N. of obs all sample N. of obs $d = 0$	Yes Yes 38,398 36,522	Yes Yes 38,398 36,522 Instrum	Yes Yes 38,398 36,522 ental-Variabl Formal	Yes Yes 38,398 36,522 e Model Informal	Yes Yes 38,398			
Year dummies Individual covariates N. of obs all sample N. of obs $d = 0$	Yes Yes 38,398 36,522 Entrep.	Yes Yes 38,398 36,522 Instrum Jobless	Yes Yes 38,398 36,522 ental-Variabl Formal employee	Yes Yes 38,398 36,522 e Model Informal employee	Yes Yes 38,398 36,522 Informa self-emp			
Year dummies Individual covariates N. of obs all sample N. of obs $d = 0$ program coverage, $\overline{d}$	Yes Yes 38,398 36,522 Entrep. 0.103*	Yes Yes 38,398 36,522 Instrum Jobless 0.003	Yes Yes 38,398 36,522 ental-Variabl Formal employee -0.053	Yes Yes 38,398 36,522 e Model Informal employee 0.008	Yes Yes 38,398 36,522 Informa self-emp -0.061			
Year dummies Individual covariates N. of obs all sample N. of obs $d = 0$ program coverage, $\overline{d}$	Yes Yes 38,398 36,522 Entrep. 0.103* (0.056) -0.038***	Yes Yes 38,398 36,522 Instrum Jobless 0.003 (0.054) 0.026	Yes Yes 38,398 36,522 ental-Variabl Formal employee -0.053 (0.082) -0.04*	Yes Yes 38,398 36,522 e Model Informal employee 0.008 (0.045) 0.011	Yes Yes 38,398 36,522 Informa self-emp -0.061 (0.051)			
Year dummies Individual covariates N. of obs all sample N. of obs $d = 0$ program coverage, $\overline{d}$ individual benefit, $d$	Yes Yes 38,398 36,522 Entrep. 0.103* (0.056)	Yes Yes 38,398 36,522 Instrum Jobless 0.003 (0.054)	Yes Yes 38,398 36,522 ental-Variabl Formal employee -0.053 (0.082)	Yes Yes 38,398 36,522 e Model Informal employee 0.008 (0.045)	Yes Yes 38,398 36,522 Informa self-emp -0.061 (0.051) 0.041**			
Year dummies Individual covariates N. of obs all sample N. of obs $d = 0$ program coverage, $\overline{d}$ individual benefit, $d$ Municipality Fixed-Effects	Yes Yes 38,398 36,522 Entrep. 0.103* (0.056) -0.038*** (0.012)	Yes Yes 38,398 36,522 Instrum Jobless 0.003 (0.054) 0.026 (0.017)	Yes Yes 38,398 36,522 ental-Variabl Formal employee -0.053 (0.082) -0.04* (0.024)	Yes Yes 38,398 36,522 e Model Informal employee 0.008 (0.045) 0.011 (0.017)	Yes Yes 38,398 36,522 Informa self-emp -0.061 (0.051) 0.041** (0.018)			
Year dummies Individual covariates N. of obs all sample N. of obs $d = 0$ program coverage, $\overline{d}$ individual benefit, $d$ Municipality Fixed-Effects Year dummies	Yes Yes 38,398 36,522 Entrep. 0.103* (0.056) -0.038*** (0.012) Yes Yes	Yes Yes 38,398 36,522 Jobless 0.003 (0.054) 0.026 (0.017) Yes Yes	Yes Yes 38,398 36,522 ental-Variabl Formal employee -0.053 (0.082) -0.04* (0.024) Yes Yes	Yes Yes 38,398 36,522 e Model Informal employee 0.008 (0.045) 0.011 (0.017) Yes Yes	Yes Yes 38,398 36,522 Informa self-emp -0.061 (0.051) 0.041** (0.018) Yes Yes			
Municipality Fixed-Effects Year dummies Individual covariates N. of obs all sample N. of obs $d = 0$ program coverage, $\overline{d}$ individual benefit, $d$ Municipality Fixed-Effects Year dummies Individual covariates N. of obs all sample	Yes Yes 38,398 36,522 Entrep. 0.103* (0.056) -0.038*** (0.012) Yes	Yes Yes 38,398 36,522 Jobless 0.003 (0.054) 0.026 (0.017) Yes	Yes Yes 38,398 36,522 ental-Variabl Formal employee -0.053 (0.082) -0.04* (0.024) Yes	Yes Yes 38,398 36,522 e Model Informal employee 0.008 (0.045) 0.011 (0.017) Yes	Yes Yes 38,398 36,522 Informa self-emp -0.061 (0.051) 0.041** (0.018) Yes			

Table B5: Indirect and Direct Effects on Occupational Choices, With and Without High School

\*\*\*, \*\*, \* represent statistical significant at the 1%, 5% and 10% levels, respectively. Standard errors in parentheses are clustered by municipality. All coefficients are estimated using Seemingly Unrelated Regressions (SUR). The indirect effect (program coverage) is estimated using the sample of non-participants, whereas the direct effect (individual benefit) is estimated using all sample and bias corrected according to Lemma 4.1. Fixed-Effect models are estimated using the within-group method. In the Instrumental-Variable models, 'program coverage' is instrumented by the interactions between municipal quotas and year dummies.