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# Constrained, or not constrained?

Identifying credit constrained firms in the Brazilian economy

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# Motivation

- ❖ Credit constraints are born from **asymmetrical information** between banks and borrowers (Stiglitz and Weiss 1981)
- ❖ Financially constrained firms have lower levels of:
  - ❖ **Investment** (Choi et al. 2018);
  - ❖ **Sales** (Banerjee and Duflo 2014; McKenzie 2017);
  - ❖ **Engagement in international trade** (Manova, Wei, and Zhang (2015), Zia (2008), and Minetti and Zhu (2011));
  - ❖ **Rentability** (Banerjee and Duflo (2014), McKenzie (2017));
- ❖ Since investment and trade flows are engines of entrepreneurship and economic growth, credit constraints jeopardize economic growth and social welfare.

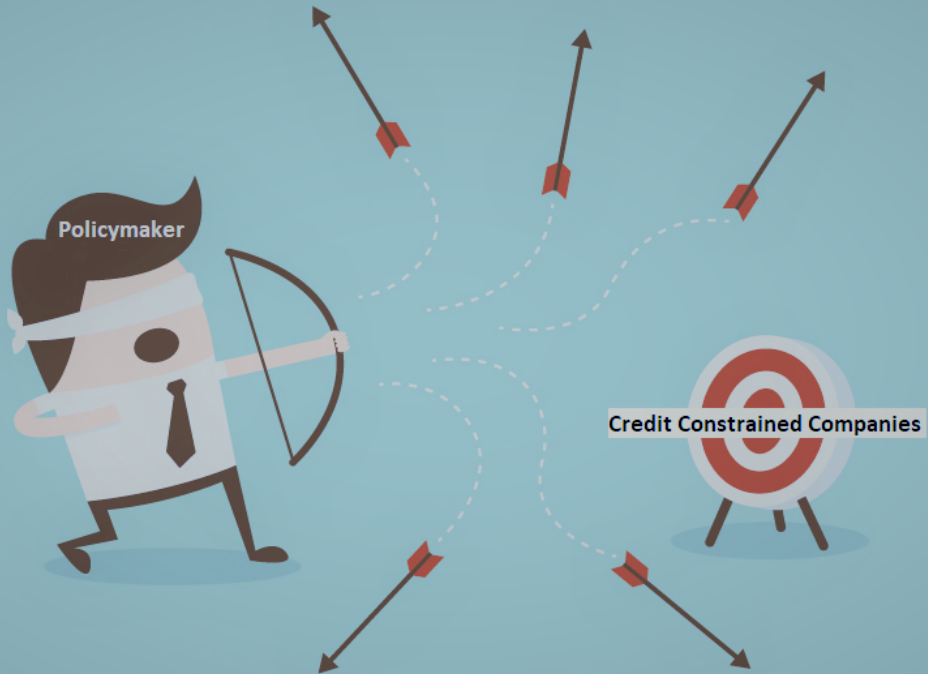
# Defining credit constraints

- ❖ Constrained companies hold two main characteristics:
  - i. access to investment opportunities with **expected returns higher enough to justify their execution**; despite that,
  - ii. they are **unable to fund these investments**.
- ❖ Formally, we can understand credit-constrained firms as those who cannot equalize the marginal productivity of capital to the opportunity cost of capital in the economy.

# Measuring credit constraints

- ❖ Several authors have proposed firm-level measures for credit constraints (Lamont, Polk, and Saa-Requejo 2001; Whited and Wu 2006; Hadlock and Pierce 2010; and Schauer, Elsas, and Breitkopf 2019).
- ❖ In the context of a developed economy, they do not seem to work as expected.

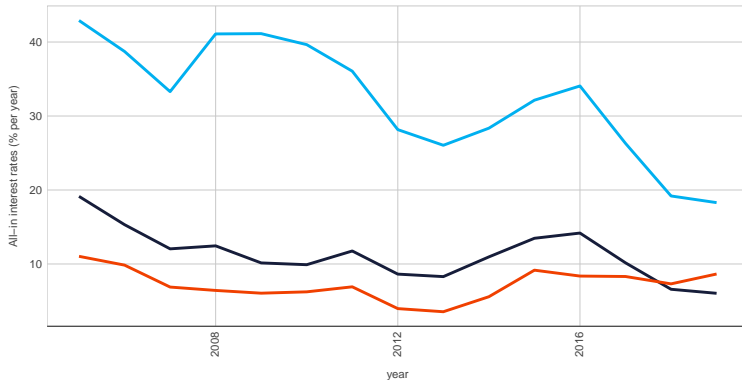
"We find that none of the five measures we evaluate is able to identify firms that behave as if they were in fact constrained." (Farre-Mensa and Ljungqvist 2016, p. 33)



# The treatment

## Credit cost in the Brazilian economy

Only operations with firms



Source: BCB-DSTAT and BNDES.

# The dataset

- ❖ RAIS (2012-2017)
  - ❖ Information about the firm (age, sector, address)
  - ❖ Information about the labor force (number of workers, qualification, earnings)
- ❖ SERASA (2012-2017)
  - ❖ Accounting information (assets, liabilities and results)
  - ❖ Covers a subset of publicly and privately traded companies
- ❖ SECEX (1998 - 2018)
  - ❖ Export and import status for all Brazilian firms
- ❖ BNDES (2002 - 2019)
  - ❖ Loans at the firm-level per credit line

# The empirical strategy (1/2)

- ❖ Triple difference estimator (3D) (Berck and Villas-Boas 2016) combined with Genetic Matching (Diamond and Sekhon 2012).
- ❖ Firms were matched within cohorts, that were defined by a time span of four years - tagged as  $t_{-2}$ ,  $t_{-1}$ ,  $t_0$ , and  $t_{+1}$ . Treatment is assigned in  $t_0$ .
- ❖ Genetic matching pairs observations based on a Generalized Mahalanobis Distance (GMD). The control vector took values observed in years  $t_{-2}$  and  $t_{-1}$  for the following variables: **number of employees, leverage ratio, year sales, operational profit, cash flow, and the total expenses with interests.**



# Balance before matching

Balance between treated and untreated units

$X$	Period	$\mu_t$	$\mu_c$	$t-ratio$	$\Gamma_{ct}$
y	-1	2.90	2.75	4.34	-0.13
employee.size	-1	4.36	3.91	12.66	-0.09
leverage.size	-1	2.23	2.10	3.87	-0.13
sales.size	-1	2.23	2.01	2.64	-1.08
operating.profit	-1	4.58	2.82	1.83	-1.33
cash.flow	-1	129.67	84.27	2.04	-0.003
interest.expenses	-1	-1.24	-2.62	3.03	-0.97
y	-2	2.75	2.66	2.63	-0.12
employee.size	-2	4.27	3.89	10.37	-0.07
leverage.size	-2	2.11	2.02	2.73	-0.12
sales.size	-2	2.23	2.06	1.59	-0.82
operating.profit	-2	4.19	4.00	0.26	-1.47
cash.flow	-2	116.27	79.12	1.78	0.02
interest.expenses	-2	-1.20	-1.71	1.49	-0.90

Untreated units are 66912.

Treated units are 1797.

Variables are in logarithms.

$\Gamma_{ct}$  calculated according to Rubin and Imbens (2015).

# Balance after matching

Balance between treated and control units

$X$	Period	$\mu_t$	$\mu_c$	$t - ratio$	$\Gamma_{ct}$
y	-1	2.92	2.91	0.22	0.01
employee.size	-1	4.39	4.38	0.07	0.01
leverage.size	-1	2.25	2.24	0.14	0.001
sales.size	-1	2.20	2.19	0.15	-0.01
operating.profit	-1	4.62	6.69	-0.91	-0.91
cash.flow	-1	133.00	115.35	0.60	0.22
interest.expenses	-1	-1.25	-1.37	0.24	0.51
y	-2	2.78	2.78	0.08	0.02
employee.size	-2	4.28	4.29	-0.24	0.02
leverage.size	-2	2.14	2.14	-0.01	0.01
sales.size	-2	2.26	2.29	-0.21	0.04
operating.profit	-2	4.24	5.87	-0.87	-1.01
cash.flow	-2	119.44	104.90	0.54	0.27
interest.expenses	-2	-1.22	-1.18	-0.10	0.32

Control units are 1669.

Treated units are 1740.

Variables are in logarithms.

$\Gamma_{ct}$  calculated according to Rubin and Imbens (2015).

# The empirical strategy (2/2)

- ❖ The cohorts are “stacked” (Gormley and Matsa 2011; Deshpande and Li 2019; and Joaquim and Doornik 2019), and the 3D estimator is implemented within the matched sample through equation (1):

$$Y_{i,t} = \alpha_{i,cohort} + \alpha_t + \beta_1 Post_t + \beta_2 BNDES_i + \beta_3 Constr_i \quad (1) \\ + \beta_4 (Post_t * BNDES_i) + \beta_5 (Post_t * Constr_i) + \\ + \beta_6 (Post_t * BNDES_i * Constr_i) + \epsilon_{i,t}$$

# Results (1/2)

	Impacts on total assets with continuous measures of credit constraints			
	KZ	WW	SA	FCP
d.post	-0.02** (0.01)	-0.03*** (0.01)	-0.01 (0.02)	-0.08*** (0.03)
d.post:d.bndes	0.13*** (0.01)	0.13*** (0.01)	0.29*** (0.04)	0.15*** (0.05)
d.post:kz.index	0.0001*** (0.0000)			
d.post:d.bndes:kz.index	0.0000 (0.0001)			
d.post:ww.index		-0.0000 (0.0000)		
d.post:d.bndes:ww.index		0.0000 (0.0000)		
d.post:sa.index			0.01 (0.01)	
d.post:d.bndes:sa.index			0.09*** (0.02)	
d.post:fcf.index				-0.02 (0.01)
d.post:d.bndes:fcf.index				0.02 (0.03)
Observations	13,636	13,496	13,636	2,908
R <sup>2</sup>	0.03	0.02	0.03	0.03
Adjusted R <sup>2</sup>	-0.30	-0.30	-0.29	-0.30

Note:

\* p<0.1; \*\* p<0.05; \*\*\* p<0.01  
Standard errors are clustered by firm-cohort.

# Results (2/2)

	Impacts on total assets with categorical measures of credit constraints			
	KZ	WW	SA	FCP
d.post	-0.02 (0.01)	-0.02 (0.01)	-0.03** (0.01)	-0.02 (0.04)
d.post:d.bndes	0.13*** (0.01)	0.13*** (0.01)	0.13*** (0.01)	0.11*** (0.02)
d.post:kz.dconstrained	-0.05*** (0.02)			
d.post:d.bndes:kz.dconstrained	-0.01 (0.03)			
d.post:ww.dconstrained		-0.11*** (0.02)		
d.post:d.bndes:ww.dconstrained		0.04 (0.03)		
d.post:sa.dconstrained			0.003 (0.02)	
d.post:d.bndes:sa.dconstrained			0.10*** (0.04)	
d.post:fcf.dconstrained				-0.09*** (0.03)
d.post:d.bndes:fcf.dconstrained				0.05 (0.04)
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Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Standard errors are clustered by firm-cohort.

# Robustness check (1/2)

- As a robustness check, we'll use two well-established findings in the literature on financial constraints to evaluate the proposed 3D estimator:
  - i. that firms engaged in international trade are typically not credit-constrained (Roberts and Tybout 1997; Zia 2008; Minetti and Zhu 2011; Manova, Wei, and Zhang 2015); and
  - ii. that smaller firms are particularly prone to suffer from financial constraints (Beck, Demirgüç-Kunt, and Maksimovic 2005; Beck and Demirguc-Kunt 2006; and Hutchinson and Xavier 2006).

# Robustness check (2/2)

Impacts on total assets of subsidized credit (stacking method)

	Heterogeneous impacts in terms of:	
	International Trade	Size
d.post	-0.03** (0.01)	-0.05* (0.03)
d.post:d.bndes	0.15*** (0.01)	0.30*** (0.04)
d.post:as.factor(international.trade.d)1	-0.004 (0.02)	
d.post:d.bndes:as.factor(international.trade.d)1	-0.06** (0.02)	
d.post:employee.size.cohort		0.004 (0.01)
d.post:d.bndes:employee.size.cohort		-0.04*** (0.01)
Observations	13,636	13,636
R <sup>2</sup>	0.03	0.03
Adjusted R <sup>2</sup>	-0.30	-0.29

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01  
Standard errors are clustered by firm-cohort.

# Final Remarks

- ❖ Except by the SA index, firms tagged as credit-constrained are not behaving differently from those that are classified as unconstrained.
- ❖ Evidently, more in-depth tests must be performed. In particular, it would be essential to use an exogenous credit shock (as Farre-Mensa and Ljungqvist (2016)) to implement a more robust estimation of equation (1).
- ❖ Finding a reliable measure is crucial for public initiatives that try to tackle this relevant market failure.



# Thank you