Business as Usual? Bank Lending under Credit Relief Programs

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- The COVID-19 pandemic caused a liquidity crunch but initial policy reactions were laid out rather rapidly
- The favorable outcome of the support policies reflects general equilibrium effects on credit. Little is known about how banks cope with the support policies in extending new credit.
- Do banks sustain lending to the economy and support viable distressed borrowers as intended, netting on the effect of the credit support programs?
 - The design of these programs needs to strike a tricky balance.
 - Scarce empirical evidence on the efficacy of credit support programs when banks are not under stress
 - Scarce empirical evidence on the externalities of credit support programs

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Target Measure Versus Blanket Measure



This Paper

- compares lending (non-public-guaranteed) of banks with different exposure to credit relief programs around the peak of policy uptakes using difference-in-differences
- addresses the endogeneity concern using the Bartik Instrument (Goldsmith-Pinkham, et al., 2020)
- We find strong evidence of policy externalities. Banks participating more in the credit guarantee programs,
 - increase lending in general as intended but decrease loan supply within the program to preserve lending to less-risky more productive firms outside the program with better conditions
 - contraction effect of moratorium programs and higher bank risk-taking behaviors

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Link to the Literature

- Public Guarantee Programs: Bachas et al. (2021), Custodio et al. (2022), de Blasio et al. (2018), Lagazio et al. (2021), and Zecchini and Ventura (2009)
- Credit Demand and Supply during COVID-19: Acharya and Steffen (2020), Li et al. (2020)), Chodorow-Reich et al. (2021), Kapan and Minoiu (2021), Greenwald et al. (2020), Berger et al. (2021), and Couaillier et al. (2022)
- Government Intervention during COVID-19: Altavilla et al. (2021), Minoiu et al. (2022), Cascarino et al. (2022), Core and De Marco (2020), Altavilla et al. (2020), Arping et al. (2010), Gourinchas et al. (2021), Kozeniauskas et al. (2022)
- Sovereign-Bank-Corporate Nexus: Acharya et al. (2014), Acharya and Steffen (2015), Acharya et al. (2018), Bottero et al. (2020), Leonello (2021), and Bonfim et al. (2022)

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Institutional Background and Data

- Significant impact of PGSs on lending in Portugal
 - By June, 2020 Newly originated loans subject to PGSs amounts to 2.1% of total loans, ranked the 2nd in the Euro zone after Spain

Public Guarantee Schemes Banco Português de Fomento Statistics

- Extended use of moratoria
 - More than 20% of their reported loans to NFCs and HHs were under moratoria up to June 2020

Moratorium EBA Report: First Insights into the COVID-19 Impacts ● Timeline

• The Portuguese banking system was in a resilient position at the onset of the COVID-19 crisis but there was a deterioration in risk perceptions

Loan Loss Provision

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- CRC: monthly loan-level data (threshold of 50 euros, PG Versus NPG)
- MFI Statistics: monthly bank-level data
- **CB:** annual firm-level data
- SPAI geodemographic infos on banks, MGSs, firms
- SIAC: in-house credit assessment system
- CITIUS: judicial restructuring (PER) process
- July 2019 March 2020, June 2020 December 2020, 251,131 firms; 55 banks; 2,079,823 bank-firm-quarter obs

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Firms under Different Credit Relief Programs

					Bank Lendir
	PGS	Moratorium	None	Total	under Credi
Firm Chars					Programs
Firm Assets (thousand)	1,366.22	2,698.86	1,721.65	1,848.95	
Risk	0.04	0.09	0.12	0.11	
Leverage	0.21	0.58	3.77	3.06	Motivation
Profitability	0.04	-0.11	-9.47	-7.45	Institutional
Industry affectedness	0.17	0.18	0.11	0.12	and Data
Region affectedness	0.07	0.08	0.08	0.08	Identification
Bank Chars					Strategy
PG Exp.	0.47	0.31	0.28	0.33	Results
Morat. Exp.	8.31	7.42	6.91	7.41	Conclusion
Bank Assets (Billion)	13.90	10.10	8.79	10.40	Appendix
Foreign Bank	0.32	0.35	0.34	0.34	Appendix
Bank Liquidity	0.03	0.02	0.02	0.02	References
NPLs	0.03	0.03	0.04	0.03	
Capital Ratio	0.14	0.16	0.17	0.16	
Obs	14,498	36,421	200,212	251,131	

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Difference-in-Differences

• We compare lending between the pre- and post-COVID periods of banks with different exposure to the two credit relief programs

$$Credit_{fbt} = \sum_{t=1}^{n} \beta_{1,t} Period_t imes P\widehat{G Exp_b} +$$

 $\sum_{t=1}^{n} \beta_{2,t} Period_t imes Morat Exp_b +$
 $\sum_{t=1}^{n} \alpha_t Period_t imes \mathbf{BankChars}_b +$
 $\gamma_{ilst} + \omega_{fb} + \epsilon_{fbt}$

- Credit: Total + NPG
- BankChars: COVID Exp., Bank Size, Liquidity, NPLs, Capital Ratio, Foreign Dummy

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• Fixed Effects: ILST (Degryse et al. (2019)), Bank-Firm

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Bank's Exposure to Credit Relief Programs

1 Bank's Exposures to Public Guarantee Programs

 $PG \ Exp_b = \frac{\sum_{f=1}^{n} PG_{fb,March-May \ 2020}}{Assets_{b,2019}}$

Ø Bank's Exposures to Moratorium Programs

$$Morat \ Exp_b = \frac{\sum_{f=1}^n Morat_{fb,March-May \ 2020}}{Assets_{b,2019}}$$

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IV for Bank Participation

• Bartik instrument (Goldsmith-Pinkham et al.,2020): shift-share predictor of policy uptake from March 2020 to May 2020 by municipality and industry

$$Bartik_b = \frac{\sum_{m=1}^{n} \sum_{s=1}^{m} Credit \ Share_{msb,2019} \times Shifter_{ms-b,March-May \ 2020}}{Assets_{b,2019}}$$

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IV for Bank Participation

Relevance

- Banks with a larger pre-pandemic credit portfolio that is affected by COVID-19 are more likely to participate and increase uptake
- Statistically important effects on PGS Participation

2 Exclusion • Disbursement of PG Loans

• Affected sectors and the expansion regions were not anticipated, thus orthogonal to bank's lending decisions

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Credit Supply

1 General **DOLS**



Figure: Bank's Exposure to Public Guarantee Programs $(\beta_{1,t})$

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Figure: Bank's Exposure to Moratorium Programs $(\beta_{2,t})$

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Credit Supply

2 Within and Outside Schemes



Figure: Firms with PG Loans



Figure: Other Firms

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Credit Supply

③ Static Specification

	Total Credit	NPG Credit	Total Credit	NPG Credit	Total Credit
	All Firms	All Firms	With PG	With PG	Without PG
Post \times PG Exp.	0.054***	-0.048***	0.268***	-0.110***	0.020***
	(0.003)	(0.003)	(0.015)	(0.017)	(0.003)
Post $ imes$ Morat. Exp.	-0.028***	-0.022***	-0.045***	-0.015	-0.023***
	(0.002)	(0.002)	(0.008)	(0.010)	(0.002)
		First Stag	ge		
$Post\timesBartik$	6.063***	6.063***	6.192***	6.192***	6.055***
	(0.007)	(0.007)	(0.033)	(0.033)	(0.007)
Bank Chars	Yes	Yes	Yes	Yes	Yes
ILST FE	Yes	Yes	Yes	Yes	Yes
Bank-Firm FE	Yes	Yes	Yes	Yes	Yes
Observations	2,079,823	2,079,823	155,639	155,639	1,908,207
F	819.405	192.384	342.443	9.448	526.333

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Lending Conditions

	Pricing				Collateral				
	All Firms	With PG	With Moratorium	None	All Firms	With PG	With Moratorium	None	
Post \times PG Exp.	-0.952***	-1.646**	-1.017***	-0.570***	-0.051**	0.106	-0.069**	-0.082***	
	(0.148)	(0.763)	(0.273)	(0.150)	(0.020)	(0.146)	(0.031)	(0.025)	
Post \times Morat. Exp.	0.627***	1.058**	0.655***	0.371***	0.047***	-0.052	0.054***	0.063***	
	(0.100)	(0.502)	(0.185)	(0.097)	(0.013)	(0.096)	(0.020)	(0.016)	
First Stage									
Post $ imes$ Bartik	0.602***	0.308***	0.732***	0.685***	0.602***	0.308***	0.732***	0.685***	
	(0.038)	(0.061)	(0.069)	(0.040)	(0.038)	(0.061)	(0.069)	(0.040)	
Bank-Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	
Time FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	
Observations	70,460	8,656	15,761	16,073	70,460	8,656	15,761	16,073	
Adjusted R^2	0.014	0.008	0.021	0.009	0.007	0.004	0.012	0.008	
F	20.410	2.062	10.992	6.237	8.040	1.843	3.773	3.015	

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Heterogeneity (Probability of obtaining a NPG loan)

1 Risk





Figure: Bank's Exposure to Public Guarantee Programs $(\beta_{1,t})$

Figure: Bank's Exposure to Moratorium $Programs(\beta_{2,t})$

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Heterogeneity

Ø Zombie Firms



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Figure: Bank's Exposure to Public Guarantee Programs $(\beta_{1,t})$

Figure: Bank's Exposure to Moratorium $Programs(m{eta}_{2,t})$

Heterogeneity

8 Productivity





Figure: Bank's Exposure to Public Guarantee Programs $(\beta_{1,t})$

Figure: Bank's Exposure to Moratorium $Programs(\beta_{2,t})$

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- Our results reveal strong credit externalities: a reallocation effect of public guarantee programs and a contraction effect of moratorium programs
- Our results are robust to
 - isolating credit demand from credit supply
 - alternative IV: Google Playstore review rating on bank's mobile app (Core and De Marco (2020))
 - controlling for COVID-related liquidity injection
 - The reallocation effect of the PGS is more prominent in firms with refinancing needs, consistent with Altavilla et al. (2021)
 - The effect is persistent at the firm level

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Credit Relief Programs

		Relief	
Guarantee Programs	- New credit with double screening		
	- Quasi-Target measure		
	- Q4, 2020 EUR 8,105 million	Motivation	
	- Guaranteed (mainly) at 80 to 90 $\%$	Institution	
	- Maximum maturity (mainly) 3 to 6 years	Backgrour and Data	
	- Cost of the guarantee (mainly) 25-175 bps	Identificat Strategy	
Moratorium	- Existent credit at firm's request	Results	
	- Blanket measure	Conclusion	
	- Applications: Mar20 to Mar21; End: Sep21 to Dec21	Appendix	
		References	
Other	Non-refundable liquidity incentives, tax relief programs, and		
	lay-off programs		

Bank Lending under Credit

Credit Relief Programs - Eligibility Criteria

Back Guarantee Programs:

- Positive equity in 2019
- No active default incidents
- No active debts to the Social Security and Tax Authority
- Not classified as "undertaking in difficulty" firms as defined in the Commission Regulation (EU) No 651/2014 article 2 number 18
- Decrease in sales of at least 40% between March and May 2020, when compared with the same period in 2019

Moratorium:

- Firms could not be in a credit default situation (more than 90 days)
- No active debts to the Social Security and Tax Authority

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• Significant impact of PGSs on lending in Portugal



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• Extended use of moratoria



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Disbursement of PG Loans







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Distribution of Bank Exposure to Credit Relief Programs

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Balance Checks on Moratorium Exposure • Back

	High Morat. Exp.	Low Morat. Exp.	Diff.	S.E.
Observations	27	28		
PG Exp.	0.429	0.047	0.382**	(0.146)
Morat. Exp.	10.943	2.357	8.586***	(1.142)
Previous PG Exp.	0.366	0.047	0.319	(0.196)
Bank Size	7.376	6.200	1.176**	(0.566)
Foreign Bank	0.370	0.357	0.013	(0.132)
Bank Liquidity	0.030	0.015	0.015	(0.010)
NPLs	0.023	0.042	-0.019	(0.015)
Capital Ratio	0.146	0.189	-0.044	(0.054)
Credit Ratio	0.716	0.662	0.054	(0.062)
Covid Exp.	39.443	29.170	10.273	(6.908)

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Balance Checks on Public Guarantee Exposure Back

	Exposed Banks	Non-Exposed Banks	Diff.	S.E.
Observations	13	42		
PG Exp.	0.991	0.000	0.991***	(0.216)
Morat. Exp.	10.102	5.480	4.622***	(1.473)
Previous PG Exp.	0.857	0.002	0.855**	(0.369)
Bank Size	9.338	5.984	3.354***	(0.570)
Foreign Bank	0.154	0.429	-0.275**	(0.130)
Bank Liquidity	0.046	0.015	0.032***	(0.010)
NPLs	0.042	0.029	0.013	(0.015)
Capital Ratio	0.113	0.185	-0.072*	(0.038)
Credit Ratio	0.685	0.690	-0.005	(0.059)
Covid Exp.	37.088	33.324	3.764	(5.074)

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Credit Supply: Bank-Firm Level OLS Results

10 05 0 -01 0 03 05 04 ---- Non-Public-Guaranteed Credit Total Credit

Figure: Bank's Exposure to Public Guarantee Programs $(\beta_{1,t})$

Figure: Bank's Exposure to Moratorium Programs $(\beta_{2,t})$

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Credit Supply: Firm Level IV Results

	Total Credit	NPG Credit	Total Credit	NPG Credit	Total Credit	Total Credit	Total Credit
	All Firms	All Firms	With PG	With PG	Without PG	With Moratorium	None
Post \times PG Exp.	0.023***	-0.035***	0.056	-0.123**	-0.005	0.002	-0.005
	(0.004)	(0.005)	(0.034)	(0.051)	(0.004)	(0.009)	(0.005)
Post \times Morat. Exp.	-0.013***	-0.013***	0.002	0.027	-0.010***	-0.021***	-0.010***
	(0.003)	(0.003)	(0.024)	(0.032)	(0.003)	(0.006)	(0.003)
Post \times Covid Exp.	-0.003	-0.020***	0.054**	-0.036	-0.012***	0.003	-0.014***
	(0.003)	(0.003)	(0.027)	(0.034)	(0.003)	(0.005)	(0.003)
			First S	tage			
Post \times Bartik	6.781***	6.781***	6.977***	6.977***	6.778***	6.875***	6.775***
	(0.012)	(0.012)	(0.081)	(0.081)	(0.013)	(0.041)	(0.014)
Bank Chars	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm Chars	Yes	Yes	Yes	Yes	Yes	Yes	Yes
ILST FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	1,055,095	1,055,095	40,203	40,203	984,512	136,374	789,992

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Who Were Granted PG and NPG Loans?





Figure: Firm Risk Deciles

Figure: Firm Productivity Deciles

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- ACHARYA, V., I. DRESCHSLER, AND P. SCHNABL (2014): "A Pyrrhic Victory? Bank Bailouts and Sovereign Credit Risk," *The Journal of Finance*, 69, 2689–2739.
- ACHARYA, V., T. EISERT, C. EUFINGER, AND C. HIRSCH (2018): "Real effects of the sovereign debt crisis in Europe: Evidence from syndicated loans," *Review of Financial Studies*, 31, 2855–2896.
- ACHARYA, V. AND S. STEFFEN (2015): "The "greatest" carry trade ever? Understanding eurozone bank risks," *Journal of Financial Economics*, 115, 215–236.
- ACHARYA, V. V. AND S. STEFFEN (2020): "The risk of being a fallen angel and the corporate dash for cash in the midst of COVID," *Review of Corporate Finance Studies*, 9, 430–471.

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- ALTAVILLA, C., F. BARBIERO, M. BOUCINHA, AND L. BURLON (2020): "The Great Lockdown: Pandemic Response Policies and Bank Lending Conditions," *ECB Working Paper*, 2465.
- ALTAVILLA, C., A. ELLUL, M. PAGANO, A. POLO, AND T. VLASSOPOULOS (2021): "Loan Guarantees, Bank Lending and Credit Risk Reallocation," SSRN Electronic Journal.
- ARPING, S., G. LÓRÁNTH, AND A. D. MORRISON (2010): "Public initiatives to support entrepreneurs: Credit guarantees versus co-funding," *Journal of Financial Stability*, 6, 26–35.
- BACHAS, N., O. S. KIM, AND C. YANNELIS (2021): "Loan guarantees and credit supply," *Journal of Financial Economics*, 139, 872–894.

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References III

- BERGER, A. N., C. H. S. BOUWMAN, L. NORDEN, R. A. ROMAN, G. F. UDELL, AND T. WANG (2021): "Is a Friend in Need a Friend Indeed? How Relationship Borrowers Fare during the COVID-19 Crisis," *Federal Reserve Bank* of Philadelphia Working Paper, 21-13.
- BONFIM, D., M. A. FERREIRA, F. QUEIRO, AND Z. SUJIAO (2022): "Sovereign-Bank Diabolic Loop: The Government Procurement Channel," *Nova SBE Working Paper Series*.
- BOTTERO, M., S. LENZU, AND F. MEZZANOTTI (2020): "Sovereign debt exposure and the bank lending channel: Impact on credit supply and the real economy," *Journal of International Economics*, 126, 103328.
- CASCARINO, G., R. GALLO, F. PALAZZO, AND E. SETTE (2022): "Public guarantees and credit additionality during the Covid-19 pandemic," *MOFiR Working Paper*, 172.

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References IV

CHODOROW-REICH, G., O. DARMOUNI, S. LUCK, AND M. PLOSSER (2021): "Bank liquidity provision across the firm size distribution," *Journal of Financial Economics*.

- CORE, F. AND F. DE MARCO (2020): "Public Guarantees for Small Businesses in Italy during Covid-19," *Working Paper*.
- COUAILLIER, C., M. LO DUCA, A. REGHEZZA, AND C. R. D'ACRI (2022): "Caution: Do Not Cross! Capital Buffers and Lending in COVID-19 Times," *ECB Working Paper.*
- CUSTODIO, C., , D. BONFIM, AND C. C. RAPOSO (2022): "Supporting small firms through recessions and recoveries," *Working Paper*.
- DE BLASIO, G., S. DE MITRI, A. D'IGNAZIO, P. FINALDI RUSSO, AND L. STOPPANI (2018): "Public guarantees to SME borrowing. A RDD evaluation," *Journal of Banking and Finance*, 96, 73–86.

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References V

- DEGRYSE, H., O. DE JONGHE, S. JAKOVLJEVIĆ, K. MULIER, AND G. SCHEPENS (2019): "Identifying credit supply shocks with bank-firm data: Methods and applications," *Journal of Financial Intermediation*, 40, 100813.
 GOURINCHAS, P.-O., KALEMLI-ÖZCAN, V. PENCIAKOVA, AND N. SANDER (2021): "COVID-19 and Small- and Medium-Sized Enterprises: A 2021 "Time Bomb"?" *AEA Papers and Proceedings*, 111, 282–286.
- GREENWALD, D. L., J. KRAINER, AND P. PAUL (2020): "The Credit Line Channel," Federal Reserve Bank of San Francisco Working Paper, 26.
- KAPAN, T. AND C. MINOIU (2021): "Liquidity Insurance vs. Credit Provision: Evidence from the COVID-19 Crisis," *SSRN Electronic Journal*, 3773328.
- KOZENIAUSKAS, N., P. MOREIRA, AND C. SANTOS (2022): "On the cleansing effect of recessions and government policy: Evidence from Covid-19," *European Economic Review*, 144, 104097.

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References VI

- LAGAZIO, C., L. PERSICO, AND F. QUERCI (2021): "Public guarantees to SME lending: Do broader eligibility criteria pay off?" *Journal of Banking Finance*, 133, 106287.
- LEONELLO, A. (2021): "Government guarantees and the two-way feedback between banking and sovereign debt crises," *Journal of Financial Economics*, 130, 592–619.
- LI, L., P. E. STRAHAN, AND S. ZHANG (2020): "Banks as lenders of first resort : evidence from the covid-21 crisis," *NBER working papaer series*, 27256, 1 37.
- MINOIU, C., , R. ZARUTSKIE, AND A. ZLATE (2022): "Motivating Banks to Lend? Credit Spillover Effects of the Main Street Lending Program," *Working Paper*.
- ZECCHINI, S. AND M. VENTURA (2009): "The impact of public guarantees on credit to SMEs," *Small Business Economics*, 32, 191–206.

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