Climate supervisory shocks and bank credit lending Empirical evidence from microdata

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Disclaimer

The views expressed in this presentation are those of the author and do not necessarily reflect the official positions of Banca d'Italia.

Introduction

Data and empirical strategy

Results

Conclusions

Appendix

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- Partial quantification of the effects of climate policy on banks' credit supply

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- Partial quantification of the effects of climate policy on banks' credit supply
 - Data gaps on SMEs and non-listed firms: limited non-financial disclosure
 - $\circ~$ IT: 47% of loans to firms, 67% of Value Added, 79% of employment
 - Under/Overestimation?

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2. Did banks apply a different cost of lending post-shock?

• NO. Predominance of credit reallocation effect



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 - The CST did not explicitly require banks to consider the exposure to climate risk with climate forward-looking data

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 - YES. Worse pricing and less credit for *green-meets-green attitude* (Degryse et al., 2023, Kacperczyk and Peydró, 2021)
 - NO. Less likely to establish new relationships and overemphasis of climate targets (greenwashing), Ehlers and De Greiff (2021), Giannetti et al. (2023), Sastry et al. (2024))

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Contribution

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Data and empirical strategy

(Micro) Data

AnaCredit

- Loan-level data from bank-firm credit registry of lending within Italy
- Performing credit lines granted by multi-banks to (non-financial) firms

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FINREP/COREP and Cerved

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Short-term effects

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Short-term effects

- Shock 1: September 2020 February 2021
- Shock 2: September 2021 February 2022

 \simeq 45K firms, 38 banking groups (11 SIs and 27 LSIs), 6 months time-window (3 before, 3 after)
• Self-disclosed information by firms and banks (as in Carbone et al. (2021))

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- Dummy variable in case the company sets targets or objectives in a given time frame to be achieved on emission reductions from business operations
- Non-Financial Reports, Sustainability Reports, Corporate Social Responsibility, Environmental Reports, and Annual Reports

Climate commitments before the shocks: 3 banks (SIs), 24 large and listed firms

Example bank commitment: Intesa San Paolo

INTESA 🖂 SANDAOLO					
	Net-Zer 	Net-Zero Commitment With the choice to pursue the Net-Zero objective by 2050 both for its own emissions and for those of the loan and investment portfolios, asset management and insurance activities, the Group has joined all the Net-Zero Initiatives of UNEP FI: Net-Zero Banking Alliance (NZBA), Net- Zero Asset Managers Initiative (NIZMAI), Net-Zero Asset Owner Initiative (NIZAA) and Net-Zero Insurance Alliance (NZIA). Carbon neutrality: for its emissions by 2030; Net-zero objective by 2050 for all main business lines: credit, investments and insurance sector;			
	<u>Carbon n</u> <u>Net-zero n</u> sector;				
	Intermed Automoti (CRE). The issues of	liate targets by 20 we and Coal Mining e sectors covered by the portfolio of non	30 for financed emissions of Oil&Gas, Power Generation, I and, from 2023, Iron&Steel and Commercial Real Estate (target settings account for more than 66% of the financed -financial companies in the sectors indicated by the NZBA;		
	 The exposi- unconven respective 	sure target is zero b itional Oil & Gas sec 2021 policies	vy 2025 for the coal mining sector and by 2030 for the .tor, in line with the exclusion policy (phase-out) in the		
	 Target ar Wealth r through t companie 	nd policy nanagement emiss he intermediate tar is active in wealth n	ions: the Net-Zero commitment has also been consolidated gets - published in October 2022 by the Intesa Sanpaolo nanagement adhering to the NZAMI and the NZAOA		

Long-term Environmental Goals (Fiscal 2021-2030) (Established in December 2019)

Priority issues		Long-term Environmental Goals (Fiscal 2021-2030)		
Creating low-carbon society		 Reduce emissions of greenhouse gases (GHG) to curtail rises in average global temperatures. <target> Reduce GHG emissions (Scope 1+2) by 35% from fiscal 2017 level by fiscal 2030.</target> 		
Efficient use of resources	Water	 Preserve water resources by reducing water withdrawal. <target> Reduce water withdrawal by 12% from fiscal 2018 level by fiscal 2030.</target> 		
	Waste	• Carry out proper waste management and treatment, and promote 3Rs (reduce, reuse, recycle) to efficiently utilize limited resources. «Targets» (i) Maintain recycling rate at 80% or higher and aim for at least 85% by fiscal 2030. (ii) Maintain final disposal rate at less than 1% and aim for less than 0.5% by fiscal 2030.		

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[•] ECB observed practices 2022

- To infer firm-level emissions within-sector variability depends only on the number of workers
- The majority of banks typically identify transition risk using:
 - ✓ **Estimated emissions** from private data providers
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 - \Rightarrow This paper exploits a similar approach to classify the firms' brownness
 - ➡ ECB observed practices 2022

Eurostat, INPS, ISTAT Cerved

 Obtained from sectoral energy consumption for different energy sources downscaled wrt the no. of employees + details

 $Y_{b,f,t,l} = \alpha_1(Brown_f \times Post_t \times Tre_b) + \alpha_2(Brown_f \times Tre_b) + \alpha_3Covid_{b,f,t} + \delta_{b,t} + \omega_{f,t} + \eta_{l,t} + \epsilon_{b,f,t,l}$

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Bank b, Firm f, Month t, Loan-type l

- $Y_{b,f,t,l}$ log of total credit granted/average credit spread
- $Brown_f \in [0,1]$, if firm-level CO2 emissions are \geq median of pre-shock distribution
- $Post_t \in [0, 1]$, 1 after the introduction of the ECB Guide / Climate Stress Test
- $Tre_b \in [0, 1]$, 1 treated bank

Loans to NFCs with Covid-19 guarantess

 $Y_{b,f,t,l} = \alpha_1(Brown_f \times Post_t \times Tre_b) + \alpha_2(Brown_f \times Tre_b) + \alpha_3Covid_{b,f,t} + \delta_{b,t} + \omega_{f,t} + \eta_{l,t} + \epsilon_{b,f,t,l}$

Bank b, Firm f, Month t, Loan-type l

- $Y_{b,f,t,l}$ log of total credit granted/average credit spread
- $Brown_f \in [0, 1]$, if firm-level CO2 emissions are \geq median of pre-shock distribution across sectors
- $Post_t \in [0, 1]$, 1 after the introduction of the ECB Guide / Climate Stress Test
- $Tre_b \in [0, 1]$, 1 treated bank (SI)

Loans to NFCs with Covid-19 guarantess

Identification with multi-bank credit relationships

$\hat{\alpha_1} \rightarrow$ Difference between two DD estimates:

$$\begin{split} &[(\overline{Y}_{SI,Brown,Post} - \overline{Y}_{SI,Brown,Pre}) - (\overline{Y}_{LSI,Brown,Post} - \overline{Y}_{LSI,Brown,Pre})] - \\ &[(\overline{Y}_{SI,NotBr,Post} - \overline{Y}_{SI,NotBr,Pre}) - (\overline{Y}_{LSI,NotBr,Post} - \overline{Y}_{LSI,NotBr,Pre})] \end{split}$$

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2 banks (b): SI (treated), LSI (control) 2 firms (f): Brown, not Brown 2 time periods (t): t=0 (before), t=1 (after)



Results

Did SIs reallocate credit to less brown firms after the shock(s)?

YES: SIs reallocated credit (-2.1%) compared to LSIs after the 2020 expectations

	$log(credit_{b,f,t,l})$		
$Brown_f \times Post_t \times Tre_b$	-0.0209**		
	(0.00996)		
$Brown_f \times Tre_b$	0.0607***		
-	(0.0175)		
$Covid_{b,f,t}$	(0.0175) , <i>t</i> 0.0180***		
	(0.00101)		
Observations	652,744		
Fixed Effects	Yes		
Bank and Firm controls	Yes		



▶ Results for price-channel: 2020 Supervisory Shock

✤ Results for quantity-channel: 2022 CST

▶ Results for price-channel: 2022 CST

	$log(credit_{b,f,t,l})$		
	$Brown_f$	$Brown \ High_f$	$CO2_i nt_f$
	\geq median	\geq Q3	$\frac{CO2emi_f}{Revenue_f}$
$Transition \ Risk_f \times Post_t \times Tre_b$	-0.0209**	-0.0234**	-0.0000606***
·	(0.00996)	(0.0109)	(0.0000235)
Transition $Risk_f \times Tre_b$	0.0607***	0.0679***	0.0000703***
·	(0.0175)	(0.0192)	(0.0000236)
$Covid_{b,f,t}$	0.0180***	0.0180***	0.0170***
	(0.00101)	(0.00101)	(0.000877)
Observations	652,744	652,744	652,744
Fixed Effects	Yes	Yes	Yes
Bank and Firm controls	Yes	Yes	Yes

	$log(credit_{b,f,t,l})$	
$Brown_f \times Post_t \times Tre_b \times Bank \ Comm_b$		-0.0229***
		(0.00771)
$Brown_f \times Post_t \times Tre_b$	-0.0135*	-0.00265
	(0.00720)	(0.00798)
$Covid_{b,f,t}$	0.0127***	0.0127***
	(0.000838)	(0.000838)
Observations	650,297	650,297
Bank-Time Fixed Effects	Yes	Yes
Firm-Time Fixed Effects	Yes	Yes
Loan Type-Time Fixed Effects	Yes	Yes
Bank-Firm Fixed Effects	Yes	Yes
Bank and Firm controls	Yes	Yes

The role of banks' commitment

• What about green banks?

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 - \rightarrow 3 SI banks (out of 11) with emission targets set in the year before the shock

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 - ightarrow 3 SI banks (out of 11) with emission targets set in the year before the shock

• *Rational*: green banks might be strongly impacted by the shocks, due to decarbonization policies of credit portfolios and the scrutiny of their investors

$$\begin{split} Y_{b,f,t,l} &= \alpha_1(Brown_f \times Post_t \times Tre_b \times Bank\ Comm_b\) + \alpha_2(Brown_f \times Post_t \times Tre_b) + \\ &+ \alpha_3(Brown_f \times Tre_b \times Bank\ Comm_b) + \alpha_4(Brown_f \times Tre_b) + \alpha_5Covid_{b,f,t} + \\ &+ \delta_{b,t} + \omega_{f,t} + \eta_{l,t} + \epsilon_{b,f,t,l} \end{split}$$

$\Rightarrow \textbf{GREEN} \text{ banks}$

	$log(credit_{b,f,t,l})$	
$Brown_f \times Post_t \times Tre_b \times Bank \ Comm_b$		-0.0261***
•		(0.0101)
$Brown_f \times Post_t \times Tre_b$	-0.0209**	-0.00901
•	(0.00996)	(0.0110)
$Brown_f \times Tre_b \times Bank \ Comm_b$		-0.0134
•		(0.0178)
$Brown_f \times Tre_b$	0.0607***	0.0671***
•	(0.0175)	(0.0194)
$Covid_{b,f,t}$	0.0180***	0.0180***
, , ,	(0.00101)	(0.00101)
Observations	652,744	652,744
Fixed Effects	Yes	Yes
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	(0.00101)	(0.00101)
Observations	652,744	652,744
Fixed Effects	Yes	Yes
Bank and Firm controls	Yes	Yes

Supervision impacted banks with decarbonization policies of credit portfolios

Focus on large and listed firms

 $Y_{b,f,t,l} = \alpha_1(Brown_f \times Post_t \times Tre_b \times Firm\ Comm_f) + \alpha_2(Brown_f \times Post_t \times Tre_b) + \\ + \alpha_3(Brown_f \times Tre_b \times Firm\ Comm_f) + \alpha_4(Brown_f \times Tre_b) + \alpha_5Covid_{b,f,t} + \\ + \delta_{b,t} + \omega_{f,t} + \eta_{l,t} + \epsilon_{b,f,t,l}$

	CST			
	$log(credit_{b,f,t})$		$spread_{b,f,t}$	
$Brown_f \times Post_t \times Tre_b \times Firm Comm_f$		-2.118**		85.37**
		(0.906)		(37.10)
$Brown_f \times Post_t \times Tre_b$	-0.128	-0.126	92.52***	92.33***
	(0.226)	(0.226)	(30.65)	(30.62)
$Brown_f \times Tre_b \times Firm Comm_f$		3.785		-16.36
		(2.469)		(116.5)
$Brown_f \times Tre_b$	1.020*	1.014*	-83.74**	-83.71**
	(0.600)	(0.595)	(40.73)	(40.74)
$Covid_{b,f,t}$	0.0230**	0.0231**	-2.970*	-2.968*
	(0.0103)	(0.0103)	(1.600)	(1.601)
Observations	3,333	3,333	3,333	3,333
Fixed Effects	Yes	Yes	Yes	Yes
Bank and Firm controls	Yes	Yes	Yes	Yes

✤ Results for Supervisory Shock

• The CST did not explicitly require banks to consider the exposure to climate risk with climate forward-looking data (i.e. emission targets or commitment)

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- Firms with targets are typically the most emitting ones in credit portfolios, based on historical CO2 emissions
- Related emissions fed in the data collected by the SSM during the exercise

 \implies Supervisory initiatives can differently affect lending policies and credit supply

• Novel evidence on the effect of climate banking supervision on credit supply with a complete assessment of banks' portfolio

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Conclusions

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- After the ECB expectations, green SIs reallocated credit away from polluting firms
 Increased efficacy of banks' commitments in the presence of supervisory shock
- Limited role of firms' emission targets in the credit process (data reliability and gaps)

Conclusions

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Policy implications

 Banking supervision should both provide the right incentives for banks to properly manage climate-related risks (Hansen, 2022)

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Policy implications

- Banking supervision should both provide the right incentives for banks to properly manage climate-related risks (Hansen, 2022)
- Avoid unintended consequences banks cut credit to brown firms needing more resources to support the transition, i.e. those committed/with targets

Thank you!

Appendix

References

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ECB Guide on Climate risk - Nov 2020

- (Non-binding) Guide where ECB expects institutions to consider climate-related risks in risk management, business strategy and governance frameworks
- Supervised banks to perform self-assessment on ECB expectations in early 2021

Climate Stress Test - Nov 2021

- Dear CEO letter to announce the official participation of banks
- The first supervisory stress test in 2022

2020 ECB Guide

Expectation 8

In their credit risk management, institutions are expected to consider climate-related and environmental risks at all relevant stages of the credit-granting process and to monitor the risks in their portfolios.

Box 2

Example of observed practice: Climate-related and environmental key performance indicators

The ECB observed an institution which had integrated the following climate-related and environmental key performance indicators (KPIs) into its strategic framework with a view to making its

strategy of reducing exposure to transition risks measurable: i) the carbon emission footprint of its assets; ii) the average energy label of its mortgage portfolios; and iii) the number of homes that saw an energy label improvement thanks to its financing. In addition to these KPIs, the institution stresses

Bank observed practices



Good practices for climaterelated and environmental risk management

Observations from the 2022 thematic review

Identification of exposures

Institutions are exposed to C&E risks through their exposures to, inter alia, corporates, small and medium-sized enterprises (SME), households and sovereigns. Depending on the type of exposure, different qualitative and quantitative approaches are used to assess the materiality of the risks. Institutions typically rely on proxybased quantification methods to identify pockets of risk based on both client data and externally sourced data. Leading institutions supplement such initial analyses with a variety of more advanced assessment methods to estimate the level of risk more accurately, including portfolio alignment approaches, sensitivity analysis and financed emissions. The observed practices described in this section include some

CST: identification of transition risk

2.1 Methodology and scenarios

The 2022 CST was a constrained bottom-up exercise, meaning that participating banks provided their own data submissions and stress test projections subject to a common methodology and common scenarios.

The methodology underlying the 2022 CST was published in October 2021, while the scenarios were published in January 2022.⁵

Three distinct modules formed the basis of the methodology (see Chart 1).

Module 1 consisted of a qualitative questionnaire aiming at providing a uniform and standardised assessment of banks' climate risk stress-testing capabilities. This module assessed banks' internal climate risk stress-testing frameworks in line with Expectation 11 (on scenario analysis and stress testing) set out in the ECB Guide on climate-related and environmental risks. Apart from including general questions regarding the existence and use of climate risk stress testing within the institution, the Module 1 questionnaire covered areas such as governance and risk appetite, integration into the institutions' business strategy, methodology and scenario design, data availability and sources, ICAAP and future plans. The aim of this was to understand interlinkages between the banks' stress-testing frameworks and other internal business processes.

Module 2 consisted of two climate risk metrics providing insights into the sensitivity of banks' income to transition risk and their exposure to carbon emission-intensive industries as of the reference date (31 December 2021). Banks were asked to split

Imputation procedure for firms' CO2 emissions

1. Estimate energy consumption (Faiella et al. (2022), Emambakhsh et al. (2023)):

$$e_{f,t} = \sum_{z=1}^{Z} w_{z,f,t}$$

where z = 1, ..., Z represents the energy sources and $w_{z,f,t}$ is defined as:

$$w_{z,f,t} = \frac{l_{f,t}}{L_t} \times E_{z,t}$$

where:

- $l_{f,t}$ denotes the number of number of employees for firm f at time t
- L_t denotes the total number of employees enrolled in the same sector as the one of the firm f at time t
- $E_{z,t}$ is the energy consumption (at the sector level of firm f) for energy source z at time t
- 2. Estimate **S1-S2 emissions** (ton of oil equivalent, *toe*) through carbon emission factors for each fuel from ISPRA and Ministero dell'Ambiente

Robustness check: Top emitting sectors

Figure 1: CO2 emissions across industries (tonnes)



Descriptive statistics: whole sample

	Mean	Std. Dev.	Q1	Median	Q3
Loan-level data					
Loan amount (log) Loan amount Covid (log) Loan spread (basis points)	11.58 1.47 442.43	1.67 3.83 343.86	10.34 0.00 196.44	11.51 0.00 343.70	12.61 0.00 589.45
Bank-level data					
Treated (bank) Bank Committed (bank)	0.79 0.36	0.41 0.48	1.00 0.00	1.00 0.00	1.00 1.00
Firm-level data					
Post (month) Brown (firm) CO2 emissions (firm) log(CO2 emissions) (firm)	0.52 0.49 1.94e+06 11.50	0.50 0.50 8.50e+07 1.66	0.00 0.00 32,235.51 10.38	1.00 0.00 83,812.32 11.34	1.00 1.00 244,989.87 12.41
Observations	909,816				

Descriptive statistics: subsample of large and listed firms

	Mean	SD	Q1	Median	Q3
Loan-level data					
Loan amount (log)	13.98	2.43	12.43	13.99	15.42
Loan amount Covid (log)	0.77	3.14	0.00	0.00	0.00
Loan spread (basis points)	313.98	290.31	138.10	238.50	399.80
Bank-level data					
Treated (bank)	0.81	0.39	1.00	1.00	1.00
Firm-level data					
Post (month)	0.50	0.50	0.00	1.00	1.00
Brown_Comm (firm)	0.08	0.27	0.00	0.00	0.00
CO2 emission (firm)	2.49e+08	1.17e+09	509,321.04	1.70e+06	5.04e+06
log(CO2 emissions) (firm)	14.57	2.41	13.14	14.35	15.43
Observations	4,546				

Evolution of credit in Covid-19 times



Notes: The chart plots the amount of loans to NFCs and loans to NFCs (rhs) with Covid 19 guarantees expressed in billion of euros, at monthly frequency, taking the end-month values. Source: supervisory data drawn from AnaCredit.

Results for price-channel: 2020 Supervisory Shock

	(1)	(2)	(3)	(4)
$Brown_f \times Post_t \times Tre_b$	-1.610	0.740	1.440	0.441
	(2.606)	(2.852)	(2.762)	(2.556)
$Brown_f \times Tre_b$	-24.08***	-17.25***	-18.37***	-17.15***
-	(4.470)	(4.687)	(4.456)	(4.171)
$Covid_{b,f,t}$	-5.653***	-9.283***	-8.987***	-3.502***
	(0.199)	(0.223)	(0.226)	(0.208)
Controls	Yes	Yes	Yes	Yes
Bank-Time Fixed Effects	Yes		Yes	Yes
Firm-Time Fixed Effects		Yes	Yes	Yes
Loan Type-Time Fixed Effects				Yes
Observations	652,744	652,744	652,744	652,744
R^2	0.083	0.289	0.318	0.528
R^2	0.073	0.306	0.324	0.541
Number of Banks	38	38	38	38
Number of Firms	26,808	26,808	26,808	26,808

Results for quantity-channel: 2022 CST

	(1)	(2)	(3)	(4)
$Brown_f \times Post_t \times Tre_b$	0.0127	0.0134	0.0108	0.00792
-	(0.00896)	(0.00845)	(0.00843)	(0.00828)
$Brown_f \times Tre_b$	0.0547**	0.0355**	0.0462***	0.0386**
	(0.0224)	(0.0171)	(0.0168)	(0.0165)
$Covid_{b,f,t}$	0.0114***	0.0274***	0.0337***	0.0165***
	(0.00116)	(0.000899)	(0.000947)	(0.000932)
Controls	Yes	Yes	Yes	Yes
Bank-Time Fixed Effects	Yes		Yes	Yes
Firm-Time Fixed Effects		Yes	Yes	Yes
Loan Type-Time Fixed Effects				Yes
Observations	703,796	703,796	703,796	703,796
R^2	0.083	0.471	0.477	0.578
Number of Banks	38	38	38	38
Number of Firms	27,404	27,404	27,404	27,404

Results for price-channel: 2022 CST

	(1)	(2)	(3)	(4)
$Brown_f \times Post_t \times Tre_b$	0.361	-1.355	-0.648	0.0162
	(1.765)	(1.926)	(1.898)	(1.748)
$Brown_f \times Tre_b$	-24.13***	-19.01***	-16.08***	-12.17***
	(3.483)	(3.575)	(3.493)	(3.161)
$Covid_{b,f,t}$	-4.417***	-6.770***	-7.114***	-2.154***
	(0.158)	(0.168)	(0.177)	(0.161)
Controls	Yes	Yes	Yes	Yes
Bank-Time Fixed Effects	Yes		Yes	Yes
Firm-Time Fixed Effects		Yes	Yes	Yes
Loan Type-Time Fixed Effects				Yes
Observations	703,796	703,796	703,796	703,796
R^2	0.083	0.471	0.477	0.556
Number of Banks	38	38	38	38
Number of Firms	27,404	27,404	27,404	27,404

Parallel Trend: the launch of the 2022 CST



 \Rightarrow Not significant impact, neither before, nor after the official launch of the 2022 Climate Stress Test one year later, in November 2021

Results with firms' commitment: Sup expectations

	Sup expectations				
	$log(credit_{b,f,t})$		sprea	$ad_{b,f,t}$	
$Brown_f \times Post_t \times Tre_b \times Firm Comm_f$	1.288 -148		-148.8*		
<i>.</i>		(1.169)		(79.41)	
$Brown_f \times Post_t \times Tre_b$	1.124	1.125*	-136.4**	-136.8**	
<i>.</i>	(0.685)	(0.667)	(65.28)	(64.98)	
$Brown_f \times Tre_b \times Firm Comm_f$		2.792		19.33	
<i></i>		(2.155)		(102.3)	
$Brown_f \times Tre_b$	0.178	0.188	90.07	90.10	
	(0.321)	(0.320)	(68.98)	(68.98)	
$Covid_{b,f,t}$	0.0273**	0.0235***	-5.776**	-5.677***	
	(0.0117)	(0.0112)	(1.911)	(1.908)	
Observations	3,337	3,337	3,337	3,337	
Fixed Effects	Yes	Yes	Yes	Yes	
Bank and Firm controls	Yes Yes Yes Yes				

Falsification Tests

	$Log(credit_{b,f,t})$	$Spread_{b,f,t}$	$Log(credit_{b,f,t})$	$Spread_{b,f,t}$
$Brown_f \times Post_t \times Fake Tre_b$	0.0037	0.4032		
	(0.0097)	(2.465)		
$Brown_f \times Fake Tre_b$	-0.048***	-1.32		
	(0.018)	(4.085)		
$Covid_{b,f,t}$	0.018***	-3.530***		
	(0.001)	(0.208)		
$Brown_f \times Fake Post_t \times Tre_b$			0.004	-0.014
U .			(0.007)	(0.017)
$Brown_f \times Tre_b$			0.0596***	-0.223***
			(0.0132)	(0.032)
$Covid_{b,f,t}$			0.0175***	-0.035***
.,,,,,			(0.00083)	(0.00176)
Observations	652,744	652,744	792,781	792,781
R^2	0.585	0.541	0.576	0.535

- Randomly assigning treated banks renders key coefficient not significant
- Assigning the month of the release of the Guide to a fake period also renders the key coefficient not significant