Synthetic, but How Much Risk Transfer?

Alex Osberghaus¹, Glenn Schepens²

 1 University of Zurich, Swiss Finance Institute 2 European Central Bank

Fifth Conference on Financial Stability — Banco de España & CEMFI June 13, 2025

The views expressed do not necessarily reflect those of the ECB or the Eurosystem.

Banks' Hot New Trade

Banks' hot new trade could burn others,

FINANCIAL TIMES

for once



March 1, 2024 1:59 PM GMT+1 - Updated 3 months ago

IMF warns of 'round-tripping' fears



If you're unfamiliar with synthetic risk transfers, there's a chance you'll hear all about them when the next financial crisis hits.

THE WALL STREET JOURNAL.

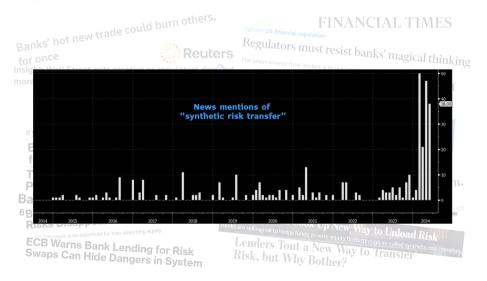
Banks Transfer Risk to Themselves By Matt Levine

Watch Out: Wall Street Is Finding New Ways to Slice and Dice Loans The people who brought you CDOs and other financing esoterica are back with SRTs, NAV loans and more

Banks' Hot New Trade...

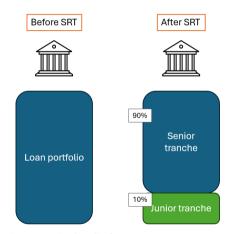


...is Getting a Lot of Media Attention

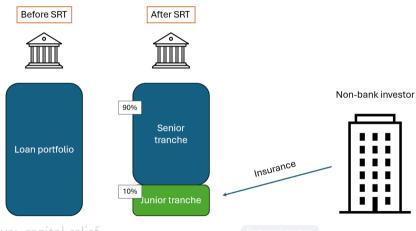




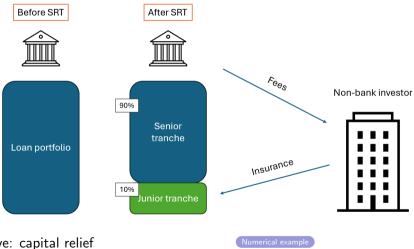
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 \Rightarrow Objective: capital relief

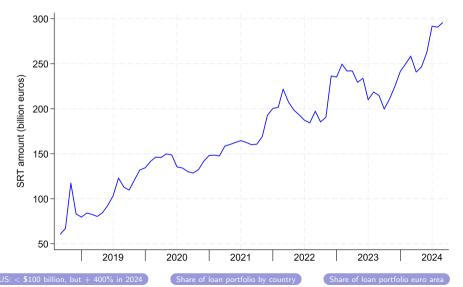


⇒ Objective: capital relief

2/18

Osberghaus, Schepens June 13, 2025

Stock of SRTs in Europe (corporate loans)



Osberghaus, Schepens Synthetic Risk Transfer June 13, 2025

What are the risks to financial stability?

- Selection of SRT loans
 - All else equal, banks choose capital-expensive loans
 - Banks deploy the freed-up capital and become less capitalized relative to the economic riskiness of their portfolio
- @ Monitoring
 - Novel monitoring measure
 - Banks reduce monitoring relative to other banks lending to the same firm
- Interconnectedness of banks and non-banks
 - Banks are more likely to sell to non-bank SRT investors to which they also grant credit
 - ightharpoonup On average 26 percent of the SRT investment is funded by banks \Rightarrow "round-tripping"

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Credit risk transfers and financial stability

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⇒ Capital optimization through credit risk transfers

Credit risk transfers and moral hazard

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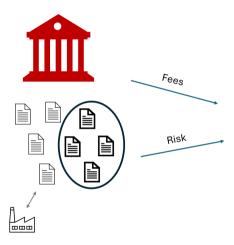
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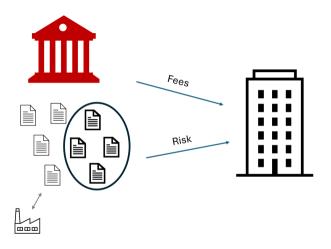


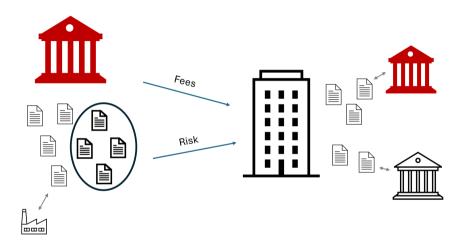




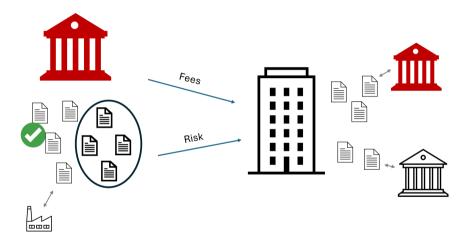




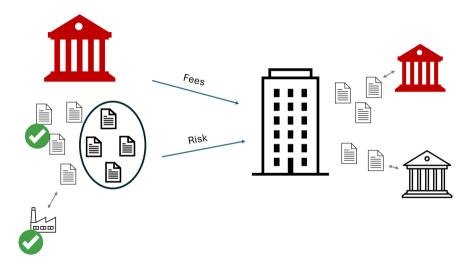




Data – what we have (AnaCredit & COREP)

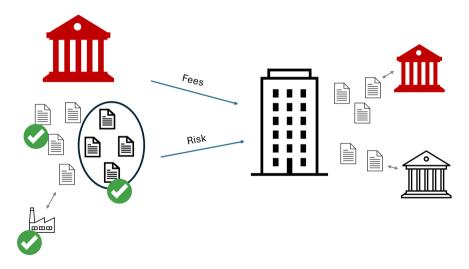


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June 13, 2025

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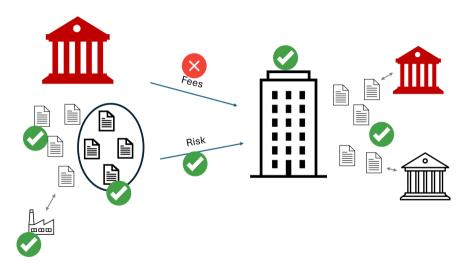
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- 71 banks that use SRTs (Riskiness of SRT banks)
- 352 SRTs
- 171.506 SRT loans Loan characteristics
- €260 billion notional value
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Loan selection



€100

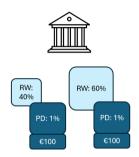
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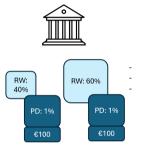




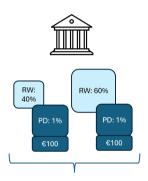


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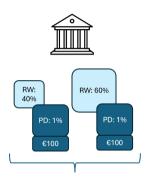




- Standardized Approach
- PD/LGD underestimation
- Regulation

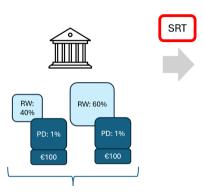


Tier 1 =
$$\frac{Tier\ 1\ Capital}{RW\ Assets} = \frac{10}{0.4*100+0.6*100} = 109$$



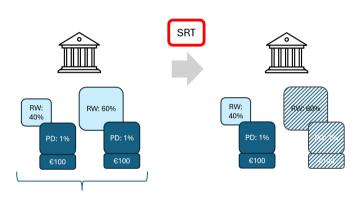
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Leverage =
$$\frac{Tier\ 1\ Capital}{Assets} = \frac{10}{100+100} = 5\%$$



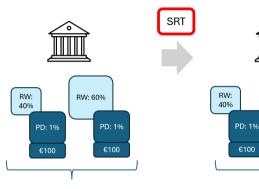
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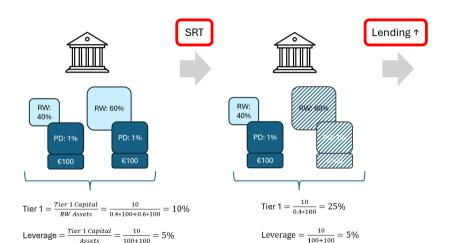


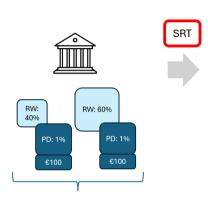
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Tier 1 =
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 = 25%

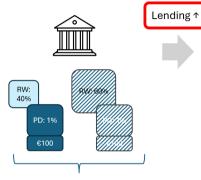
Leverage =
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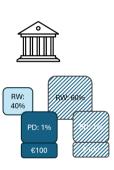
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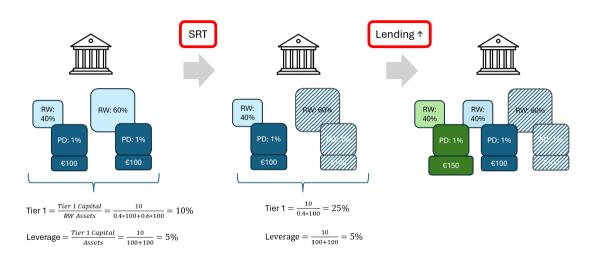
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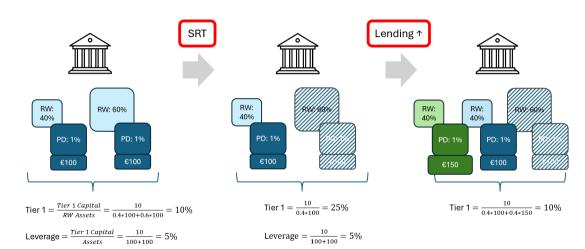
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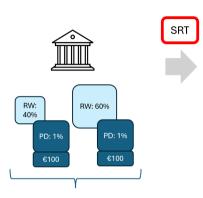
Leverage =
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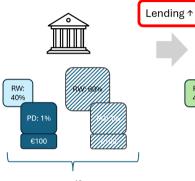
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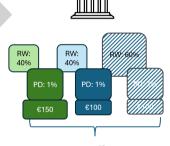
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Leverage =
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 = 5%



Tier 1 =
$$\frac{10}{0.4*100+0.4*150}$$
 = 10%

Leverage =
$$\frac{10}{150+100+100}$$
 = 3%

9/18

Loan selection — laboratory

Problem: Diff. in economic and assigned riskiness endogenous to loan characteristics

Solution: Discontinuity in assigned riskiness (SME supporting factor)

Loan selection — laboratory

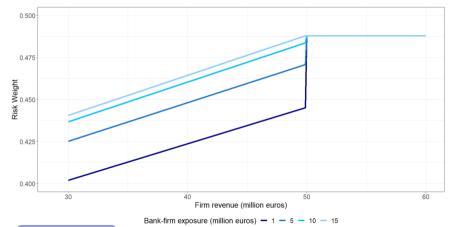
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IRBA formula Smoothness of cov

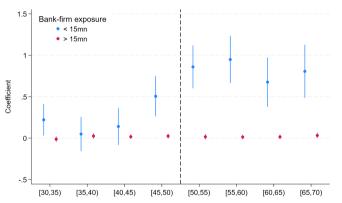
Banks choose capital-expensive loans

$$SRT$$
-loan_{i,t} = β_1 Borrower revenue $bins_{i,t} + \delta X_{i,t} + \varepsilon_{i,t}$

Control: PD, loan rate, log loan size. FE: $Bank \times year \times loan \ type \times interest \ rate \ type \times loan \ purpose \times borrower \ industry \times residual \ maturity \ above 1 \ year.$ SE: wild-bootstrapped at bank-level.

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This matters for all firms with revenues < €50 million

bo tests with €40 and €60 million threshold 🤇

Capital deployment

$$Y_{b,t} = \beta_1 SRT$$
-intensity_{b,t} + $\beta_2 X_{b,t} + \delta_b + \gamma_t + \varepsilon_{b,t}$.

Standard errors in parenthese

^{*} p < 0.1, ** p < 0.05, *** p < 0.05

Capital deployment

$$Y_{b,t} = \beta_1 SRT$$
-intensity_{b,t} + $\beta_2 X_{b,t} + \delta_b + \gamma_t + \varepsilon_{b,t}$.

	(1)	(2)	(3)	(4)
	Loan growth	Loan growth	Leverage ratio (percent)	Tier 1 capital ratio (percent)
SRT intensity $[t]$	4.497**	3.389*	-5.345**	0.00660
	(2.098)	(1.917)	(2.174)	(0.0322)
SRT intensity $[t-1]$		5.167***		
		(1.547)		
Mean	3.350	3.337	7.665	0.202
Fixed effects	Bank, quarter	Bank, quarter	Bank, year	Bank, year
Frequency	Quarterly	Quarterly	Yearly	Yearly
Controls	Size bins \times bank size capital ratio	Size bins \times bank size capital ratio	Size bins \times bank size	Size bins \times bank size
SE cluster	Bank	Bank	Bank	Bank
Adj. R-squared	0.395	0.393	0.801	0.925
N	10,660	10,651	1,949	2,119

Standard errors in parentheses

New lending is not particularly capital-expensive



^{*} p < 0.1, ** p < 0.05, *** p < 0.01

Moral hazard

Monitoring measure

Assumptions

- Firm PD fluctuates over time
- ② Higher monitoring ⇒ closer mapping of actual PD fluctuations into PD estimates

Quarterly monitoring measure

- PD update
- ② SD(PD)

Validation

 PD-updating banks (i.e., monitoring banks) are better at predicting actual default one year ahead

Monitoring measure

Assumptions

- Firm PD fluctuates over time
- ② Higher monitoring ⇒ closer mapping of actual PD fluctuations into PD estimates

Quarterly monitoring measure

- PD update
- SD(PD)

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SRT \Rightarrow decline in monitoring by 15-30%

$Monitoring_{f,b,t} = SRT\ loan_{f,b,t-1} imes Post\ SRT_{f,b,t-1} + \gamma_f + \delta_b + \omega_t + \varepsilon_{f,b,t}$

	Firm, quarter, bank	Firm × quarter, bank		
			Bank	Bank
N				

Standard errors in parentheses

^{*} p < 0.1, ** p < 0.05, *** p < 0.05

SRT \Rightarrow decline in monitoring by 15-30%

$$Monitoring_{f,b,t} = SRT\ loan_{f,b,t-1} \times Post\ SRT_{f,b,t-1} + \gamma_f + \delta_b + \omega_t + \varepsilon_{f,b,t}$$

	(1)	(2)	(3)	(4)		
	PD update	PD update	SD(PD)	SD(PD)		
SRT loan=1 \times Post SRT=1	-0.0495**	-0.0458**	-0.0356**	-0.0323**		
	(0.0209)	(0.0231)	(0.0162)	(0.0157)		
Average PD	0.00782*	0.00153	0.0571***	0.0488***		
	(0.00470)	(0.00377)	(0.0187)	(0.0118)		
Mean	0.332	0.315	0.116	0.108		
Fixed effects	Firm, quarter, bank	$Firm \times quarter, bank$	Firm, quarter, bank	$Firm \times quarter, bank$		
SE cluster	Bank	Bank	Bank	Bank		
Controls	Firm revenue	Firm revenue	Firm revenue	Firm revenue		
Adj. R-squared	0.356	0.510	0.209	0.433		
N	4,086,342	3,396,856	4,007,636	3,321,116		

Standard errors in parentheses

June 13, 2025

^{*} p < 0.1, ** p < 0.05, *** p < 0.01

Bank-non-bank nexus

14/18

Are banks and non-bank SRT investors interconnected?



Banks' hot new trade could burn others, for once March 1, 2024 1:59 PM GMT+1 - Updated 3 months ago

FINANCIAL TIMES

IMF warns of 'round-tripping' fears

Banks Transfer Risk to Themselves By Matt Levine

Banks sell SRTs to familiar investors

- SRT investment dummy_{i,b,j,t} = $\beta_0 + \beta_1$ Credit rel. dummy_{b,j,t} + $\delta_b + \eta_j + \omega_t + \varepsilon_{i,b,j,t}$
- i: SRT issue; b: bank; j: investor; t: year

		Bank, investor, year			
	No Government/ EIF	No Government/ EIF	No Government/ EIF		
		Bank	Bank		
N					

Standard errors in parentheses

^{*} p < 0.1, ** p < 0.05, *** p < 0.01

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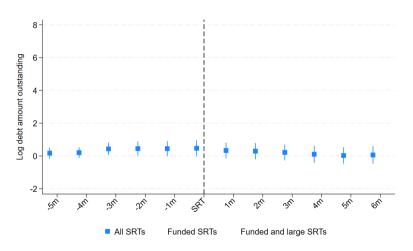
	(1)	(2)	(3)		
	SRT investment	SRT investment	SRT investment		
Credit rel.	0.0853**	0.0743**			
	(0.0372)	(0.0291)			
Credit rel. (group)			0.0855**		
			(0.0384)		
Mean	0.135	0.130	0.130		
Fixed effects	Year	Bank, investor, year	Bank, investor, year		
Restriction	No Government/ EIF	No Government/ EIF	No Government/ EIF		
SE cluster		Bank	Bank		
Adj. R-squared	0.0248	0.153	0.154		
N	1,055	1,050	1,050		

Standard errors in parentheses

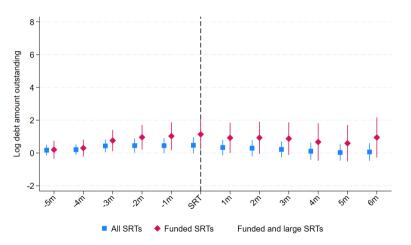
^{*} p < 0.1, ** p < 0.05, *** p < 0.01

Log debt amount outstanding_{i,j,t} = $\beta_0 + \beta_1$ Months to SRT investment_{i,j,t} + $\theta_i + \omega_t + \varepsilon_{i,j,t}$

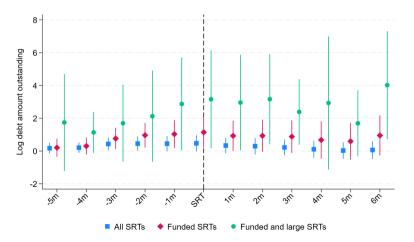
Log debt amount outstanding $i,j,t=eta_0+eta_1$ Months to SRT investment $i,j,t+ heta_i+\omega_t+arepsilon_{i,j,t}$



Log debt amount outstanding_{i,j,t} = $\beta_0 + \beta_1$ Months to SRT investment_{i,j,t} + $\theta_i + \omega_t + \varepsilon_{i,j,t}$



 $\textit{Log debt amount outstanding}_{i,j,t} = \beta_0 + \beta_1 \textit{Months to SRT investment}_{i,j,t} + \theta_i + \omega_t + \epsilon_{i,j,t}$



Pre-trends

- lacktriangledown Objective: capital relief \Rightarrow loan selection "optimizes" effective ex-post capitalization
 - \Rightarrow capital saving of 315 million euros for the average bank
 - ⇒ New floor on risk weights could be "traded away" with SRTs
- ② Decline in monitoring
- ② 26 percent of SRT financing comes from banks (Higher in the US?)
 - ⇒ Could make SRTs procyclica

 \Rightarrow The amount of capital relief for SRTs should account for these three channels

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Appendix

18/18

Literature I

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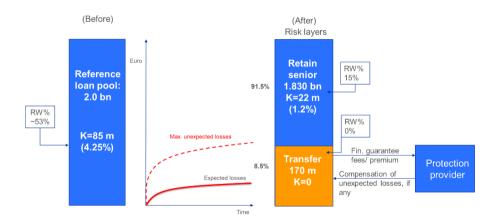
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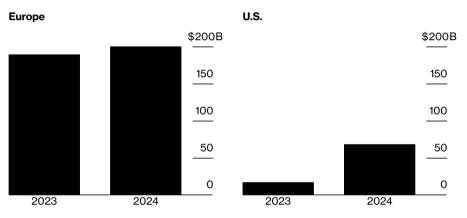
Rajan, Uday, Amit Seru, and Vikrant Vig (2015). "The failure of models that predict failure: Distance, incentives, and defaults". In: *Journal of Financial Economics* 115.2, pp. 237–260.

Wang, Yihui and Han Xia (2014). "Do lenders still monitor when they can securitize loans?" In: *The Review of Financial Studies* 27.8, pp. 2354–2391.

Numerical example from Gonzalez & Triandafil (2023)

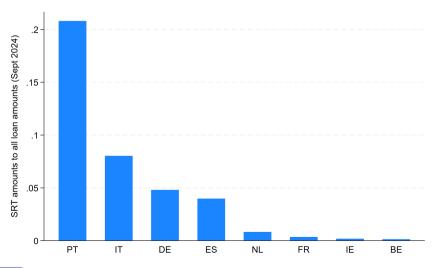


Market size in the U.S.



Source: KKR & Co., Bank of America estimates as of March 2024 Note: Figures refer to assets securitized. 2024 numbers are estimated volume.

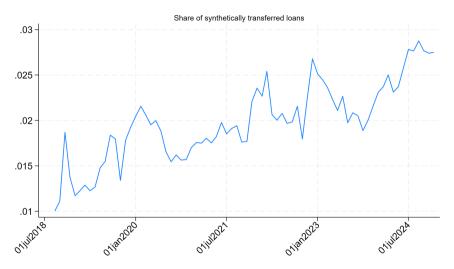
Share of all corporate loans by country



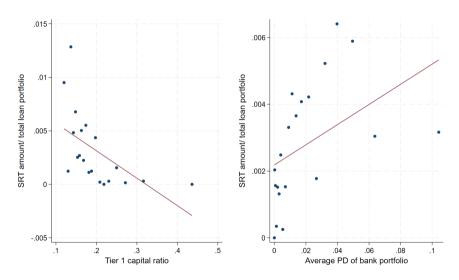




Share of synthetically transferred loans – euro area



SRT banks



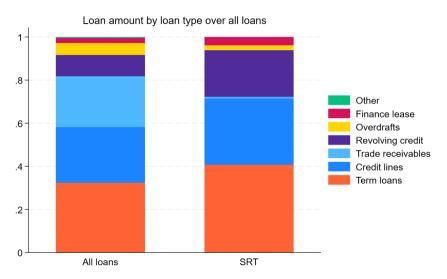
June 13, 2025

Sample: synthetically transferred loans

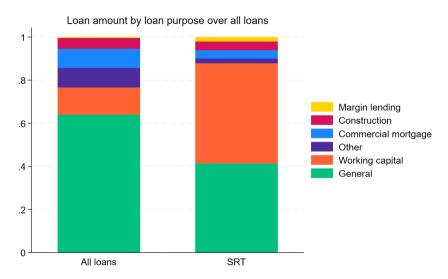
		Synthetica	lly transf	erred loans		All loans						
			171,506				June 2024: 13,600,000					
	Mean	SD	10^{th}	Median	90 th	Mean	SD	10^{th}	Median	90 th		
Loan amount	943,665	2,860,711	38,772	200,000	2,000,000	364,597	937,721	30,000	100,000	750,000		
Loan maturity (years)	5.6	3.6	2.0	5.0	10.0	8.2	6.5	1.0	6.0	18.0		
Loan rate	0.028	0.019	0.008	0.023	0.059	0.039	0.028	0.007	0.040	0.074		
Borrower revenue (mn.)	45	209	0.2	4	74	27	105	0.04	1.6	53		
Borrower PD	0.014	0.020	0.002	0.008	0.032	0.080	0.230	0.001	0.009	0.115		
Loan payments are overdue	0.014					0.045						
Loan is delinquent	0.009					0.028						
Share fixed rate	0.61					0.40						

18/18

Which loan types get synthetically transferred?

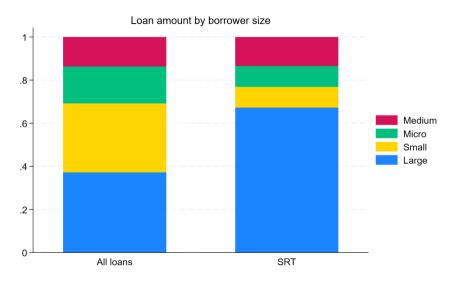


Of which loan purpose?



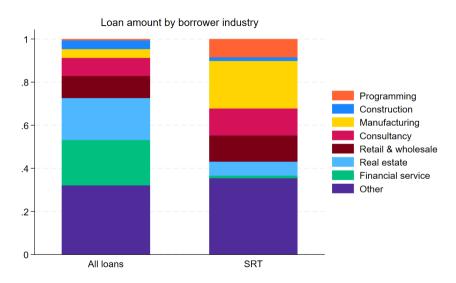
18/18

To which debtor size?

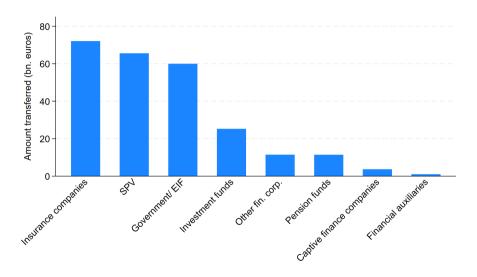


18/18

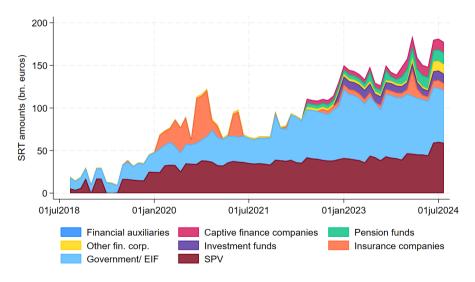
Of which debtor industries?



Who are the non-bank SRT investors?



Who are the non-bank SRT investors?



Descriptive statistics of SRT investor

	All SRT investments						SRT investments excl. Government / EIF 142					
	282											
	Mean	SD	10 th	Median	90 th	N	Mean	SD	10 th	Median	90 th	
Group level												
Total assets (mn. euros)*	168,000	309,000	32	3,040	5,690	133	134,000	393,000	9	23	220,000	3
Bank loans outstanding (mn.)	6,130	30,200	0	386	5,090	282	5,500	45,100	0	0.1	56	14
Bank loans committed (mn.)	6,550	30,200	0	1,250	6,070	282	5,190	40,500	0	0.5	146	1
Bank debt outstanding to assets*	0.25	0.32	0	0.06	0.79	282	0.04	0.07	0	0.02	0.17	1
SRT investment (first losses) to assets*	0.03	0.04	0	0	0.07	105	0.04	0.07	0	0.01	0.07	:
Number of bank relationships	18.2	33.6	1	7	28	282	7.3	31.0	1	1	3	1
Number of bank relationships during sample	35.4	60.3	2	10	55	282	12.8	52.6	2	3	7	1
Share of investors that receive credit from SRT-bank during sample	0.50					282	0.22					1
Unconsolidated level												
Total assets (mn. euros)	36,000	199,000	9	2,490	2,490	112	92,900	342,000	5	37	781,000	
Bank loans outstanding (mn.)	505	1,690	0	0.6	381	282	21	135	0	0.1	8	1
Bank loans committed (mn.)	678	2,910	0	2	394	282	23	126	0	0.2	22	1
Loans & securities (mn.)	20,300	211,000	0	5	381	282	49	193	0	0.2	61	1
Liabilities to assets	0.08	0.12	0	0.05	0.15	102	0.04	0.10	0	0	0.12	
SRT investment (first losses) to assets	0.06	0.11	0	0.02	0.13	103	0.10	0.15	0	0.05	0.36	
Number of bank relationships	4.6	7.6	1	1	8	282	1.4	0.9	1	1	2	1
Number of bank relationships during sample	8.2	11.2	2	3	12	282	3.0	2.0	1	3	4	1
Share of investors that receive credit from SRT-bank during sample	0.18					282	0.15					1

18/18

Internal Ratings Based Approach Formula

For loans to firms with annual revenues < 50 million euros:

$$RW = \mathscr{F}(PD\uparrow, M\uparrow, LGD\uparrow, \frac{\min\{\max\{5, \frac{S}{S}\}, 50\} - 5}{45}\uparrow) \times \frac{\min\{\cancel{E}; \leqslant 2, 500, 000\} \cdot 0.7619 + \max\{\cancel{E} - \leqslant 2, 500, 000; 0\} \cdot 0.85}{\cancel{E}}$$

PD: probability of default

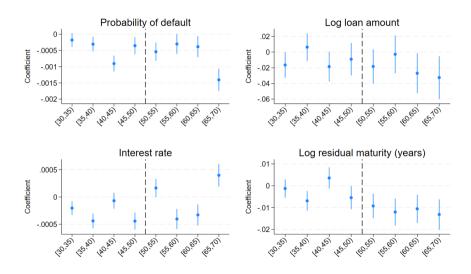
M: maturity

LGD: loss given default

5: annual revenue

E: bank-firm exposure

Covariates are smooth



19/19

Correlational result

$$SRT$$
-loan_{i,t} = $\beta_1 Revenue_{i,t} + \beta_2 Log \ bank \ firm \ exposure_{i,t} + \delta X_{i,t} + \varepsilon_{i,t}$

	(1)	(2)	(3)	(4)
	SRT loan (= 100)	SRT loan (= 100)	SRT loan (= 100)	SRT loan (= 100
Revenue (million euros)	0.00608***	0.00608***	-0.000392	-0.000392
	(0.000131)	(0.00205)	(0.000267)	(0.000723)
Log bank firm exposure	0.0210***	0.0210**	-0.00391	-0.00391
	(0.000662)	(0.00965)	(0.00277)	(0.00507)
Mean	0.359	0.359	0.344	0.344
Revenue (million)	[0,50)	[0,50)	[50,100)	[50,100)
Fixed effects	FE	FE	FE	FE
SE cluster		Bank		Bank
	PD, Loan rate,	PD, Loan rate,	PD, Loan rate,	PD, Loan rate,
Controls	loan size bins \times	loan size bins \times	loan size bins \times	loan size bins $ imes$
	log loan amount	log loan amount	log loan amount	log loan amount
Adj. R-squared	0.173	0.173	0.365	0.365
N	31,282,991	31,282,991	1,993,606	1,993,606
Frequency	Yearly	Yearly	Yearly	Yearly

Standard errors in parentheses

FE: $Bank \times year \times loan \ type \times interest \ rate \ type \times loan \ purpose \times borrower \ industry \times residual \ maturity$ above 1 year

^{*} p < 0.1, ** p < 0.05, *** p < 0.01

Placebo tests

SRT-loan_{i,t} = $\beta_1 \mathbb{1}$ (revenue $\geq \in 50$ mn)_{i,t} \times Log bank firm exposure_{i,t} $+ \delta X_{i,t} + \varepsilon_{i,t}$

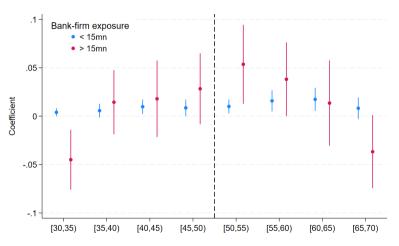
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	SRT loan (=100)	SRT loan (=100)	SRT loan (=100)	SRT loan (=100)	SRT loan (=100)	SRT loan (=100)	SRT loan (=100)	SRT loan (=100
Revenue > 50 mn=1	0.0389**	0.231**			0.0346***	0.245***		
	(0.0156)	(0.101)			(0.0102)	(0.0642)		
Revenue $>$ 50 mn=1 \times Log bank firm exposure		-0.0116+				-0.0127***		
		(0.00600)				(0.00381)		
Revenue > 40 mn=1			-0.135				0.0116	
			(0.0853)				(0.0640)	
Revenue > 40 mn=1 × Log bank firm exposure			0.00782				0.000564	
			(0.00521)				(0.00398)	
Revenue > 60 mn=1				-0.00266 (0.0962)				-0.146** (0.0710)
Revenue $>$ 60 mn=1 \times Log bank firm exposure				-0.000427 (0.00547)				0.00720° (0.00413)
Mean	0.387	0.387	0.00389	0.301	0.364	0.364	0.431	0.330
Revenue (million)	[45,55]	[45,55]	[35,45]	[55,65]	[40,60]	[40,60]	[30,50]	[50,70]
Fixed effects	FE	FE	FE	FE	FE	FE	FE	FE
SE cluster	WCR: Bank	WCR: Bank	WCR: Bank	WCR: Bank	WCR: Bank	WCR: Bank	WCR: Bank	WCR: Bank
	PD, loan rate	PD, Ioan rate	PD, loan rate	PD, Ioan rate	PD, loan rate	PD, loan rate	PD, loan rate	PD, loan rate
Controls	loan size bins \times	loan size bins ×	loan size bins \times	loan size bins \times	loan size bins >			
	log loan amount	log loan amount	log loan amount	log loan amount	log loan amount	log loan amount	log loan amount	log loan amoun
Adj. R-squared	0.450	0.450	0.400	0.448	0.393	0.393	0.366	0.406
N	615,408	615,408	794,296	530,886	1,295,035	1,295,035	1,618,225	1,021,100
Frequency	Yearly	Yearly	Yearly	Yearly	Yearly	Yearly	Yearly	Yearly

Standard errors in parenthese

* $\rho < 0.1$, ** $\rho < 0.05$, *** $\rho < 0.01$

New lending is not particularly capital-expensive

Capital-expense of firms that receive net increase in lending



Strategic loan selection – quantification

Calculate the additional benefit from the SME supporting factor with and without SRT:

$$\begin{split} \Delta \mathsf{Cap. benefit}_{b,t} &= \mathsf{Cap. benefit}_{b,t}^{\textit{with SRT}} - \mathsf{Cap. benefit}_{b,t}^{\textit{without SRT}} \cdot \mathsf{scaling factor}_{b,t} \\ &= \left(\widehat{\mathit{CAP}}_{b,t}^{\textit{with SRT}} - \widehat{\mathit{CAP}}_{b,t}^{\textit{with SRT}} \right) \\ &- \left(\widehat{\mathit{CAP}}_{b,t}^{\textit{without SRT}} - \widehat{\mathit{CAP}}_{b,t}^{\textit{without SRT}} \right) \cdot \frac{\widehat{\mathit{CAP}}_{b,t}^{\textit{with SRT}}}{\widehat{\mathit{CAP}}_{b,t}^{\textit{without SRT}}} \\ &= \widehat{\mathit{CAP}}_{b,t}^{\textit{without SRT}} \cdot \frac{\widehat{\mathit{CAP}}_{b,t}^{\textit{with SRT}}}{\widehat{\mathit{CAP}}_{b,t}^{\textit{without SRT}}} - \widehat{\mathit{CAP}}_{b,t}^{\textit{with SRT}}. \end{split}$$

⇒ Additional capital savings: 315 million euros (mean bank); 7.2 billion euros (all SRT banks)

Back

PD-updating banks are better at predicting default

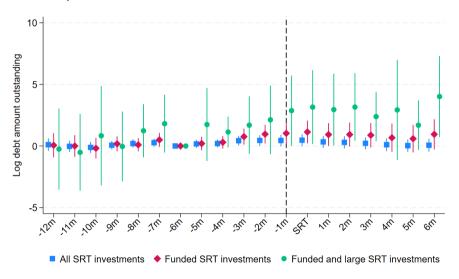
Firm defaults_{f,b,t} = Average $PD_{f,b,t-4} \times Monitoring measure_{f,b,t-4} + \gamma_f + \delta_b + \omega_t + \varepsilon_{f,b,t}$

	(1)	(2)	(3)	(4)
	Firm defaults	Firm defaults	Firm defaults	Firm defaults
Average PD \times PD update=1	0.00656***	0.0162***		
	(0.00223)	(0.00442)		
Average PD × SD(PD)			0.155*	0.301***
,			(0.0936)	(0.0941)
Average PD	0.0173***	0.0135***	0.0407***	0.0186***
	(0.00311)	(0.00233)	(0.00762)	(0.00340)
PD update=1	0.000252*	-0.0000160		
	(0.000130)	(0.000193)		
SD(PD)			0.0862***	0.0233**
			(0.0141)	(0.0106)
Fixed effects	Firm, quarter, bank	Firm × quarter, bank	Firm, quarter, bank	Firm × quarter, bank
SE cluster	Bank	Bank	Bank	Bank
Adj. R-squared	0.156	0.701	0.278	0.703
N	71,585,011	24,198,394	70,568,226	23,811,546
Frequency	Quarterly	Quarterly	Quarterly	Quarterly
Standard errors in parentheses				

Standard errors in parentheses

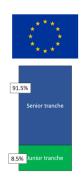
^{*} p < 0.1, ** p < 0.05, *** p < 0.01

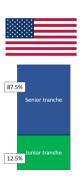
Debt dynamics: pre-trends



Predictions for the U.S.

Collins Amendment to the Dodd-Frank Act ⇒ lower bound to the risk weights on assets





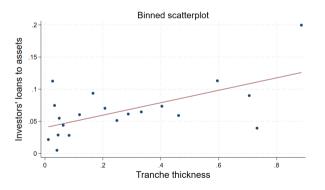
To get the same return

- investors have to be more leveraged.
- SRTs have to contain riskier loans.

Osberghaus, Schepens Synthetic Risk Transfer June 13, 2025

Thicker first-loss tranches are bought by more leveraged investors

- Debt (loans) to assets_{s b,i} = $\beta_0 + \beta_1$ Thickness tranche sold_{s,b,i} + $\varepsilon_{s,b,i}$
- s: SRT issue; b: bank; j: investor



	(1)	(2)
	Bank debt to assets	Bank debt to assets
Thickness of tranche sold	0.0969***	0.127
	(0.0290)	(0.0847)
Mean	0.0643	0.0295
Fixed effects	Investor type, country	Investor type, country
Restriction		No Gvrnmt/ EIF
SE cluster	Bank	Bank
Controls	SRT size, Avg PD	SRT size, Avg PD
Adj. R-squared	0.437	0.850
N	102	37

Standard errors in parentheses

Thickness of tranche sold \uparrow by 1 SD \rightarrow bank debt over assets \uparrow 48 percent. Back

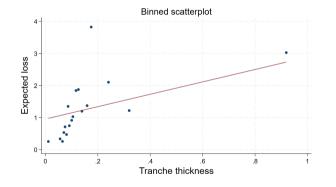


Osberghaus, Schepens Synthetic Risk Transfer June 13, 2025

^{*} p < 0.1. ** p < 0.05. *** p < 0.01

Thicker tranches are associated with riskier underlying loans

- Expected loss_{s,b,j} = $\beta_0 + \beta_1$ Thickness tranche sold_{s,b,j} + $\varepsilon_{s,b,j}$
- s: SRT issue; b: bank; j: investor

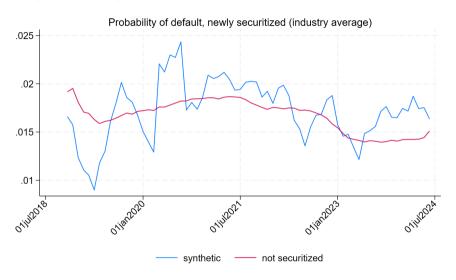


	(1)	(2)
	Expected losses	Expected losses
Thickness of tranche sold (COREP)	0.114***	0.119***
	(0.0297)	(0.0200)
Mean	0.0141	0.0151
Fixed effects		Bank, year
Restriction		
SE cluster	Bank	Bank
Controls		SRT size
Adj. R-squared	0.405	0.761
N	103	92

Other

18/18

Riskiness of synthetically securitized loans



18/18

From April 2021, SRT banks received more capital relief

