Systemic Risk and Vulnerabilities of Bank Networks

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Outline

- Microprudential vs macroprudential regulation
- Description of EBA data
- Shock propagation model
- Results for various scenarios
- Discussion of phase space
- Conclusion

Micro vs. Macro

The macro- and microprudential perspectives compared

	Macroprudential	Microprudential		
Proximate objective	limit financial system-wide distress	limit distress of individual institutions		
Ultimate objective	avoid output (GDP) costs	consumer (investor/depositor) protection		
Characterisation of risk	Seen as dependent on collective behaviour ("endogenous")	Seen as independent of individual agents' behaviour ("exogenous")		
Correlations and common exposures across institutions	important	irrelevant		
Calibration of prudential controls	in terms of system-wide risk; top-down	in terms of risks of individual institutions; bottom-up		

BIS Working Papers, No 128 "Towards a macroprudential framework for financial supervision and regulation?" Claudio Borio (2003)

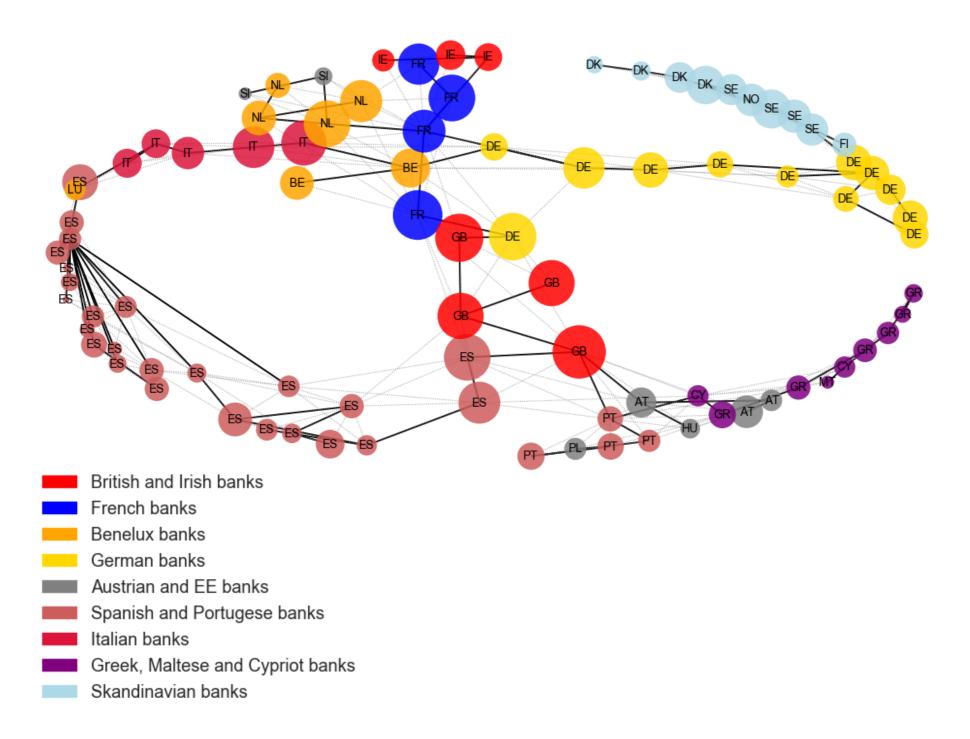


EBA data

- 90 banks
- initial capitalizations of banks from 2011
- 33 sovereign debts
- 7 asset classes (sov. debt, financial institutions, corporations, retail residential, retail SME, retail revolving, commercial real estate)
- Assumption: sovereign debt is a proxy for where a bank does its business.



Regional Bias in Banks' Portfolios



Network of European Banks

Nodes are banks in the EBA dataset.

Link weights are given by the similarity of portfolio in sovereign debts.

Planar Maximally Filtered Graph to capture most meaningful information.

Size of nodes corresponds to log of total asset exposure

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Bank Assets

- Tier 1 Capital C
- Risk Weighted Assets WaAa

a	Item		w_a range	A_a	w_a	$w_a A_a$
1	Sovereign Debt		[0.002, 0.1]	27,267	0.002	55
2	2 Financial institutions		[0.5, 1.0]	25,044	0.5	12522
3	Corporate		[0.5, 1.3]	61,237	0.5	42866
4	Retail:	Residential Mortgages	[0.5, 0.8]	36,663	0.5	14665
5		Revolving	[0.8, 1.2]	23,153	0.8	18522
6	SME		[1.0, 1.3]	3,467	1.0	3467
7	Commercial real estate		[1, 2]	22,228	1.0	22228
Total RWA W , according to Eq. (2)						114325

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Tier 1 Capital Ratio R

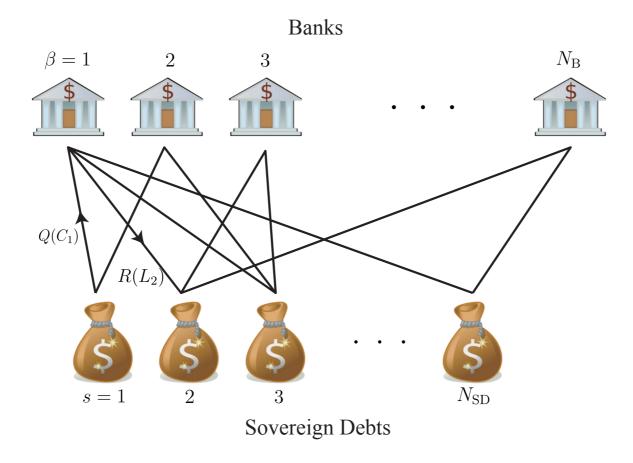
Risk weights

- Risk weighted assets describe the exposure of a bank to its assets and their risk
- The more accurate bank assesses risk, the more loans it can give out with the same amount of capital
- Internal rating-based approaches common to assess counterparty credit risk



Model

 We propose shock propagation model on a bipartite network between assets, like sovereign debts, on one side and banks on the other side



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Model

- At t=0, the risk weight of a sovereign debt (SD) is increased to reflect a readjustment of risk perception
- At t=1, all banks who own the SD see an increase in their risk weighted assets and thus a decrease in their Tier 1 Capital Ratio
- At t=2, decrease in tier 1 capital ratio of some banks creates credit pressure, amplifying the risk weights of SD:

r_s(t=2) = r_s(t=1) / credit pressure

Continue back and forth



Credit pressure

r_s(t+1) = **r**_s(t) / credit pressure

$$\Omega_s(t+1) = 1 - Q(D_s) \left(1 - \frac{\sum_{\beta=1}^{N_{\rm B}} S_{\beta,s} P\left(\frac{R_{\beta}(t)}{R_{\beta}(t-1)}\right)}{\sum_{\beta=1}^{N_{\rm B}} S_{\beta,s}} \right)$$

How strongly is a shock to a bank propagated back to the asset *s*?

- asset specific parameter $Q(D_s)$;
- how much of this asset is held by the affected banks;
- how the banks react to a reduction in their Tier 1 Capital Ratio R.



Inherent riskiness of asset

Credit pressure depends on Ω_s

 Use CDS spread as a parameter for riskiness of sovereign debt

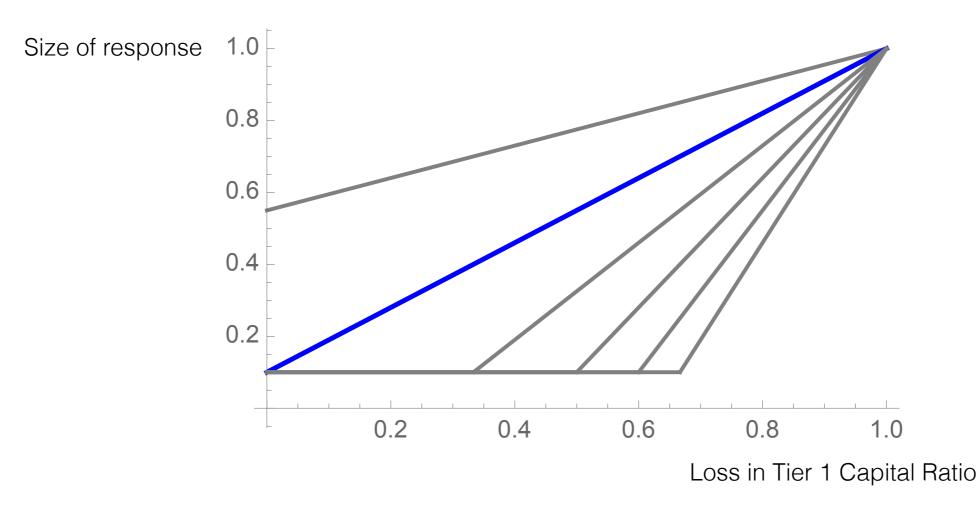
> $Q(D_s) = 1 - 2^{-D_s/100}$ AT - 20 - 0.129 CY - 130 - 0.594 DE - 5 - 0.034 GR - 1400 - 0.999 PT - 400 - 0.938

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Bank Response to Shock

- P(x) bank response function
- How strongly does a bank respond to a loss?



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What happens to sovereign debts?

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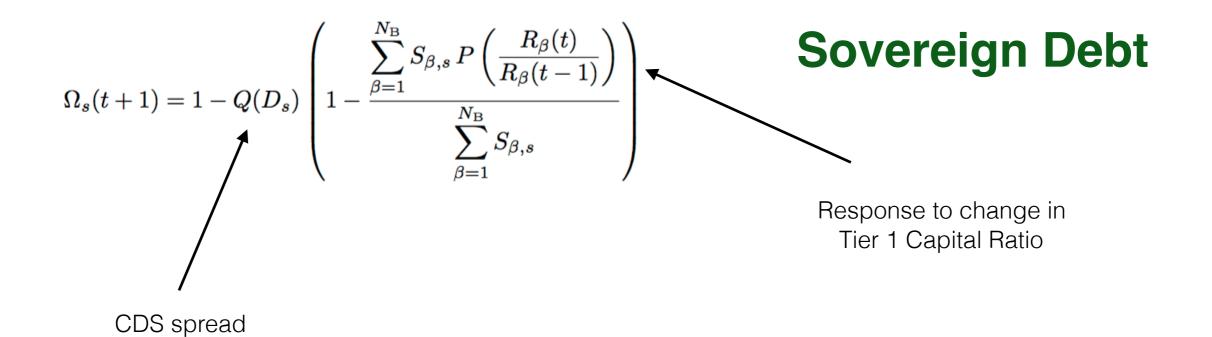
What happens to sovereign debts?

\boldsymbol{a}	Item		w_a range	A_a	w_a	$w_a A_a$
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While exposure is significant in absolute terms, the risk weights for sovereign debt are magnitudes lower than for other asset classes!



Spillover to other asset classes



Other Asset Classes
Asset spread
$$\Omega_{a}(t+1) = 1 - Q_{a} \left(1 - \frac{\sum_{\beta=1}^{N_{B}} A_{\beta,a} P\left(\frac{R_{\beta}(t)}{R_{\beta}(t-1)}\right)}{\sum_{\beta=1}^{N_{B}} A_{\beta,a}} \right)$$

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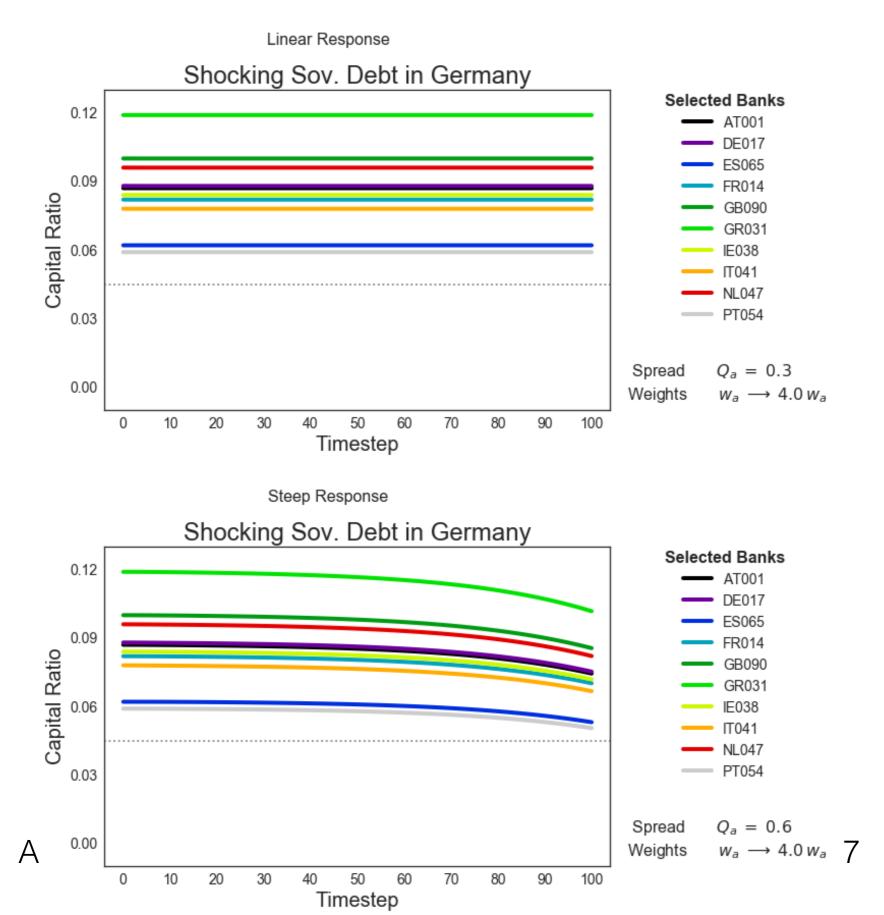
Shock Scenarios

- Increase in risk weights by sector and / or by country
- Reduce bank capital
- Vary spreading parameter
- Consider different bank response functions

Where does the Tier 1 Capital Ratio end up?

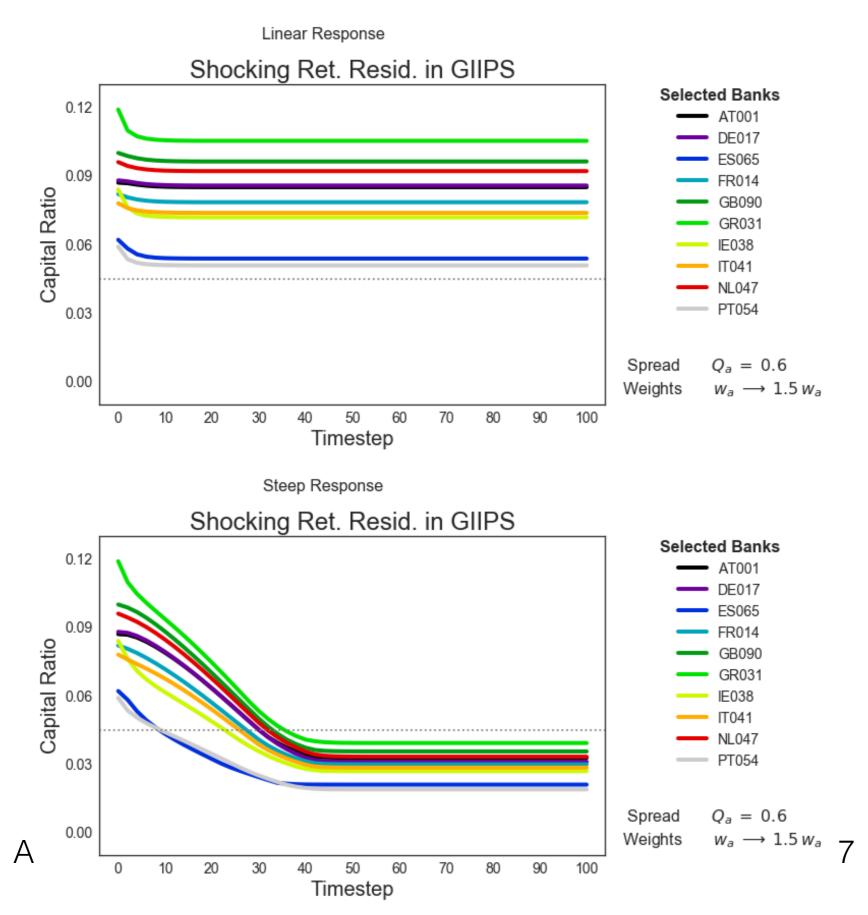


Exploring Different Scenarios





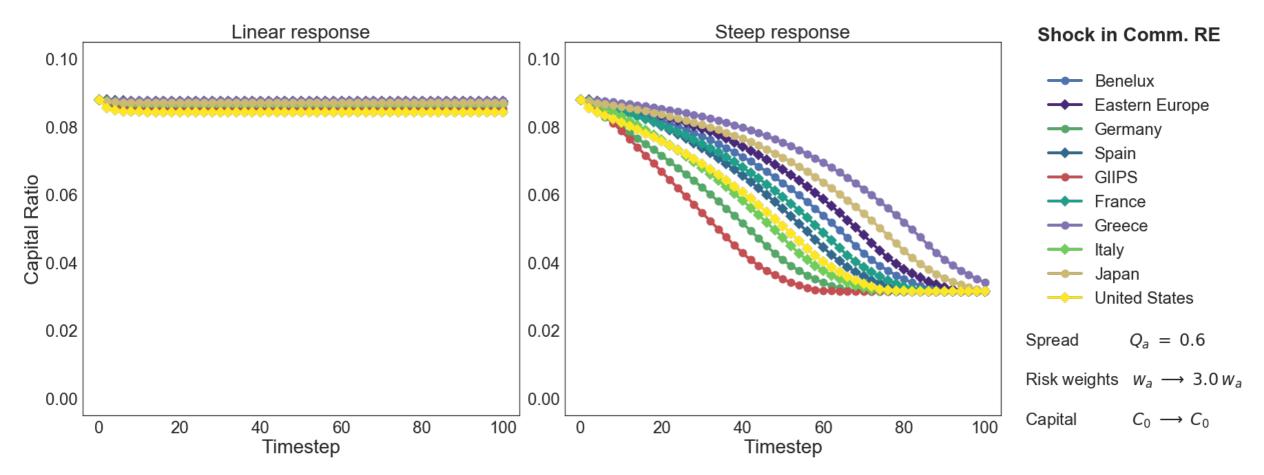
Exploring Different Scenarios



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One Bank's Tale

Deutsche Bank, given a shock in Commercial Real Estate



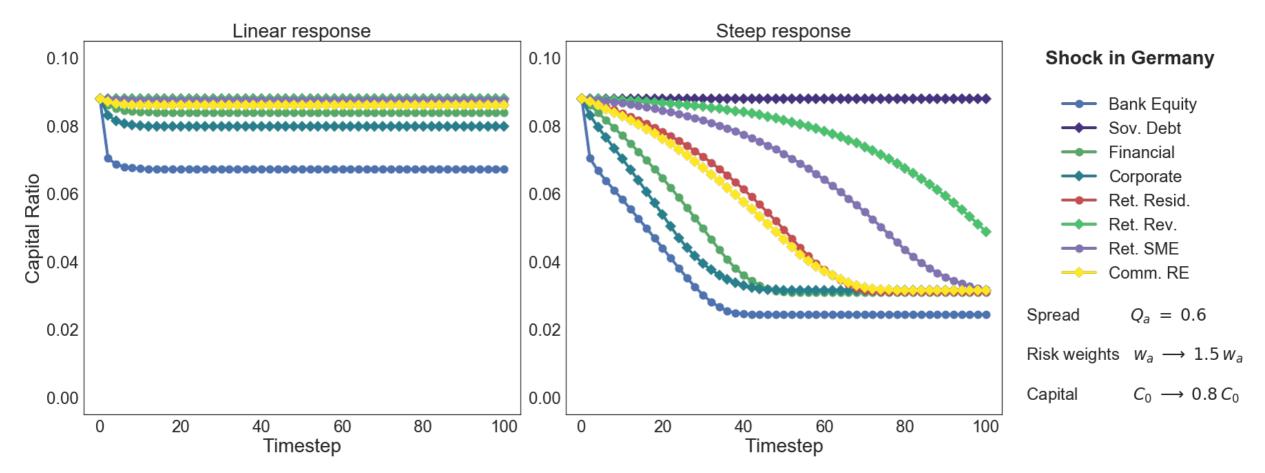
DEUTSCHE BANK AG (DE)

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One Bank's Tale

Deutsche Bank, given a shock in Germany



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Phase diagrams

- Study the outcome for banks after a fixed number of time steps for different parameters in different scenarios
 - initial shock size
 - spreading parameter
 - bank response function

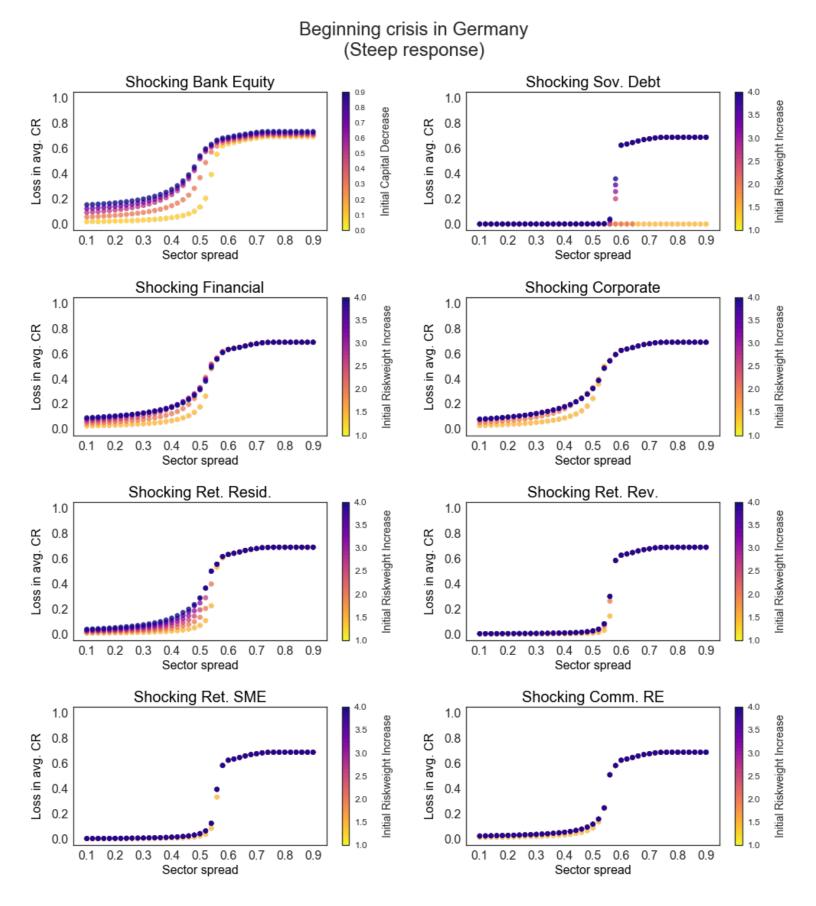


Beginning crisis in Germany (Linear response) Shocking Bank Equity Shocking Sov. Debt 40 1.0 1.0 0.8 Initial Riskweight Increase 0.8 CR 0.8 Loss in avg. CR 0.7 Initial Capital Decrea 0.6 3.0 Loss in avg. 0.6 0.6 0.5 25 0.4 0.4 0.4 0.3 2.0 0.2 0.2 0.2 1.5 0.1 0.0 0.0 0.0 1.0 0.3 0.5 0.6 0.7 0.8 0.9 0.6 0.7 0.8 0.9 0.1 0.2 0.4 0.2 0.3 0.4 0.5 0.1 Sector spread Sector spread Shocking Financial Shocking Corporate 4.0 40 1.0 1.0 Initial Riskweight Increase 3.5 3.5 Loss in avg. CR 0.8 Loss in avg. CR 0.8 Initial Riskweight Incr 3.0 3.0 0.6 0.6 2.5 2.5 0.4 0.4 2.0 2.0 0.2 0.2 1.5 1.5 0.0 0.0 1.0 1.0 0.4 0.5 0.6 0.7 0.8 0.9 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 0.1 0.2 0.3 Sector spread Sector spread Shocking Ret. Resid. Shocking Ret. Rev. 4.0 40 1.0 1.0 3.5 Riskweight Increas 3.5 Loss in avg. CR 0.8 0.8 Loss in avg. CR Riskweight Incr 3.0 3.0 0.6 0.6 2.5 2.5 0.4 0.4 2.0 2.0 0.2 0.2 Initial Initial 1.5 1.5 ********************** 0.0 0.0 1.0 1.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 0.4 0.5 0.6 0.7 0.8 0.9 0.1 0.2 0.3 Sector spread Sector spread Shocking Ret. SME Shocking Comm. RE 4.0 4.0 1.0 1.0 Riskweight Increase 3.5 3.5 8.0 USS IN AVE 2.5 CK 2 Riskweight Increa Loss in avg. CR 0.8 3.0 3.0 0.6 2.5 25 0.4 2.0 2.0 0.2 Initial nitial 1.5 1.5 0.0 0.0 1.0 1.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 Sector spread Sector spread

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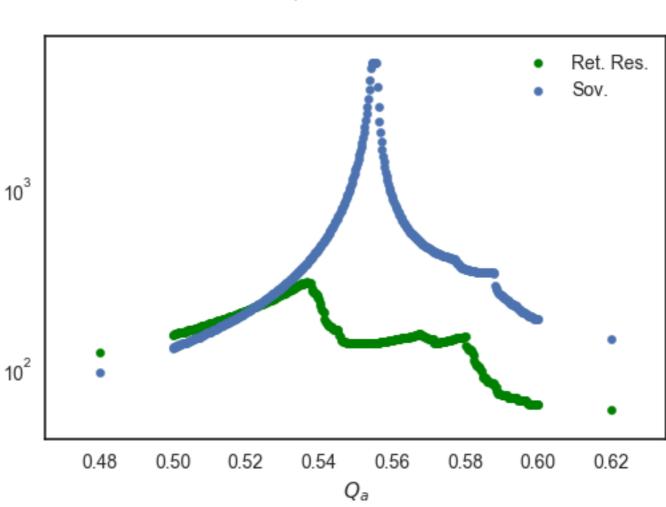
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Simulation run time

runs

- As system nears critical point, the time to reach the final configuration diverges
- Especially pronounced for sovereign debt: very sharp transition
- Non-monotonic decline of relaxation time after first transition; indicates further spread in network.



Shock in Germany

Conclusion

- Significant overlap in portfolios, especially through regional bias
- Outcome depends on spreading parameter: measure of contagion!
- Common exposures are more dangerous if the response to a shock is more risk-averse.

==> macroprudential approach is essential!

