

Modelling Financial Contagion to Examine Systemic Risks

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seminar: *Modelling and Simulating Social Systems with MATLAB*
ETH Zurich

December 15, 2015

Outline

1 Introduction and Motivation

2 Model

3 Capital Buffers

4 Bail-outs

5 Summary and Outlook

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5 Summary and Outlook

The recent crisis has shown the interconnectedness of banks as key issue w.r.t. financial stability

1. Which banks are better able to decrease their probability of being exposed to contagious defaults when adjusting their equity ratio?
2. Which bail-out strategy is the most effective w.r.t. impeding contagion and costs?

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Structure of Financial Network

- adjacency matrix A , A_{ij} exposure of i w.r.t j , A_{ji} liability of i towards j
- types of default:
 - 1 fundamental default
 - 2 contagious default

assets	liabilities
interbank assets ($A_i^{IB}(t)$)	interbank liabilities ($L_i^{IB}(t)$)
external assets ($A_i^E(t)$)	external liabilities ($L_i^E(t)$)
	equity ($E_i(t)$)

Network generation (Gabrieli2014)

- use real data from balance sheet
 - 1 interbank assets
 - 2 external assets
 - 3 equity
- core-periphery structure
- for adjacency matrix: assumption of perfectly diversified exposure

Fictitious Default Algorithm (Eisenberg2001)

- aim: calculate clearing vector for payment each round
- pure-default based algorithm
- algorithm
 - 1 assume that in $t = 0$ only banks default due to external shock
 - 2 round computing starts: solve linear equations for defaulting banks (assets=payment) and get their clearing payment
 - 3 iterate until no new default occurs

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Two strategies to adjust equity ratio

$$\text{equity ratio } ER_i(t) = \frac{E_i(t)}{A_i^{IB}(t) + A_i^E(t)}$$

- 1 periphery banks:** sell external assets and pay back external liabilities trying to get to required ER
alternatively: reduce interbank assets and reduce interbank liabilities \Rightarrow interconnectedness decreases
- 2 core banks:** can issue equity since more liquid but market with frictions ($\Delta E'_i(t)$ is additional equity amount needed):

$$\Delta E_i(t) = \min \left(\Delta E'_i(t), (E_i(0))^{1/t} \right), \quad \forall t \neq 0.$$

Simulation

- two situations: pre (no ER before shock) and post (ER before shock) crisis
- low ($a = 1, b = 10$) and high-risk ($a = 1, b = 1$) scenario drawn from $\text{Beta}(a, b)$
- simulation steps: 250
- ER varies from 0 to 0.4 in 0.01 steps
- those 50 banks with highest number of incoming links (interbank liabilities)

Low-risk scenario

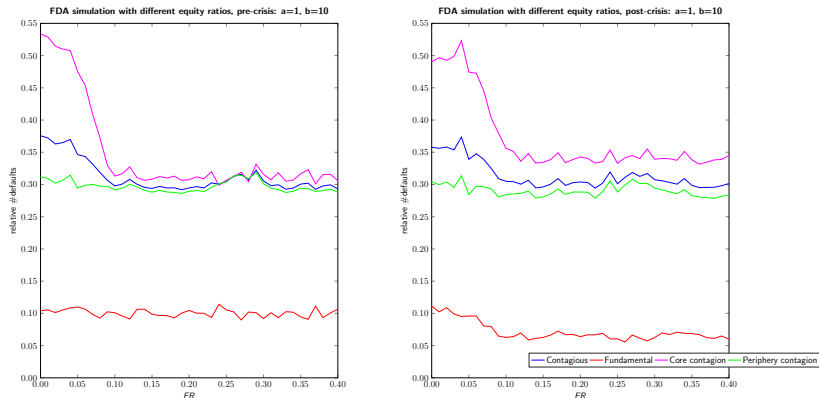


Figure: Plot for default probabilities in dependence of minimum capital requirement ER . Core (periphery) contagion means the fraction of core (periphery) banks that default due to contagion. Contagious (fundamental) default means the overall probability of defaults due to contagion (an exogenous shock).

High-risk scenario

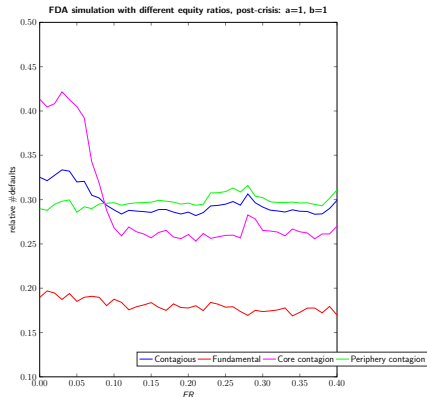
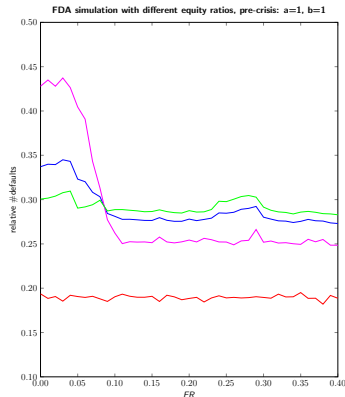


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Model

- lender of last resort can bail out defaulting banks to prevent the network from collapsing because of contagion
- Bailing out starts if 5% of banks fail through both fundamental and contagious defaults
- the equity injection is added onto external assets, so that the regulatory equity ratio of $ER=4.5\%$ is reestablished

$$\text{injection}_i(t) = \frac{L_i^{IB}(t) + L_i^E(t) - (1 - ER)(A_i^{IB}(t) + A_i^E(t))}{(1 - ER)}$$

Bail-out default simulation, high stress scenario

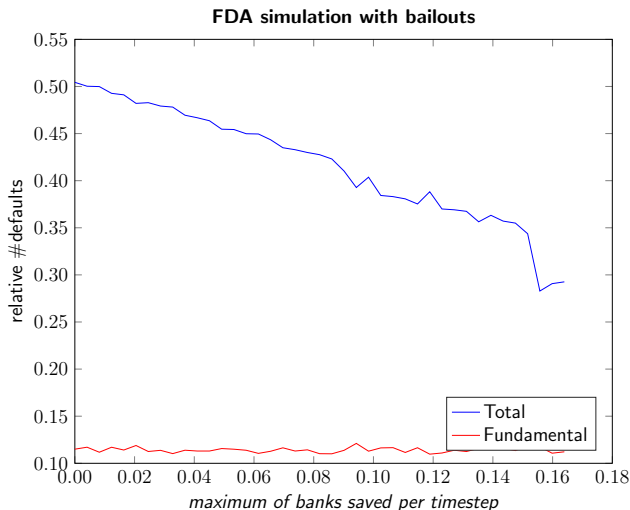


Figure: Plot for total defaults using eigenvector centrality as selection scheme in dependence of maximum number of banks bailed-out per timestep.

Bail-out cost simulation, high stress scenario

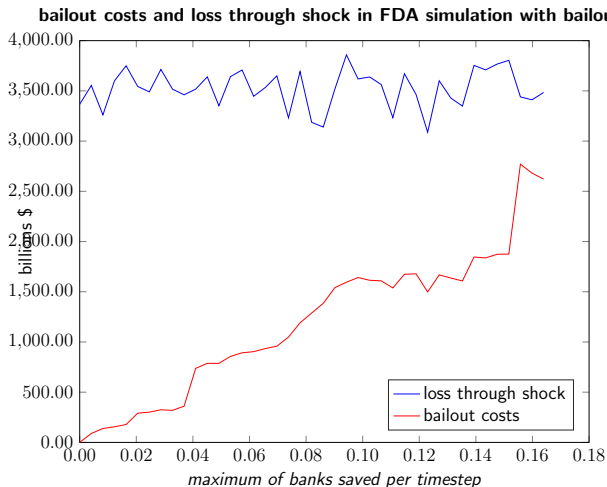


Figure: Plot for total bail-out cost and the loss from the initial shock using eigenvector centrality as selection scheme in dependence of maximum number of banks bailed-out per timestep.

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Policy implications I

- core banks better able to pursue a strategy to lower their risk of contagious default
- does not hold for periphery banks: alternative strategy
- policy makers have to specify how banks should adapt to capital requirements

Policy implications II

- bail-out: no significant differences w.r.t. bail-out costs in terms of selection scheme
- effect of an increase of max number of banks to save per time step is significant
- use mixed strategies, conduct simulation in a non pure-default model, e.g. DebtRank algorithm

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Data summary

country	count	Average tot asset	Min	Max	Sum
AT	7	68	22,5	189,1	472,7
BE	7	135	33,5	275,2	943,5
CH	15	151	20,0	884,7	2270,0
DE	34	245	21,6	2240,7	8324,5
DK	11	110	20,5	464,6	1215,4
ES	18	196	20,0	1266,3	3528,0
FI	4	134	29,7	346,2	537,0
FR	43	292	20,1	2077,8	12535,1
GB	39	277	20,5	1746,8	10803,1
GR	4	88	72,9	115,5	353,2
IE	8	52	20,9	129,8	419,5
IT	25	115	20,1	844,2	2864,8
LU	8	40	20,3	85,5	320,7
NL	8	302	20,0	828,6	2418,3
PT	6	82	22,5	189,1	492,4
SE	7	148	24,7	299,9	1032,8
244					48530,9

Table: Summary statistics on banking data from Bankscope. Interbank market characterisation w.r.t. total asset position in \$bn.