



Bank networks, interbank liquidity runs and the identification of banks that are Too Interconnected to Fail

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What do we do?

Basic idea of the paper

- 1. Identify the scenarios that are sufficient to simulate real interbank market crises
- 2. Use this methodology to calculate the potential contribution of bank_{i,t} to contagion in period t.
- 3. Identify the systemically important banks (SIFI or superspreaders) using only data on the position of the bank in the network, as opposed to size.

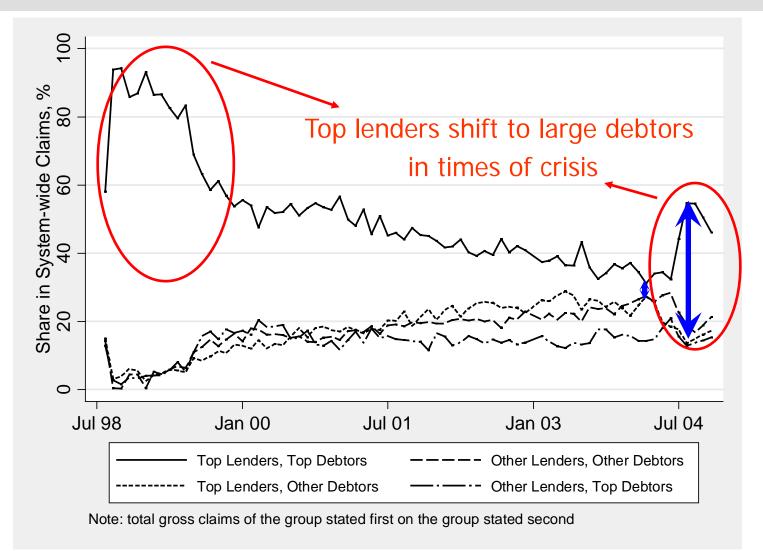
Basic findings

- Capital contagion, funding liquidity losses from infected banks and haircuts are not sufficient
- We need liquidity hoarding to reliably simulated real banking crises (preferential detachment)
- The superspreaders (SIFI) are best identified by their position in the network (K-shell index)
- This is NOT the same as size
- Incomplete network data already does a good job

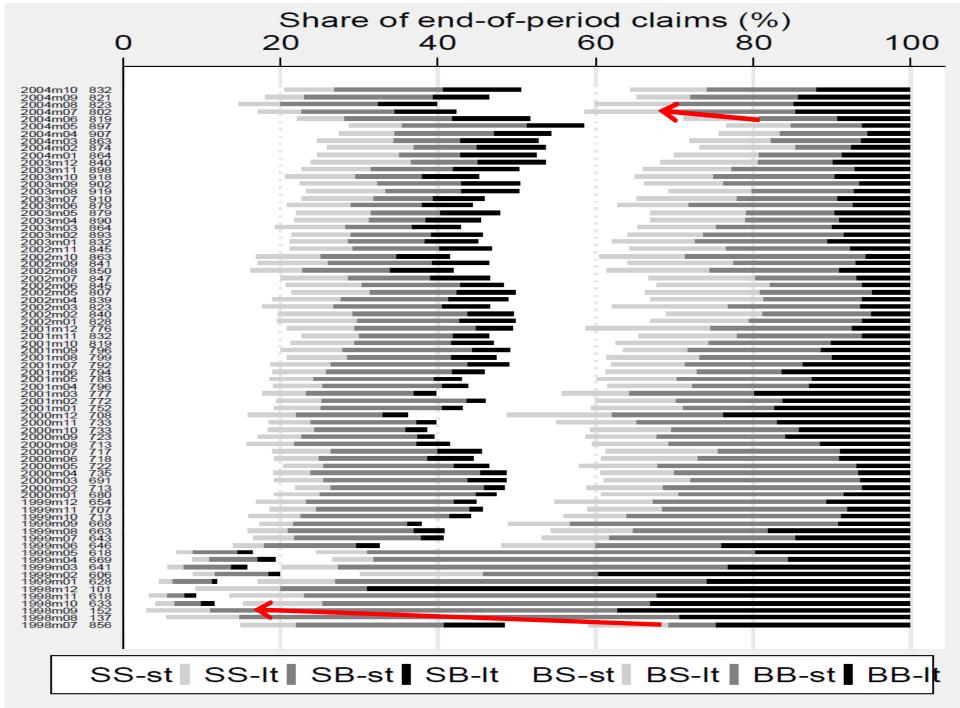
We use Russian data as a training dataset

- 75 months of complete bilateral contract data (98-04)
 - Identity of both parties
 - Contract types
 - Volumes
 - Maturities
 - Prices
- Monthly bank balances and P&L (Interfax, Mobile)
 - Capital, liquidity, reserves, securities
- Two real but very different interbank market crises
 - The infamous 1998 default
 - The 2004 panic that was only stopped by deposit insurance
- An almost experimental setting

Flight to quality in crisis time



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Interbank market contagion

Scenario 1

- Credit losses deplete a bank's capital
- Default on interbank obligations
- Potential domino effects via credit losses of other banks
- Contagion propagates until it stops

Scenario 2

- We add funding liquidity losses
- The borrowers of the initial failing bank lose funding that can only partially be replaced
- If the loss >liquid assets, we get haircuts on fire sales
- More banks fail in the further rounds

Interbank market contagion

Scenario 3

- If a bank severely hit by scenario 1/2, it may face a run on total interbank obligations by uninfected banks as in Rochet and Vives (2004).
- Preferential detachment from banks that are hit but still solvent and liquid
- The network structure itself changes endogenously
- This does the trick
- Scenario 4
 - Panic and complete liquidity hoarding
 - All banks run on each other regardless fundamentals

The early literature

- Early theoretical literature was based on capital channel
 - Allen and Gale (2000)
- Early empirical literature was based on the capital channel
 - Sheldon and Maurer (1998) for Switzerland,
 - Furfine (2003) for the U.S.,
 - Upper and Worms (2004) for Germany,
 - Lelyveld and Liedorp (2006) for the Netherlands,
 - Degryse and Nguyen (2007) for Belgium

New channels

- Fire sales, haircuts and asset prices
 - Eisenberg and Noe, 2001
 - Cifuentes et al. (2005), Shin (2008)
- Liquidity hoarding and rund
 - Rochet and Vives (2004): large well-informed investors don't renew interbank credit if a large adverse shock to one bank creates uncertainty about other banks
 - Also Müller, 2006
- Overview of possible channels in Upper (2001)
- Recent theoretical contributions of Gai, Haldane and Kapadia (2010, 2011)

Bilateral simulations

- Krause and Giansante (2011)
 - Generate theoretical networks and attack them
 - Draw conclusions about contagion

Our approach

- Start from real endogenously formed network
- Attack it allowing increasingly more damaging channels
- Random attack (we also did correlated attacks)
- Till you reproduce the real crises
- Then use the scenario to calculate the SIFI banks (those with largest contributions to contagion)
- And identify them with more limited information

Formal bank balance sheet

Panel A. Simplified bank balance sheet identity

$$r_i + \sum_{j=1}^n y_{ij}^{st} + \sum_{j=1}^n y_{ij}^{lt} + s_i + a_i = c_i + \sum_{j=1}^n y_{ji}^{st} + \sum_{j=1}^n y_{ji}^{lt} + l_i$$

Formal condition set (solvency, liquidity, Infection)

Panel B. Conditions for being insolvent (S), illiquid (L) and infected (I)

Irreplacable funding liquidity loss Remaining liquid assets S1 $c_i < \lambda \sum_{j=1}^n \theta_j y_{ij}$ $S1 = c_i < \lambda \sum_{j=1}^{n} \theta_j y_{ij} + \max \left\{ 0, \delta \left[\rho \sum_{j=1}^{n} \theta_j (y_{ji}^{st} + y_{ji}^{lt}) - r_i - \sum_{j=1}^{n} (1 - \theta_j) (y_{ij}^{st} + y_{ij}^{lt}) \right] \right\}$ $S3 = c_i < \lambda \sum_{j=1}^{n} \theta_j y_{ij} + \max \left\{ 0, \delta \left[\sum_{j=1}^{n} (y_{ji}^{st} + y_{ji}^{lt}) - r_i - \sum_{j=1}^{n} (1 - \theta_j) (y_{ij}^{st} + y_{ij}^{lt}) \right] \right\}$ $L1 \quad r_i + \sum_{j=1}^n (1 - \theta_j) (y_{ij}^{st} + y_{ij}^{lt}) + (1 - \frac{\delta}{1 + \delta}) s_i < \rho \sum_{j=1}^n \theta_j (y_{ji}^{st} + y_{ji}^{lt})$ $L2 \underbrace{r_i + \sum_{j=1}^n (1 - \theta_j)(y_{ij}^{st} + y_{ij}^{lt})}_{I1 \quad 0 < \lambda \sum_{j=1}^n \theta_j y_{ij}} \underbrace{+ (1 - \frac{\delta}{1+\delta})s_i}_{Market value securities after haircut}$ $0 < \rho \sum_{i=1}^{n} \theta_i (y_{ii}^{st} + y_{ii}^{lt})$ I2 $\max\left[0, (1-\mu)c_{i}\right] < \lambda \sum_{j=1}^{n} \theta_{j} y_{ij} + \max\left\{0, \delta\left[\rho \sum_{j=1}^{n} \theta_{j} (y_{ji}^{st} + y_{ji}^{lt}) - r_{i} - \sum_{j=1}^{n} (1-\theta_{j})(y_{ij}^{st} + y_{ij}^{lt})\right\}\right\}$ I3 $I4 \quad (1-\mu)r_i < \rho \sum_{j=1}^n \theta_j (y_{ji}^{st} + y_{ji}^{lt})$ where: $\theta_j = 1$ if bank j has defaulted, and 0 otherwise λ - loss given default (LGD) on interbank assets ρ - fraction of lost funding from failed banks that cannot be replaced δ - fire sale asset haircut: selling assets worth $(1 + \delta)$ a bank takes a loss of δ $(1-\mu)$ - fraction of capital c_i / reserves r_i needed to be destroyed to trigger a run

Scenario's

Panel C. Default rules for different contagion scenarios

Contagion scenario 1a: credit loss 2a: credit + funding loss 3a: credit + funding loss + run on infected 4a: credit + funding loss + run on all 2s, 3s, 4s:

 $\begin{array}{l} \mbox{Default rule} \\ S1 \& I1 \\ (S2 \ or \ L1) \& \ (I1 \ or \ I2) \\ \{(S2 \ or \ L1) \& \ (I1 \ or \ I2)\} \ or \ \{(S3 \ or \ L2) \& \ (I3 \ or \ I4)\} \\ S3 \ or \ L2 \\ \mbox{same as } 2a, \ 3a, \ 4a \ but \ all \ y^{lt} = 0 \end{array}$

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On haircuts

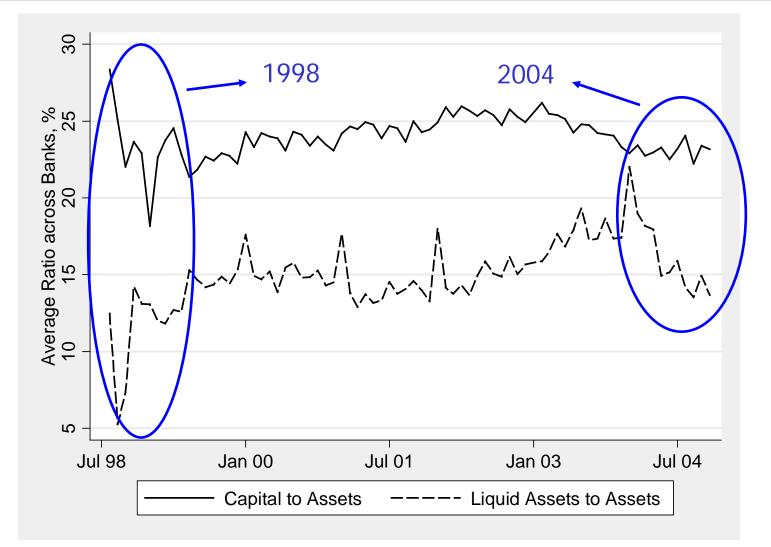
Why not endogenous?

- We could increase the haircut in function of results of previous rounds (spirit of Eisenberg and Noe; Müller)
- But this would only reinforce results

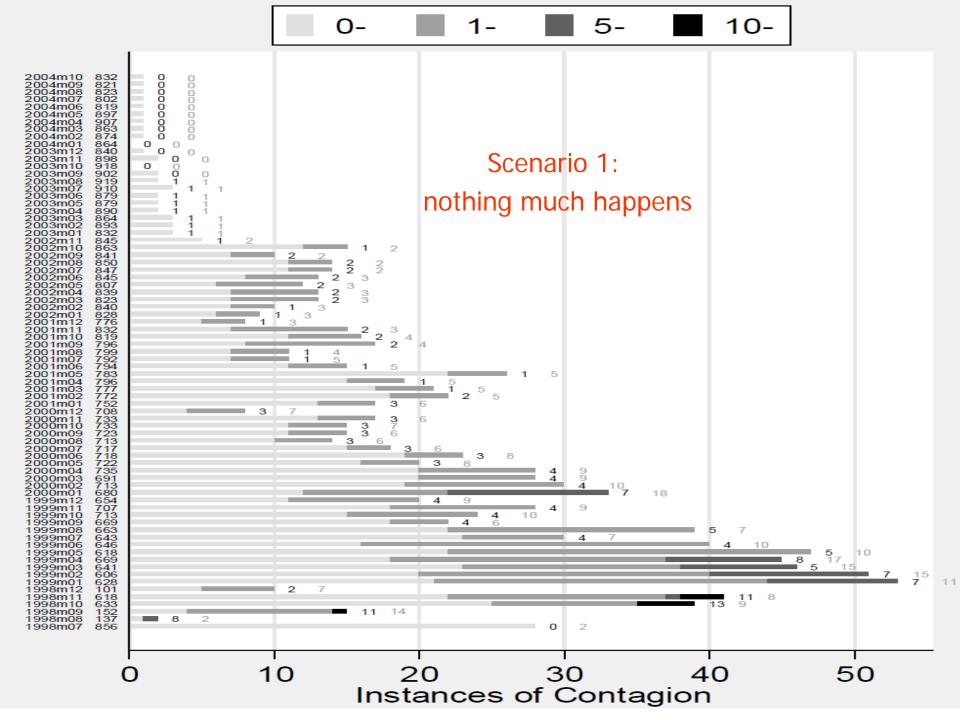
Why not after liquidity hoarding?

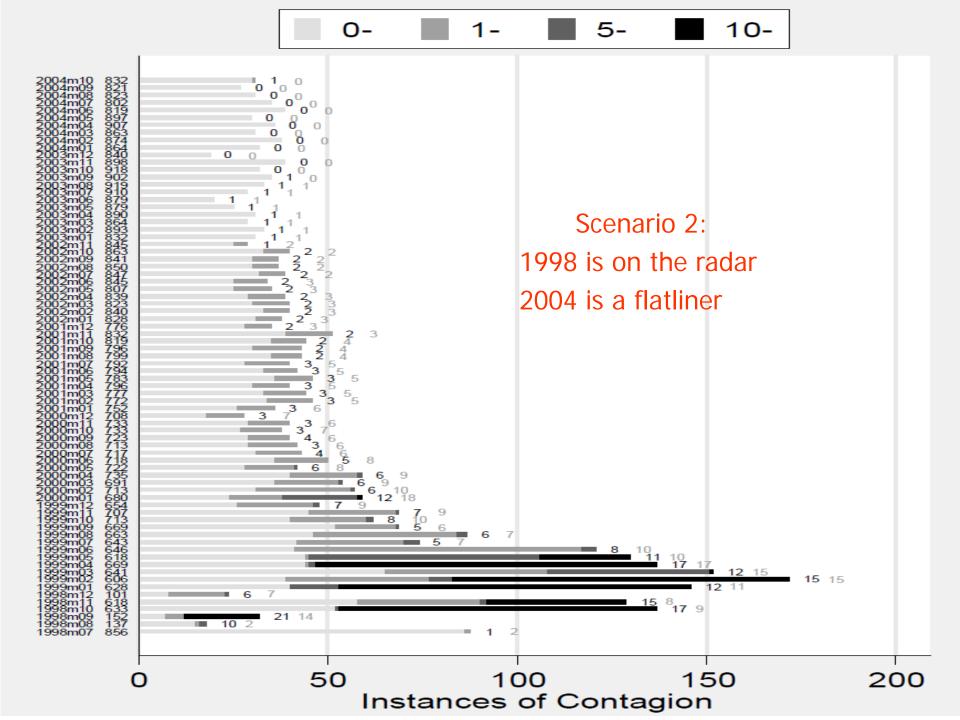
- We could also change the order,
- but the scenario with hoarding, but no haircut yet, would suffice to get contagion
- Haircut would then drop from the simulation scenario
- More important in more developed markets?

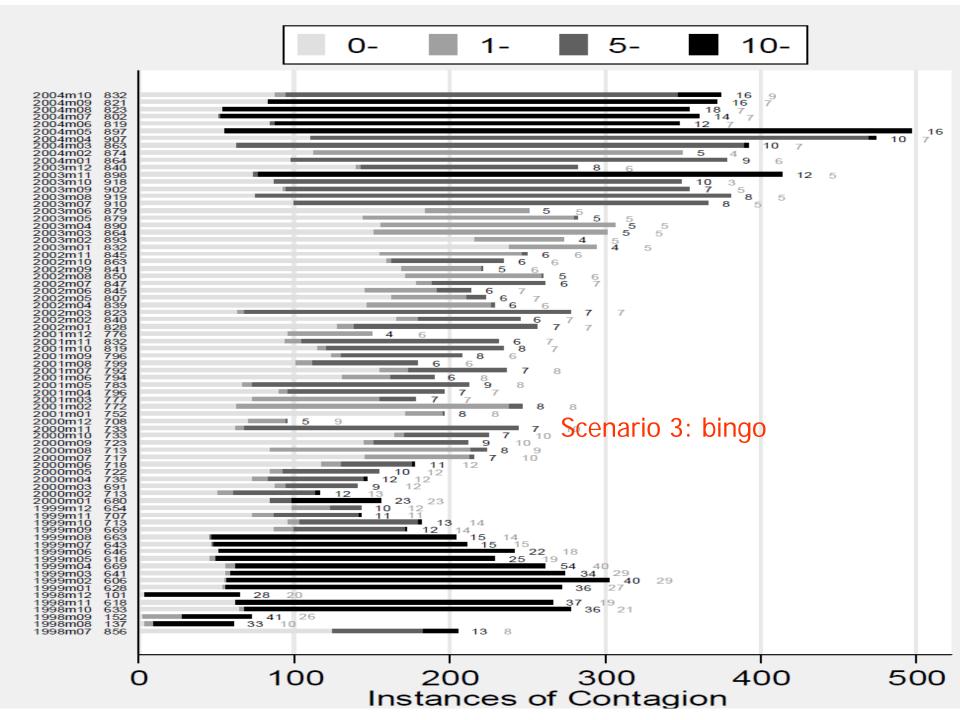
Financial crises and bank health Capital versus liquidity



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Intermediate conclusion

- The capital channel does not suffice
 - The 1998 crisis is somewhat predicted by it
 - The 2004 crisis is off the screen
- Funding liquidity and asset sales don't do it either
 - 1998 is now really on the screen
 - 2004 is still flat
- Scenario 3 captures both crisis periods
 - Liquidity runs and preferential detachment are essential
 - We will use this scenario to calculate individual banks' contributions to contagion in a second step

Identifying the spreaders of contagion

- We have identified by simulation the banks that contribute most to contagion
- The question: can we identify the "SIFI" by
 - Looking at the structure of the network
 - And at the position of banks in the network
- Conventional wisdom
 - Degree, centrality indices, betweenness
- Our contribution
 - K-core centrality

Last step:

Table 2: Centrality Indices

Index	Formula	Description		
Valued Outdegree	$0 \leq VO_i = rac{\sum_{j=1}^n y_{ij}}{ ext{System-wide Assets}} \leq 1$	bank share in system-wide interbank assets		
Valued Indegree	$0 \leq VI_i = rac{\sum_{j=1}^n y_{ji}}{ ext{System-wide Liabilities}} \leq 1$	bank share in system-wide interbank liabilities		
Non-valued Outdegree	$0 \leq NO_i = \frac{\sum_{j=1}^n (y_{ij} > 0)}{n-1} \leq 1$	% of market participants a bank has as counterparties on its asset side		
Non-valued Indegree	$0 \le NI_i = \frac{\sum_{j=1}^n (y_{ji} > 0)}{n-1} \le 1$	% of market participants a bank has as counterparties on its liability side		
Betweenness Centrality	see Miura (2011) whose Stata Graph Library we use	% of shortest paths linking institutions other than bank i passing through bank i		
where	$y_{ij}-$ gross claims of bank i on bank j			

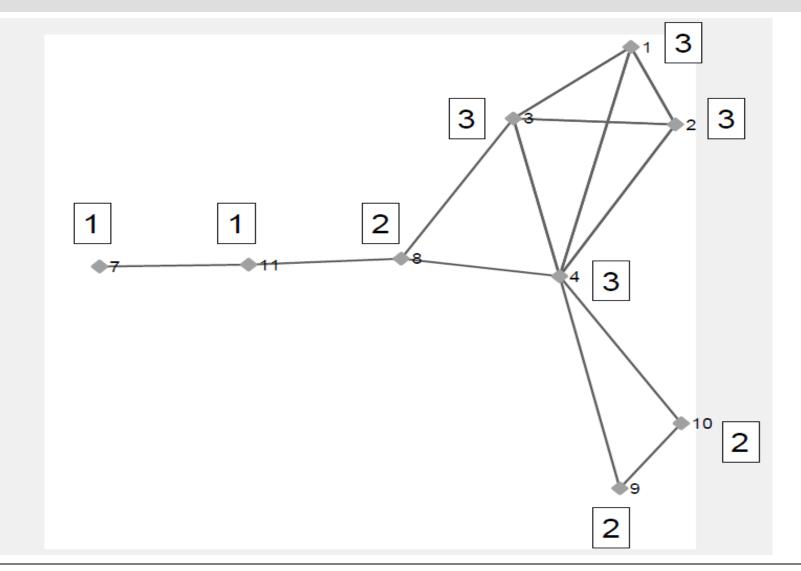
 $(y_{ij} > 0)$ evaluates to 1 if bank *i* has claims on bank *j*; and 0 otherwise (n-1)- max number of links a bank can have

i

Concepts from econophysics

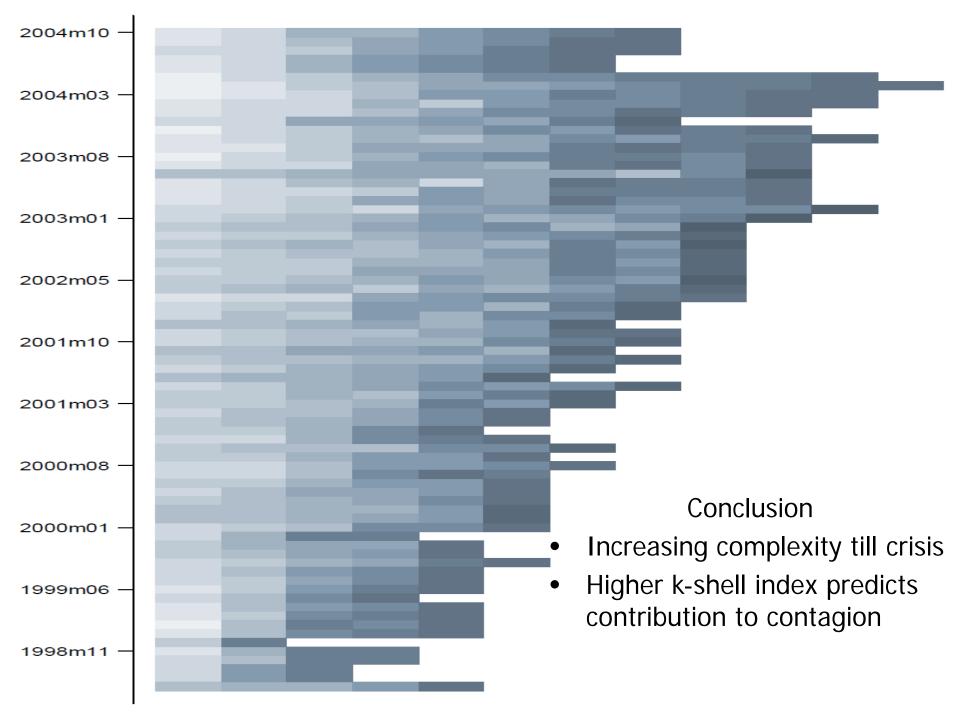
- Conventional wisdom
 - Centrality of a node in a network predicts the node's potential to spread contagion
- Kitsak et al.
 - Challenge this view for a variety of networks
 - Shows that the K-shell index (result from K-core decomposition) beats any traditional network variable
 - We introduce this concept to the banking literature
 - The measure is unweighted and undirected

K-core decomposition analysis



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$C_{it} = \alpha + \beta' Bank_{it} + \lambda_t + \varepsilon_{it}$

Table 3: Identifying Influential Spreaders

	C = Share of failed banks			C = Share of failed assets			
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	
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NI	1.11***		-0.71***	2.88***		-2.00***	
	(8.6)		(-5.6)	(7.7)		(-5.4)	
NO	2.32^{***}		0.17^{*}	5.55^{***}		-0.22	
	(11.2)		(1.9)	(9.9)		(-1.0)	
VI	0.38^{**}		0.30^{***}	1.15^{**}		0.93^{***}	
	(2.6)		(3.0)	(2.5)		(2.9)	
VO	0.09		0.10*	0.33*		0.35	
	(1.6)		(1.7)	(1.7)		(1.6)	
Betw	-0.74***		0.56^{***}	-1.89***		1.59^{***}	
	(-6.9)		(5.4)	(-6.4)		(5.3)	
Size	0.04		-0.02	0.11*		-0.04	
	(1.5)		(-1.4)	(1.6)		(-0.7)	
K-shell index		0.01***	0.01***		0.02^{***}	0.02^{***}	
		(47.6)	(33.2)		(42.6)	(32.1)	
Constant	-0.04***	-0.05***	-0.05***	-0.11***	-0.14***	-0.14***	
	(-21.6)	(-29.4)	(-29.4)	(-20.5)	(-27.7)	(-27.9)	
Observations	56,782	56,782	56,782	56,782	56,782	56,782	
AIC	-35266	-39023	-40119	3026	-443.9	-1297	
BIC	-34532	-38334	-39376	3760	245.0	-554.1	
ML (Cox-Snell) $R2$	0.268	0.315	0.328	0.233	0.278	0.289	
McKelvey-Zavoina's $R2$	0.328	0.397	0.409	0.287	0.355	0.365	

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K-shell index versus size

- K-shell index is unweighted and undirected
- Consider the simple weighted K(a)-index, that consider only the a% largest edges
- Standard K = K(100)
- Calculate K(50)
 - Correlation K(50), K(100) = 0.85
 - In the regressions K(50) is clearly weaker than K(100)
 - But still far stronger than anything else
- More complex weighing schemes give same result

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	K-core	NI	NO	VI	VO E	Betw :	Size
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Policy implications

- Basel III capital conservation and countercyclical buffers, fully effective on 1 January 2019.
- Higher loss absorbency requirements for SIFI
- Basel SIFI: an indicator-based approach
 - size,
 - interconnectedness,
 - lack of readily available substitutes
 - Global (cross-jurisdictional) activity
 - complexity.
- it has been suggested that size is the main indicator of systemic importance.

Our analysis challenges this wisdom

- Liquidity runs and preferential detachment are at the heart of banking panics
- By consequence a bank's position in the network (K-shell) may be more important for its "coreness" to the system than size
- Data on the biggest bilateral links may suffice to identify the SIFI (the K(50) results)
- It may be wise for the guardians of financial stability to invest in this

Concluding remarks

- Liquidity hoarding
 - Is relevant to financial stability
 - though theoretical effects are poorly understood
- Supervisors who knows interbank market structure
 - can predict the stability of the interbank market
 - can identify SIFI who are too interconnected to fail
 - Can demand from them higher capital buffers
- The lender of last resort
 - Can solve the problem by timely and targeted injections,
 - As to keep upright the 'too central to fail' banks in the heat of the moment.