



OESTERREICHISCHE NATIONALBANK
EUROSYSTEM

Structure and Stability in Complex Systems

An empirical analysis of the interbank payment system ARTIS

Evidence

Claus Puhr, Stefan W. Schmitz, OeNB

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Agenda

Motivation

Related Research

ARTIS Simulations

Network Indicators

Panel Data Analysis

Conclusion

Motivation and objectives

- **Motivation**
 - OeNB is in charge of payment system oversight
 - ESCB/OeNB objective: smooth functioning of the payment system
- **Objectives**
 - Statistical analysis for a better understanding of ARTIS
 - Analyse impact of operational risk in payment systems
 - On the system and the individual bank level
- **Policy implications?**

Research questions

- **Based on Schmitz / Puhr (2006, 2007, 2009)**
- **We operate with real rather than simulated liquidity data**
- **We run various operational stress scenarios**
- **We observe large variations across scenarios and days**
- **We try to uncover systemically important accounts**
- **We try to explain variations across scenarios and / or days**
- **In particular with regard to payment system structure / topology**

Methodological approach

- **Network theory**
 - Robustness studies
 - Typology of flow processes
 - Measures of network structure
- **Simulation studies**
 - Simulations of operational shocks generate contagion
- **Panel data econometrics**
 - Variations of contagion across scenarios and / or days

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Network theory

- **Robustness studies (Albert et al. 1999, 2000: Internet)**
 - Nature of shock: removal of nodes and links from network
 - Measure of impact: Connectivity measured by size of largest cluster and average path length

- **Robustness in ARTIS**
 - ARTIS is a physically complete network
 - Flow of liquidity not equal to flow of information in the internet
 - Connectivity inappropriate conceptualisation of network stability
 - Incoming links to stricken account not removed

Payment system research

- **Simulation studies (Leinonen (ed.) 2005, 2007, 2009)**
 - Many conducted with the Bank of Finland Payment System Simulator
 - Publicly available tool to simulate payment systems
 - Based on real and / or simulated transaction data
- **Network topology studies (Soramäki et al. 2006)**
 - The topology of interbank payment flows in FedWire
 - For a comparison of network indicators across networks refer to Schmitz et al. 2008

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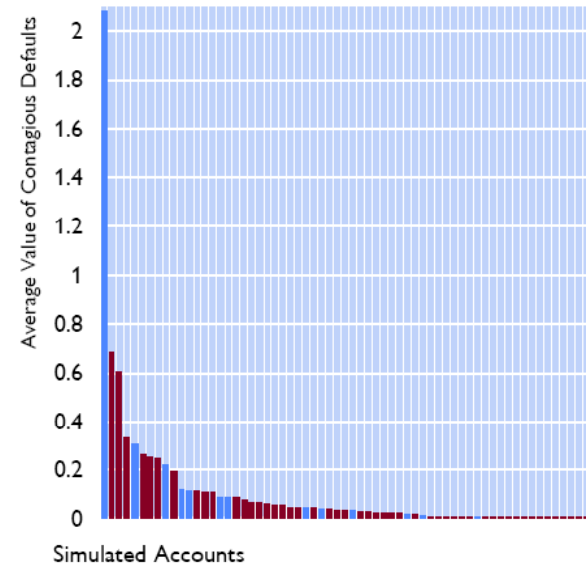
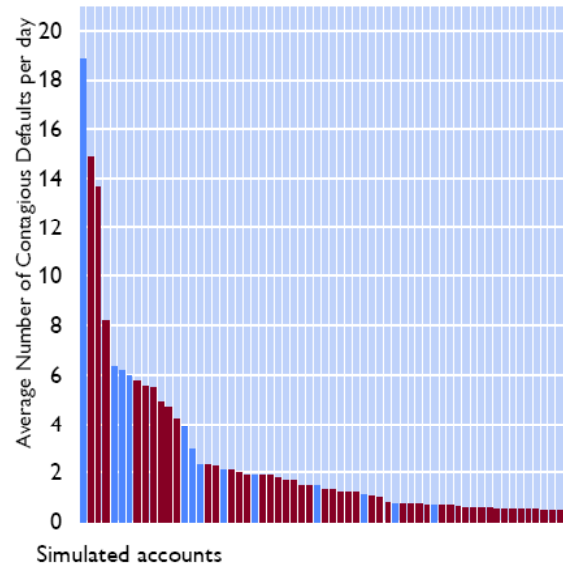
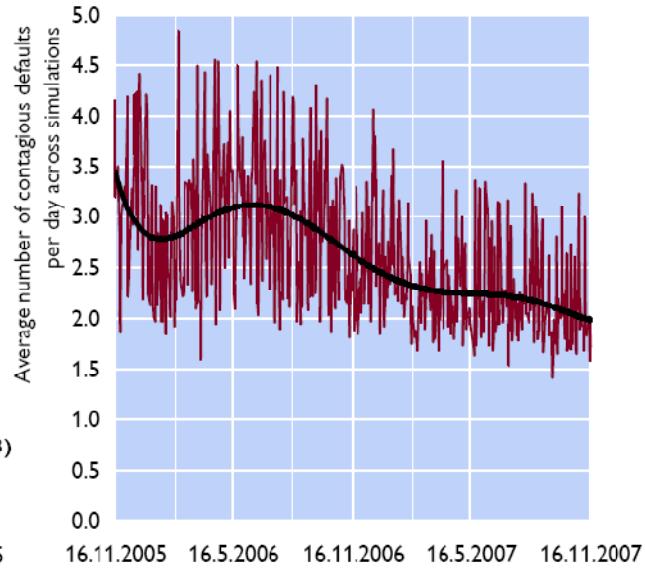
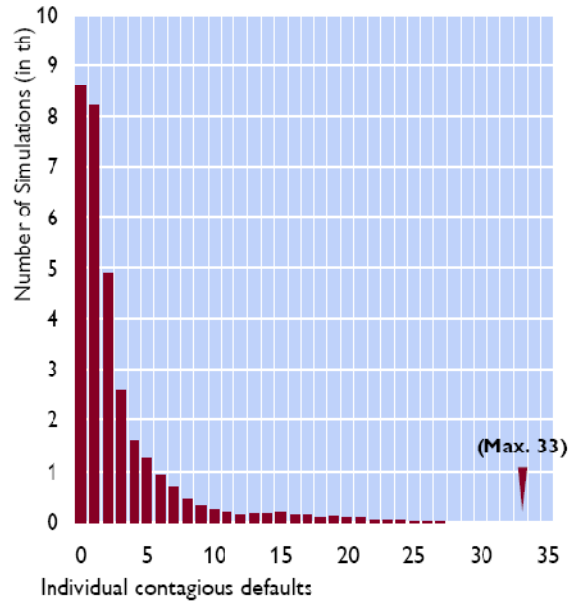
Data requirements

- **Transaction data**
 - Payer, payee, value, time stamp
 - If available prioritisation and underlying economic purpose
- **Liquidity data**
 - Beginning of day balance and collateral
- **Institutional data**
 - Settlement algorithm and attribution of accounts to economic entities
 - Institutional features: e.g. stop-sending rule, through-put rules, ...
- **Additional data**
 - Qualitative information (e.g. experience of operators)

Simulations

- **Assumption: one day incapacitation to submit payments**
- **Sample period: 16 November 2005 to 16 November 2007 (497 days)**
- **63 scenarios**
 - 50 banks which are in GSCC on all days in the sample period
 - 13 transfer accounts which are part of the system on all days
- **Matlab based simulation tool**
 - Accounts for all institutional features of ARTIS (e.g. stop-sending rule, direct debit)
- **31 311 simulations (63×497) with 650 mn transactions**

Simulation results



■ Bank accounts ■ Transfer accounts

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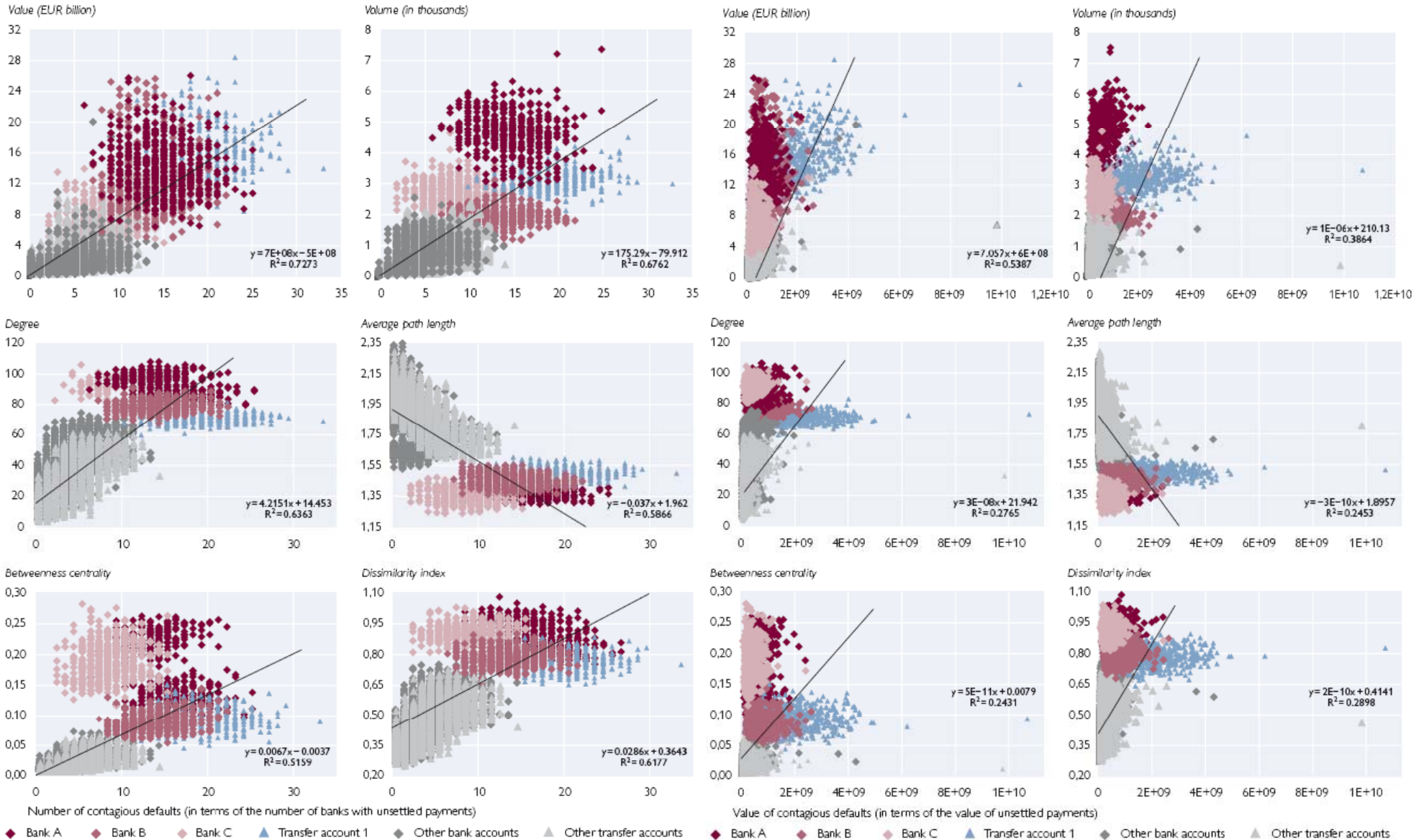
Panel Data Analysis

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Network indicators

- **Appropriate measure of network structure?**
 - Directed / undirected? Weighted / unweighted?
 - Network level (44) & node level (stricken bank) (71)
- **Selection based on**
 - **Comparability (e.g. indicators that were used in other studies)**
 - Albert et al. (1999, 2000: Internet) – average path length
 - Boss et al. (2004: Interbank liabilities) – betweenness centrality
 - **Theoretical considerations**
 - Driven by typology of flow processes (Borgatti 2005)
 - Route / transfer characteristics
 - Liquidity follows a walk and is transferred

Node-level network indicators vs. unsettled payments



Correlation between volume, value and network indicators

	Volume	Value	Avg. PL	Degree	Conn.	Clust.	Btw. C.	Dissim.
Volume	100%	89%	-77%	84%	83%	-57%	89%	85%
Value		100%	-70%	76%	75%	-52%	77%	78%
Avg. PL			100%	-96%	-97%	62%	-79%	-85%
Degree				100%	99%	-77%	85%	95%
Conn.					100%	-72%	85%	93%
Clust.						100%	-56%	-78%
Btw. C.							100%	87%
Dissim.								100%

Source: OeNB. Average Path Length (Avg. PL), Connectivity (Conn.), Clustering Index (Clust.), Betweenness Centrality (Btw. C.), Dissimilarity Index (Dissim.).

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Dependent variables

- **Measures of contagion (excl. stricken bank)**
 - Value of unsettled payments at end of day
 - Number of unsettled payments at end of day
 - Number of banks with unsettled payments at end of day

Independent variables

- **Network level (constant across panels but not across time [Z])**
 - Aggregate liquidity
(BoD balances + unencumbered collateral across banks)
 - Network indicators at the network level
- **Node level (varies across panels and across time [X])**
 - Liquidity loss due to operational problem at stricken bank
(liquidity sink/drain, unreceived payments)
 - Network indicators at the node level
- **Dummy for transfer accounts (D_{\times} unreceived payments)**

Model

$$\begin{bmatrix} y_1 \\ y_2 \\ \cdot \\ y_{63} \end{bmatrix} = \begin{bmatrix} X_1 \\ X_2 \\ \cdot \\ X_{63} \end{bmatrix} \beta_1 + [Z] \beta_2 + \begin{bmatrix} v_1 \\ v_2 \\ \cdot \\ v_{63} \end{bmatrix} + \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \cdot \\ \varepsilon_{63} \end{bmatrix}$$

$$E[\varepsilon_{it} | x_i, v_i] = 0 \quad T = 1 \dots 497, \quad N = 1 \dots 63.$$

$$Var[\varepsilon_{it} | x_i, v_i] = \sigma^2 I_T$$

$$Cov[\varepsilon_{it}, \varepsilon_{js}] = 0 \quad \text{if } t \neq s \quad \text{or } i \neq j.$$

Assumptions

- **Cross-panel conditional homoskedasticity**
 - Variance of error terms constant across panels (and across time)
- **Serial independence**
 - Error terms are serially uncorrelated within panels
- **Cross-panel independence**
 - Error terms are independent across panels
- **Strict exogeneity**
 - Error terms and explanatory variables are independent

Estimation procedure

- **Estimate fixed-effects model (yields inconsistent standard errors)**
- **Correct for cross-section conditional heteroskedasticity, autocorrelation and cross-section-dependence**
- **Prais-Winsten regression, PCSE (Panel-Corrected Standard Errors)**
 - Accounts for heteroskedasticity and cross-panel correlation
 - Additional option panel-specific autocorrelation
- **Estimate three models without network indicators**
 - One for each measure of contagion
 - Add individual network indicators at node- and network-level
 - One at a time, due to high correlation between network indicators

Basic models used for tests

Model 1

$$\begin{aligned} \text{simdefaults}_{it} = & \alpha_{it} + \beta_1 \text{Liquidity}_t + \beta_2 \text{simunrvol}_{it} + \beta_3 \text{TransUnrVol}_{it} + \\ & + \beta_4 \text{nodeavgpath}_{it} + \beta_5 \text{netavgpath}_t + u_i + \varepsilon_{it} \end{aligned}$$

Model 2

$$\begin{aligned} \text{simqueuednum}_{it} = & \alpha_{it} + \beta_1 \text{Liquidity}_t + \beta_2 \text{simunrdvol}_{it} + \beta_3 \text{TransUnrVol}_{it} + \\ & + \beta_4 \text{nodeavgpath}_{it} + \beta_5 \text{netavgpath}_t + u_i + \varepsilon_{it} \end{aligned}$$

Model 3

$$\begin{aligned} \text{simqueuedvol}_{it} = & \alpha_{it} + \beta_1 \text{Liquidity}_t + \beta_2 \text{simunrvol}_{it} + \beta_3 \text{TransUnrVol}_{it} + \\ & + \beta_4 \text{nodeavgpath}_{it} + \beta_5 \text{netavgpath}_t + u_i + \varepsilon_{it} \end{aligned}$$

Estimation results (1/2)

- **Explanatory value of models is high (40 to 70 per cent)**
 - Much higher for between than for within panel variation
- **Results robust across specifications & estimation methods**
- **Higher liquidity reduces contagion effect**
- **Higher liquidity loss increases contagion effect**
 - Impact highest for value of unsubmitted payments
 - Less for liquidity drain and liquidity sink
 - Variable has very high explanatory power
- **Transfer accounts cause significantly more contagion**

Estimation results (2/2)

- **Higher value of transactions in the network reduces contagion**
 - Time trend?
- **At the network level no indicator is significant in all three models**
- **At the node level three network indicators are significant in all models**
 - Higher node degree and connectivity increase contagion
 - Higher average path length decreases contagion
 - More central nodes cause more contagion
 - Additional explanatory value:
 - 3 per cent regarding the number of contagiously defaulting banks
 - Negligible in the other two models

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- **Number of systemically important accounts is low**
 - At least one default / day: 11 transfer accounts & 28 banks
 - 0.1% of average total value / day: 7 transfer accounts & 17 banks
- **Network indicators in payment systems**
 - Degree seems to be adequate indicator
- **Panel approach yields high explanatory value**
 - Higher between scenarios than within
- **Most of the variation is explained by**
 - Aggregate liquidity, liquidity loss, and impact of transfer accounts
- **Some network indicators at node level (!) are significant**
 - Their explanatory contribution is low

Thanks a lot for your attention!

Presentation is based on:

Stefan W. Schmitz & Claus Puhr

“Structure and stability in payment networks – a panel data analysis of ARTIS simulations”
in

Harry Leinonen (ed.)

“Simulation analyses and stress testing of payment networks”
Bank of Finland, Scientific Monographs E42, 2009

Also available at SSRN:

http://papers.ssrn.com/sol3/papers.cfm?abstract_id=1400883

Annex

Network Indicators

Panel Data Analysis:

Descriptive Statistics

Assumptions & Tests

Results

Definition of network indicators (1/3)

The *average degree* k of the network is calculated by summing across all (active) links originating from each node (out-degree k_i^{out}) or terminating at each node (in-degree k_i^{in}) and then averaging across nodes:

$$k = \frac{1}{n} \sum_i k_i^{out} = \frac{1}{n} \sum_i k_i^{in} = \frac{m}{n}$$

We calculated the *average path length* for each (active) originating node i by averaging across terminating nodes j and then averaged across originating nodes i to derive the average path length ℓ of the entire network.

$$\ell_i = \frac{1}{n-1} \sum_{j \neq i} d_{ij}$$

$$\ell = \frac{1}{n} \sum_i \ell_i$$

Considering the maximum eccentricity \mathcal{E} (the maximum path length between any originating and any terminating node) across nodes defines the *diameter* D :

$$D = \max_i \mathcal{E}_i$$

The *connectivity* of the network is defined by the number of actual directed links m over the number of possible directed links $n(n-1)$.

Definition of network indicators (2/3)

An indicator of the *distance* d_{ij} between nodes is the lowest possible number of links that connects each (active) node i with each other (active) node j in the network. It is referred to as shortest path length.

The *betweenness centrality* $C_B(h)$ of node h provides a measure of how many shortest paths d_{ij} pass through this node. Let $s_{ij}(h)$ be the number of shortest paths between all pairs of nodes i and j that pass through the node h and let s_{ij} the number of all shortest paths between all pairs of nodes i and j then

$$C_B(h) = \sum_{s \neq i \neq j} \frac{s_{ij}(h)}{s_{ij}}$$

$C_B(h)$ is sometimes normalised by dividing it by the number of pairs of nodes not including the node h . The betweenness centrality of the network is

$$C_B = \frac{1}{n} C_B(h)$$

Definition of network indicators (3/3)

The *dissimilarity index* of two neighbours nodes i and j in a network is defined as

$$\Delta_{ij} = \frac{\sqrt{\sum_{h \neq i, j}^N [d_{ih} - d_{jh}]^2}}{(N-2)},$$

where d_{ik} are distance measures from nodes i and j to node h . It provides a comparison of the viewpoints of the entire network from the perspective of the all pairs of neighbouring nodes. For the entire network the dissimilarity index is

$$\Delta = \frac{1}{n(n-1)/2} \Delta_{ij}$$

The *clustering coefficient* $C_C(h)$ of an individual node h with k_h neighbours measures how well the latter are connected among each other. The number of potential links between the k_h neighbours is $k_h(k_h - 1)/2$. Let the actual number of nodes between them be E_h so that

$$C_C = \frac{E_h}{k_h(k_h - 1)/2}$$

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Descriptive statistics – dependent variables

Variable	Mean	Std. Dev.	Min	Max	Observations	
simnum~s	overall	2.607678	3.765892	0	33	N = 31311
	between		3.534751	.4507042	18.85714	n = 63
	within		1.373103	-6.301779	16.75053	T = 497
simque~m	overall	7.554757	21.71519	0	1172	N = 31311
	between		14.0254	.4507042	76.49899	n = 63
	within		16.67194	-55.94424	1145.718	T = 497
simque~l	overall	1.12e+08	3.35e+08	0	1.07e+10	N = 31311
	between		2.84e+08	3680408	2.08e+09	n = 63
	within		1.81e+08	-1.54e+09	9.79e+09	T = 497

Descriptive statistics – aggregate liquidity

Variable	Mean	Std. Dev.	Min	Max	Observations	
Liqbod-e	overall	7.47e+09	8.43e+08	5.49e+09	1.13e+10	N = 31311
	between		0	7.47e+09	7.47e+09	n = 63
	within		8.43e+08	5.49e+09	1.13e+10	T = 497
Liqcol-l	overall	1.08e+10	2.89e+09	6.10e+09	2.53e+10	N = 31311
	between		0	1.08e+10	1.08e+10	n = 63
	within		2.89e+09	6.10e+09	2.53e+10	T = 497
Liquid-y	overall	1.83e+10	3.23e+09	1.17e+10	3.24e+10	N = 31311
	between		0	1.83e+10	1.83e+10	n = 63
	within		3.23e+09	1.17e+10	3.24e+10	T = 497

Descriptive statistics – liquidity loss

Variable	Mean	Std. Dev.	Min	Max	Observations
Liquid-s overall	1.39e+09	3.07e+09	12032	2.86e+10	N = 31310
between	2.92e+09	2.92e+09	7869281	1.57e+10	n = 63
within	1.02e+09	1.02e+09	-7.11e+09	1.96e+10	T-bar = 496.984
simliq-n overall	7.23e+08	1.69e+09	0	1.60e+10	N = 31310
between	1.59e+09	1.59e+09	452661.6	8.80e+09	n = 63
within	6.16e+08	6.16e+08	-3.91e+09	1.00e+10	T-bar = 496.984
simliq-k overall	6.67e+08	1.53e+09	524	1.29e+10	N = 31310
between	1.45e+09	1.45e+09	1026011	6.92e+09	n = 63
within	5.02e+08	5.02e+08	-3.32e+09	9.59e+09	T-bar = 496.984

Descriptive statistics – network indicators node level

Variable	Mean	Std. Dev.	Min	Max	Observations
nodede~e overall between within	25.44499	19.89985	2	105	N = 31311
		19.83007	6.428571	89.84708	n = 63
		3.000303	10.34036	43.26994	T = 497
nodeco~y overall between within	.1930355	.1510905	.0153	.7949	N = 31311
		.1503863	.0490459	.6821545	n = 63
		.0238868	.075481	.3372814	T = 497
nodeav~h overall between within	1.86558	.1817904	1.2137	2.3356	N = 31311
		.1782929	1.328216	2.102365	n = 63
		.0419877	1.67934	2.169161	T = 497
nodecl~x overall between within	.5401702	.2014584	.1333	1	N = 31311
		.1873986	.1753167	.9800881	n = 63
		.0776117	.1056149	1.013537	T = 497
nodebe~y overall between within	.013652	.0349052	0	.2761	N = 31311
		.0340873	6.04e-07	.1833795	n = 63
		.0086509	-.0796849	.1063725	T = 497
nodedi~x overall between within	.4387912	.1369233	.2603	1.0754	N = 31311
		.13323	.312793	.9046616	n = 63
		.0357625	.281154	.626754	T = 497

Descriptive statistics – network indicators network level

Variable	Mean	Std. Dev.	Min	Max	Observations
netavg-e overall	12.36032	.3990639	11.2609	14.35	N = 31311
between		0	12.36032	12.36032	n = 63
within		.3990639	11.2609	14.35	T = 497
netcon-y overall	.0375596	.0027451	.0294	.0462	N = 31311
between		0	.0375596	.0375596	n = 63
within		.0027451	.0294	.0462	T = 497
netavg-h overall	2.546561	.0594155	2.4025	2.6833	N = 31311
between		0	2.546561	2.546561	n = 63
within		.0594155	2.4025	2.6833	T = 497
netavgc~ overall	.4382753	.0279867	.3612	.5217	N = 31311
between		0	.4382753	.4382753	n = 63
within		.0279867	.3612	.5217	T = 497
netavg-y overall	.0047867	.0002584	.0038	.0055	N = 31311
between		0	.0047867	.0047867	n = 63
within		.0002584	.0038	.0055	T = 497
netavg.. overall	1.269688	.9284118	.5946	5.228	N = 31311
between		0	1.269688	1.269688	n = 63
within		.9284118	.5946	5.228	T = 497

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Panel approach – the assumptions

- **Cross-panel conditional homoskedasticity**
 - **Variance of error terms constant across panels (and across time)**
- **Serial independence**
 - **Error terms are serially uncorrelated within panels**
- **Cross-panel independence**
 - **Error terms are independent across panels**
- **Strict exogeneity**
 - **Error terms and explanatory variables are independent**

Assumption of conditional homoscedasticity across panels

- Likelihood ratio test

- Compares log-likelihoods under restricted and unrestricted model based on iterated generalised least square estimates.

Model 1	LR chi2 (62) = 18501.32	Prob. = 0.00
Model 2	LR chi2 (62) = 103014.77	Prob. = 0.00
Model 3	LR chi2 (62) = 74980.00	Prob. = 0.00

• **Assumption of conditional homoscedasticity rejected!**

Assumption of serial independence

- Wooldridge test

- Based on residuals of regressions in first differences which are then regressed on lagged value t-1

- Test is robust to conditional heteroskedasticity

Model 1	F (1, 62) = 14.388	Prob. = 0.00
Model 2	F (1, 62) = 3.076	Prob. = 0.08
Model 3	F (1, 62) = 23.636	Prob. = 0.00

- Assumption of serial independence rejected!

Assumption of cross-panel independence

- Pesaran, Friedman, Frees tests

Model 1	Frees = 11.116	Prob. = 0.00
	Pesaran = 363.108	Prob. = 0.00
	Friedman = 11728.05	Prob. = 0.00
Model 2	Frees = 7.06	Prob. = 0.00
	Pesaran = 147.08	Prob. = 0.00
	Friedman = 7378.70	Prob. = 0.00
Model 3	Frees = 4.81	Prob. = 0.00
	Pesaran = 120.80	Prob. = 0.00
	Friedman = 5744.16	Prob. = 0.00

- Assumption of cross-sectional independence rejected!

Random- versus fixed-effects

- High correlation btw individual level effects and explanatory variables
- Breusch-Pagan LR test of random effects
 - Compares log-likelihoods under restricted and unrestricted model based on iterated generalised least square estimates

Model 1	LR test = 3.06E05	Prob. = 0.00
Model 2	LR test = 4.33E04	Prob. = 0.00
Model 3	LR test = 2.30E05	Prob. = 0.00

• Random-effects rejected!

Assumption of strict exogeneity

- **Fundamental assumption**
 - **Error terms are not influenced by past, current or future values of explanatory variables**
 - **Values of explanatory variables are not influenced by past, current or future values of error terms**
- **Simulation design ensures strict exogeneity**
 - **Values of explanatory variables are empirical observations**
 - **Error terms cannot influence values of explanatory variables**
 - **E.g. banks cannot adjust liquidity holdings, node or network characteristics in response to observed error terms**

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# contagious bank defaults	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Constant	3.28	4.02	3.38	2.32	14.73	6.97	4.63	-0.93
Liquidity	-1.05E-10 <i>-8.19</i>	-9.69E-11 <i>-8.05</i>	-9.91E-11 <i>-9.16</i>	-8.39E-10 <i>-7.87</i>	-9.31E-11 <i>-7.56</i>	-9.40E-11 <i>-16.00</i>	-9.70E-11 <i>-7.67</i>	-1.18E-10 <i>-9.65</i>
SimUnrVol	1.65E-09 <i>59.21</i>	1.66E-09 <i>58.92</i>	7.77E-10 <i>28.64</i>	8.06E-10 <i>30.80</i>	1.01E-09 <i>38.61</i>	1.43E-09 <i>55.18</i>	1.07E-09 <i>0.00</i>	9.37E-10 <i>33.17</i>
Transfer*SimUnrVol	2.51E-10 <i>8.99</i>	2.52E-10 <i>8.93</i>	6.12E-10 <i>23.28</i>	6.00E-10 <i>23.92</i>	5.27E-10 <i>20.57</i>	2.58E-10 <i>9.66</i>	5.44E-10 <i>19.29</i>	5.06E-10 <i>19.42</i>
Nodedegree			9.20E-02 <i>59.92</i>					
Nodeconnectivity				1.22E+01 <i>59.88</i>				
Nodeavgpath					-8.09E+00 <i>-56.71</i>			
Nodeclusterindex						-4.00E+00 <i>-51.65</i>		
Nodebetweenness							3.38E+01 <i>26.37</i>	
Node dissimilarity								1.11E+01 <i>40.07</i>
Netvolume		-1.88E-11 <i>-7.76</i>						
Netavgdegree			-1.60E-01 <i>-2.24</i>					
Netconnectivity				-3.25E+01 <i>-3.25</i>				
Netavgpath					1.51E+00 <i>2.52</i>			
Netavgclusterindex						-3.60E+00 <i>-3.42</i>		
Netavgbetweenness							-3.42E+02 <i>-3.00</i>	
Netavgdissimilarity								3.00E-02 <i>0.82</i>
R ²	69.23	69.35	69.96	72.41	71.58	70.88	66.54	68.69
Relative impact of Transfer Account (in %)	15%	15%	79%	74%	52%	18%	51%	54%

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# contagious unsettled payments	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Constant	7.32 <i>7.95</i>	9.90 <i>10.67</i>	-3.06 <i>-0.93</i>	0.57 <i>0.34</i>	35.54 <i>6.38</i>	12.55 <i>6.93</i>	4.73 <i>2.13</i>	-0.36 <i>-0.35</i>
Liquidity	-2.79E-10 <i>-5.65</i>	-2.49E+00 <i>-5.44</i>	-2.70E-10 <i>-6.18</i>	-1.97E-10 <i>-4.57</i>	-2.25E-10 <i>-4.75</i>	-2.52E-10 <i>-5.43</i>	-2.59E-10 <i>-5.29</i>	-2.91E-10 <i>-6.24</i>
SimUnrVol	6.61E-09 <i>41.91</i>	6.67E-09 <i>41.84</i>	4.97E-09 <i>24.31</i>	4.94E-09 <i>24.17</i>	5.32E-09 <i>28.22</i>	6.29E-09 <i>38.61</i>	5.59E-09 <i>22.59</i>	5.41E-09 <i>26.62</i>
Transfer*SimUnrVol	2.40E-09 <i>8.36</i>	2.38E-09 <i>8.29</i>	3.10E-09 <i>10.53</i>	3.10E-09 <i>10.55</i>	2.95E-09 <i>10.08</i>	2.45E-09 <i>8.49</i>	2.91E-09 <i>9.52</i>	2.84E-09 <i>9.72</i>
Nodedegree			1.80E-01 <i>13.51</i>					
Nodeconnectivity				2.46E+01 <i>13.85</i>				
Nodeavgpath					-1.71E+01 <i>-13.92</i>			
Nodeclusterindex						-6.25E+00 <i>-10.84</i>		
Nodebetweenness							5.40E+01 <i>5.65</i>	
Node dissimilarity								1.90E+01 <i>10.02</i>
Netvolume		-6.67E-11 <i>-8.15</i>						
Netavgdegree			5.10E-01 1.93					
Netconnectivity				3.09E+01 0.86				
Netavgpath					1.21E+00 0.56			
Netavgclusterindex						-5.29E+00 -1.40		
Netavgbetweenness							4.27E+02 1.09	
Netavgdissimilarity								5.00E-02 0.52
R ²	39.79	39.81	39.90	40.01	40.02	39.77	39.72	39.67
Relative impact of Transfer Account (in %)	36%	36%	62%	63%	55%	39%	52%	52%

Value of contagious unsettled payments	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Constant	5.22E+07 <i>5.30</i>	7.37E+07 <i>7.16</i>	2.47E+07 <i>0.59</i>	3.63E+07 <i>1.70</i>	2.87E+08 <i>4.08</i>	5.94E+07 <i>2.63</i>	4.24E+07 <i>1.44</i>	5.35E+07 <i>4.59</i>
Liquidity	-2.70E-03 <i>-5.01</i>	-2.19E-03 <i>-4.29</i>	-2.70E-03 <i>-5.08</i>	-2.50E-03 <i>-4.72</i>	-2.10E-03 <i>-3.57</i>	-2.68E-03 <i>-4.77</i>	-2.62E-03 <i>-4.80</i>	-2.72E-03 <i>-4.97</i>
SimUnrVol	9.56E-02 <i>47.60</i>	9.60E-02 <i>47.77</i>	9.22E-02 <i>35.69</i>	9.04E-02 <i>35.17</i>	9.09E-02 <i>38.17</i>	9.72E-02 <i>45.99</i>	9.65E-02 <i>33.87</i>	9.56E-02 <i>36.91</i>
Transfer*SimUnrVol	1.59E-01 <i>34.54</i>	1.59E-01 <i>34.48</i>	1.60E-01 <i>34.24</i>	1.61E-01 <i>34.39</i>	1.61E-01 <i>34.55</i>	1.58E-01 <i>34.60</i>	1.58E-01 <i>33.08</i>	1.59E-01 <i>34.04</i>
Nodedegree			3.33E+05 <i>2.34</i>					
Nodeconnectivity				6.85E+07 <i>3.63</i>				
Nodeavgpath					-5.76E+07 <i>-4.42</i>			
Nodeclusterindex						2.78E+07 <i>3.66</i>		
Nodebetweenness							-4.65E+07 <i>-0.53</i>	
Nodedissimilarity								-2.52E+06 <i>-0.12</i>
Netvolume		6.20E-04 <i>-5.49</i>						
Netavgdegree			1.76E+06 <i>0.52</i>					
Netconnectivity				7.28E+07 <i>0.16</i>				
Netavgpath					-5.32E+07 <i>-1.92</i>			
Netavgclusterindex						-5.39E+07 <i>-1.11</i>		
Netavgbetweenness							1.93E+07 <i>0.36</i>	
Netavgdissimilarity								5.25E+05 <i>0.41</i>
R ²	70.62	70.63	70.64	70.65	70.69	70.68	70.60	70.61
Relative impact of Transfer Account (in %)	166%	166%	174%	178%	177%	163%	164%	166%