

An agent-based model for bank formation, bank runs and interbank

networks

Matheus R.

Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

# An agent-based model for bank formation, bank runs and interbank networks

Matheus R. Grasselli

Mathematics and Statistics - McMaster University Joint work with Omneia Ismail (McMaster)

UCSB, June 2, 2011



An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

Matheus R. Grasselli

Introduction

Theoretical

underpinnings

Bank formation

Bank runs

Interbank networks Introduction

2 Theoretical underpinnings

Bank formation

4 Bank runs



#### The quest to understand banking crises

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

#### Introductio

Theoretical underpinnings

Bank formation

Bank runs

- Financial crises in the past 800 years encompass:
  - sovereign defaults
  - currency debasement and inflation
  - exchange rate crises
  - banking crises
- Graduating from banking crises has eluded developed and developing countries alike Reinhart and Rogoff (2009).
- Individual banks are subject to runs, largely addressed by deposit insurance, capital requirements, and regulation.
- However, the principles that govern individual prudence do not necessarily apply to systems as a whole.
- Financial innovation and integration leads to highly interconnected, complex and potentially fragile banking systems.
- Systemic crises are essentially stories of contagion, interdependence, interaction and trust - Kirman (2010).



#### Agent-Based Models in Economics

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

#### Introductio

Theoretical underpinnings

Bank formation

Bank runs

- Modern macroeconomic theory (e.g 'sophisticated' DSGE models) is hopeless to deal with banking crises.
- Representative agents, neutrality of money, stationarity of expectations, and assumed equilibrium states are non-starters for the problem at hand.
- Agent-based computational economics (ACE) has emerged as an alternative
- Agents have rational objectives, but realistic computational devices (inductive learning, bounded memory, limited information, war games, etc).
- Interactions are modelled directly, without fictitious clearing mechanisms.
- Hierarchical structures (i.e, banks are agents, but so are their clients, as well as the government).
- Equilibrium is just one possible outcome, not assumed a priori.



### Network paradigm

An agent-based model for bank formation, bank runs and interbank

networks

Matheus R.
Grasselli

#### Introductio

Theoretical underpinnings

Bank formation

Bank runs

- Focus on the relationships between different entities as well as the entities themselves
- Well suited to study systems where complexity arises from both the interactions among units and the anatomy of the system.
- Provide unifying principles for ecosystems, power transmission, infectious diseases, etc.
- In the context of banking, networks can help explain:
  - the effect of network structure on system stability
  - the dynamic evolution of interbank links in order to reduce exposure to risk
- The bulk of recent work on systemic risk focuses on the first aspect.



## Banking networks

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

#### Introductio

Theoretical underpinnings

Bank formation

Bank runs

- Financial institutions are connected through:
  - direct links in the interbank market
  - indirect links through similar portfolio exposure
- Shocks come from assets or liabilities.
- For example, Allen and Gale (2000) investigate links of the first type and conclude that fully connected networks are robust to liquidity (liability) shocks.
- Alternatively, Cifuentes, Ferrucci and Shin (2005) consider exposure to common assets under market-to-market and minimal capital requirements and reach different conclusions.
- Unifying the effects of both types of links and shocks is still largely open.
- Most studies define failure as default and loss of capital.
- Systemic failure should also include cases where the network does not provide its social and economic function.

#### Liquidity preferences

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

- An asset is illiquid if its liquidation value at an earlier time is less than the present value of its future payoff.
- For example, an asset can pay  $1 \le r_1 \le r_2$  at dates T = 0, 1, 2.
- The lower the ratio  $r_1/r_2$  the less liquid is the asset.
- At time t = 0, consumers don't know in which future date they will consume.
- The consumer's expected utility is

$$wU(r_1) + (1 - w)U(r_2),$$

where w is the proportion of early consumers (type 1).

Sufficiently risk-averse consumers prefer the liquid asset.

## Example: Diamond (2007)

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

- Let  $A = (r_1 = 1, r_2 = 2)$  represent an illiquid asset and  $B = (r_1 = 1.28, r_2 = 1.813)$  a liquid one.
- Assume investors with power utility  $u(c) = 1 c^{-1}$  and w = 1/4.
- The expected utility from holding the illiquid asset is

$$E[u(c)] = \frac{1}{4}u(1) + \frac{3}{4}u(2) = 0.375$$

 By comparison, the expected utility from holding the liquid asset is

$$E[u(c)] = \frac{1}{4}u(1.28) + \frac{3}{4}u(1.813) = 0.391$$

 Observe, however, that risk-neutral investors would prefer the illiquid asset, since:

$$E[A] = 1.75 > 1.68 = E[B]$$

### Liquidity risk sharing

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks • Consider an economy with dates T=0,1,2 and an illiquid asset A=(1,R) and consumer preferences given by

$$U(c_1^j, c_2^j, \omega) = \begin{cases} u^j(c_1) & \text{if j is of type 1 in state } \omega \\ u^j(c_2) & \text{if j is of type 2 in state } \omega \end{cases}$$
 (1)

• Denoting by w the fraction of early consumers (type 1), the optimal risk sharing for *publicly* observed preferences is

$$u'(c_1^{1*}) = Ru'(c_2^{2*})$$
 (2)

$$(1-w)c_2^{2*} = (1-wc_1^{1*})R$$
 (3)

 However, liquidity preferences are private unverifiable information!



### A model for banks - Diamond and Dybvig (1983)

An agent-based model for bank formation.

bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

- Suppose now that a bank offers a fixed claim  $r_1$  per unit deposited at time 0.
- Assume that withdrawers are served sequentially in random order until bank runs out of assets.
- Denoting by  $f_j$  the fraction of withdrawers before j and by f their total fraction, the payoffs per unit deposited are

$$V_1(f_j, r_1) = r_1 \mathbf{1}_{\{f_j < r_1^{-1}\}}$$
  
$$V_2(f, r_1) = [R(1 - r_1 f)/(1 - f)]^+$$

- Setting  $r_1 = c_1^{1*}$ , a good equilibrium corresponds to f = w, since this leads to  $V_2 = c_2^{2*} > c_1^{1*} = V_1$ .
- However, it is clear that f=1 (run) is also an equilibrium leading to  $V_1 \le c_1^{1*}$  and  $V_2=0 < c_2^{2*}$ .

# Example revisited: Diamond (2007)

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

- Let the illiquid asset be A=(1,2),  $u(c)=1-c^{-1}$  and w=1/4
- ullet Then the marginal utility condition becomes  $c_2^{2*}=\sqrt{R}c_1^{1*}$ .
- Substituting into the budget constraint (3) gives

$$c_1^{1*} = \frac{\sqrt{R}}{1 - w + w\sqrt{R}} = 1.28, \qquad c_2^{2*} = 1.813.$$

- Suppose the bank offers the liquid asset B = (1.28, 1.813) to 100 depositors each with \$1 at 0 and invests in A.
- If f = 1/4, the bank needs to pay  $25 \times 1.28 = 32$  at t = 1.
- At t=2 the remaining depositors receive  $\frac{68\times2}{75}=1.813$ .
- Therefore a forecast  $\hat{f} = 1/4$  is a Nash equilibrium.
- However, the forecast  $\hat{f} = 1$  is another Nash equilibrium.



# A model for interbank loans - Allen and Gale (2000)

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

- Consider an economy with 4 banks (regions) A, B, C, D.
- There is a continuum of agents with unit endowment at time 0 and liquidity preferences given according to (1).
- The probability w of being an early consumer varies from one region to another conditional on two states  $S_1$  and  $S_2$  with equal probabilities:

Table: Regional Liquidity Shocks

$$S_1 = \begin{array}{ccccc} A & B & C & D \\ \hline w_H & w_L & w_H & w_L \\ S_2 & w_L & w_H & w_L & w_H \end{array}$$

• Each bank can invest in a liquid asset (1,1) and an illiquid asset (r < 1, R > 1) and promises consumption  $(c_1, c_2)$ .



#### The central planner solution

An agent-based model for bank formation, bank runs and

interbank

networks

Matheus R.
Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

- The **central planner solution** consists of the best allocation (x, y) of per capita amounts invested in the illiquid and liquid assets maximizing the consumer's expected utility.
- This is easily seen to be given by

$$\gamma c_1 = y, \quad (1 - \gamma)c_2 = Rx,$$

where  $\gamma = \frac{w_H + w_L}{2}$  is the fraction of early consumers.

- Once liquidity is revealed, the central planner moves resources around.
- For example, in state  $S_1$ , A and C have excess demand  $(w_H \gamma)c_1$  at t = 1, which equals the excess supply  $(\gamma w_I)c_1$  from B and D.
- At t=2 the flow is reversed, since the excess supply  $(w_H \gamma)c_2$  from A and C equals the excess demand  $(\gamma w_I)c_2$  from B and D.

### Optimal interbank loans

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

- In the absence of a central planner, interbank loans can overcome the maldistribution of liquidity.
- Suppose that the network is completely connected (i.e links between all banks).
- To achieve the optimal allocation, it is enough for banks to exchange deposits  $z_i = (w_H \gamma)/2$  at time t = 0.
- ullet At t=1, a bank with high liquidity demand satisfies

$$\left[w_H + \frac{w_H - \gamma}{2}\right]c_1 = y + \frac{3(w_H - \gamma)c_1}{2},$$

which reduces to  $\gamma c_1 = y$ .

• At t = 2, the same bank satisfies

$$[(1 - w_H) + (w_H - \gamma)]c_2 = Rx,$$

which reduces to  $(1 - \gamma)c_2 = Rx$ .



## Shocks and stability

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

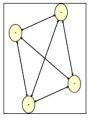
Theoretical underpinnings

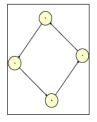
Bank formation

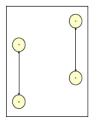
Bank runs

Interbank

 Allen and Gale (2000) then analyze the effects of small shocks to interbank markets with networks of the form:







- They show that the complete network absorbs shocks better than the incomplete one.
- Their analytic model is difficult to generalize to arbitrary (asymmetric) networks.



#### Our model - the summarized story

An agent-based model for bank formation, bank runs and interbank

networks Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

- Society
- Liquidity Preference
- Searching for partners
- Learning and Predicting
- Bank birth
- Interbank Links
- Contagion

# Society

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

- We have a society of individuals investing at the beginning of each period (t = 0).
- For each individual i, an initial preference is drawn from a continuous uniform random variable  $U_i$
- If  $U_i < 0.5$  the agent is deemed to be liquid asset investor (short-term, early consumer), otherwise the agent is an illiquid asset investor (long-term, late consumer).
- There is a mid-period (t = 1) shock to their preferences:

$$\widetilde{U}_i = U_i + (-1)^{ran_i} \frac{\epsilon_i}{2}$$

- If  $\widetilde{U}_i < 0.5$  the investor wants to be a short term investor, otherwise he wants to be long term investor.
- If the shock is big enough the individual wishes to have invested differently.
- Because of anticipated shocks, individuals explore the society searching to partners to exchange investments.



#### Searching for partners

An agent-based model for bank formation, bank runs and interbank

networks

Matheus R.
Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

- We impose some constrains on the individual capacity to go around and seek other individuals to trade.
- This reflects the inherited limited capability of information gathering and environment knowledge of individual agents.
- We use a combination of Von Neumann and Moore neighborhood:



#### Inductive reasoning

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

- We follow the inductive reasoning proposed by Arthur (2000) for individuals with bounded rationality dealing with complex environments.
- We assume agents make predictions using a memory of 5 periods.
- All agents have a set of 7 predictors as follows:
  - Today would be the same as last period.
  - 2 Today would be the same as two periods ago.
    - Today would be the same as three periods ago.
  - Today would be the same as four periods ago.
  - **5** Today would be the same as five periods ago.
  - **1** Today would be the same as the mode for the last three periods.
  - Today would be the same as the mode for the last five periods.

### Learning and Predicting

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

- Each predictor makes one of the following forecasts:
  - $\mathbf{0}$  N = agent will not need a partner
  - ② G = agent will need a partner and will find one
  - 3 B = agent will need a partner and will not find one
- Depending on the realized outcome, a predictor's strength gets updated by

$$\Delta S = \left\{ \begin{array}{ll} +1 & \text{if the forecast is correct} \\ -1 & \text{if the forecast is incorrect} \end{array} \right.$$



#### Learning simulation

An
agent-based
model for
bank
formation,
bank runs and
interbank

networks

Matheus R.
Grasselli

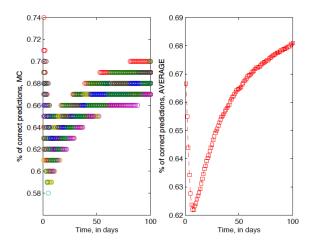
Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks We use 400 persons over a time span of 100 periods in a simulation with 100 realizations:





#### Bank birth

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

- We follow the work of Howitt and Clower (1999, 2007) on the emergence of economic organizations.
- A randomly selected agent i is hit by the 'idea of entrepreneurship' and makes an initial estimate  $W^i = Z^i/8$  of the fraction of early consumers, where  $Z^i$  is a random integer in [0,8] and reflects the entrepreneur's 'animal spirits'.
  - The bank is establish if there are x and y such that  $x + y \le 1$  and

$$y = c_1 W_i$$
  
 
$$Rx = c_2 (1 - W_i),$$

where  $(c_1, c_2)$  is the promised consumption.

- Individuals become aware of bank existence only if the bank lies in their neighbourhood
- In addition we give the bank the reach of its new members



#### To join or not to join a bank

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

formation

Bank runs

Interbank networks

- Agents need to decide between trading directly either in the liquid asset (1,1) or the illiquid asset (r < 1, R > 1) or joining the bank and receiving  $(c_1 > 1, c_2 < R)$ .
- For example, an agent who current has late preferences might have the following payoff table:

	forecast	strength	payoff (join)	payoff (not join)
1	N	-2	<i>c</i> <sub>2</sub>	R
2	G	0	$c_1$	1
3	Ν	+1	<i>c</i> <sub>2</sub>	R
4	В	-1	$c_1$	r
5	G	+1	$c_1$	1
6	Ν	0	$c_2$	R
7	В	+2	$c_1$	r

• The decision is based on the weighted sum of payoffs.



#### Experiment: bank formation

An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs



#### Experiment (continued): established banks

An agent-based model for bank formation, bank runs and interbank

networks

Matheus R.
Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

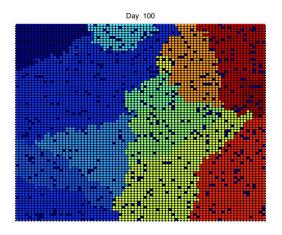


Figure: Banks at T=100 with  $c_1 = 1.1$ ,  $c_2 = 1.5$  and R = 2



#### Experiment (continued): number of depositors

An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

Matheus R. Grasselli

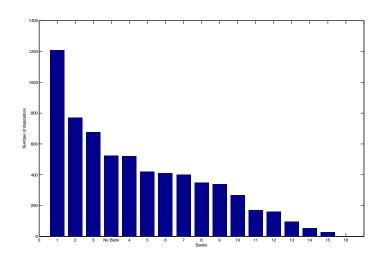
Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank



## Dynamic allocation

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

- In the previous section we assumed that an agent never leaves a bank after joining.
- To model bank failures and runs we need a learning mechanism for banks themselves.
- Having made the allocation  $(x_t^i, y_t^i)$  based on  $W_t^i$ , banks accumulates reserves according to the realized  $\overline{W}_t^i$ :

$$C_t^i = [y_t^i - c_1 \overline{W}_t^i] + [Rx_t^i - c_2(1 - \overline{W}_t^i)].$$

• Banks update their estimate of early consumers through

$$W_{t+1}^i = \max\left\{W_t^i + \alpha(\overline{W}_t^i - W_t^i), \frac{1 - c_2/R}{c_1 - c_2/R}\right\}, \quad (4)$$

reflecting both adaptation through a parameter  $\alpha \in (0,1)$  and the budget constraint  $x_{t+1}^i + y_{t+1}^i \leq 1$  where

$$y_{t+1}^i = c_1 W_{t+1}^i, \quad Rx_{t+1}^i = c_2 (1 - W_{t+1}^i).$$



#### A run on the bank

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank

- We say that a bank is subject to a run if late consumers receive less than  $c_1$  at the end of the period.
- If the bank underestimates the fraction of early consumers, there is a run provided

$$(\overline{W}_t^i - W_t^i)c_1 > \left\lceil \frac{(1 - W_t^i)c_2}{R} - \frac{(1 - \overline{W}_t^i)c_1}{R} \right\rceil r + C_t^i$$

• Conversely, if the bank overestimates  $\overline{W}_t^i$ , the amount available to late consumers (without using reserves) is

$$egin{split} rac{c_2(1-W_t^i)+c_1(W_t^i-\overline{W}_t^i)}{1-\overline{W}_t^i} &= c_2-(c_2-c_1)rac{W_t^i-\overline{W}_t^i}{1-\overline{W}_t^i}\ &= c_1+(c_2-c_1)rac{1-W_t^i}{1-\overline{W}_t^i}. \end{split}$$

• The banks uses reserves to bring this as close as possible



#### Experiment: bank formation and runs

An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs



# Experiment: established banks (with possible runs)

An agent-based model for bank formation, bank runs and interbank

networks

Matheus R.
Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

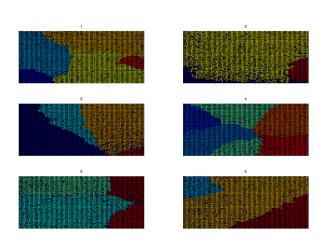


Figure: Banks at T=100 with  $c_1 = 1.1$ ,  $c_2 = 1.5$  and R = 2



#### Banks and learning

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

- As before, banks update their estimate of the fraction of early consumers according to (4).
- In addition, they deem the estimate to be adequate if the fraction of reserves lost in a given period is less than a certain threshold.
- They use the same set of predictors as clients to forecast the adequacy of their estimates as being 'adequate', 'inadequate' or 'undetermined'.
- Banks with inadequate or undetermined estimates have an incentive to exchange deposits with other banks and try to protect their reserves.



#### Experiment: adequacy of estimates through time

An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

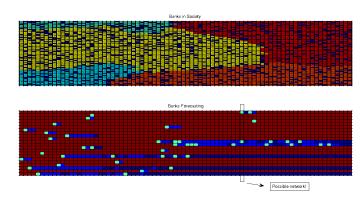


Figure: Banks at T=100 with  $c_1 = 1.1$ ,  $c_2 = 1.5$  and R = 2 and adequacy of estimates over time.



#### Experiment: possible network

An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

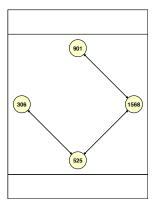


Figure: Snapshot of possible interbank loans



#### Correlated liquidity shocks

An
agent-based
model for
bank
formation,

bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank

- As in Allen and Gale (2000), we consider regional liquidity shocks in a society with no overall shortage of liquidity.
- We form 2C different regions (communities) as follows:
  - 1 Select 2C cells at random to be the base
  - Choose the largest reach M around the base
  - Randomly select 2M<sup>2</sup> cells around the base to form a community
  - Alter half of the communities to early preferences (i.e  $\widetilde{U}_i = 0.2$ ) and half of the communities to late preferences (i.e  $\widetilde{U}_i = 0.8$ ).



#### Examples of correlated liquidity shocks

An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

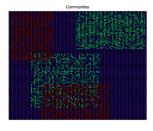
Matheus R. Grasselli

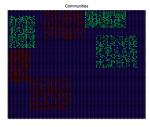
Introduction

Theoretical underpinnings

Bank formation

Bank runs







# Experiment: bank formation and runs with correlated shocks

An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs



# Experiment: adequacy of estimates through time (with correlated shocks)

An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

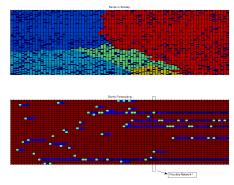


Figure: Banks at T=100 with  $c_1 = 1.1$ ,  $c_2 = 1.5$  and R = 2 and adequacy of estimates over time.



#### Experiment: another possible network

An
agent-based
model for
bank
formation,
bank runs and
interbank
networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

Interbank networks

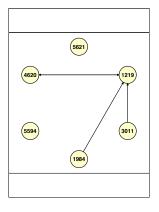


Figure: Snapshot of possible interbank loans with correlated liquidity shocks



#### Concluding remarks

An agent-based model for bank

formation, bank runs and interbank networks

Matheus R. Grasselli

Introduction

Theoretical underpinnings

Bank formation

Bank runs

- We modelled individual liquidity preferences in a society.
- Changes in preferences lead agents to search for trading partners.
- Banks arise as providers of liquidity, but are inevitably subject to possible runs.
- Interbank loans redistributed the effect of correlated liquidity shocks across the society.
- Robustness of the model is being tested through extensive simulations.
- Ultimately want to adjust model parameters to reproduced different observed networks and use it as a testbed for policy implications.
- Thank you.