Conclusion

### Strategic complementarity in banks' funding liquidity choices and financial stability

André Silva

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- Insufficient bank liquidity buffers were one of the main causes of the financial crisis (Brunnermeier, JEP 2009).
- ► Funding liquidity risk is inherently systemic one agent's liquid asset is another agent's liquid liability → funding arrangements link banks with other financial institutions and the non-financial sector.
- ► Liquidity requirements in most regulatory initiates (e.g., Basel III LCR/NSFR) are idiosyncratic in nature → abstract from any formal or informal interconnections between banks.
- Competitors matter for bank liquidity (Bonfim and Kim, 2014), bank credit (Uchida and Nakagawa, JFI 2007), capital structure (Leary and Roberts, JF 2014), compensation (Shue, RFS 2013), investment policies (Dougal et al., JF 2015).

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- 1. Why and how are liquidity holding choices of a bank influenced by the behaviour of its peers?
  - Why? Learning i.e., free-riding in information acquisition (Banerjee, QJE 1992)? Or collective moral-hazard arising from LOLR bailout commitment (Ratnovski, JFI 2009; Farhi and Tirole, AER 2012)?
    - How? Through direct responses to peers' liquidity decisions? Or through changes in other peers' characteristics?

- Do strategic funding liquidity risk management decisions have an impact on financial stability?
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  - ► Restrict coverage to largest 100 commercial banks in each country i.e., exclude smaller (mostly regional) banks in the US and Japan.
- ▶ Bank ownership data → manually collected from various sources:
  - BvD ownership database, banks' annual reports and websites, newspaper articles. Data is further cross-checked with the Claessens and van Horen (JMCB 2014) bank ownership database.
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▶ Endogeneity problem: if peers liquidity choices affect the liquidity decisions of a specific bank, the decision of this bank may also in turn affect the choice made by the peers (Manski, RES 1993).

- Solution: explore systematic differences in peer group composition to identify peer effects (Bramoullé et al., JE 2009) → heterogeneity allows to use liquidity holdings of the "peer's peer" as an instrument, thus extracting the exogenous part of the variation.
  - Strategy solves reflection problem and causes potential bias from weak instruments to fall away (Angrist, LE 2014).

#### ► How?

- Large cross-border banking groups manage liquidity on a global scale (e.g., Cetorelli and Goldberg, JF 2012).
- Identifying assumption: in addition to liquidity choices of its direct competitors, a foreign-owned subsidiary also takes into account the funding liquidity risk management policies of its parent bank-holding group when determining its own.

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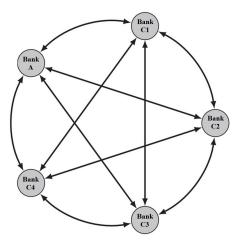
Main Results

### Identification strategy

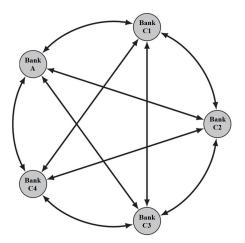
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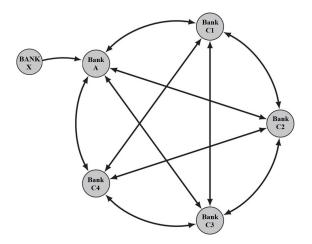
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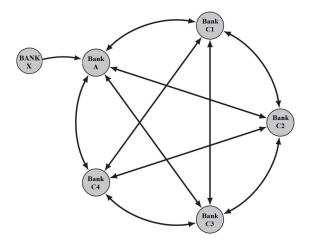
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### Criteria to specify peer groups

#### 1. Country and Year:

- Within-country banks expected to have higher incentives to mimic their peers since they share same LOLR.
- Learning also more likely to occur within countries where information for bank managers is more accessible.
- 2. Business Model: only commercial banks included in the sample
  - ▶ Most cooperative and saving banks are domestically owned
- Bank Size: each peer group in each country j in each year t has a
  maximum of 20 banks in the benchmark case.
  - We need to have at least 1 foreign-owned subsidiary within the 20 banks to identify the remaining 19
  - Bizjak et al. (JFE 2011) → average peer group size when setting executive compensation is 17.3 for S&P 500 firms

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# Empirical Model 2

#### Baseline model to examine impact of peer effects on financial stability

### Step 1:

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- $ightharpoonup eta_{j,t}$  is now allowed to vary across countries and over time.
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#### Step 2.

$$Stability_{i,j,t} = \kappa + \delta \hat{\beta}_{j,t} + \gamma' X_{i,j,t-1} + \nu_{j,t} + u_{i,j,t}$$

► Stability<sub>i,j,t</sub> is a measure of (i) individual banks' financial stability: Z-Score or Merton's Distance-to-Default; or (ii) systemic risk: MES or SRISK (Acharya et al., 2010, 2012).

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## Baseline model to examine impact of peer effects on financial stability

#### Step 1:

$$Liq_{i,j,t} = \omega + \beta_{j,t} \overline{Liq}_{-i,j,t} + \lambda' \overline{X}_{-i,j,t-1} + \gamma' X_{i,j,t-1} + \eta' Z_{j,t-1} + \mu_i + v_t + \varepsilon_{i,j,t}$$

- $\triangleright$   $\beta_{i,t}$  is now allowed to vary across countries and over time.
- ► e.g., UK in 2010:

$$Liq_{i,j,t} = \omega + [\beta_0 + (\beta_1 \times I_{UK} \times I_{2010})] \overline{Liq}_{-i,j,t} + \lambda' \bar{X}_{-i,j,t-1} + \gamma' X_{i,j,t-1} + \eta' Z_{j,t-1} + \mu_i + v_t + \varepsilon_{i,j,t}$$

#### Step 2:

$$Stability_{i,j,t} = \kappa + \delta \hat{\beta}_{j,t} + \gamma' X_{i,j,t-1} + \nu_{j,t} + u_{i,j,t}$$

➤ Stability<sub>i,j,t</sub> is a measure of (i) individual banks' financial stability: Z-Score or Merton's Distance-to-Default; or (ii) systemic risk: MES or SRISK (Acharya et al., 2010, 2012).

Dep Var: Liquidity Creation				
Peer Banks' Liquidity Creation	0.455** (0.222)	0.522*** (0.134)	0.532*** (0.194)	0.462*** (0.157)
Peer Banks' Total Assets	0.004 (0.005)	0.009**	0.004 (0.004)	0.007** (0.003)
Peer Banks' Capital Ratio	0.110	0.123**	0.121**	0.084
Peer Banks' Return-on-Assets	(0.068) 0.093	(0.051) 0.195	(0.062) 0.053	(0.053) -0.035
	(0.374)	(0.291)	(0.373)	(0.278)
Peer Banks' Provisions	-0.009 (0.030)	0.030 (0.026)	0.004 (0.027)	0.043* (0.026)
4.4.4	(0.000)	(0.020)	(0.021)	(0.020)
Bank-level controls	Y	Υ	Y	Υ
Country-level controls	Υ	Υ	=	=
Year FE	Υ	Υ	N	N
Country FE	Υ	-	N	=
Bank FE	N	Υ	N	Υ
Country-Year FE	N	N	Υ	Υ
IV (1st stage)	0.129***	0.160***	0.141***	0.125***
	(0.013)	(0.014)	(0.013)	(0.011)

Dep Var: Liquidity Ratio				
Peer Banks' Liquidity Ratio	0.574*** (0.152)	0 474*** (0 102)	0.596*** (0.159)	0.250** (0.110)
Peer Banks' Total Assets	-0.018 (0.027)	0.011 (0.019)	-0.010 (0.025)	0.018 (0.019)
Peer Banks' Capital Ratio	0.456	-0.181	0.639*	-0.233
Peer Banks' Return-on-Assets	(0.358) 3.841*	(0.249) 0.581	(0.357) 3.722*	(0.251) 1.837
Peer Banks' Provisions	(1.982) -0.046	(1 486) -0 283**	(2.005) -0.069	(1.418) -0.264**
r eer Danks T Tovisions	(0.176)	(0.140)	(0.163)	(0.132)
<u> </u>				
Bank-level controls	Υ	Υ	Υ	Υ
Country-level controls	Υ	Υ	-	-
Year FE	Υ	Y	N	N
Country FE	Υ	-	N	-
Bank FE	N	Υ	N	Υ
Country-Year FE	N	N	Υ	Υ
IV (1st stage)	0.216***	0.202***	0.203***	0.178***
	(0.010)	(0.012)	(0.010)	(0.012)

- ► Peer banks play an important role in determining individual banks' liquidity holding policies:
  - e.g., one standard deviation change in peers' liquidity creation (0.15) is associated with change in liquidity creation of bank i of 0.07-0.08.
- Banks' liquidity decisions are in large part direct responses to the liquidity choices of peer banks and, to a lesser extent, to changes in their characteristics.
- ▶ These peer effects are one of the most important factors for liquidity holding determination → together with bank-specific capital and loans/total assets (untabulated).

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#### 1. Alternative peer group definitions:

- Form peer groups using peer-weighted averages based on size similarity - inverse of Euclidean distance i.e., the smaller the distance between two banks, the more weight it has.
- Split within-country-year banks into small and large banks; small, medium and large banks; or groups of 25 banks by size, . . .

#### 2. Alternative econometric specifications

► Include lagged liquidity ratio or liquidity creation as an explanatory variable and estimate the model with S-GMM, . . .

#### 3 Alternative IVs

- Regress liquidity holdings of parent bank-holding group with country-level characteristics and country and time FE → use the residual to instrument peer firms' liquidity choices.
- Instrument peer firms' liquidity choices with the lagged idiosyncratic component of peers' equity returns (Leary and Roberts, JF 2014).

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# Result 2: Who mimics who?

	Peer Effect:	Liq Creation	Peer Effect	Liq Ratio
$Large\ banks \to Large\ banks$	0.981*** (0.164)	0.773*** (0.179)	0 909** (0 396)	1.185*** (0.327)
$Large\ banks \to Small\ banks$	0.227 (0.300)	0.045 (0.293)	-0.059 (0.212)	0.218 (0.173)
Small banks $ o$ Small banks	1.332*** (0.379)	0.803** (0.373)	0.943*** (0.285)	0 428** (0 209)
Small banks $ ightarrow$ Large banks	0.765*** (0.211)	0.886*** (0.192)	1.155** (0.530)	1.178*** (0.453)
Peer Characteristics	Υ	Υ	Y	Y
Bank-level controls	Υ	Υ	Υ	Υ
Country-level controls	Υ	Υ	Υ	Υ
Year FE	Υ	Υ	Υ	Υ
Country FE	Υ	-	Υ	-
Bank FE	N	Υ	N	Υ

# Result 3.1: Peer effects and default risk

$\ln( extsf{Z-Score})$ – 3-year window: $\ln[(E/A+ROA)/\sigma(ROA)_{3y}]$				
Peer Effect: Liq. Creation - $\widehat{eta_{j,t}^{LC}}$	-0.319** (0.142)			
Peer Effect:			-0.442***	-0.366***
Liq. Ratio - $\widehat{eta_{j,t}^{LR}}$			(0.132)	(0.118)
No observations	10,051	10,051	10,049	10,049
No. banks	1,406	1,406	1,407	1,407
Adj. $R^2$	0.269	0.126	0.269	0.127
Bank-level controls	Υ	Y	Υ	Υ
Country-level controls	Υ	-	Υ	-
Year FE	Υ	N	Υ	N
Bank FE	N	Υ	N	Υ
Country FE	Υ	-	Υ	-
Country-Year FE	N	Y	N	Y

 Conclusions do not change when using a 5-year window to compute Z-Scores, or the market-based Merton Distance-to-Default.

# Result 3.2: Peer effects and systemic risk

	Marginal Ex	pected Shortfall	SRISK	
Peer Effect:	1.761***		1.945*	
Liq. Creation - $\widehat{eta_{j,t}^{LC}}$	(0.492)		(1.005)	
Peer Effect:		0.598***		0.698**
Liq. Ratio - $\widehat{eta_{j,t}^{\widehat{LR}}}$		(0.175)		(0.283)
No observations	2,201	2,207	2,092	2,098
No. banks	316	317	313	314
Adj. $R^2$	0.161	0.157	0.245	0.243
Bank-level controls	Υ	Υ	Υ	Υ
Country-level controls	-	-	-	
Bank FE	Υ	Υ	Υ	Υ
Country-Year FE	Υ	Υ	Y	Υ

Introduction

# •

- 1. Liquidity holding *choices* of competitor banks *do matter* for funding liquidity risk management policies of individual banks.
- Both learning and collective moral-hazard seem to be at play.
  - ► A well functioning resolution and bail-in framework is essential to mitigate banks' bail-out expectations.

- Strategic liquidity risk management decisions increase (i) individual banks' default risk and (ii) overall systemic risk.
  - ► The effect is economically significant e.g., one standard deviation increase in peer effect (0.24 to 0.30) leads to a decrease in the Z-score of bank i of 0.08 to 0.14 (where mean of Z-Score is 3.46).
  - From a macro-prudential perspective, results highlight the importance of dealing with the systemic component of funding liquidity risk.

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# Thank you

Any comments or suggestions are more than welcome.

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"When the music stops, in terms of liquidity, things will be complicated. But as long as the music is playing, you've got to get up and dance. We're still dancing."

Chuck Prince, former chief executive of Citigroup - FT, July 2007