

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

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# Motivation

- ▶ *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 initiatives (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* (Rajan, QJE 1994; Uchida and Nakagawa, JFI 2007), *capital structure* (Leary and Roberts, JF 2014), *compensation* (Shue, RFS 2013), *investment policies* (Dougal et al., JF 2015).

# Research Questions

## 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?

## 2. Do strategic funding liquidity risk management decisions have an impact on financial stability?

- ▶ Collective risk-taking increases likelihood that banks fail altogether due to higher correlation of defaults (Allen et al., JFE 2012).

# Main Findings and Contribution

1. *Strategic liquidity risk management decisions increase both individual banks' default risk and overall systemic risk.*
  - ▶ To the best of my knowledge, no study so far empirically examine the impact of banks' strategic balance-sheet decisions on financial stability.
  
- 2a. *Both learning and collective moral-hazard channels seem to be at play.*
  - ▶ Consistent with literature analysing direct effects of bailout guarantees on bank risk-taking (e.g., Dam and Koetter, RFS 2012) → *But moral-hazard may not be confined to situations in which banks choose to bear excessive exogenous risk.*
  - ▶ Guarantees to a bank also affect risk-taking of competitors (Gropp et al., RFS 2011; Mariathasan et al., 2014) → *But collective risk-taking may not only be driven by bailout incentives.*

# Main Findings and Contribution

2b. *Banks' liquidity choices are determined directly by the decisions of competitors and, to a lesser extent, their other characteristics.*

- ▶ Bonfim and Kim (2014) analyse behaviour of European and North-American banks from 2002 to 2009 → evidence of competitors affecting individual banks' liquidity risk management policies.
- ▶ *But silent on how and why these peer effects materialise, and their consequences for individual banks' and system-wide stability.*

# Data

- ▶ Sample: 17,831 bank-year observations corresponding to 2,058 commercial banks in 32 OECD countries from 1999 to 2013.
- ▶ Banks' balance-sheets and income statements → Bankscope
  - ▶ 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.
- ▶ Daily stock prices and no. shares outstanding → Datastream
- ▶ Country/sector equity market indices → MSCI
- ▶ Country-level data → World Bank WDI and Doing Business database, IMF International Financial Statistics

# Empirical Model 1

## Baseline model to capture peer effects

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

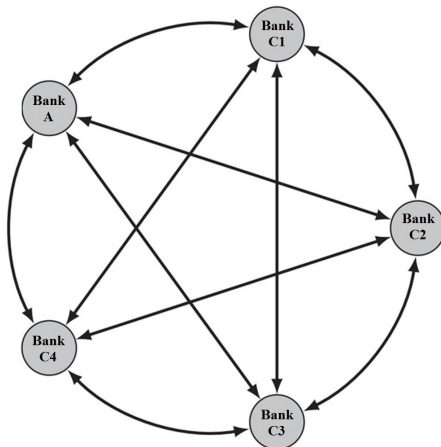
- ▶ Peer effects are captured by coefficient  $\beta \rightarrow$  influence of peer banks' funding liquidity choices on those of bank  $i$ .
- ▶  $Liq_{i,j,t}$  is either the Liquidity Ratio (Acharya and Mora, JF 2015) or the Berger and Bowman (RFS 2009) Liquidity Creation measure.
- ▶ **Endogeneity 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).*

# Identification strategy

- ▶ **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.

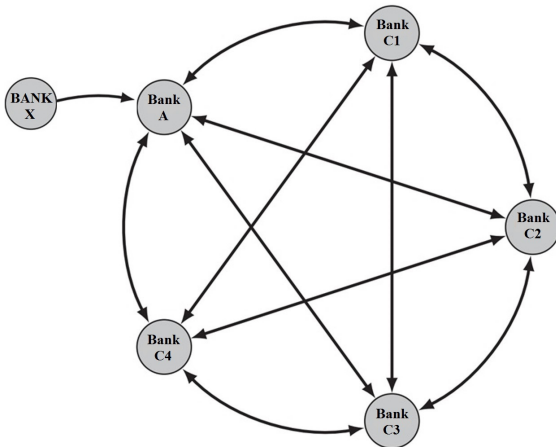


# Identification strategy



- ▶ A “complete network” (Acemoglu et al., AER 2015) of banks operating in the same country where (i) *Bank A* is a foreign-owned subsidiary; (ii) *Banks Cs* are its domestic competitors - similar size and business model.

# Identification strategy



- ▶ Funding liquidity risk profile of a bank-holding group (*Bank X*) based in country  $f$  can be viewed as an instrument for all banks in country  $j$  (*Banks Cs*) that belong to peer group of its foreign subsidiary (*Bank A*).

# 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 (Acharya, JFS 2009).
- ▶ 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.

## 3. *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.

# Empirical Model 2

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

### Step 1:

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

- ▶  $\beta_{j,t}$  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:

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

- ▶  $Stability_{i,j,t}$  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).

# Result 1: Peer effects in banks' liquidity choices

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** (0.003)	0.004 (0.004)	0.007** (0.003)
Peer Banks' Capital Ratio	0.110 (0.068)	0.123** (0.051)	0.121** (0.062)	0.084 (0.053)
Peer Banks' Return-on-Assets	0.093 (0.374)	0.195 (0.291)	0.053 (0.373)	-0.035 (0.278)
Peer Banks' Provisions	-0.009 (0.030)	0.030 (0.026)	0.004 (0.027)	0.043* (0.026)
Bank-level controls	Y	Y	Y	Y
Country-level controls	Y	Y	-	-
Year FE	Y	Y	N	N
Country FE	Y	-	N	-
Bank FE	N	Y	N	Y
Country-Year FE	N	N	Y	Y
IV (1st stage)	0.129*** (0.013)	0.160*** (0.014)	0.141*** (0.013)	0.125*** (0.011)

# Result 1: Peer effects in banks' liquidity choices

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 (0.358)	-0.181 (0.249)	0.639* (0.357)	-0.233 (0.251)
Peer Banks' Return-on-Assets	3.841* (1.982)	0.581 (1.486)	3.722* (2.005)	1.837 (1.418)
Peer Banks' Provisions	-0.046 (0.176)	-0.283** (0.140)	-0.069 (0.163)	-0.264** (0.132)
Bank-level controls	Y	Y	Y	Y
Country-level controls	Y	Y	-	-
Year FE	Y	Y	N	N
Country FE	Y	-	N	-
Bank FE	N	Y	N	Y
Country-Year FE	N	N	Y	Y
IV (1st stage)	0.216*** (0.010)	0.202*** (0.012)	0.203*** (0.010)	0.178*** (0.012)

# Result 1: Peer effects in banks' funding liquidity choices

- ▶ *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).

# Result 1: Peer effects in banks' liquidity choices - robustness

## 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).



# Result 2.1: Which banks strategically mimic their peers?

	Peer Effect: Liq. Creation		Peer Effect: Liq. Ratio	
Low Capital Ratio	0.898*** (0.337)	1.114*** (0.400)	0.383** (0.195)	0.444** (0.177)
High Capital Ratio	0.354* (0.207)	0.194 (0.203)	0.203 (0.199)	0.154 (0.185)
Low Profitability	0.476** (0.217)	0.497** (0.217)	0.426** (0.204)	0.503*** (0.166)
High Profitability	0.342 (0.214)	0.447** (0.213)	0.073 (0.201)	0.188 (0.189)
Low share of wholesale funding	0.374** (0.188)	0.292 (0.179)	0.191 (0.217)	0.241 (0.194)
High share of wholesale funding	0.942*** (0.302)	1.085*** (0.313)	0.544*** (0.196)	0.521*** (0.185)
Low loan-to-assets ratio	0.374** (0.175)	0.354** (0.175)	0.212 (0.190)	0.201 (0.189)
High loan-to-assets ratio	0.675*** (0.200)	0.743*** (0.224)	0.801*** (0.250)	0.928*** (0.226)
Foreign-owned banks	0.182 (0.313)	0.410 (0.288)	0.174 (0.195)	0.310* (0.159)
Non-foreign-owned banks	0.739*** (0.169)	0.663*** (0.178)	0.485*** (0.153)	0.565*** (0.138)

## Result 2.2: Who mimics who?

	Peer Effect: Liq. Creation		Peer Effect: Liq. Ratio	
Large banks → Large banks	0.981*** (0.164)	0.773*** (0.179)	0.909** (0.396)	1.185*** (0.327)
Large banks → Small banks	0.227 (0.300)	0.045 (0.293)	-0.059 (0.212)	0.218 (0.173)
Small banks → Small banks	1.332*** (0.379)	0.803** (0.373)	0.943*** (0.285)	0.428** (0.209)
Small banks → Large banks	0.765*** (0.211)	0.886*** (0.192)	1.155** (0.530)	1.178*** (0.453)
Peer Characteristics	Y	Y	Y	Y
Bank-level controls	Y	Y	Y	Y
Country-level controls	Y	Y	Y	Y
Year FE	Y	Y	Y	Y
Country FE	Y	-	Y	-
Bank FE	N	Y	N	Y

# Result 3.1: Peer effects and default risk

ln(Z-Score) – 3-year window: $\ln[(E/A + ROA)/\sigma(ROA)_{3y}]$				
Peer Effect:		-0.319**	-0.360**	
Liq. Creation - $\widehat{\beta}_{j,t}^{LC}$		(0.142)	(0.144)	
Peer Effect:			-0.442***	-0.366***
Liq. Ratio - $\widehat{\beta}_{j,t}^{LR}$			(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	Y	Y	Y	Y
Country-level controls	Y	-	Y	-
Year FE	Y	N	Y	N
Bank FE	N	Y	N	Y
Country FE	Y	-	Y	-
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 Expected Shortfall		SRISK	
Peer Effect:	1.761***		1.945*	
Liq. Creation - $\widehat{\beta}_{j,t}^{LC}$	(0.492)		(1.005)	
Peer Effect:		0.598***		0.698**
Liq. Ratio - $\widehat{\beta}_{j,t}^{LR}$		(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	Y	Y	Y	Y
Country-level controls	-	-	-	-
Bank FE	Y	Y	Y	Y
Country-Year FE	Y	Y	Y	Y

# Summary

1. Liquidity holding *choices* of competitor banks *do matter* for funding liquidity risk management policies of individual banks.
2. 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.
3. *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.

# 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