Data

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

André Silva

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  - 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).

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- 1. Strategic liquidity risk management decisions increase (i) individual banks' default risk and (ii) 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. While large banks' liquidity decisions are only sensitive to their large counterparts, small banks' liquidity choices are affected by the decisions of both small and large banks.
- 2b. Banks' liquidity choices are determined directly by the decisions of competitors and, to a lesser extent, their other characteristics.
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- 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.
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$$Liq_{i,j,t} = \omega + \beta \overline{Liq}_{-i,j,t} + \lambda' \bar{X}_{-i,j,t-1} + \gamma' X_{i,j,t-1} + \eta' Z_{j,t-1} + \mu_i + v_t + \varepsilon_{i,j,t}$$

- ▶ Peer effects are captured by coefficient  $\beta \rightarrow$  influence of peer banks' funding liquidity choices on those of bank i.
- Liqi,j,t is either the Liquidity Ratio (Acharya and Mora, JF 2015) on 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).

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- 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).

- 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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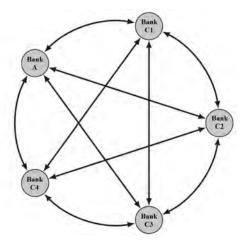
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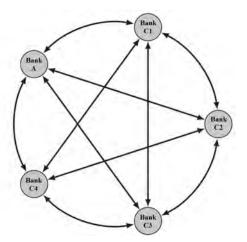
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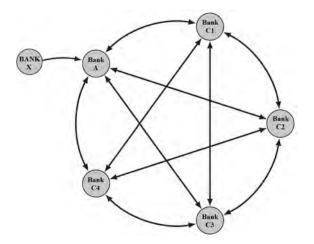
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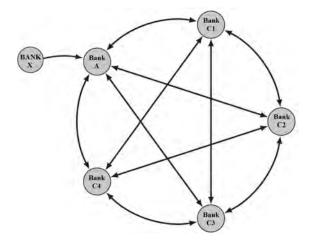
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Identification Strategy

- 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.

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#### 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.
- ▶ e.g., UK in 2010:

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

#### 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). Data

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

#### 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	0.009**	0.004	0.007**		
Peer Banks' Capital Ratio	(0.005) 0.110	(0 003) 0 123**	(0 004) 0 121**	(0.003) 0.084		
Peer Banks' Return-on-Assets	(0.068) 0.093	(0.051) 0.195	(0.062) 0.053	(0.053) -0.035		
Peer Banks' Provisions	(0.374) -0.009	(0.291) 0.030	(0.373) 0.004	(0.278) 0.043*		
r cer banks i rovisions	(0.030)	(0.026)	(0.027)	(0.026)		
Bank-level controls	Y	Y	Y	Y		
Country-level controls	Ϋ́	Ϋ́	-	-		
Year FE	Y	Y	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)		

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Peer Banks' Provisions	-0.009 (0.030)	0.030 (0.026)	0.004 (0.027)	0.043* (0.026)
	(0.030)	(0.020)	(0.021)	(0.020)
Bank-level controls	Υ	Υ	Υ	Y
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
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	(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
r eer Danks Netuni-on-Assets	(1.982)	(1.486)	(2.005)	(1.418)
Peer Banks' Provisions	-0.046	-0.283**	-0.069	-0.264**
	(0.176)	(0.140)	(0.163)	(0.132)
Bank-level controls	Y	Y	Υ	Y
Country-level controls	Y	Ϋ́	ī	ı
Year FE	<u>т</u> Ү	<u>т</u> Ү	N	N .
Country FE	Ÿ	-	N	-
Bank FE	N	Υ	N	Υ
Country-Year FE	N	N	Y	Ϋ́
IV (1st stage)	0.216***	0.202***	0.203***	0 178***
	(0.010)	(0.012)	(0.010)	(0.012)

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	(0.170)	(0.140)	(0.103)	(0.132)
Bank-level controls	Υ	Υ	Υ	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
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- 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
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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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### Which banks strategically mimic their peers?

	Peer Effect:	Liq Creation	Peer Effect	Liq Ratio
Low Capital Ratio	0.898***	1 114***	0.383**	0.444**
	(0.337)	(0.400)	(0.195)	(0.177)
High Capital Ratio	0.354*	0.194	0.203	0.154
	(0.207)	(0.203)	(0.199)	(0.185)
Low Profitability	0.476**	0.497**	0.426**	0.503***
•	(0.217)	(0.217)	(0.204)	(0.166)
High Profitability	0.342	ò.447**	0.073	0.188
,	(0.214)	(0.213)	(0.201)	(0.189)
Low share of wholesale funding	0.374**	0.292	0.191	0.241
ů	(0.188)	(0.179)	(0.217)	(0.194)
High share of wholesale funding	0 942***	1.085***	0.544***	0 521***
5	(0.302)	(0.313)	(0.196)	(0.185)
Low loan-to-assets ratio	0.374**	0.354**	0.212	0.201
	(0.175)	(0.175)	(0.190)	(0.189)
High loan-to-assets ratio	0.675***	0 743***	0 801***	0.928***
8	(0.200)	(0.224)	(0.250)	(0.226)
Foreign-owned banks	0.182	0.410	0.174	0.310*
Ŭ	(0.313)	(0.288)	(0.195)	(0.159)
Non-foreign-owned banks	0 739***	0.663***	0.485***	0 565***
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Conclusion

### Who mimics who?

	Peer Effect:	Liq. Creation	Peer Effect	Liq Ratio
Large banks $ ightarrow$ Large banks	0.981***	0.773***	0.909**	1.185***
	(0.164)	(0.179)	(0.396)	(0.327)
Large banks $ o$ Small banks	0.227	0.045	-0.059	0.218
0	(0.300)	(0.293)	(0.212)	(0.173)
Small banks $ ightarrow$ Small banks	1.332***	0.803**	0.943***	0.428**
Silvan Sanks , Silvan Sanks	(0.379)	(0.373)	(0.285)	(0.209)
Small banks $ ightarrow$ Large banks	0.765***	0.886***	1.155**	1.178***
· ·	(0.211)	(0.192)	(0.530)	(0.453)
Peer Characteristics	Y	Υ	Y	Y
Bank-level controls	Ý	Ý	Ý	Ý
Country-level controls	Υ	Υ	Υ	Υ
Year FE	Υ	Υ	Υ	Υ
Country FE	Υ	-	Υ	-
Bank FE	N	Υ	N	Υ

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Year FE	Υ	Υ	Υ	Υ
Country FE	Υ	-	Υ	-
Bank FE	N	Υ	N	Υ

### Peer effects and financial stability

n(Z-Score) - 3-ye	ear window:	$\ln[(E/A +$	$ROA)/\sigma(RO)$	$[OA)_{3y}]$
Peer Effect: $\widehat{\rho_LC}$	-0.319** (0.142)			
Liq. Creation - $ar{eta}_{j,t}^{LC}$	(0.142)	(0.144)		
Peer Effect:			-0.442***	-0.366***
Liq Ratio $\widehat{eta}_{j,t}^{L\widehat{R}}$			(0.132)	(0.118)
	10.051	100=1	10010	10010
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	Υ	=	Υ	=
Year FE	Υ	N	Υ	N
Bank FE	N	Υ	N	Υ
Country FE	Υ	-	Υ	-
Country-Year FE	N	Υ	N	Υ

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

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Bank-level controls	Y	Y	Υ	Υ		
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Bank FE	N	Υ	N	Υ		
Country FE	Υ	-	Υ	-		
Country-Year FE	N	Υ	N	Υ		

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### Peer effects and systemic risk

	Marginal Ex	pected Shortfall	SR	ISK
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 $\widehat{eta_{j,t}^{LR}}$		(0.175)		(0.283)
$J, \iota$		, ,		. ,
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	Υ	Υ	Υ
Country-level controls	-	-	-	-
Bank FE	Υ	Υ	Υ	Υ
Country-Year FE	Υ	Y	Υ	Υ

Identification Strategy

### Peer effects and systemic risk

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Bank FE	Υ	Υ	Υ	Υ
Country-Year FE	Y	Υ	Υ	Υ

Identification Strategy

Introduction

# 1. Liquidity holding *choices* of competitor banks *do matter* for funding

- 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.

- 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).
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# Thank you

Any comments or suggestions are more than welcome. andre.silva.3@cass.city.ac.uk

"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