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Taking regulation seriously: Fire sales under solvency and liquidity constraints

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Columbia University NYC, February 2018

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Asset shock: variants of 2017 stress test Funding shock Asset and Funding shocks

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"During the early 'liquidity phase' of the financial crisis that began in 2007, many credit institutions, despite maintaining adequate capital levels, experienced significant difficulties because they had failed to manage their liquidity risk prudently... (Such) credit institutions were then forced to liquidate assets in a fire-sale which created a self-reinforcing downward price spiral and lack of market confidence triggering a solvency crisis."

(European Commission, 2015)

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Motivation

- Liquidity issues during the crisis
- Multiple regulatory constraints
- Macroprudential stress tests

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Motivation

- Liquidity issues during the crisis
- Multiple regulatory constraints
- Macroprudential stress tests
- Objectives:
 - Build a quantitative model of fire sales to assess the interaction between liquidity and solvency constraints that banks simultaneously face.
 - Which types of financial shocks and regulatory requirements combine to produce fire sales?
 - How do banks optimally liquidate their portfolios when they are forced to do so?



Literature review

• Fire-sale models:

[Greenwood et al., 2015], [Cont and Schaanning, 2017], [Duarte and Eisenbach, 2013]

• Constraints and optimal deleveraging:

[Cecchetti and Kashyap, 2016], [Braouezec and Wagalath, 2016]

• Liquidity:

[Hellwig, 2009], [Gorton and Metrick, 2012], [Pierret, 2015] , [Acharya and Merrouche, 2012]

• Macro-stress tests:

[Dees and Henry, 2017], [Bank of England, 2017], [Bardoscia et al., 2017], [Fique, 2017], [Puhr and Schmitz, 2014], [Calimani et al., 2017]

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Model overview





Bank balance sheets

- Marketable securities $M_{i,k}$, k = 1...310 and i = 1...7Bonds and equity holdings that are available for sale and suffer a price impact.
- Other assets $O_{i,k}$, k = 1, 2: loans, intangible goods, and off-balance sheet items, which are **not** available for deleveraging.
- **Cash** or cash-like assets $C_{i,k}$, k = 1, 2.



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- **Cash** or cash-like assets $C_{i,k}$, k = 1, 2.
- Liabilities $L_{i,k}$, k = 1...12. These include classic retail customer deposits, institutional deposits, short-term whole-sale funding, and issued debt.
- Capital E_i.



Regulatory constraints

• Risk-weighted Capital Ratio:

$$CAP^{i}(A, E) := rac{E^{i}}{
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Regulatory constraints

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• Liquidity Coverage Ratio:

$$LCR^{i}(A, C, L) := \frac{\lambda^{\top}M^{i} + \mathbf{1}^{\top}C^{i}}{\omega_{out}^{\top}L^{i} - \omega_{in}^{\top}A^{i}} \geq REG_{LCR}.$$

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Shocks

We consider three type of shocks:

- Asset shock (ϵ_A): $A_0^{i,k} = A^{i,k}(1 \epsilon_A^k)$. (k = 1...314)
- **2** Funding shock (ϵ_L) : $L_0^{i,k} = L^{i,k}(1 \epsilon_L^k)$. (k = 1..12)
- **3** Combined asset and funding shock.

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3 Combined asset and funding shock.

$$E_0^i = (E^i - \epsilon_A^\top A^i)^+.$$

$$C_0^i = (C^i - \epsilon_L^\top L^i)^+.$$

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Bank deleveraging



Figure: Shrinking a bank's balance sheet

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Price evolution under fire sales

$$P_{t+1}^{k} = P_{t}^{k} \left(1 - \delta_{k}^{-1} \sum_{i=1}^{N} S_{t}^{i,k} \right),$$

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Price evolution under fire sales

$$P_{t+1}^{k} = P_{t}^{k} \left(1 - \delta_{k}^{-1} \sum_{i=1}^{N} S_{t}^{i,k} \right),$$

Two forms of loss:

• Mark-to-market losses



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Price evolution under fire sales

$$P_{t+1}^{k} = P_{t}^{k} \left(1 - \delta_{k}^{-1} \sum_{i=1}^{N} S_{t}^{i,k} \right),$$

Two forms of loss:

Mark-to-market losses



• Implementation shortfall

$$\frac{1}{2}\sum_{k=1}^{K} S_t^{i,k} \sum_{j=1}^{N} \delta_k^{-1} S_t^{j,k}.$$



$$\min_{\mathbf{S}^{i},\mathbf{R}^{i}}(M^{i}-\frac{1}{2}S^{i})^{\top}(\frac{S^{i}}{\delta}),$$



subject to the constraints

 $CAP^{i}(A, E; \mathbf{S}) \geq REG_{CAP}$ $LEV^{i}(A, C, E; \mathbf{S}, \mathbf{R}) \geq REG_{LEV}$ $LCR^{i}(A, C, L; \mathbf{S}, \mathbf{R}) \geq REG_{LCR}$ $CASH^{i}(A, C; \mathbf{S}, \mathbf{R}) \geq 0.$



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Note: banks only internalise the effects of their own sales, and not the spillover effects of sales by other banks.



Calibration

- Balance sheet data taken from regulatory returns (COREP and FINREP) and Bank of England stress test data.
- **Regulatory weights** based on Basel guidance, European legislation and firms' annual statements.
- **Regulatory ratios & constraints** taken from regulatory returns.
- Market depths based on national authorities' published statistics on average trading volumes and S&P price indices for government bonds, and BoAML prices and oustanding volumes for corporate bonds.

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Stress scenarios

We consider three scenarios:

- Asset shock (ϵ_A) : Bank of England 2017 Stress scenario and shocks of increased intensity.
- **2** Funding shock (ϵ_L) : Depositor run (20%, 40% and 60% deposit outflows).
- Combined asset and funding shock: Bank of England 2017 Stress scenario and 20% deposits outflows.



Asset shock

- Risk-weighted capital requirements tend to be more tightly binding than leverage constraints.
- Banks constrained by risk-weighted capital constraints sell on average more illiquid assets, and in larger amounts, than when constrained by the leverage ratio.
- The size of unexpected losses, which are not internalized by banks, can be as important as the size of expected losses.



Asset sales: leverage ratio only





Asset sales: capital ratio only





Asset sales: all constraints









Fire-sale losses: decomposition





Funding shock: deposit outflows

- Banks prefer to use cash and sell highly liquid assets first to minimise losses.
- However, as the shock becomes larger, banks are forced to sell less liquid assets.
- When banks defend their LCRs to keep them above 100%, they need to sell less liquid assets in larger amounts.
- Hence fire-sale losses are significantly larger relative to the case when banks do not defend their LCRs.

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Funding shock					

Asset sales



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Fire-sale losses: decomposition



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Asset and Funding shocks





Conclusions

- Both risk-weighted capital and liquidity constraints can become binding and generate significant fire sales losses, by incentivising sales of larger amounts of less liquid assets.
- Models that only account for a leverage constraint might then under-estimate fire sale losses.



Conclusions

- Both risk-weighted capital and liquidity constraints can become binding and generate significant fire sales losses, by incentivising sales of larger amounts of less liquid assets.
- Models that only account for a leverage constraint might then under-estimate fire sale losses.
- Unexpected fire sales losses, e.g. losses due to deleveraging by other banks, can be larger than banks' expected losses from their own sales.
- Relaxing banks' regulatory constraints during stress may be a possible mitigating action to avoid fire sales. For example, allowing banks to draw down their LCR.

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- Run more rounds of fire sales.
- Explore solvency-liquidity nexus by running asset and funding shocks (both at the same time and sequentially).
- Sensitivity analysis: market depths, price function, targeting vs threshold.
- Constraints: UK leverage framework, LCR with limits to reserves usability.

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Thank you



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