Fire sales, price-mediated contagion and systemic risk.

Financial Risk & Network Theory - Cambridge September 2016 Eric Schaanning^{a,b}

Joint work with Rama Cont^a

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- 2 Modelling fire sales
- 3 Empirical application: European Banking Network





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Goal: Develop models for macro stress testing that can quantify such second round effects in a realistic and robust way.

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2 Modelling fire sales





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- **② Deleveraging** of marketable assets by some institutions
- **§ Feedback effects** via price-mediated contagion
 - \rightarrow potentially triggers more deleveraging (cascade).

Modelling fire sales

Model balancesheet

Illiquid assets	
Residential mortgage exposures	
Commercial real estate exposure	
Retail exposures: Revolving credits, SME, Other	
Indirect sovereign exposures in the trading book	
Defaulted exposures	
Residual exposures	
Marketable assets	
Corporate bonds	
Sovereign debt	
Direct sovereign exposures in derivatives	
Institutional client exposures: interbank, CCPs,	

Table: Stylized representation of asset classes in bank balance sheets.

- A stress scenario is defined by a vector ε ∈ [0, 1]^K whose components ε_μ are the percentage shocks to asset class μ.
- Gradual increase of the shock from 0% to 20%.
- Four scenarios:
 - 1. Spanish residential and commercial real estate losses
 - 2. Northern Europe residential losses
 - 3. Southern Europe commercial real estate losses
 - 4. Eastern Europe commercial real estate losses

Response functions



Figure: Leverage targeting response function (dashed) and two variants of the threshold (full and circles) response functions.

Price impact

The price of an asset undergoing a forced liquidation at *t*:

$$S_{t+1}^{\mu} = S_{t}^{\mu} \exp\left(-\delta_{\mu}^{-1} \sum_{j=1}^{M} \Pi_{t}^{j\mu} \Gamma_{t+1}^{j}\right)$$
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Figure: Large variation in estimated liquidity of different assets.



2 Modelling fire sales





Empirical application: European Banking Network

Fire sales losses and market depth



Indirect exposures and stress test outcomes



Liquidity weighted overlap

Our model shows that losses are proportional to the liquidity weighted overlap

$$\omega_{ij} := \sum_{\mu=1}^{M} \frac{\Pi_{i\mu} \Pi_{j\mu}}{\delta_{\mu}} \tag{2}$$

This leads to a network of portfolio overlaps:

$$\Omega := \Pi D^{-1} \Pi^{\top}, \tag{3}$$

which can can be studied with network analysis tools.

Empirical application: European Banking Network



Figure: European banking system: liquidity weighted overlap network



Omega1: EV 1

Figure: European banking system: Liquidity weighted overlaps



Omega2: EV 1

Figure: European banking system: Nominal overlaps

Distribution of fire sales losses



Figure: \log_{10} (fire sales loss) for different scenarios and different model combinations.

Sensitivity to initial stress scenario

Scenario combination	Sample correlation coefficient
1 & 2	0.0840
1 & 3	0.2130
1 & 4	-0.1449
2 & 3	-0.0509
2 & 4	0.0394
3 & 4	-0.0149

Table: Sample correlations between the initial loss vectors from the stress scenarios. The four stress scenarios are very different in terms of which banks are hit by the corresponding shock.

Empirical application: European Banking Network

Sensitivity to initial stress scenario



Figure: The pairwise sample correlation between the fire sales loss vectors of different scenarios as a function of the initial shock. Threshold model full lines - leverage targeting dashed lines.

Sensitivity to initial stress scenario



Figure: The evolution of the pairwise sample correlation during the fire sales cascade for a given scenario. Threshold full - leverage targeting dashed.

Empirical application: European Banking Network

Risk management for whales (Cont and Wagalath 2016)



Figure 6: 95% 5-month VaR for positions in CDX IG9 (size in Bn \$).



2 Modelling fire sales







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- Including fire sales and endogenous mechanisms (with realistic parameter estimates) can change the outcome of stress tests: Next generation stress testing models must include such feedback effects.
- Seemingly innocent modelling choices on response functions and liquidity estimates have a significant effect on results!

Conclusions for modelling

• Important to account for heterogeneity in agent resilience and asset liquidity. Any meaningful fire sales stress test needs to include a sensitivity analysis on the market depth parameter.

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- The threshold model generates more realistic short term dynamics under stress. Leverage targeting models seem better suited to capture long term dynamics.
- Leverage targeting models produce counter-intuitive short term dynamics.
- Singular value decompositions of liquidity weighted overlap matrices can provide valuable information for monitoring purposes and policy responses.

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