

# Market Structure, Counterparty Risk, and Systemic Risk

Dale W.R. Rosenthal<sup>1</sup>

UIC, Department of Finance

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<sup>1</sup>daler@uic.edu; tigger.uic.edu/~daler

# Counterparty Risk

- *Counterparty*: other side of ongoing financial agreement.
  - A bank enters into a swap with you on the S&P 500.
- Counterparty Risk
  - Risk resulting from default/bankruptcy of a counterparty.
  - Strictly: Risk to you from one of your counterparties.
  - Broadly: Includes effects on overall market (our concern).
- This broad definition we refer to as *systemic risk*.

# Counterparty Risk to Systemic Risk

- Counterparty risk affects market when large failure looms:
  - Near-bankruptcy of Bear Stearns (May 2008)
  - Bankruptcy of Lehman Brothers (Sep 2008)
  - Bankruptcy of Refco Inc? (Oct 2005, owned #1 CME broker)
- Outstanding notional at CME before ceasing trading:

Bear	Lehman	Refco LLC
\$761 BB	\$1,150 BB	\$130 BB

- N.B. No defaults or trade halts at CME for these events.
- Other bankruptcies: Askin (1994), LTCM (1998, why I care).
- Is counterparty risk an “accelerant” in financial crises?

# Systemic Risk

- Distress increases volatility sharply and significantly.
  - Widens spreads: transactions costs  $\uparrow$ ; market liquidity  $\downarrow$ .
  - Volatility is pushed onto the survivors (externality).
- Crisis bankruptcies have real costs:
  - Virtuous, vicious circles of market and funding liquidity<sup>2</sup>.
  - Reduced funding liquidity affects non-financial firms also.
  - Less invested in risky assets; allocative inefficiency?
  - Higher unemployment: harder job searches, lower tax revenue.
  - Bernanke (1983): affects credit markets; possible depression.

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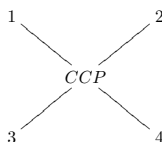
<sup>2</sup>Brunnemeier and Pedersen (2009).

# Results Preview

- Market structure affects contagion and exposure to defaults.
- Specifically: complete networks magnify systemic risk.
  - Disagrees with Allen and Gale (2000), Nier et al (2007).
  - Difference due to differing creation of complete networks.
  - Also: financial, banking networks differ (cf Acemoglu).
- Market fragility estimable with a few metrics of market core.
- Can price distress volatility of differing structures.

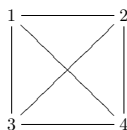
# Model: Market Structures

- Investigate two extremes of  $n$ -counterparty networks.



Star network

(Market with CCP<sup>3</sup>)



Complete network

(Bilateral "OTC" market)

- Each node is a counterparty (capital  $K$ , risk aversion  $\lambda$ ).
- Each edge is a contract<sup>4</sup> linking counterparties  $i$  and  $j$
- Contract exposure:  $q_{ij} = -q_{ji}$ ;  $q_{i<j} \stackrel{iid}{\sim} N(0, \eta^2)$
- Counterparty  $i$ 's net exposure:  $Q_i = \sum_{j \neq i} q_{ij}$ .
- Same net exposures ( $Q_i$ 's) in both networks.

<sup>3</sup>Central counterparty.

<sup>4</sup>A swap or forward on a risky asset.

# Model: Event Timing

To study counterparty risk, events occur at discrete times.

$t = 0$ : Bankruptcy of counterparty  $n$  occurs.

- All contracts with counterparty  $n$  are invalidated.
- Pushes unwanted exposure onto other  $n - 1$  counterparties.

$t = 1$ : Living counterparties trade in response to bankruptcy.

$t = 2$ : Living counterparties close out bankruptcy-induced exposure.

Order of trading in a period is random, not strategic.

# Model: Price Impact of Trading

- Each counterparty  $i$  trades  $x_i$  shares at time  $t = 1$ .
- Huberman and Stanzl (2004) arbitrage-free price impact.
  - Impact has linear permanent component<sup>5</sup>.
  - Permanent component impacts prices for later traders.
- Trade ordering, price impact create low and high prices.
- Time periods are very short; two simplifying assumptions:
  - 1 Prices have no drift other than price impact due to trading.
  - 2 Price diffusion is Gaussian (not log-normal).
- Defer handling crisis-related adverse selection.

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<sup>5</sup>Price impact could arise from inventory risk cost, non-crisis adverse selection.



# Effects of Invalidated Contracts

- Suppose counterparty A is net long the market.
- $\Rightarrow$  Other counterparties are net short the market.
- These are their preferred equilibrium positions.
- Thus when counterparty A defaults:
  - Survivors must re-create exposure from counterparty A.
  - Survivors become net sellers.
- CCP market: only CCP trades; net sell.
- OTC market: some counterparties will sell, some will buy.
- However, counterparties trade in own interest.
  - Do they re hedge immediately? Push market further?

# Large Bankruptcy

- Consider bankruptcy of a large financial firm.
- Assume large market move  $r_0$  at  $t = 0$  induces bankruptcy.
- Net exposure  $Q_n$  probably large; estimate via EVT<sup>6</sup>.

$$\hat{Q}_n = \frac{-K}{r_0} + \frac{\eta\sqrt{n-1}}{c_n(1 - e^{-e^{-c_n\kappa_1-d_n}})} \sum_{k=1}^{\infty} \frac{(-1)^{k+1} e^{-k(c_n\kappa_1+d_n)}}{kk!} \quad (1)$$

where  $\kappa_1 = \frac{-K}{r_0\eta\sqrt{n-1}}$  (minimum exposure causing death),

$$c_n = \frac{1}{\sqrt{2\log(n)}}, \text{ and } d_n = \sqrt{2\log(n)} - \frac{\log\log(n) + \log(16\tan^{-1}(1))}{2\sqrt{2\log(n)}}.$$

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<sup>6</sup>Equivalent: endow all counterparties with perfect information, examine most likely  $Q_n|r_0$ .

# Large Bankruptcies

- For large  $Q_n$ , trading at  $t = 1, 2$  will move market a lot.
- Move will be further in direction that caused bankruptcy.
- This raises two distressing possibilities:
  - Contagion: move may cause other counterparties to fail; or,
  - Checkmate: hedging may bankrupt the hedger.
- Counterparties anticipate these, respond selfishly.
- For bilateral OTC market, all counterparties may trade.
  - All hedge anticipated follow-on bankruptcy exposure  $\hat{Q}_f$ .
  - Trouble:  $\nu > 1$  (overtrading at  $t = 1$ ) to be expected.
  - Longs, shorts may largely self-segregate re hedge timing.
- Thus network structure matters.

# Large Bankruptcy: Equilibrium CCP Trade

- CCP anticipates follow-on bankruptcies; equilibrium yields
- Follow-on bankruptcy exposure  $\hat{Q}_f$  (**distress exposure**):

$$\hat{Q}_f = (n-1)^{3/2} \eta \frac{\phi(\kappa_2) - \phi(\kappa_1)}{\Phi(\kappa_1)} \quad \text{where} \quad (2)$$

$$\kappa_2 = \frac{-Kp_0 / [\eta \sqrt{n-1}]}{p_0 r_0 - \pi(\hat{Q}_n + \hat{Q}_f)} = \text{min exposure for follow-on death.}$$

- # follow-on bankruptcies  $\hat{b}$  (**distress pervasiveness**):

$$\hat{b} = (n-1) \frac{\int_{\kappa_2}^{\kappa_1} \phi(z) dz}{\int_{-\infty}^{\kappa_1} \phi(z) dz} = (n-1) \left( 1 - \frac{\Phi(\kappa_2)}{\Phi(\kappa_1)} \right) \quad (3)$$

# Large Bankruptcy: Equilibrium OTC Net Trade

- OTC traders anticipate one another, follow-on bankruptcies.
- However: those most at-risk re hedge quickly, others delay.
- Random trade sequence  $\Rightarrow$  uncertain low of rehedging  $\underline{S}_{n-1}$ .
- Use these to solve for equilibrium OTC net trade.

$$\kappa_2 = \frac{-Kp_0}{\eta\sqrt{n-1}(p_0r_0 + \pi E(\underline{S}_{n-1}|\nu))}, \quad (4)$$

$$\hat{Q}_f = (n-1)^{3/2}\eta \frac{\phi(\kappa_2) - \phi(\kappa_1)}{\Phi(\kappa_1)}. \quad (5)$$

- Important to note that  $\nu \geq 1$  (in  $E(\underline{S}_{n-1})$ ).
- Finding  $\nu$  is hard:  $n$ -player (random) game; usually c1.75.

# Bad Behavior? Checkmate and Hunting

## Proposition (Checkmate)

*A large enough initial bankruptcy may yield a follow-on bankruptcy in expectation — despite any finite effort by the troubled counterparty.*

## Proposition (Hunting)

*For a complete network of 3 or more counterparties and a large enough initial bankruptcy, two or more other counterparties may profit by driving a survivor into (follow-on) bankruptcy.*

# The Other Extreme: A Separating Equilibrium?

- Another (extreme) possibility exists in bilateral OTC markets:
  - Buyers and sellers may separate when they trade.
- Those who are same side as net rehedger rush to hedge first.
- Those on other side wait to allow maximum distress.
- If net rehedger makes sellers panic, net sale in period 1 is:

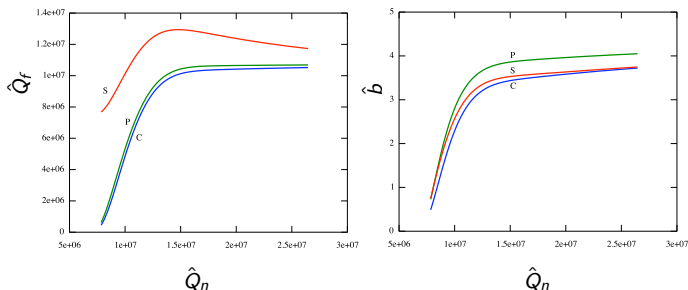
$$-E\left(\sum_{i=1}^{n-1} [x_i]^- \mid \sum_{i=1}^{n-1} x_i = -\hat{Q}_n - \hat{Q}_f\right) \quad (6)$$

$$\approx -(n-1)^{3/2} \eta \phi(\mu^*) - (\hat{Q}_n + \hat{Q}_f)(1 - \Phi(\mu^*)) \quad (7)$$

where  $\mu^* = \frac{\hat{Q}_n + \hat{Q}_f}{(n-1)^{3/2} \eta}$  (net rehedger in std devs/survivor)  
and  $\phi, \Phi$  are standard normal pdf, cdf.

# Large Bankruptcies: Indicative Distress

- Consider large bankruptcy for  $n = 10$  counterparties<sup>7</sup>.
- Std deviation of bilateral contract exposure  $\eta = 1,000,000$ .
- Distress exposure  $\hat{Q}_f$  and pervasiveness  $\hat{b}$  vs.  $\hat{Q}_n$ .



Lines: (P)ooled OTC; (S)eparated OTC; (C)CP

*P – S: Envelopes of distress exposure, pervasiveness*

<sup>7</sup>Price impact parameters are as in Almgren and Chriss (2001).



# Large Bankruptcies: Example of Market Impact

- Suppose  $\hat{Q}_n = 10,000,000$ ; GARCH variance decay of 0.9.
- For CCP market:
  - Expected market impact:  $-\$30$ .
  - Effective annual volatility goes from 30% to 38%.
- If pooled OTC buyers, sellers overtrade  $1.75\times$  at  $t = 1$ .
  - Expected market impact:  $-\$31$ .
  - Annual volatility  $\uparrow$  to 328% (instant.), 146% (effective).
- If OTC buyers and sellers separate, at  $t = 1$ :
  - Expected market impact:  $-\$41$ .
  - Annual volatility  $\uparrow$  to 596% (instant.), 268% (effective).

# Large Bankruptcies: Example of Real Effects

- Suppose  $\hat{Q}_n = 10$  MM, market size of \$40 MM<sup>8</sup>.
- If 8% equity premium and mean risk aversion of  $\hat{\lambda} = 3$ :
  - Equilibrium allocation to risky asset: 29% (71% cash).
  - Post-crisis: 19% (CCP), 1.2% (OTC pool), 0.4% (OTC sep).
- Cost of distress externality:
  - \$3.2MM (CCP), \$123 MM (OTC pool), \$425 MM (OTC sep).
  - Cost of OTC market distress is 3–11 $\times$  market size.
- Given 2–3 bankruptcies; mean employees, compensation:
  - 260,000–400,000 unemployed; \$33–\$49 billion pay loss.
  - At 40% total taxes: revenue loss of \$13–\$20 billion.
- Also affects credit markets, overall macroeconomy.

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<sup>8</sup>Approximately  $2(\hat{Q}_n + \hat{Q}_f)$ .

# Large Bankruptcies: Not So Random

- Complete networks admit two destabilizing events:
  - Checkmate: weak counterparty may have no beneficial trade.
  - Hunting: counterparties force others into bankruptcy.
- Worse, hunting is a full equilibrium behavior.
  - Market may be pushed far beyond one follow-on bankruptcy.
- Are counterparties selfishly amoral/evil? Maybe not.
  - Trade amount may pre-hedge expected follow-on bankruptcies.
  - This reduces surprise need for trading in period 2.
- CCP markets have fewer such destabilizing events.
  - Suggests central clearing reduces OTC market volatility.

## Difference from Allen and Gale (2000)

- Allen and Gale (2000): complete networks are more robust.
- I disagree: complete networks are more fragile.
- Why? Differing methods of network construction.
- Allen and Gale approach: top-down.
  - Net exposure:  $Q_i \stackrel{iid}{\sim} N(0, (n-1)\eta^2)$
  - Contract exposure:  $q_{ij} = Q_i / (n-1)$ . (all same sign)
- My approach: bottom-up.
  - Contract exposure:  $q_{i < j} \stackrel{iid}{\sim} N(0, \eta^2)$ ;  $q_{ij} = -q_{ji}$ ;
  - Net exposure:  $Q_i = \sum_{j \neq i} q_{ij}$ ;  $Q_i \stackrel{iid}{\sim} N(0, (n-1)\eta^2)$ .
- Same net exposures  $Q_i$ 's, different contract exposures  $q_{ij}$ 's.
- Strategic separation of buyers, sellers unlikely in A&G.

# Conclusion

- Even small bankruptcies temporarily increase volatility.
- For a large bankruptcy in a bilateral OTC market:
  - Counterparties may be unable to save themselves (checkmate).
  - Counterparties may hunt their weakest peers for profit.
  - Volatility externality (and thus cost) higher than CCP market.
- Self-segregating buyers, sellers in OTC markets can be nasty:
  - Externality distress cost  $\gg$  market size. (market failure?)
- Suggests benefits to centralized clearing in OTC markets<sup>9</sup>.
- Volatility externality cost  $\Rightarrow$  when to move markets to CCP.
- May be able to measure when markets are more/less brittle.
  - $n, \eta, \bar{K}$  for part of market like complete network.

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<sup>9</sup>Biais, Heider, Hoerova (2011) suggests CCP is capital efficient.