

# How Likely is Contagion in Financial Networks?

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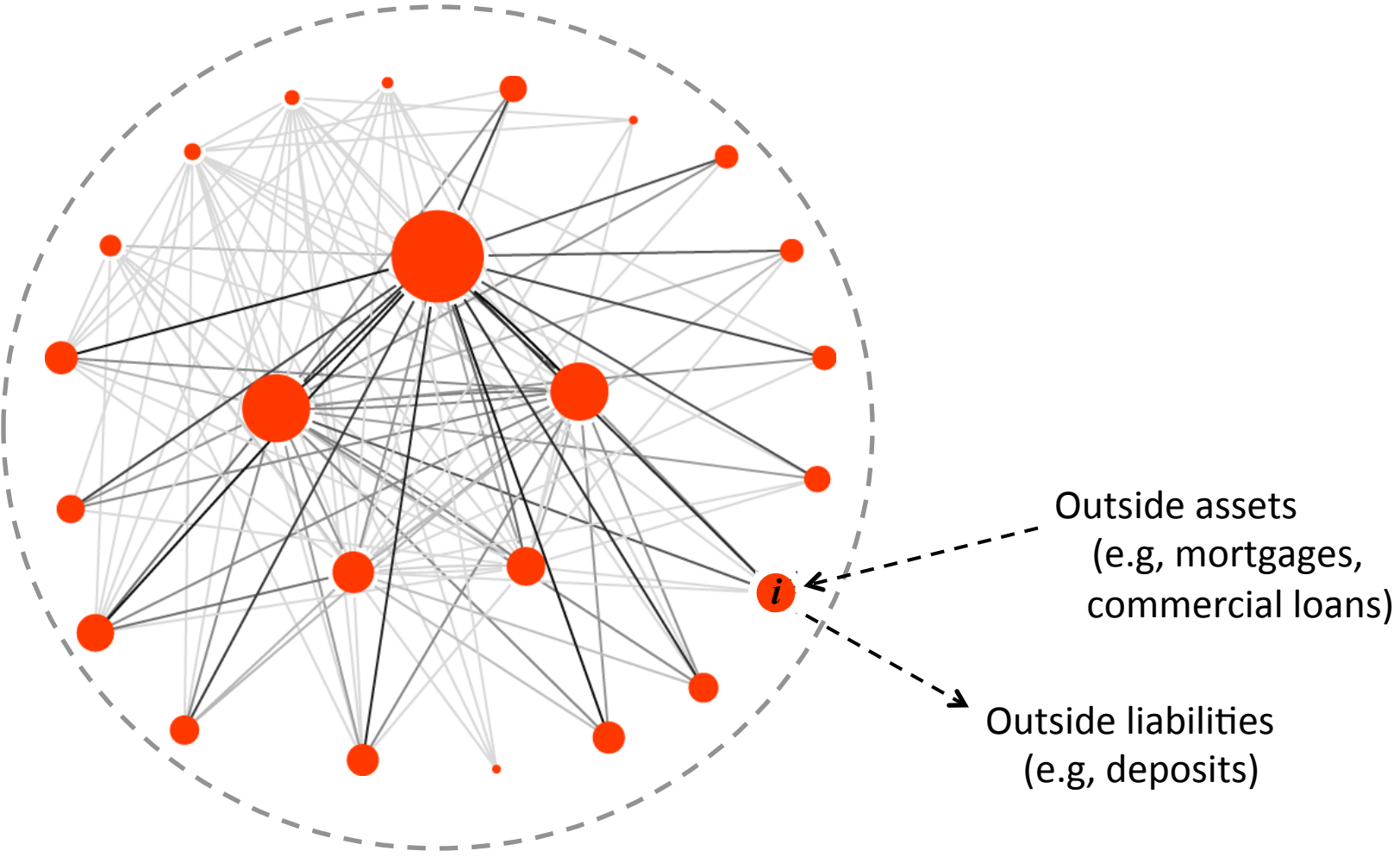
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# Overview of What We Do

- We look at “interbank” networks in which
  - Nodes are financial institutions
  - Edges are defined by payment obligations
  - Assets are subject to exogenous stochastic shocks
- Networks are opaque – nobody has a comprehensive view
  - What can we say using just node-level data?
  - What node-level metrics should be monitored?
- How much do network effects add to system-wide losses?
  - We bound contagion probabilities and loss amplification

# The Financial Network And The Outside World



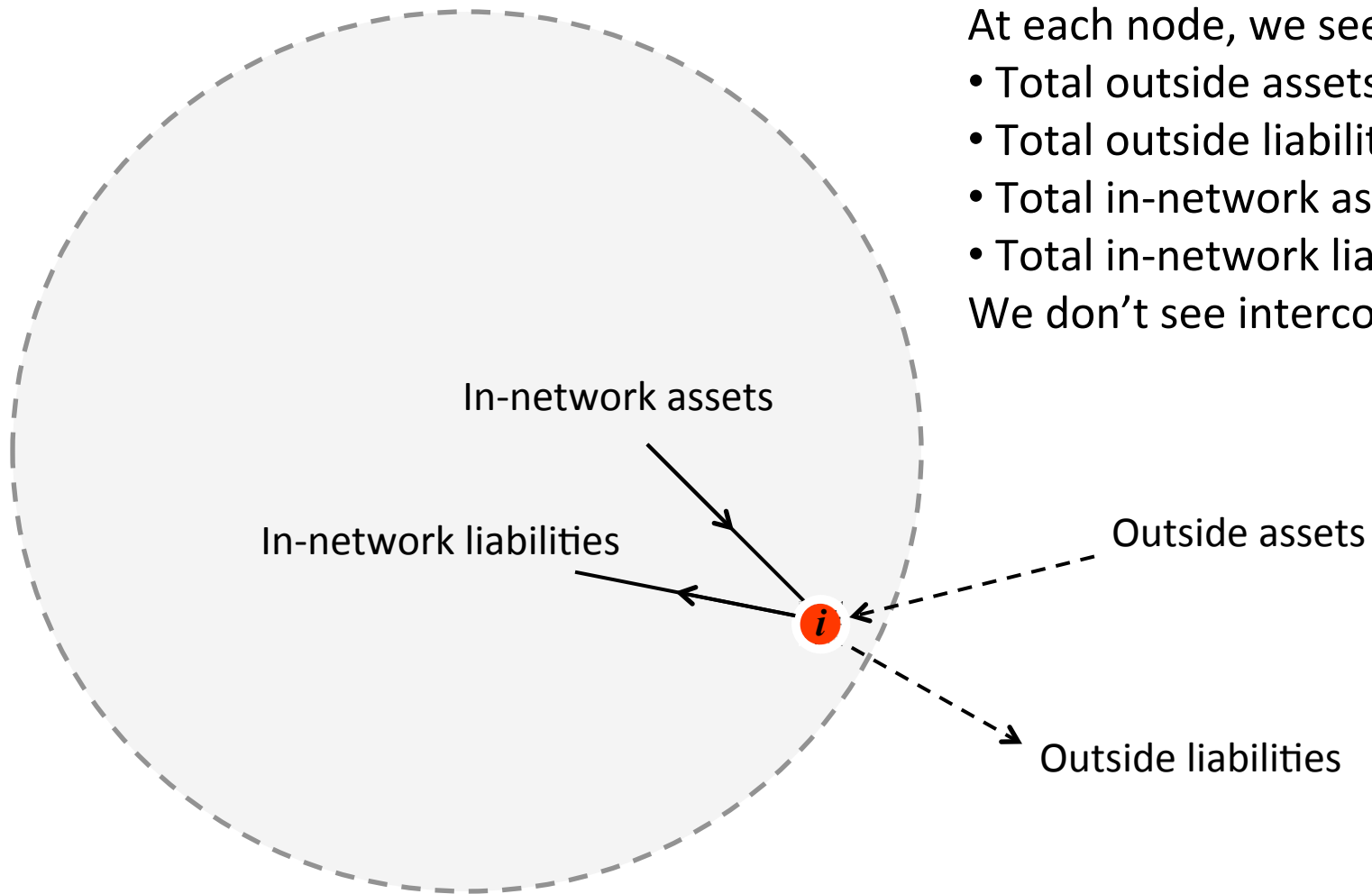
Network figure from Bank of England

# What We See

At each node, we see

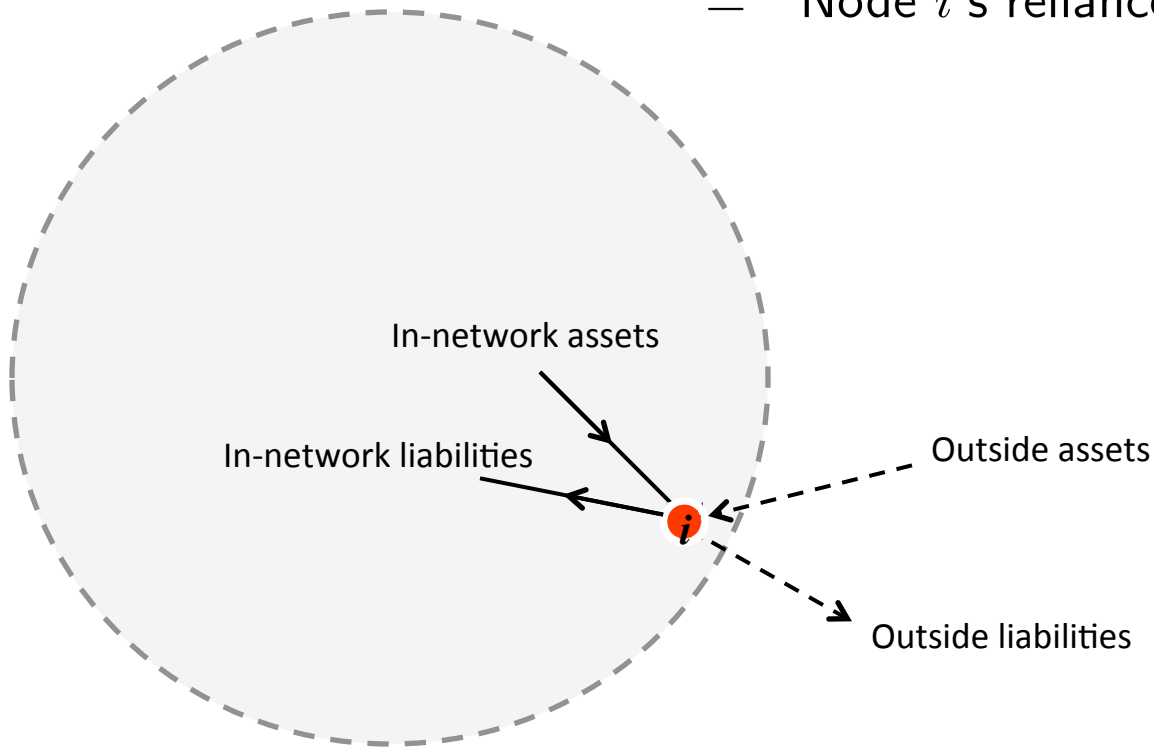
- Total outside assets
- Total outside liabilities
- Total in-network assets
- Total in-network liabilities

We don't see interconnections



# Financial Connectivity

A key quantity:  $\beta_i = \frac{\text{Node } i\text{'s in-network liabilities}}{\text{Node } i\text{'s total liabilities}}$   
 $= \text{Node } i\text{'s reliance on in-network funding}$



# Comparison of Expected Losses with a Network and in a System of Isolated Nodes

$c_i$  = value of node  $i$ 's outside assets

$\delta_i$  = probability that  $i$  defaults from a direct shock to its assets

For a broad class of shock distributions

$$\frac{E[\text{losses with network}]}{E[\text{losses w/o network}]} \leq 1 + \frac{\sum_i \delta_i c_i}{(1 - \beta_{\max}) \sum_i c_i}$$

where  $\beta_{\max} = \max_i \beta_i$ .

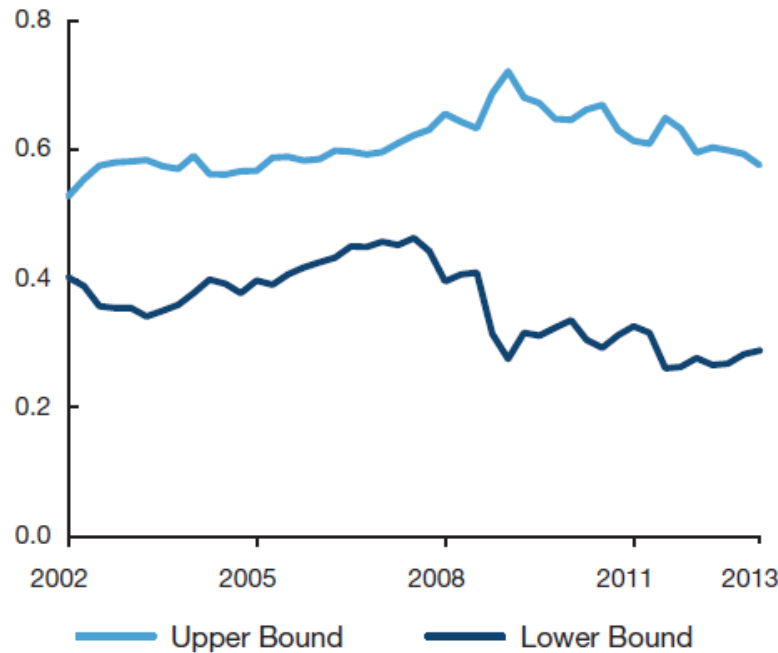
If  $\delta_i \leq 1\%$  and  $\beta_{\max} \leq 0.50$ , then the network multiplier is at most 1.02.

If the risk at each node is managed and the nodes are not overly connected, the network can't do a lot of damage

# Rough Estimates of Financial Connectivity

From OFR's 2013  
Annual Report

Figure 44. Trends in Financial Connectivity of 50 Largest U.S. Bank Holding Companies (ratio)



Note: Top 50 U.S. bank holding companies by assets as of March 2013. Upper bound calculated as  $(\text{Noncore Funding} + \text{Off-Balance-Sheet Derivative Liabilities}) / (\text{Total Liabilities} + \text{Off-Balance-Sheet Derivative Liabilities})$ . Lower bound calculated as  $(\text{Noncore Funding}) / (\text{Total Liabilities} + \text{Off-Balance-Sheet Derivative Liabilities})$ .

Sources: FR Y-9C data from Federal Reserve System (2013).

# Network Effects on Contagion

- We say that contagion from node  $i$  to a set of nodes  $D$  is *weak* if the nodes in  $D$  are more likely to default from direct shocks than through contagion from node  $i$
- Define the node-level measure  
**Contagion Index** $_i = \beta_i \times (\text{Net Worth}_i) \times (\text{Leverage of Outside Assets}_i - 1)$
- We show that contagion is weak if the Contagion Index is small, relative to the average net worth and leverage of the nodes in  $D$ .
- In other words, we bound network contagion effects using node-level information: connectivity, net worth, and leverage
- At reasonable parameter values, contagion is typically weak



# Summary

- We can bound network effects using node-level balance sheet data
- The bounds point to quantities to monitor at each node:
  - Financial connectivity
  - Contagion index
- But network effects from spillovers or domino effects are not very large at realistic parameter values
- Need other mechanisms (e.g., funding runs, fire sales) to get crisis dynamics in network models