

A General Framework for Optimization in Curbing Risk Contagion among Financial Networks

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★ RISK CONTAGION

- Financial institutions are interconnected by holding debt claims against each other.
- The interconnection is a key contributing factor to the worldwide financial crisis and the European sovereignty debt crisis.
- A default bank may cause its creditors to default, and the risk may be further propagated to up-stream institutes.

\star OBJECTIVES

\$15 E \$75 \$45

★ MARKOV DECISION PROCESS MODEL

- We propose a possible role that the CB may take in curbing contagion: arbitrating the liquidation among banks in the system during the economic crisis and providing required compensation to achieve fairness.
- Allowing different liquidation schemes, we may reduce the system's loss and save banks from defaulting.
- This problem can be formulated as a nonlinear optimization problem with equilibrium constraints:

★ MIN COST FLOW MODEL



S: source node, B: outside liability node, T: sink node

s.t. $x = \min[l, \alpha + xP]$



How to curb risk contagion? What is the role of a Central Bank? What is the optimal liquidation scheme?

★ KEY WORDS

Direct-Comparison Based Optimization, Discrete Event Systems, Financial Network, Markov Decision Problems, **Risk Contagion, Policy Iteration, Minimum Cost Flow**

★ PROBLEM FORMULATION

$$p_{i,i} = 0, p_{i,j} \ge 0, \sum_{j} p_{i,j} = 1 - b_i / l_i,$$

$$\forall i, j = 1, 2, ..., n, i \ne j.$$

- This is a Bi-Level (Leader-Follower) problem with:
 - **Non-Convex Regions**
 - **Non-Linear Constraints**
 - Large Dimensions

★ DIRECT-COMPARISON APPROACH



Policy Iteration and Gradient Based Algorithm (Resembles a Markov Decision Problem)

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41

42

Find an optimal *P**

to maximize η

50

60

- Initialization
- Policy Evaluation
- Policy Update
- Stopping Rule

 $u_{iT} = \max[\alpha_i + \sum_{j=1}^n L_{j,i} - \sum_{j=1}^n L_{i,j} - b_i, 0], i = 1, 2, ..., n$

- We can formulate this system into a Min Cost Flow model shown as the above network (G = (N, A)).
- The system should satisfy the following conditions: 1) All data (cost, demand, and capacity) are integral; 2) The flow of inside arc (i, j) satisfies: $X_{i,j} \leq L_{i,j}$; 3) The cost of inside arc (i, j) satisfies: $c_{i,j} = -1$; 4) Total demand equals total supply, i.e., $\sum_{i \in N} d(i) = 0$.
- This problem can be formulated as a linear optimization problem with equilibrium constraints:



The Min Cost Flow problem can be solved by the Network Simplex algorithm.

★ NUMERICAL EXAMPLES

Capacity





Bank *i* has an obligation $L_{i,k}$ to bank *k*, a claim $L_{i,i}$ on bank j, outside assets α_i , and outside liabilities b_i

\star CONTRIBUTIONS

- We develop a sensitivity based view of systemic risk modeling to characterize analytically how the mechanism of default liquidation affects the total wealth of the financial system.
- We put forward two formulations, i.e., the Markov Decision Process model and the Min Cost Flow model.
- We derive efficient iteration algorithms to this highly nonlinear problem.
- We demonstrate the effectiveness of our approach for 4. reducing systemic risk through examples.
- Our work provides a new direction in curbing risk contagion in financial networks.
- This work illustrates the advantages of the direct-6.

\star NUMERICAL EXAMPLES

- 3-Bank Example
- Default Set
- *D*={1, 2}, *D*'={Ø}
- Performance
- η=162.26, η'=190
- Achieves the local as well as **GLOBAL** optimal;
- Reduces **100%** of the system's loss;
- By only **ONE** iteration.
- 50-Bank Example



- Default Set
- *D*={1, 3, 4, 5}, *D*'={3, 4, 5}
- Performance
- *η=400, η'=499*
- Achieves the local as well as **GLOBAL** optimal;
- Reduces **36.95%** of the system's loss;
- By only **10** iterations.

\star REFERENCES

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• This is achieved **ONLY** by improving the liquidation scheme;







