

# Trade Credit Optimization

**Kimmo Soramäki**

**7 December 2023**



# Payment System optimization has resulted in Billions of liquidity savings

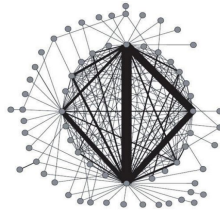
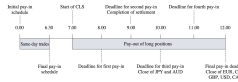
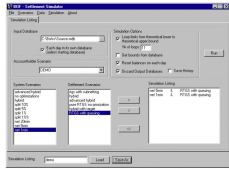
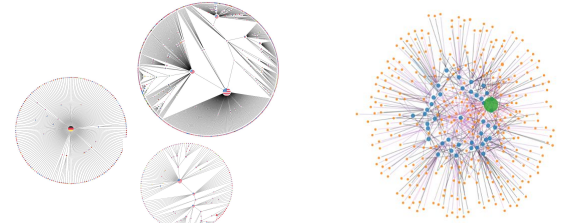
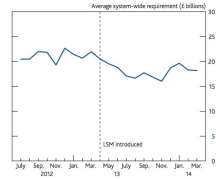


Chart 3 Average value of intraday liquidity required by CHAPS settlement banks for sterling payment systems



May - August 2022



1999

Bank of Finland developed the first Payment System Simulator to assess **liquidity impact of joining Target**

2002

CLS goes live after a **regulatory approval process**, setting a precedent of the critical role of simulations

2004-2006

FRBNY and U.S. National Laboratories develop Agent Based Model to assess **attacks on critical infrastructure**

2013

Bank of England **introduces new liquidity savings mechanisms** in CHAPS with FNA - £4b collateral savings for members

2018

First **GSIB** optimizes its group payments with FNA

2023-

First **wCBDC simulator**.  
A bottom-up approach, focused on design options, adoption, and macro financial impact.

# Key Features of the Trade Credit FMI

## Objectives

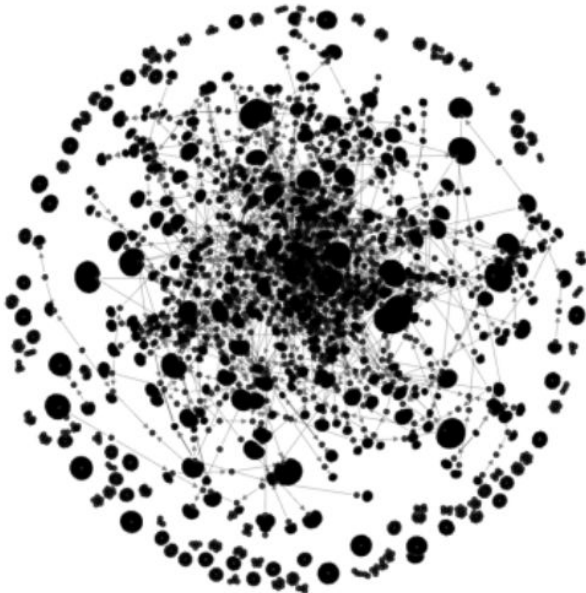
- Save 25%+ of liquidity/operating capital for firms
- Reduce substantially payment delays
- Improve economic efficiency and resiliency
- Significant downstream effects for the economy

## Design questions

- Data access
- Liquidity Saving Mechanism
- Credit Provision & Pricing
- Settlement Mechanisms (smart contracts, de/centralised, synchronization, frequency of Settlement)
- Payment Failure Recovery
- Incentives to join the FMI (banks & firms)
- Regulatory & supervisory handling

-> **Phased approach with early wins and full vision achieved in stages**

# Trade Credit / Receivables are (payment) networks that can be optimized



Using network science / graph theory we can model and optimize these networks

Example:

Trade credit data from Huangdao Zone, Shandong Province, China, sourced from Cui (2021)

8,704 obligations between firms

# Network Motifs

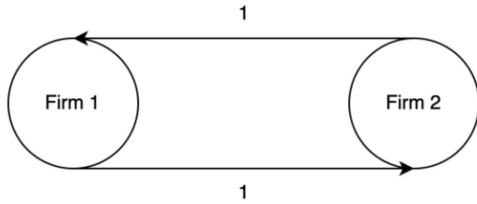


Diagram 1. Reciprocal payment flows.

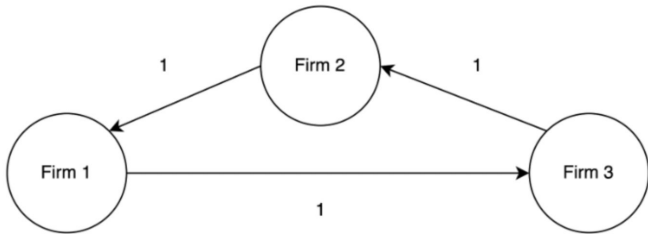


Diagram 3. Cyclical payment flows.

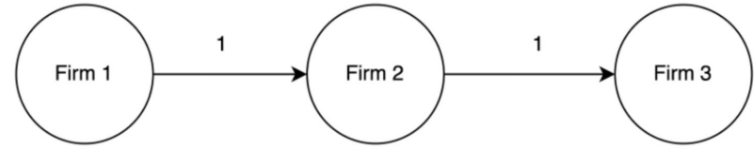


Diagram 2. Chain payment flows.

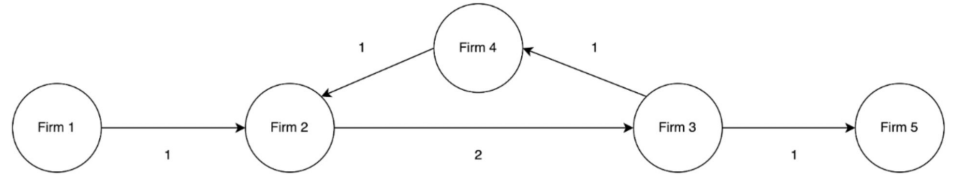


Diagram 4. Integrated payment flows. Source: Fleischman, et al. (2020)

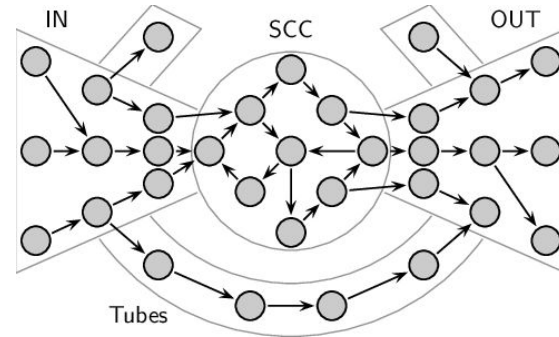


Diagram 5: Generalized bow-tie network

# Basic idea of resequencing and positive impact on liquidity

Example: We have two obligations: A to C of value 10 and C to B of value 10.  
And two orders to settle these.

Original order: C  $\xrightarrow{10}$  B and A  $\xrightarrow{10}$  C **Credit need: 20**

**Balances:**

A		-10
B	10	
C	-10	0

Resequenced order: A  $\xrightarrow{10}$  C and C  $\xrightarrow{10}$  B **Credit need: 10**

**Balances:**

A	-10	
B		10
C	10	0

**50% Reduction of credit need using Resequencing**

# Until now impossible to solve for large systems

Using brute force to find the sequence of payments which minimises liquidity is impossible

A sequence of 2 payments results in  $2! = 2$  possible sequences

A sequence of 20 payments results in  $20! = 2.4 * 10^{18}$  possible sequences

= Three quintillion four hundred quadrillion

A sequence of 200 payments results in more sequences than there are atoms in the universe

**Quantum computing can solve up to 120 payments (McMahon et al. 2022)**

**FNA Re-sequence algorithm based on Graph Optimization works in a real-time environments with up to 25 million payments in 10 minutes,**

# Large potential liquidity savings

## Huangdao Zone Credit network

Baseline CNY 1.61 Billion  
Optimal sequence with FNA: CNY 1.09 Billion

This is a **32%** liquidity saving, significant in relative terms—and potentially significant in absolute terms if extrapolated across a region or an economy.

## CHIPS interbank payment system (US)

Baseline USD 195 Billion  
CHIPS algorithms USD 68 Billion  
Optimal sequence with FNA USD 53 Billion

This is a **22%** saving to CHIPS algorithm and **73%** saving to Baseline

