

# Topology Optimization for Maximizing Feasible Wealth Volume in Payment Channel Networks

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## ABSTRACT

We investigate the problem of finding payment channel network topologies that maximize  $r(G) = |W_G|/|\mathcal{W}(C, n)|$ , the ratio of feasible off-chain wealth distributions to total on-chain distributions. Through exhaustive enumeration over all connected graphs for  $n \leq 5$  and evolutionary search for  $n = 6$ , we find that the complete graph ( $K_n$ ) maximizes  $r(G)$  for  $n \leq 4$  with  $r(K_4) = 0.441$ , while for  $n \geq 5$ , sparser topologies with moderate connectivity dominate. Across all tested configurations, cycle graphs consistently achieve the highest  $r(G)$  among standard graph families for  $n \geq 5$ , with  $r(C_6) = 0.100$  compared to  $r(K_6) = 0.001$ . The optimal edge count for  $n = 4$  is consistently 6 (complete graph) regardless of capacity, while the optimal degree sequence transitions from regular to near-regular as  $n$  grows. These findings provide practical guidance for designing payment channel networks that maximize off-chain payment feasibility.

## 1 INTRODUCTION

Payment channel networks (PCNs) such as the Lightning Network [3] enable scalable off-chain transactions. Pickhardt [2] defined  $r(G)$  as the ratio of feasible wealth distributions to all possible distributions, measuring how well a network topology supports off-chain payments. Finding the topology maximizing  $r(G)$  remains open.

This work presents the first systematic computational study of  $r(G)$  optimization, combining exhaustive enumeration, evolutionary search [1], and analytical bounds.

## 2 METHODS

### 2.1 Exhaustive Enumeration

For  $n \leq 5$  nodes, we enumerate all connected graphs on  $n$  vertices, computing  $r(G)$  for each via exact enumeration of liquidity assignments. For  $n = 4$ , this yields 38 distinct connected topologies.

### 2.2 Evolutionary Search

For  $n = 6$ , we employ an evolutionary algorithm with tournament selection, edge-flip mutation (rate 0.15), and elitism. The population of 10 connected graphs evolves over 15 generations, using  $r(G)$  as fitness.

## 3 RESULTS

### 3.1 Optimal Topologies

For  $n = 3$ : The cycle  $C_3$  ( $= K_3$ ) is optimal with  $r(G) = 0.673$ .

For  $n = 4$ : The complete graph  $K_4$  achieves  $r(G) = 0.441$ , the highest among all 38 connected graphs. The optimal degree sequence is  $[3, 3, 3, 3]$ .

For  $n = 5$ : A graph with degree sequence  $[3, 3, 2, 2, 2]$  achieves  $r(G) = 0.228$ , outperforming both  $K_5$  ( $r = 0.044$ ) and  $C_5$  ( $r = 0.202$ ).

Table 1: Best  $r(G)$  by graph family and node count (cap=3).

Family	$n = 3$	$n = 4$	$n = 5$	$n = 6$
Path	0.571	0.291	0.141	0.066
Cycle	0.673	0.385	0.202	0.100
Star	0.571	0.291	0.141	0.066
Complete	0.673	0.441	0.044	0.001
Best found	0.673	0.441	0.228	0.100

### 3.2 Edge Count Analysis

The optimal number of edges for  $n = 4$  is consistently 6 (the complete graph) across capacities 2–5, with  $r(G)$  stable at approximately 0.441. This stability suggests the optimal topology is robust to capacity variations.

### 3.3 Scaling Behavior

All graph families show decreasing  $r(G)$  with  $n$ , but the rate of decrease varies dramatically. Complete graphs decay fastest (from 0.673 to 0.001 for  $n = 3$  to 6), while cycles decay most slowly (0.673 to 0.100). This crossover between  $n = 4$  and  $n = 5$  marks a critical transition in the optimal topology structure.

## 4 DISCUSSION

The key finding is a phase transition in optimal topology: for small networks ( $n \leq 4$ ), maximum connectivity is optimal, while for larger networks ( $n \geq 5$ ), moderate connectivity preserves a higher fraction of feasible distributions. This transition occurs because the denominator  $|\mathcal{W}(C, n)|$  grows faster with total capacity  $C = |E| \cdot \text{cap}$  than  $|W_G|$  grows with additional edges.

For practical network design, cycle-like topologies with degree close to 2 offer the best feasibility-to-capacity trade-off at scale, consistent with the routing structure used in real payment channel networks [4, 5].

## 5 CONCLUSION

We identified a phase transition in the topology maximizing  $r(G)$ : from complete graphs for  $n \leq 4$  to sparser, cycle-like topologies for  $n \geq 5$ . Cycle graphs achieve the highest  $r(G)$  among standard families for larger networks. These results provide the first computational characterization of optimal PCN topologies for wealth feasibility.

## REFERENCES

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