

# Dynamic Network Analysis of Bitcoin’s Lightning Network

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## Abstract

**Keywords:** Bitcoin, Lightning Network, Temporal Network Analysis, Simulation, Network Topology, Payment Channel Network

## 1 Introduction

### 1.1 Related work

Our contribution:

## 2 Background

### 2.1 Network properties

Let  $E(t)$  and  $N(t)$  denote the number of edges and nodes at time  $t$  respectively.

### 2.2 Bitcoin and Lightning Network

## 3 Data

## 4 Experiments

### 4.1 Simulating transactions on LN

### 4.2 LN’s evolution

#### 4.2.1 Macroscopic evolution

Ever since LN had been launched its popularity steadily grew, causing its average degree increasing over time. In [1] it was shown that densification of real networks follow power-law distribution, i.e.:

$$E(t) \propto N(t)^a, \tag{1}$$

where we call  $a$  the power law densification exponent, where  $1 \leq a \leq 2$ . If  $a = 1$ , then average degree of the network is constant over time, on the other hand if  $a = 2$ , the network is an extremely dense graph, where each node has, on average, edges to a constant fraction of all nodes [1]. We found that LN is no exception to the DPL rule (4.2.1).

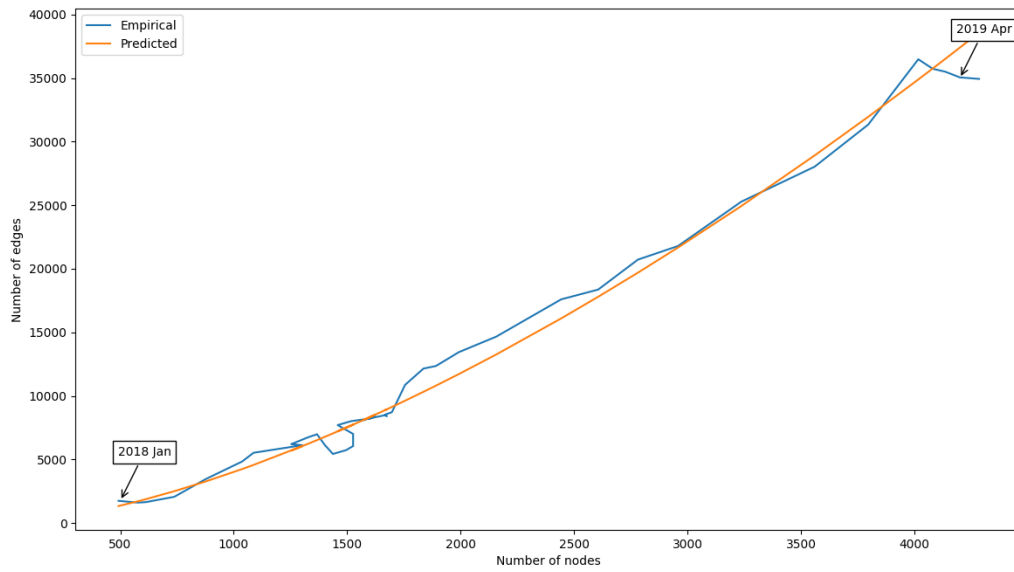


Figure 1: LN follows the Densification Power Law relation with exponent  $a = 1.55634117$ . Goodness-of-fit:  $R^2 = 0.98$ .

#### 4.2.2 Densification of LN

#### 4.2.3 Link prediction

## 5 Results

## 6 Acknowledgements

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## References

- [1] Jurij Leskovec. *Dynamics of large networks*. PhD thesis, Carnegie Mellon University, School of Computer Science, Machine Learning . . . , 2008.