

# A Petri Nets Model for Blockchain Analysis

Seminar for the course of CMCS

Luca Lombardo

*Based on the work of: Andrea Pinna, Roberto Tonelli, Matteo Orrú, Michele Marchesi*

# Structure of the Presentation

## 1 Introduction: A Petri Net Model for Blockchain Analysis

- Bitcoin Blockchain
- Petri Nets: an overview

## 2 Address Petri Net

- Constructing the Incidence Matrices
- Entities Petri Net and algorithm to manage them

## 3 Results

- Results of the Address Petri Net
- Results of the Entities Petri Net

## 4 Analysis

## 5 Conclusions

# A Petri Net Model for Blockchain Analysis

## Key Contributions

- **Addresses Petri Net (APN):** Maps Bitcoin addresses to *places* and transactions to *transitions* in a P/T Net
- **Entities Petri Net (EPN):** Groups addresses into owner entities using a clustering algorithm

## Advantages of PN Formalism

- Algebraic structure enables formal analysis of transaction patterns
- Native representation of blockchain architecture through PN semantics
- Enables dynamic simulations for network behavior forecasting

## Model Features

- Preserves transaction topology
- Supports address clustering
- Enables multi-scale analysis

## Paper Structure

- System overview
- PN model definition
- APN/EPN construction
- Blockchain analysis results

# Bitcoin Blockchain: Architecture & Transactions

## Overview

- **Distributed Public Ledger:** A global database that stores every validated Bitcoin transaction.
- **Chain of Blocks:** Transactions are grouped into blocks that form an ordered sequence (the Blockchain).

## Transaction Structure

- **Inputs:** Reference previous unspent outputs (UTXOs) from earlier transactions.
- **Outputs:** List one or more addresses (e.g., strings starting with "1" or "4") along with their associated values.
- **UTXO Model:** An address's balance equals the sum of its unspent outputs.

## User Interaction

- **Digital Wallets:** Bitcoin clients store public/private key pairs that manage one or more addresses.
- **Anonymity:** Only addresses are recorded; no personal identity information is required.

# Bitcoin Blockchain: Architecture & Transactions

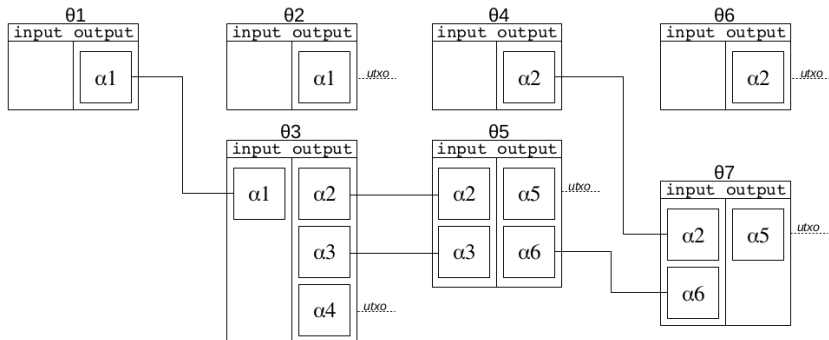


Figure: Simplified transaction schema

## Transaction Validation

- **Broadcast & Pending:** Transactions are shared in the peer-to-peer network and wait to be validated.
- **Validation Rules:**
  - Each input must reference a valid UTXO.
  - The total value of outputs must not exceed that of inputs.

## Mining Process

- **Proof-of-Work:** Miners solve a computational puzzle by finding a hash.
- **Block Contents:** A block includes the selected transactions, the previous block's hash, its block height, and miner information.
- **Incentives:** The first miner to solve the puzzle receives a block reward

## Overview

A **Petri Net** is a formal model for distributed systems built on a bipartite graph consisting of:

- **Places:** Represent conditions, resources, or system states.
- **Transitions:** Represent events that change these states.

## Graph Structure

- **Bipartite Graph:** Only two types of nodes are allowed, and connections occur only between nodes of different types.
- **Arcs:** Directed arcs link places and transitions:
  - *Pre-arcs:* Arcs from places to transitions (inputs).
  - *Post-arcs:* Arcs from transitions to places (outputs).

## Algebraic Description

A Petri Net is formally defined as a quadruple:

$$N = (P, T, Pre, Post)$$

where:

- $P = \{p_1, p_2, \dots, p_m\}$  is the set of places.
- $T = \{t_1, t_2, \dots, t_n\}$  is the set of transitions.
- $Pre : P \times T \rightarrow \mathbb{N}$  is the *pre-incidence* function.
- $Post : P \times T \rightarrow \mathbb{N}$  is the *post-incidence* function.

These incidence functions are typically represented as  $m \times n$  matrices.



## Markings & Firing Rule

- A **marking**  $M$  is a vector assigning tokens to places, thus representing the system state.
- **Firing:** When a transition fires, it:
  - ① Consumes tokens from its pre-connected places.
  - ② Produces tokens in its post-connected places.
- The complete system is denoted as  $\langle N, \mathbf{M}_0 \rangle$ , where  $\mathbf{M}_0$  is the initial marking.
- In our work on Blockchain analysis, we focus on the net structure, without defining a specific marking.

# Addresses Petri Net: Overview and Definitions

## Motivation

- Blockchain transactions move bitcoins between addresses.
- The inherent bipartite structure (addresses and transactions) suggests a natural mapping to a Petri Net.

## Definitions

- $\mathcal{A} = \{\alpha_1, \alpha_2, \dots, \alpha_m\}$ : Finite set of addresses (inputs/outputs).
- $\Theta = \{\theta_1, \theta_2, \dots, \theta_n\}$ : Set of validated transactions.

## Addresses Petri Net Structure

- Define  $N_\alpha = (P_\alpha, T, \mathbf{PreA}, \mathbf{PostA})$  where:
  - $P_\alpha = \{p\alpha_1, p\alpha_2, \dots, p\alpha_m\}$  associates one place per address.
  - $T = \{t_1, t_2, \dots, t_n\}$  associates one transition per transaction.
  - **PreA** and **PostA** are the pre- and post-incidence matrices.
- These sets are built by scanning the Blockchain for new addresses and transactions.

# Addresses Petri Net: Constructing the Incidence Matrices

## Transaction Representation

- Each transaction  $\theta$  is split into:
  - $\mathbf{In}(\theta) \subseteq \mathcal{A}$ : Set of input addresses.
  - $\mathbf{Out}(\theta) \subseteq \mathcal{A}$ : Set of output addresses.
- The corresponding transition  $t$  represents  $\theta$  in the Petri Net.

## Matrix Construction

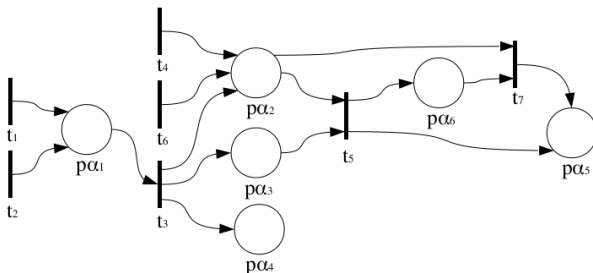
- For every  $\alpha \in \mathbf{In}(\theta)$ :
  - Add a **pre-arc** from place  $p_\alpha$  to transition  $t$ .
  - Set  $\mathbf{PreA}(p_\alpha, t) = 1$ .
- For every  $\alpha \in \mathbf{Out}(\theta)$ :
  - Add a **post-arc** from transition  $t$  to place  $p_\alpha$ .
  - Set  $\mathbf{PostA}(p_\alpha, t) = 1$ .
- The matrices  $\mathbf{PreA}$  and  $\mathbf{PostA}$  are of dimension  $m \times n$ .

# Addresses Petri Net: Example and Analysis

## Example Overview

- Consider a simplified set of transactions (e.g. as in the previous slide)
- The Addresses Petri Net has:
  - 6 places ( $P_\alpha = \{p\alpha_1, \dots, p\alpha_6\}$ ).
  - 7 transitions ( $T = \{t_1, \dots, t_7\}$ ).

Below a graphical representation of an addresses Petri Net equivalent to the simplified transaction chains.



## Incidence Matrices and Analysis

- **Pre-incidence matrix  $\text{PreA}$ :** Captures the number of times an address appears as an input.
- **Post-incidence matrix  $\text{PostA}$ :** Captures the outputs associated with each transaction.
- By computing the difference ( **$\text{PostA} - \text{PreA}$** ) for each row, one can:
  - Determine the number of UTXOs per address.
  - Infer whether an address balance is null (zero tokens).
- Transactions sharing identical input and output sets may be merged into one transition with a firing clock, aiding dynamic analyses.

# Pre-incidence matrix of the simplified transaction

$$\mathbf{PreA} = \begin{matrix} & \begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} & \begin{matrix} p\alpha_1 \\ p\alpha_2 \\ p\alpha_3 \\ p\alpha_4 \\ p\alpha_5 \\ p\alpha_6 \end{matrix} \end{matrix}$$

$t_1 \quad t_2 \quad t_3 \quad t_4 \quad t_5 \quad t_6 \quad t_7$

**Figure:** Pre-incidence matrix of the Petri net for the example of the simplified transaction schema.

# Post-incidence matrix of the simplified transaction

$$\mathbf{PostA} = \begin{matrix} & \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} & \begin{matrix} p\alpha_1 \\ p\alpha_2 \\ p\alpha_3 \\ p\alpha_4 \\ p\alpha_5 \\ p\alpha_6 \end{matrix} \end{matrix}$$

$t_1 \quad t_2 \quad t_3 \quad t_4 \quad t_5 \quad t_6 \quad t_7$

**Figure:** Post-incidence matrix of the Petri net for the example of the simplified transaction schema.

# Entities Petri Net: Introduction

## Motivation

- Bitcoin users often control multiple addresses to manage exchanges and preserve anonymity.
- We define an **entity** as the person, organization, or group that controls a set of addresses.
- **Key Property:** All addresses appearing in the input of a single transaction must belong to the same entity (since transferring funds requires control over all associated private keys).

## Why Group Addresses?

- Enables high-level analysis of the Blockchain.
- Reduces complexity by clustering addresses that are likely controlled by the same user.
- Provides a more natural mapping to real-world economic actors.



# From Addresses to Entities: Definitions & Mapping

## Mapping Addresses to Entities

- Let  $\mathcal{A} = \{\alpha_1, \alpha_2, \dots, \alpha_m\}$  be the set of addresses.
- Let  $\Theta = \{\theta_1, \theta_2, \dots, \theta_n\}$  be the set of transactions.
- In the Addresses Petri Net  $N_\alpha = (P_\alpha, T, \mathbf{PreA}, \mathbf{PostA})$ , each place  $p_\alpha$  corresponds to an address  $\alpha$ .

## Defining an Entity

- An **entity**  $\epsilon$  is a set of addresses:  $\epsilon \subseteq \mathcal{A}$ .
- Denote the set of entities as  $E = \{\epsilon_1, \epsilon_2, \dots, \epsilon_k\}$ .
- In the **Entities Petri Net**  $N_\epsilon = (P_\epsilon, T, \mathbf{PreE}, \mathbf{PostE})$ , each place  $p_\epsilon$  represents one entity.

## Algorithm 1 Compute Entities from the Addresses Petri Net

```
1:  $T^* \leftarrow T$  ▷ Set of unexplored transactions
2:  $E \leftarrow \emptyset$  ▷ Set of entities
3: while  $T^* \neq \emptyset$  do
4:   Select and remove a transaction  $t$  from  $T^*$ 
5:    $e \leftarrow \emptyset$  ▷ Initialize new entity
6:   for all  $p_i$  such that  $\mathbf{PreA}(p_i, t) = 1$  do
7:      $e \leftarrow e \cup \{p_i\}$ 
8:   end for
9:    $e^* \leftarrow e$  ▷ Set of unexplored places within  $e$ 
10:  while  $e^* \neq \emptyset$  do
11:    Select a place  $p$  from  $e^*$ 
12:    for all transactions  $t'$  with  $\mathbf{PreA}(p, t') = 1$  do
13:      for all  $p_h$  such that  $\mathbf{PreA}(p_h, t') = 1$  do
14:         $e \leftarrow e \cup \{p_h\}$ ,  $e^* \leftarrow e^* \cup \{p_h\}$ 
15:      end for
16:      Remove  $t'$  from  $T^*$ 
17:    end for
18:    Remove  $p$  from  $e^*$ 
19:  end while
20:   $E \leftarrow E \cup \{e\}$ 
21: end while
```

# Entities Petri Net: Incidence Matrices and Aggregation

## Constructing the Entities Petri Net

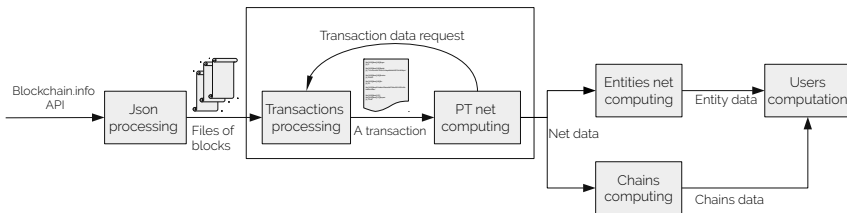
- Each computed entity  $e \in E$  is represented by a unique place  $p_e$  in the Entities Petri Net  $N_e = (P_e, T, \mathbf{PreE}, \mathbf{PostE})$ .
- The set of transitions  $T$  remains the same as in the Addresses Petri Net.

## Aggregating Incidence Matrices

- For each entity  $e$ , identify all corresponding addresses  $p_\alpha \in e$  from the Addresses Petri Net.
- **PreE** and **PostE** are obtained by summing the rows in **PreA** and **PostA** for all places in  $e$ .
- This aggregation captures the cumulative input and output interactions of all addresses within the entity.

# Data Acquisition and Processing Pipeline

- **Approach:** Downloaded formatted JSON blocks from *blockchain.info*.
- **Dataset:** Parsed the first 180,000 blocks (Jan 2009 – Mar 2012)
- **Implementation:** Data processing executed in R using RStudio IDE.
- **Performance:** Total processing time  $\approx$  250 hours (avg. 5 sec per block).



# Addresses Petri Net: Global Statistics and CCDF Analysis

- **Addresses:** 3,730,480 distinct addresses.
- **Transactions:** 3,142,019 transactions (columns in **PreA**/**PostA**).
- **Arcs:** 4,575,888 pre-arcs and 7,352,494 post-arcs.
- The number of nonzero elements in a row of **PreA** (or **PostA**) indicates the number of input (or output) transactions for that address.
- CCDF plots reveal a power-law distribution: many addresses with few transactions and a few addresses with very high activity.

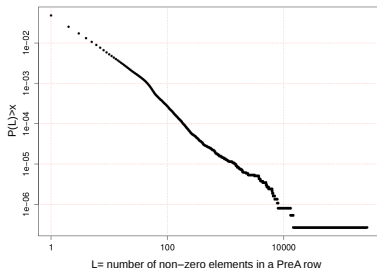


Figure: CCDF of  $L$  for **PreA**.

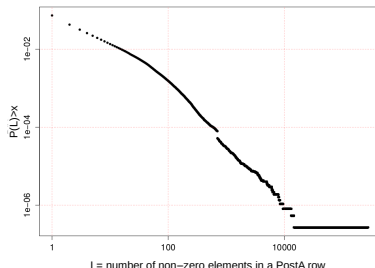


Figure: CCDF of  $L$  for **PostA**.

# Addresses Petri Net: Most Used and Imbalanced Addresses

## Usage Ranking

The most used addresses are identified by summing the nonzero elements in both **PreA** and **PostA** rows.

## Imbalanced Addresses

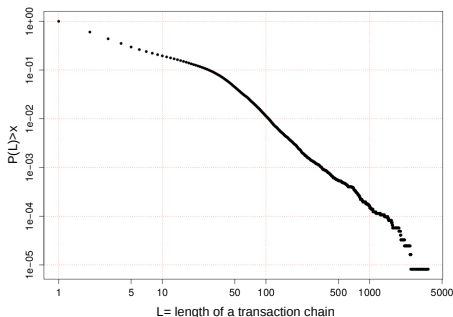
609,295 addresses show zero entries in **PreA** but at least one nonzero in **PostA**. These addresses have never been used to spend bitcoins but only to receive them. Table 1 lists the top 5 addresses by incoming transactions along with their current balances.

Address	L post	current balance BTC
15S1TFTosxrgZxkqJR2n1AFJ22ZJE2rTck	3,853	120.85215349
1PtnGiNvhAKbuUQ6nZ7nF3CDKCKGfeMsCX	1,199	0
129FTwWoi5H5ujasMZ6M6VjJzBJfsXVQGw	1,138	0.78425567
1FN9kKsZA9XttrAwuDDgsXjs6CXUR2fzmt	1,111	0
1DYvtKtZ2Ay9vTjzjb9BiRauMgXdjRDaD	973	14.5601

**Table:** Summary of first 5 most imbalanced addresses

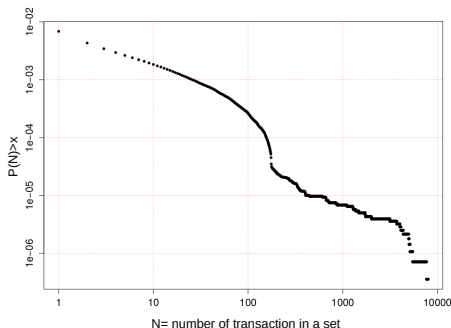
# Addresses Petri Net: Disposable Address Chains

- **Disposable Addresses:** Defined as addresses used exactly twice (once to receive and once to send all bitcoins).
- Transactions involving disposable addresses typically have:
  - One pre-arc (single input) and two post-arcs (output to two distinct addresses).
- The model easily identifies these by analyzing **PreA** and **PostA**.
- 122,155 disposable address chains have been detected, involving 1,350,010 different addresses/transactions.



# Addresses Petri Net: Repeated Transaction Patterns

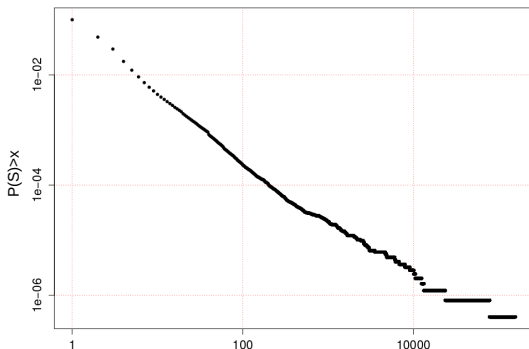
- **Repeated Transfers:** Users sometimes execute transactions with identical sets of input and output addresses.
- In the Petri Net model, such repetitions appear as transitions with identical pre- and post-arc configurations.
- Approximately 11% of transactions are repetitions, reflecting steady bitcoin flows between fixed groups of addresses.
- The CCDF of grouped transaction sizes (i.e., the number of repetitions) illustrates this phenomenon.





# Entities Petri Net: Address Distribution and Entity Control

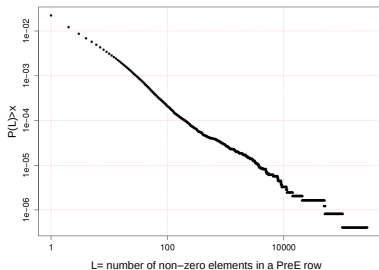
- **Entities and Addresses:** 2,461,010 entities control 3,730,480 addresses.
- Distribution is highly non-uniform and follows a power-law pattern:
  - Many entities hold a single address.
  - Few entities control a large number of addresses and influence a significant portion of bitcoin transactions.



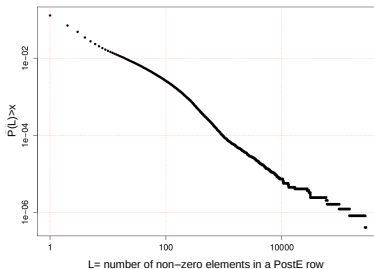
S = size of a set of addresses

# Entities Petri Net: Transaction Distributions for Entities

- **Transaction Involvement:** The number of non-zero elements in the rows of **PreE** and **PostE** indicates how many transactions an entity is involved in.
- **Power-law Distribution:** Transactions among entities show a power-law behavior for both input and output transactions.



**Figure:** CCDF of the length  $L$  for **PreE**.



**Figure:** CCDF of the length  $L$  for **PostE**.

# Entities Petri Net: Most Active Entities

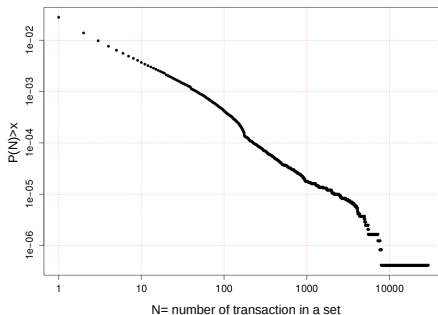
- **Top Entities:** The 5 most active entities are identified by the sum of non-zero elements in both **PreE** and **PostE**.
- Their balances are calculated by summing the balances of all addresses belonging to each entity.
- **Tags:** Some entities are associated with known tags (e.g., deepbit.net, ilovethebtc).

Entity number	L pre	L post	size	tags
95237	270,204	275,398	2	deepbit.net
2	102,186	283,973	156,725	ilovethebtc
37	51,228	147,712	78,251	jmm5699
11	49,959	97,732	10,37	- unknow -
130	20,857	58,350	23,649	Instawallet

Table: Summary of first 5 most active entities

# Entities Petri Net: Repeated Transaction Patterns

- **Repeated Transactions:** About 22.6% of transactions are repetitions of previous transactions between the same input and output entities.
- These repetitions indicate steady fluxes of bitcoins at the entity level.
- **Visualization:** The CCDF for grouped transaction sizes demonstrates this repetitive behavior.



**Figure:** CCDF of the size  $L$  of grouped transaction sets for the Entities Petri Net.

# Analyzing Bitcoin Users with Petri Nets

## Petri Net Model

Clusters Bitcoin addresses into entities (246,660 owners control 1.5M addresses) and traces transaction chains (122,155 owners linked to 1.35M addresses).

## User Classification

368,815 engaged owners use disposable addresses or multiple addresses (72.6% of transactions). 609,295 addresses are likely "deposit addresses" for engaged users. 255,045 addresses are owned by occasional users.

## Case Studies

Most active input address (270k transactions): *DeepBit* mining pool (now defunct).  
Largest output entity (156k addresses): Tags include *ilovethebtc*, *mikeo*, and links to unreachable domains.

## Insight

Strongly uneven transaction distribution (CCDFs) suggests miner pools use high-activity addresses for reward redistribution.

# Advantages of Petri Net Formalism

## Non-Deterministic Dynamics

Models UTXO-enabled transactions as independent events. Natively handles probabilistic block inclusion (miner fees, validation).

## Simulation Power

Captures concurrent non-conflicting transactions (e.g., same-block validations). Enables statistical modeling of future states (e.g., miner pool growth predictions).

## Sequential Analysis

Automatically identifies chains of disposable addresses (privacy-preserving flows).

## Matrix-Driven Insights

Pre/Post matrices reveal never-spent addresses (609k output-only), single-use addresses, and transaction volumes.

## State Equations

Enable probabilistic forecasting of Bitcoin fluxes between entities (exchanges, pools) using historical data.

# Conclusions

## Novel Petri Net Architecture

Processed first 180,000 blocks ( $\approx 3.5$  years of Bitcoin history).

Addresses  $\rightarrow$  Places (2.7M+), Transactions  $\rightarrow$  Transitions.

Pre/Post matrices capture all input-output relationships.

## Key Discoveries

Universal power-law distributions were observed in transaction arcs (pre/post connections), disposable address chain lengths, and repeated input-output transaction clusters.

Entity Network Reconstruction: Built a secondary Petri Net of address owners through matrix analysis.

## Computational Considerations

The current blockchain (480k+ blocks, 240M+ transactions) challenges full-scale analysis.

Flexible partial analysis allows investigation of specific address subsets or time windows.