

Combinatorial Optimization for Self Contained Blockchain: An Example of Useful Synergy

Presented by: Milan Todorović

Mathematical Institute of the Serbian Academy of Sciences and Arts

Seminar Decision making - Theory, technology and practice
Oct. 07, 2021



Acknowledgements

Authors: Dušan Ramljak (PSU), Tatjana Davidović, Dragan Urošević, Tatjana Jakšić Krüger, Luka Matijević, Milan Todorović, Đorđe Jovanović (MISANU)

- Grant AI4TrustBC: Advanced Artificial Intelligence Techniques for Analysis and Design of System Components Based on Trustworthy Blockchain Technology
- Miloš Simić, University of Belgrade
- Nouf Albarakati, Temple University
- West Texas A&M University, College of Engineering

Preliminary version presented at SYM-OP-IS 2021.



Overview

- 1 Blockchain background
- 2 Consensus Protocol and Deanonymization
- 3 Clustering
- 4 Proposed consensus protocol
- 5 Concluding remarks



Blockchain (BC) overview

- Represents a special type of distributed data storage (ledger)
- BC types:
 - Public (permissionless)
 - Private (permissioned)
- Users (nodes, agents, participants, entities) can have different roles
- Applications: cryptocurrency, smart contracts, IoT, voting, DRM, healthcare, etc.



BC maintenance

- Issues:
 - Security of data and users (Majority attack, DDoS, Eclipse attack, etc.)
 - Users' privacy (data is public, user's identity is hidden)
 - Data consistency
 - Wasteful usage of resources
- Our focus: energy efficiency and deanonymization
- We propose a self contained BC based on:
 - Proof-of-Useful-Work concept
 - Community detection in directed graphs



BC background

- Autonomous (unsupervised) append-only distributed data storage
- Autonomous: the removal of external authority (third party)
- Data is submitted in a form of (append-only) transactions
- Consensus protocol controls transaction additions
- Distributed: each participant has a copy of the whole database
- Data is stored in form of blocks of transactions

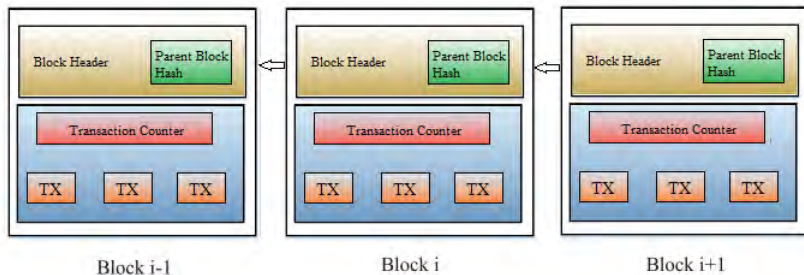


Block structure

- Data is stored in form of blocks of transactions
- Blocks consist of header and body
- Header includes:
 - Block number
 - Hash value of the previous block
 - Timestamp
 - Validity code (nonce), etc.
- Body contains transactions data



An illustration of BC



Procedure of appending a block

- Transactions are submitted by users to transaction pool
- Special users of the BC (miners):
 - Form a block by selecting transactions from the pool
 - Execute consensus protocol
 - Publish the new block
- Block validity is checked
- Valid blocks are appended to BC
- Potential forks are resolved periodically
- Successful miners receive reward



Consensus protocol

- Enables verification of transactions and blocks
- Eliminates the central authority
- Involves certain overhead
- Accepts or rejects each transaction from the pool
- Must be followed by all users of BC
- Ensures the agreement of all users on a common set of valid blocks



Types of consensus protocols

- Proof-of-Work (PoW) - inversion of hash function
- Proof-of-Stake (PoS) - cryptocurrency amount guarantees trustworthiness
- Proof-of-Useful-Work (PoUW) - swaps PoW task with real-world problems
- Other consensus algorithms:
 - Proof-of-Capacity
 - Proof-of-Elapsed-Time
 - Proof-of-Burn, etc.



Proof-of-Useful-Work (PoUW)

- Proof-of-Useful-Work (PoUW) is a novel BC consensus protocol
- PoUW supports BC and contributes to solving real-world problems
- PoUW:
 - Improves efficiency of BC
 - Preserves security of BC provided by PoW
 - Provides users' equity missing in PoS
- Two sources of reward for miners:
 - Mining a block
 - Solving a problem instance
- Combination of PoW and PoUW (hybrid)



PoW Optimization Insights

- PoW involves optimization problems:
 - Travelling Salesman Problem
 - Artificial Neural Network training
 - Medical Image Processing, etc.
- Optimization problems can be addressed by:
 - Exact methods (Branch-and-... , CPLEX, Gurobi, etc)
 - Problem-specific heuristics
 - Metaheuristics (VNS, BCO, etc.)
- Our idea: Useful work of self contained BC involves its maintenance



Pseudoanonymity vs. Deanonymization

- Pseudoanonymity: pseudonyms are known, identities should be hidden
- Transactions are made via e-wallet
- Users can generate a new e-wallet public key for every transaction
- Only public key of e-wallet available for everyone to see
- Anonymity is desirable for users but may hide malicious behaviour
- Deanonymization involves:
 - Defining similarity measures
 - Identifying patterns
 - Grouping transactions
- Clustering problem can model Deanonymization process



Clustering problem

- Helps to discover patterns in observed data
- Cluster analysis determines the quality of clustering
 - Could be incorporated in method (agglomerative, hierarchical, etc.)
 - Done separately (centroid-based, distribution-based, density-based, etc.)
- In clustering of graph structured data similarity depends on:
 - Graph topology
 - Location of the nodes
 - Weights on nodes or edges, etc.



Community detection

- Community detection in case of graph-based clustering
 - Incorporates cluster analysis
 - Reveals underlying community structure
- Patterns of community detection involve
 - More interaction among nodes in the same community
 - Less interaction between the nodes that are in different communities
 - Edge directions contain useful information (e.g. sender/receiver)
- Set of BC transactions can be modeled as directed graph
- Community detection in directed graphs could be useful part of PoUW

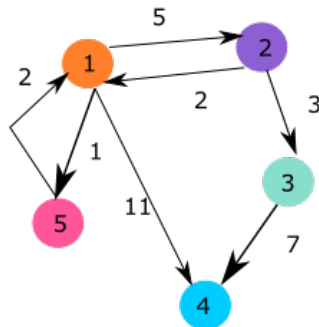


List of transactions

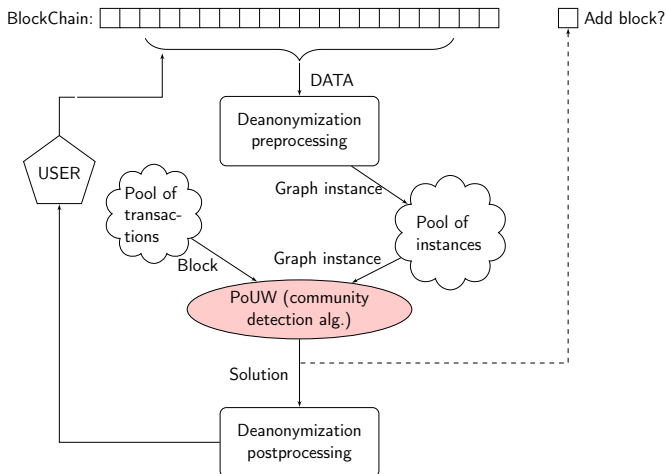
Tx1: Addr1 -> Addr2 (5)
 Tx2: Addr2 -> Addr1 (2)
 Tx3: Addr2 -> Addr3 (3)
 Tx4: Addr3 -> Addr4 (7)
 Tx5: Addr1 -> Addr4 (11)
 Tx6: Addr1 -> Addr5 (1)
 Tx7: Addr5 -> Addr1 (2)



Transaction graph



Methodology



Formulation of Community Detection Problem: Parameters

Given is a graph $G = (V, E, W)$.

Vertices: $V = \{1, 2, \dots, n\}$,

Arcs: $E \subseteq V^2, |E| = m$

$W : E \rightarrow R_+$ weights on arcs; $w_{i,j}$ is weight of arc (i, j) .

C denotes the maximum number of communities

d_i^{in} , $i = 1, 2, \dots, n$ - number of arcs ending in i

d_i^{out} , $i = 1, 2, \dots, n$ - number of arcs beginning at i



Formulation of Community Detection Problem: Variables

$$x_{il} = \begin{cases} 1, & \text{if node } i \text{ is assigned to community } \ell, \\ 0, & \text{otherwise.} \end{cases}$$

$$y_{i,j,\ell} = \begin{cases} 1, & \text{if nodes } i \text{ and } j \text{ are assigned to the same community } \ell, \\ 0, & \text{otherwise.} \end{cases}$$

$$x_{il}, y_{i,j,\ell} \in \{0, 1\}, \quad 1 \leq i, j \leq n, \quad 1 \leq \ell \leq C.$$

Introduce y variables to linearize problem formulation, i.e. $y_{i,j,\ell} = x_{il} \cdot x_{jl}$



Formulation of Community Detection Problem: Objective function

Mixed Integer Linear Programming formulation maximizes the *modularity* Q defined as:

$$Q = \frac{1}{m} \sum_{\ell=1}^C \sum_{i,j=1}^n \left[w_{ij} - \frac{d_i^{in} d_j^{out}}{m} \right] y_{ij\ell} \quad (1)$$

Formulation of Community Detection Problem: Constraints

$$y_{ijl} \leq x_{il}, \quad 1 \leq i, j \leq n, \quad 1 \leq l \leq C \quad (2)$$

$$y_{ijl} \leq x_{jl}, \quad 1 \leq i, j \leq n, \quad 1 \leq l \leq C \quad (3)$$

$$y_{ijl} \geq x_{il} + x_{jl} - 1, \quad 1 \leq i, j \leq n, \quad 1 \leq l \leq C \quad (4)$$

$$\sum_{l=1}^C x_{il} = 1, \quad 1 \leq i \leq n, \quad (5)$$

$$2 \leq \sum_{i=1}^n x_{il} \leq n - 2(C - 1), \quad 1 \leq l \leq C, \quad (6)$$

$$x_{il} \leq \sum_{j=1}^n x_{jl-1}, \quad 1 \leq i \leq n, \quad 1 \leq l \leq C \quad (7)$$

Basic Variable Neighborhood Search Algorithm

procedure BVNS(Problem input data, k_{max} , STOP)

$x \leftarrow \text{InitSolution}()$

$x_{best} \leftarrow x$

repeat

$k \leftarrow 1$

repeat

$x' \leftarrow \text{RandomSolution}(x_{best}, \mathcal{N}_k)$

$x'' \leftarrow \text{LS}(x')$

if ($f(x'') < f(x_{best})$) **then**

$x_{best} \leftarrow x''$

$k \leftarrow 1$

else

$k \leftarrow k + 1$

end if

$terminate \leftarrow \text{StoppingCriterion}(\text{STOP})$

until ($k > k_{max} \vee terminate$)

until ($terminate$)

return ($x_{best}, f(x_{best})$)

end procedure

▷ Shaking

▷ Local Search

▷ Move or Not (Neighborhood Change)



VNS-based approach to Community Detection

Solution representation

$$Sol = [s_1, s_2, \dots, s_n]$$

$Sol[i] = s_i$, $s_i \in \{1, 2, \dots, c\}$, c is current number of communities

Initial solution is determined randomly

$$Sol[i] = rand(1, C), i = 1, 2, \dots, n$$

Neighborhood: move a node from the current community to some other (including $c + 1$).

If community ℓ is emptied c decreases.

Explored systematically once (at distance 1) in LS.

Performed randomly k times in Shaking.



Auxiliary data - reducing the complexity of calculating Q

Modularity matrix $B_{n \times n}$

$$B_{i,j} = A_{i,j} - \frac{d_i^{\text{in}} d_j^{\text{out}}}{2m}.$$

Truth matrix $T_{n \times n}$

$$T_{i,j} = \delta(\text{Sol}[i], \text{Sol}[j]), \quad \delta(\text{Sol}[i], \text{Sol}[j]) = \begin{cases} 1, & \text{if exists } \ell \text{ s.t. } y_{i,j\ell} = 1 \\ 0, & \text{otherwise} \end{cases}$$

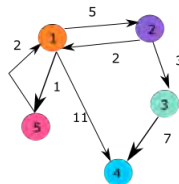
Q is determined for initial solution, and then modified by at most $2n - 2$ operations (instead of n^2).



List of transactions

Tx1: Addr1 -> Addr2 (5)
 Tx2: Addr2 -> Addr1 (2)
 Tx3: Addr2 -> Addr3 (3)
 Tx4: Addr3 -> Addr4 (7)
 Tx5: Addr1 -> Addr4 (11)
 Tx6: Addr1 -> Addr5 (1)
 Tx7: Addr5 -> Addr1 (2)

Transaction graph



Adjacency matrix

	2			2
5				
	3			
11		7		
1				

Modularity matrix

		1.03			
4.03					
		2.03			

Solution

1	1	2	2	1
---	---	---	---	---

General Issues/Future work

- Correspondence between problem instance and the composed block
- The efficient exploration of miners' hardware
- Format for stating optimization problem instances
- Increasing the efficiency by parallelization (pool of miners)



Summary and conclusion

- We considered energy efficiency and security in BC maintenance
- Energy consumption is directed to optimization problems by PoUW
- We selected MAX-SAT as a useful optimization problem
- Security problem (deanonymization) can be modeled by the community detection in directed graphs
- We combined solution approaches into a single procedure:
- Deanonymization represents the useful part of PoUW
- The resulting BC becomes self contained framework



Thank you for your attention!

Questions?

Milan Todorović
mtodorovic@mi.sanu.ac.rs