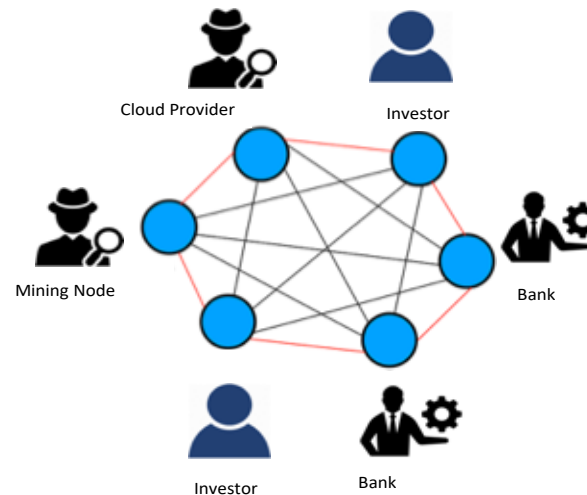




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A Reputation Management Framework for Knowledge-Based and Probabilistic Blockchains

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Tara Salman, Raj Jain, and Lav Gupta, "A Reputation Management Framework for Knowledge-Based and Probabilistic Blockchains," IEEE 1st International Workshop on Advances in Artificial Intelligence for Blockchain (AIChain 2019), held in conjunction with the 2019 IEEE International Conference on Blockchain, Atlanta, July 14, 2019, <http://www.cse.wustl.edu/~jain/papers/rpmcewa.htm>.

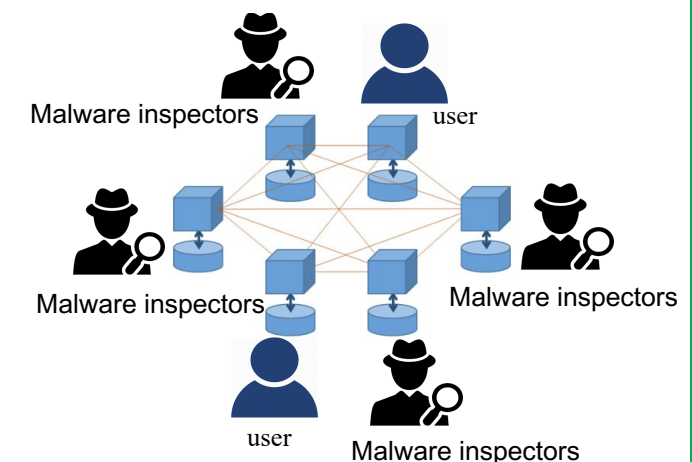


Outline

- ❑ Knowledge-based and probabilistic blockchains
- ❑ Reputation management requirements
- ❑ Our approach
- ❑ Application to malicious node detection
- ❑ Limitations and work extensions

Blockchains and Knowledge-based Blockchain

- ❑ Blockchains are distributed, secure, and immutable systems
- ❑ Getting popular in building distributed AI and decision making applications → *Knowledge-based blockchains*
 - Transform the chain from a data storage to a processing and decision making platform
- ❑ Example: Malware detection system
 - Multiple malware inspectors analyze a file submitted by a user. The blockchain returns a “consensus” decision about that malware



A malware detection system example



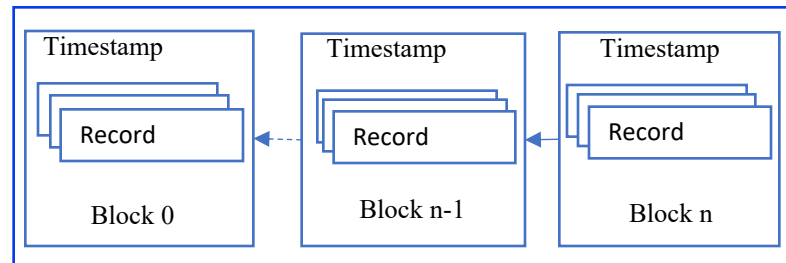
Need for Blockchain Extensions

- ❑ Knowledge-based blockchains require:
 1. Transactions → store probabilistic decisions
 2. Blockchains should be able to summarize decisions and achieve consensus
 3. Consensuses should be visible to others and possibly updated whenever needed
- ❑ Challenge
 - Current blockchain systems cannot efficiently do that
- ❑ Solution:
 - Extend blockchain systems to meet these requirements
 - *probabilistic blockchain*



Probabilistic Blockchain

Traditional chain

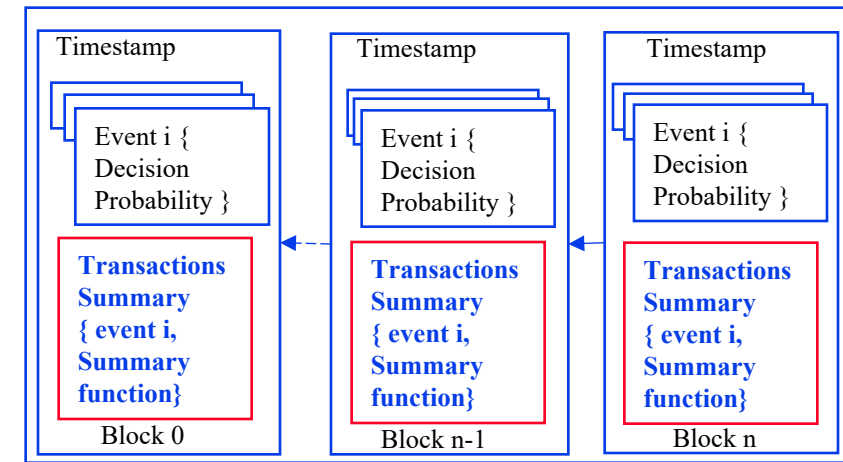


Users broadcast transactions or contracts (Records)

Miners validate transactions and generate blocks

Blockchain nodes validate the block and construct the chain

Probabilistic chain



Users broadcast transactions (**Decisions/opinions**)

Miners validate transactions, **create a knowledgeable summary** and generate blocks

Blockchain nodes validate the block and construct the chain



The Need for Agent Reputation Framework

- ❑ The challenge is that **not all agents contributing to the group decision** may be treated equally
 - Agents contribution to the collaborative decision should be based on their past performance

- ❑ **A reputation management framework is needed** to assign reputations to agents
 - Good decisions → Reputation close to 1
 - Bad decisions → Reputation close to 0
 - Just starting → Reputation is 0.5



Reputation Function Requirements

1. Time-dependent

- A mistake in the past should not be treated the same as a recent one

2. Configurable increase and decrease

- Suit different applications

3. Proportional update

- The decrease after a wrong decision is proportional to how good an agent has performed so far; thus, provide fairness



Reputation Function

- Our work extends Exponentially Weighted Average function (EWA) to **Rated Proportional Multi-Configurable EWA (RPMC-EWA)**
- EWA does not meet **configurable increase and decrease and proportional update**
- p is the number of correct decisions, n is the number of wrong decisions before time t , α is configurable increase, β configurable decrease

$$R_{i,t} = \begin{cases} 0 & t = 0 \\ \alpha X + (1 - \alpha)R_{i,t-1} & t \neq 0 \end{cases}$$

EWA

$$R_t = \frac{\begin{cases} 0 & p=0, n=0, t=0 \\ \alpha \frac{p}{p+n} + (1-\alpha)(2R_{t-1}-1) & t>0, \text{ correct decision} \\ \beta \frac{-n}{p+n} + (1-\beta)(2R_{t-1}-1) & t>0, \text{ wrong decision} \end{cases}}{2} + 0.5$$

RPMC-EWA



Reputations in Probabilistic Blockchains

- ❑ How it is done?
 - Miner get latest reputations and use it for calculating the consensus
 - Validators recalculate the consensus with their latest reputations and use it for validating
 - Everyone update the reputation value based on the last decision made

- ❑ Applications
 - Malicious node detection
 - Efficient collaborative decision approach



Application to Malicious Node Detection

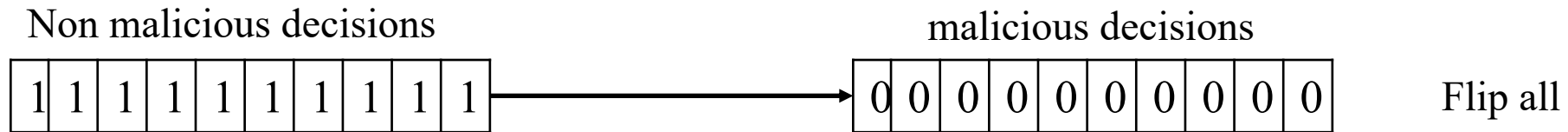
- ❑ Evaluate the effect of malicious node detection on the collaborative decision in probabilistic blockchains
 - Summary function for probabilistic blockchains: mean

- ❑ Setup:
 - Ten agents are participating with one malicious.
 - Honest nodes give 1 as a decision while a malicious node flips the decision.

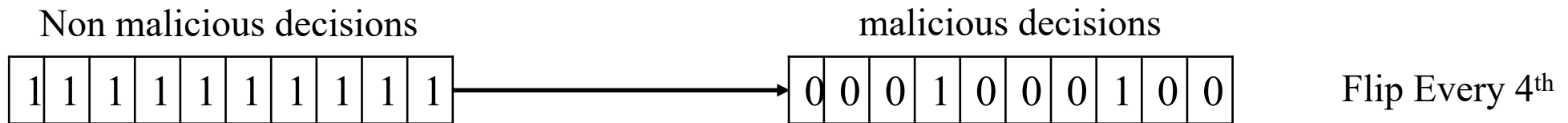
- ❑ Metric:
 - When the consensus decision return back to 1 (i.e., malicious node with a reputation of less than 0.5 is eliminated)

Malicious Node Strategies

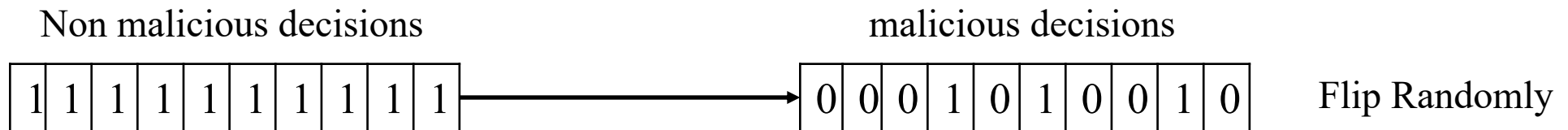
Continuous-flipping strategy



Pattern-flipping strategy

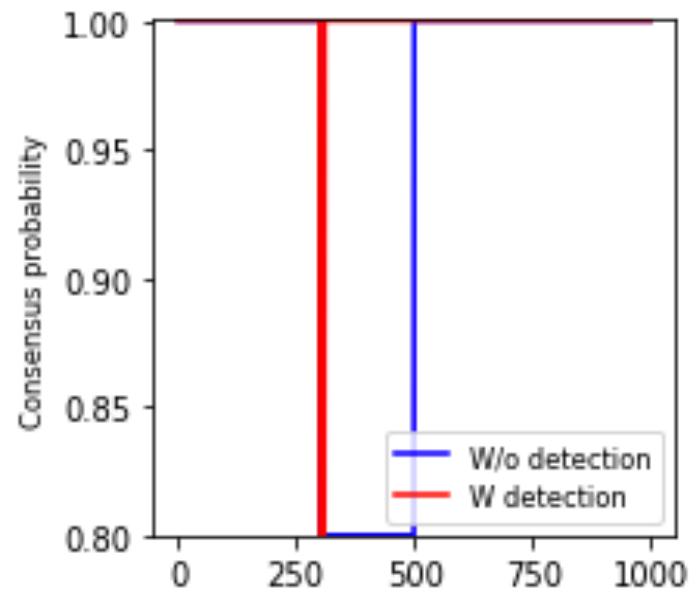


Random-flipping strategy

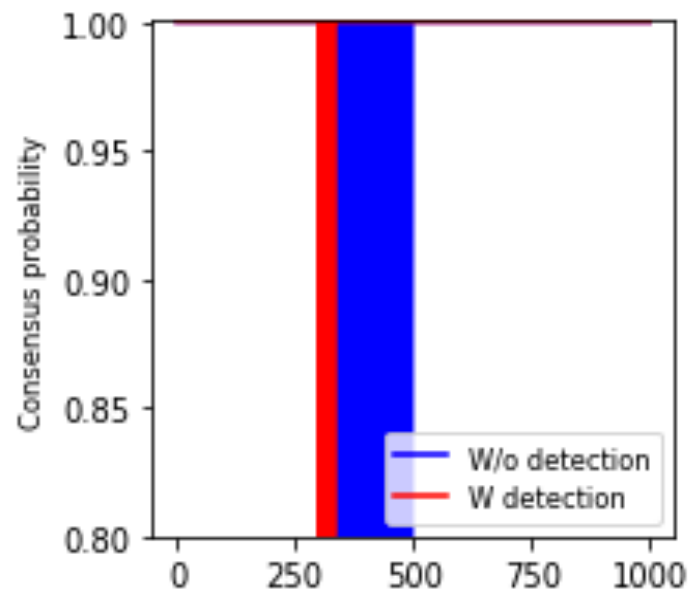


Results

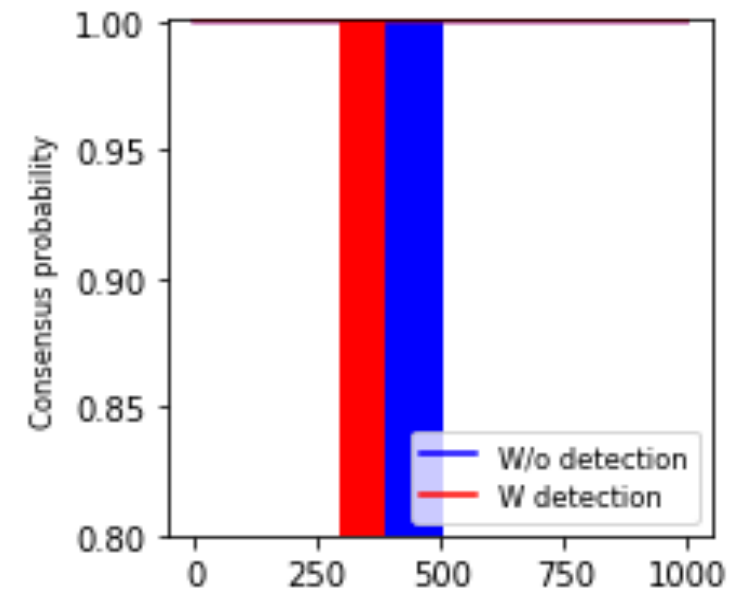
A node turned malicious from decision 300 to decision 500



Continuous flipping
RPMC-EWA has recovered
after less than 10 decisions



Random flipping
RPMC-EWA has recovered
after 25 decisions



Pattern flipping
RPMC-EWA has recovered
after 50 decisions



Limitations and Further Extensions

- ❑ **Limitation 1:** Empirical analysis (off-chain)
 - The reputation framework integration with a blockchain platform

- ❑ **Limitation 2:** The framework security, computation, and overhead analyses are not considered
 - Effects of consensus and reputation calculation on throughput and delay should be measured
 - Theoretical and practical security analyses need to be considered



Conclusion

- ❑ Probabilistic blockchains transform blockchains from data storage to data analyses and processing engine
- ❑ A reputation management framework is essential for better decision making and malicious node detection
- ❑ This work proposed as an extension of EWA that meets our requirements of reputation functions
- ❑ Empirical validations showed that the proposed approach outperform the baselines
- ❑ Extensions include implementation, and proper security and performance analyses



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