Future of Blockchain DLT and Reputation Consensus

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Speed + Cheapness + Security + Decentralization = CONST

Optimistic rollup

Speed + Cheapness + Security + Decentralization = CONST

Zero-Knowledge Succinct Non-Interactive Argument of Knowledge

(ZK-SNARK)

Speed + Cheapness + Security + Decentralization = CONST

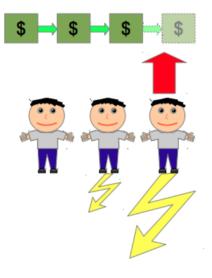


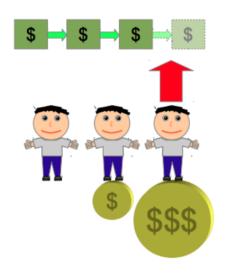
Reputation Consensus as a Liquid Democracy (2017)

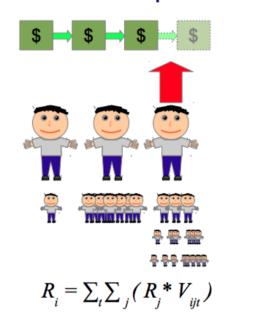


Proof-Of-Stake

Proof-Of-Reputation







Those who own more computing power govern the network.

Those who have more money govern the network.

Those who earn deeper reputation and greater long-term audience base govern the network.

Reputation Engine

Algorithm 1 Weighted Liquid Rank (simplified version) Inputs:

- 1) Volume of rated transactions each with financial value of the purchased product or service and rating value evaluating quality of the product/service, covering specified period of time;
- 2) Reputation ranks for every participant at the end of the previous time period.

Parameters: List of parmeters, affecting computations - default value, logarithmic ratings, conservatism, decayed value, etc.

Outputs: Reputation ranks for every participant at the end of the previous time period.

- 1: foreach of transactions do
- let rater_value be rank of the rater at the end of previous period of default value
- let rating_value be rating supplied by trasaction rater (consumer) to ratee (supplier)
- 4: **let** rating_weight be financial value of the transaction of its logarithm, if logarithmic ratings parameter is set to true
- 5: **sum** rater_value*rating_value*rating_weight for every ratee
- 6: end foreach

- do normalization of the sum of the muliplications per ratee to range 0.0-1.0, get differential_ranks
- 8: **do** blending of the old_ranks known at the end of previous peiod with differential_ranks based on parameter of conservatism, so that new_ranks = (old_ranks*conservatism+N*(1-differential_ranks)), using decayed value if no rating are given to ratee during the period
- 9: **do** normalization of *new_ranks* to range *0.0-1.0* 10:**return** *new_ranks*

- R_d default initial reputation rank;
- R_c decayed reputation in range to be approached by inactive agents eventually;
- C conservatism as a blending "alpha" factor between the previous reputation rank recorded at the beginning of the observed period and the differential one obtained during the observation period;
- FullNorm when this boolean option is set to True the reputation system performs a full-scale normalization of incremental ratings;
- LogRatings when this boolean option is set to True the reputation system applies log10(1+value) to financial values used for weighting explicit ratings;
- Aggregation when this boolean option is set to True the reputation system aggregates all explicit ratings between each unique combination of two agents with computes a weighted average of ratings across the observation period;
- Downrating when this boolean option is set to True the reputation system translates original explicit rating values in range 0.0-0.25 to negative values in range -1.0 to 0.0 and original values in range 0.25-1.0 to the interval 0.0-1.0.
- UpdatePeriod the number of days to update reputation state, considered as observation period for computing incremental reputations.

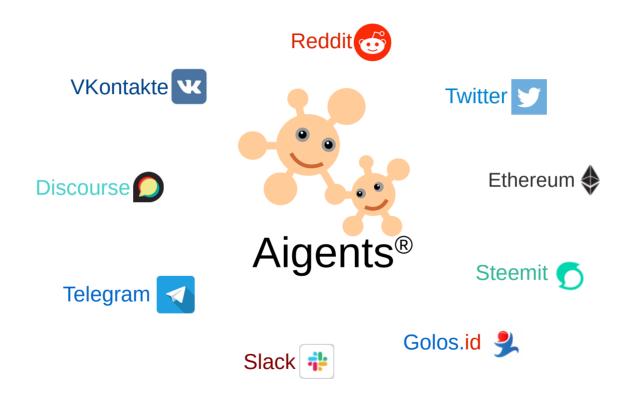
A Reputation System for Multi-Agent Marketplaces

Anton Kolonin, Ben Goertzel, Cassio Pennachin, Deborah Duong, Matt Iklé, Nejc Znidar, Marco Argentieri https://arxiv.org/pdf/1905.08036.pdf

https://github.com/singnet/reputation https://github.com/aigents/aigents-java/blob/master/src/main/java/net/webstructor/peer/Reputationer.java

Reputation System for Social Platforms (2014-2022)

Unified Liquid Rank Reputation computation across diverse social platforms

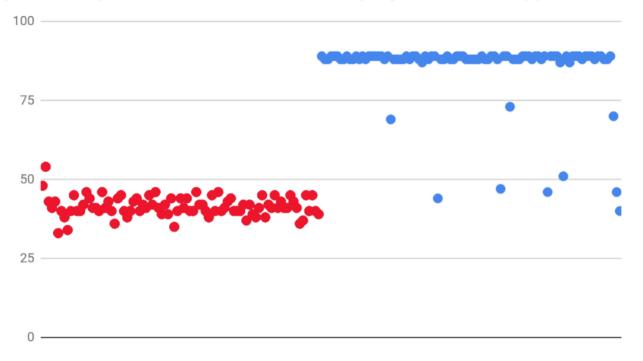


https://arxiv.org/abs/1912.00176

https://aigents.medium.com/aigents-bot-for-telegram-groups-1dba32140047

"Weighted Liquid Rank" for Fraud Resistance (2019)

Using Reputation System for protection from scam identifying dishonest suppliers on online marketplaces



Ranks of Suppliers, dishonest Supplier (including alias) in red and honest suppliers in blue

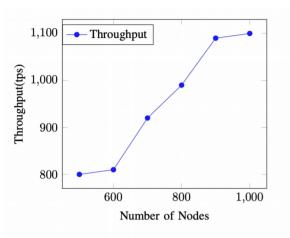
https://arxiv.org/pdf/1905.08036.pdf

https://blog.singularitynet.io/minimizing-recommendation-fraud-7dabbee8fc00

https://aiforgood2019.github.io/papers/IJCAI19-AI4SG_paper_28.pdf

Reputation Consensus for Distributed Ledger (2021)

International Journal of Network Security & Its Applications (IJNSA) Vol.13, No.4, July 2021



Average Block Time

10
20
30
400
500
Block Size
Figure 2: Average Block Time as the number of

transactions in a single Block is varied

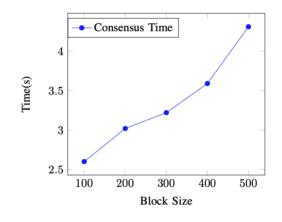


Figure 1: Throughput vs Number of network nodes

Figure 3: Consensus Time as the number of transactions in a single Block is varied

Proof-of-Reputation: An Alternative Consensus Mechanism for Blockchain Systems

https://arxiv.org/abs/2108.03542 https://aircconline.com/ijnsa/V13N4/13421ijnsa03.pdf Table 1 shows the performance of our scheme against other existing consensus mechanisms.

Under certain conditions when the number of participating nodes is increased, our scheme can achieve up to 1,100 transactions per second.

Table 1. Comparison with other consensus mechanisms

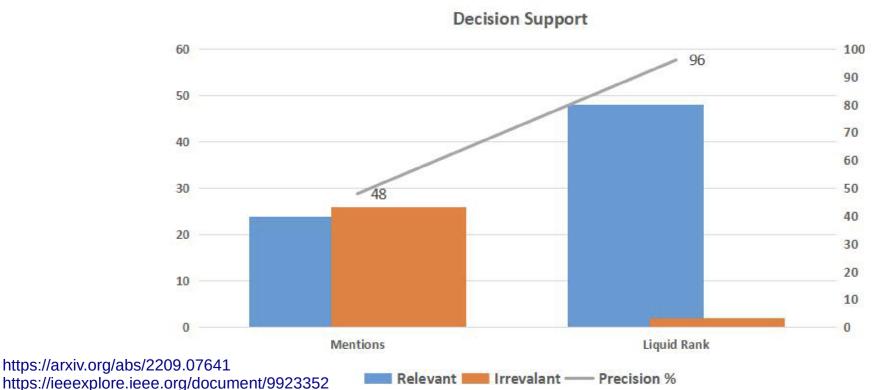
Consensus Mechanism	Throughput(TPS)
Proof-of-Work	7
Proof-of-Stake	60
Proof-of-Reputation(Baseline)	800
Proof-of-Burn	854
Proof-of-Reputation	1,100

Reputation System for Recommendation (2022)

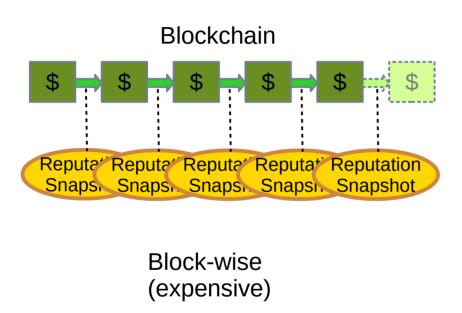
Application of Liquid Rank Reputation System for Content Recommendation

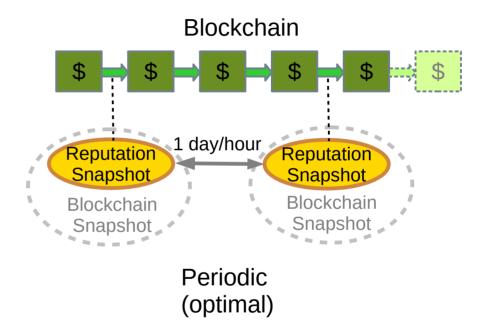
RESULTS QUALITATIVE ANALYSIS: DECISION SUPPORT





Reputation Consensus – Synchronization Options

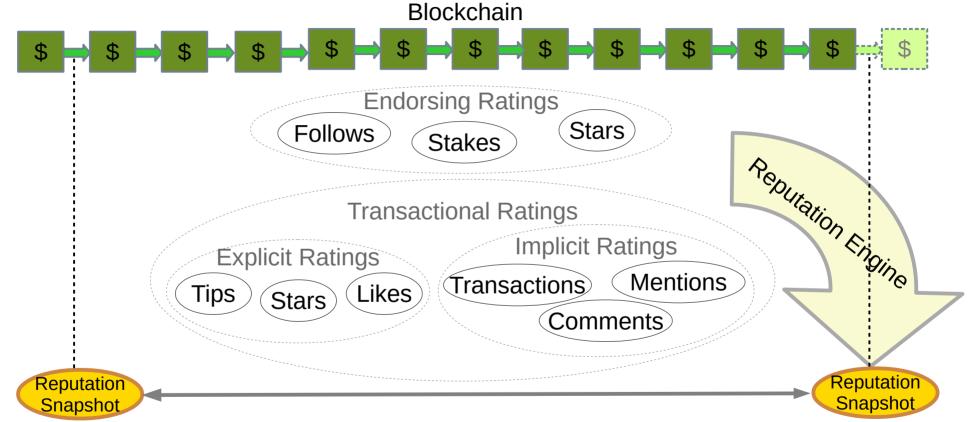




A Reputation System for Artificial Societies

Anton Kolonin, Ben Goertzel, Deborah Duong, Matt Ikle https://arxiv.org/pdf/1806.07342.pdf

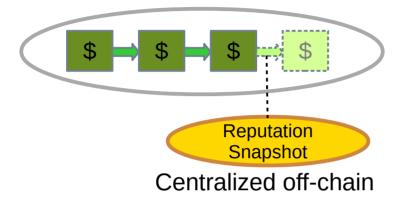
Reputation Consensus – Rating Sources

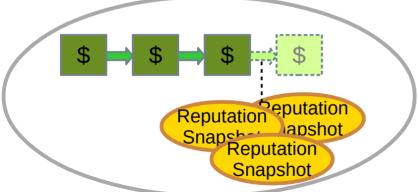


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Reputation Consensus – Engine Design Options

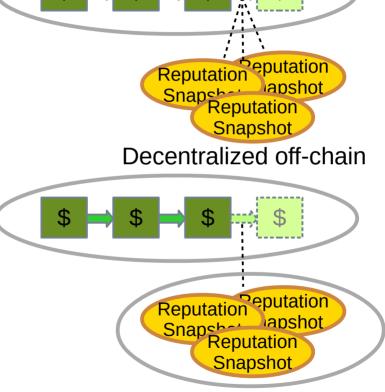




Decentralized on-chain (reputation mining)

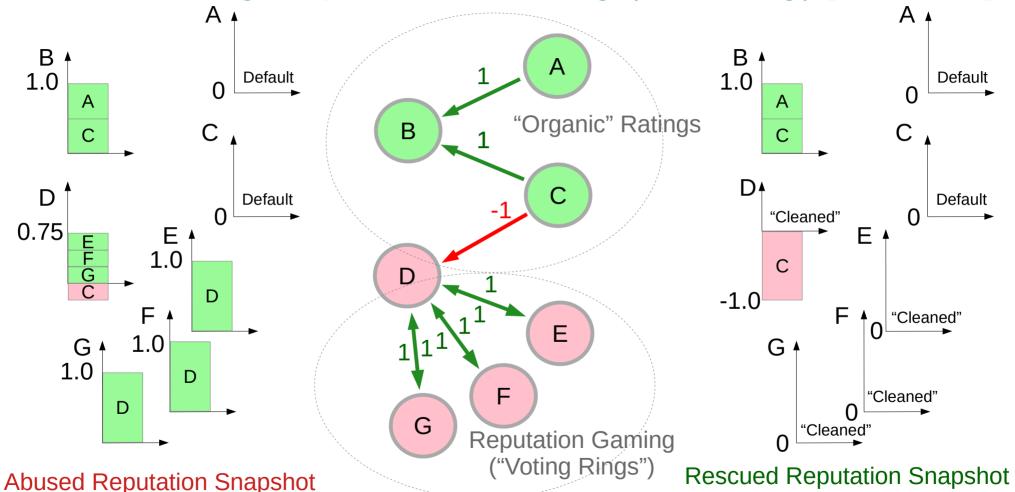
Reputation Reputation lapshot Reputation Snapshot A Reputation System for Artificial Societies

Anton Kolonin, Ben Goertzel, Deborah Duong, Matt Ikle https://arxiv.org/pdf/1806.07342.pdf

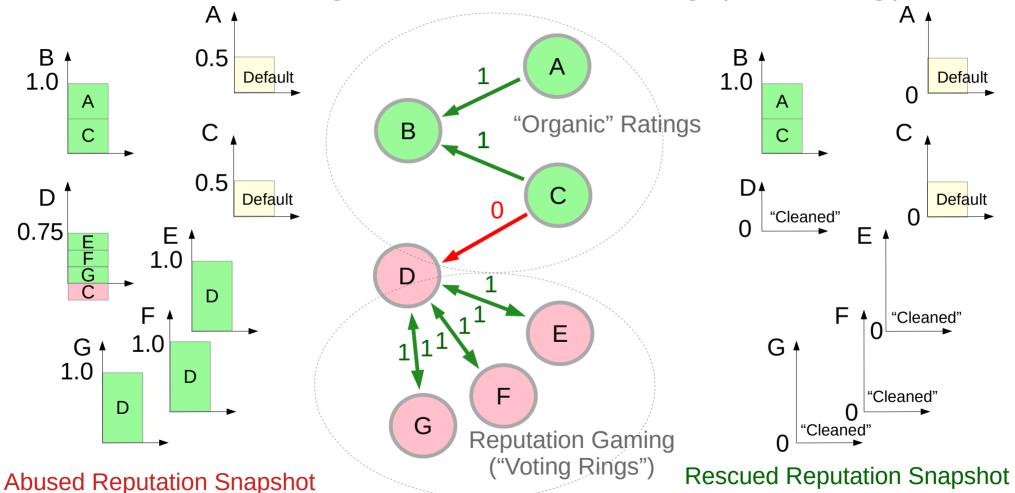


Decentralized side-chain (reputation consensus)

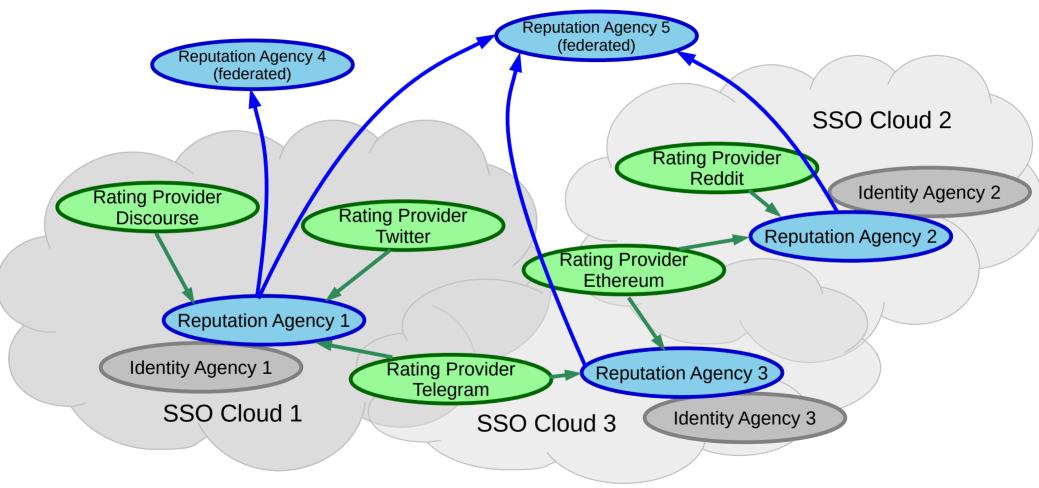
Next: Resisting Reputation Gaming (Churning) [1.0..-1.0]



Next: Resisting Reputation Gaming (Churning)



Multiple Identity and Reputation Agencies



Thank You and Welcome!

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