Proof-of-Reputation

as a Liquid Democracy for Distributed Systems

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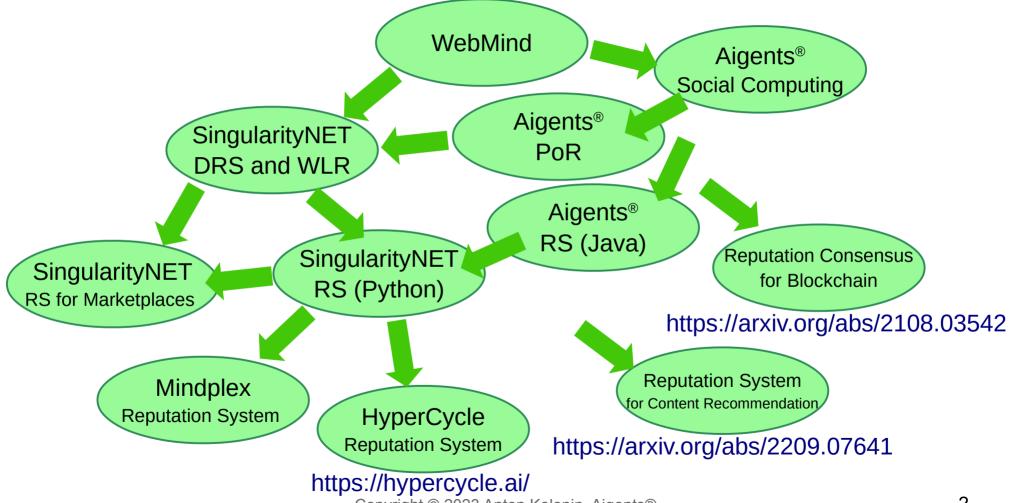
Telegram: akolonin



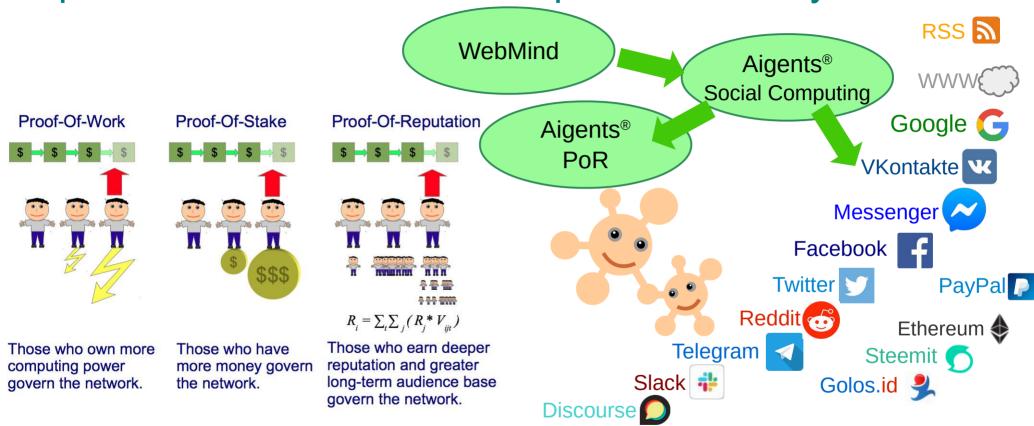




Reputation System and Applications 1999-2022

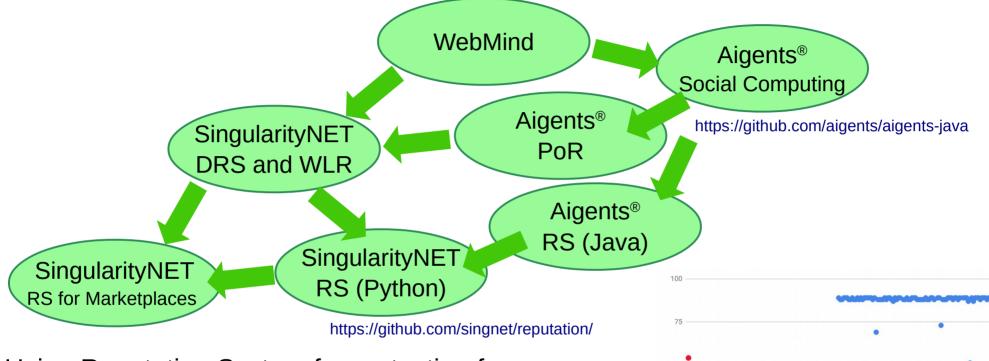


Reputation Consensus as a Liquid Democracy 1999-2017



https://steemit.com/blockchain/@aigents/proof-of-reputation-as-liquid-democracy-for-blockchain https://research.nsu.ru/en/publications/reputation-systems-for-human-computer-environments https://ieeexplore.ieee.org/document/8109887

"Weighted Liquid Rank" for Fraud Resistance 1999-2019



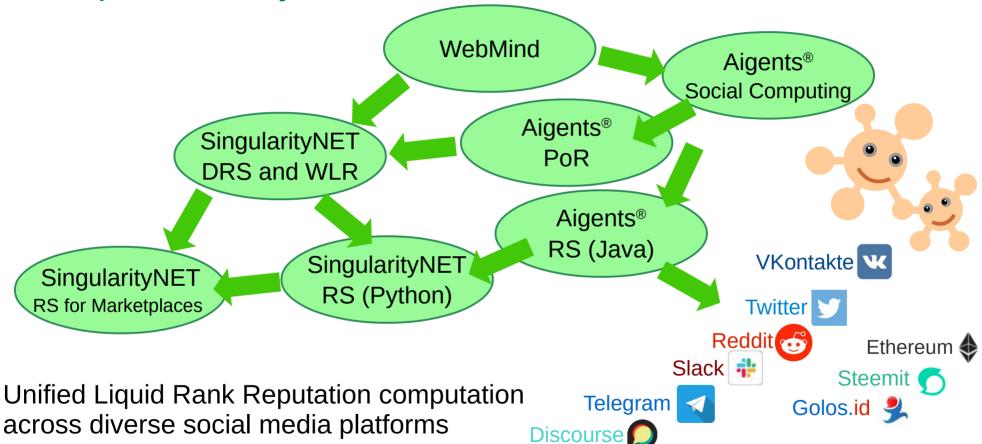
Using Reputation System for protection from scam identifying dishonest suppliers on online marketplaces

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

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

Reputation System for Social Platforms 1999-2020



https://arxiv.org/abs/1912.00176

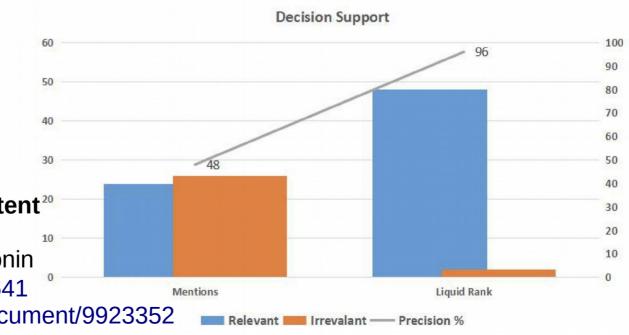
https://aigents.medium.com/aigents-bot-for-telegram-groups-1dba32140047

Reputation System for Recommendation 2017-2022



RESULTS QUALITATIVE ANALYSIS: DECISION SUPPORT





Application of Liquid Rank Reputation System for Content²⁰ Recommendation

Abhishek Saxena, Anton Kolonin

https://arxiv.org/abs/2209.07641

https://ieeexplore.ieee.org/document/9923352

Reputation Consensus for Distributed Ledger 2017-2021



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

Oladotun Aluko, Anton Kolonin

https://arxiv.org/abs/2108.03542

https://aircconline.com/ijnsa/V13N4/13421ijnsa03.pdf

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

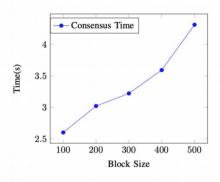


Figure 3: Consensus Time as the number of transactions in a single Block is varied

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

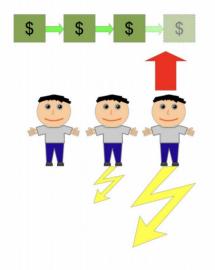
Consensus Mechanism	Throughput(TPS)	
Proof-of-Work	7	
Proof-of-Stake	60	
Proof-of-Reputation(Baseline)	800	
Proof-of-Burn	854	
Proof-of-Reputation	1,100	

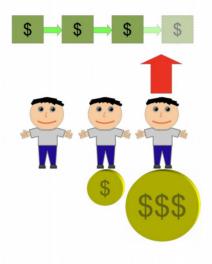
Reputation Consensus for Blockchain

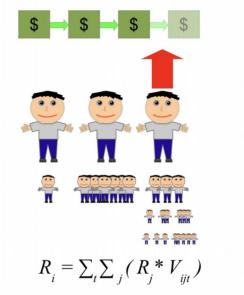
Proof-Of-Work

Proof-Of-Stake

Proof-Of-Reputation







Force is Power:

Those who own more computing resources govern the network.

Money is Power:

Those who have more money govern the network.

Reputation is Power:

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

Reputation Consensus for Blockchain - Experiments

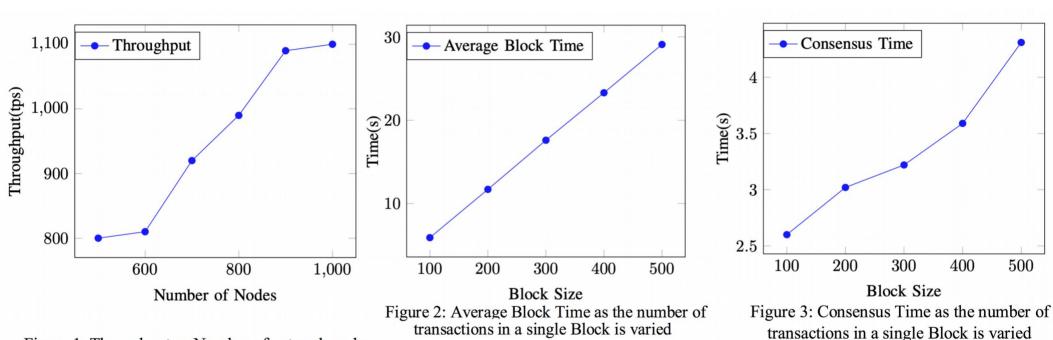


Figure 1: Throughput vs Number of network nodes

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

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Reputation Consensus – Throughput / Performance

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

Proof-of-Reputation: An Alternative Consensus

Mechanism for Blockchain Systems

Oladotun Aluko, Anton Kolonin

https://arxiv.org/abs/2108.03542

https://aircconline.com/ijnsa/V13N4/13421ijnsa03.pdf

Reputation Consensus – Attack Resilience

Table 2. Attack Resilience

Attacks	PoW(Bitcoin)	PBFT(ByzCoin)	PoR
Liveness	✓	X?	✓
Flash Attacks	X	X?	✓
Selfish Mining Attacks	X	X?	✓
Eclipse Attacks	X	✓	✓

Proof-of-Reputation: An Alternative Consensus

Mechanism for Blockchain Systems

Oladotun Aluko, Anton Kolonin

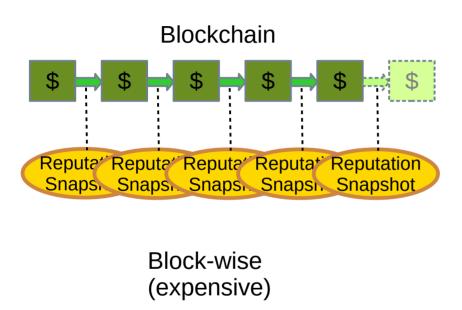
https://arxiv.org/abs/2108.03542

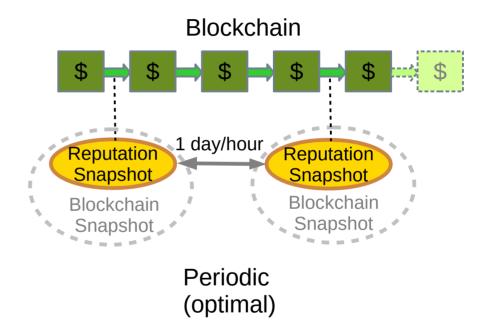
https://aircconline.com/ijnsa/V13N4/13421ijnsa03.pdf

PBFT:

https://pmg.csail.mit.edu/papers/osdi99.pdf

Reputation Consensus – Synchronization Options

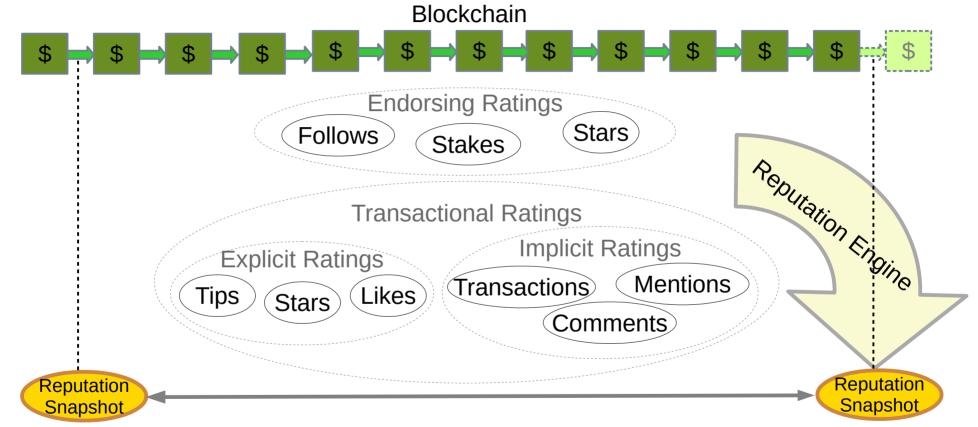




A Reputation System for Artificial Societies

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

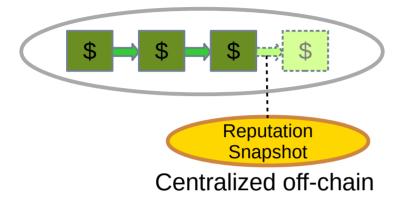
Reputation Consensus – Rating Sources

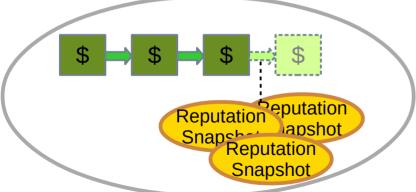


A Reputation System for Artificial Societies

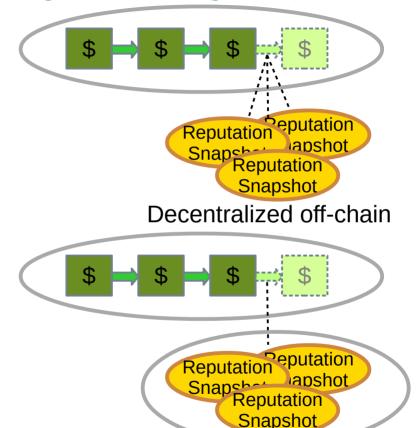
Anton Kolonin, Ben Goertzel, Deborah Duong, Matt Ikle https://arxiv.org/pdf/1806.07342.pdf

Reputation Consensus – Engine Design Options





Decentralized on-chain (reputation mining)



Decentralized side-chain (reputation consensus)

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

- 7: 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*

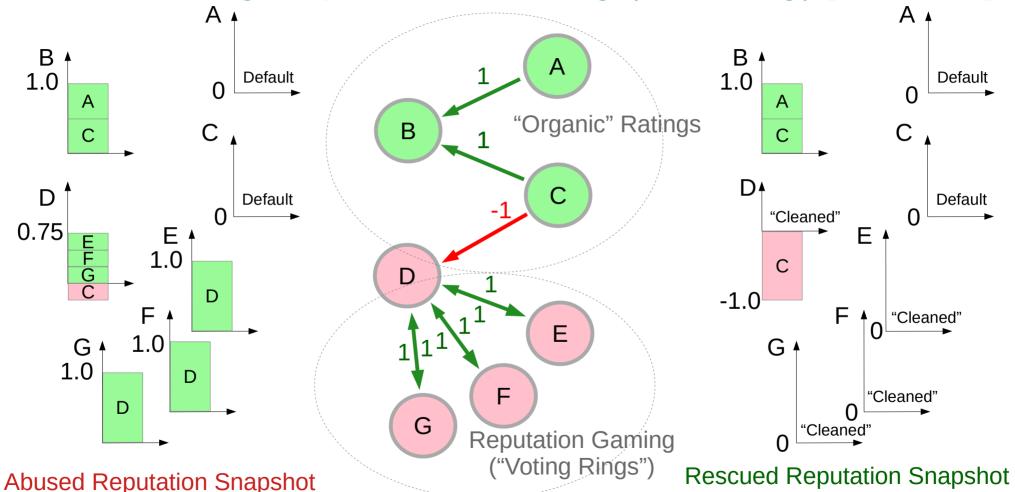
- R_d default initial reputation rank;
- R_c decayed reputation in range to be approached by inactive agents eventually;
- C conservatism as a blending "alpha" factor between the previous reputation rank recorded at the beginning of the observed period and the differential one obtained during the observation 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

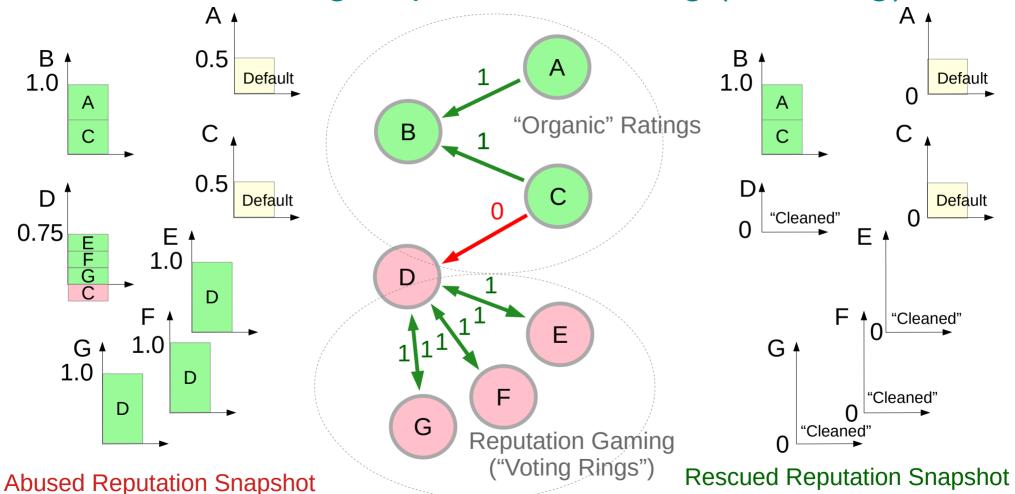
Anton Kolonin, Ben Goertzel, Cassio Pennachin, Deborah Duong, Matt Iklé, Nejc Znidar, Marco Argentieri https://arxiv.org/pdf/1905.08036.pdf

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

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



Next: Resisting Reputation Gaming (Churning)



Thank You and Welcome!

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