A Flexible Peer Management Architecture for Blockchain Systems

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Abstract. For a blockchain network to be secure and efficient, it is essential that nodes follow a disciplined peer management strategy, trying to avoid malicious peers and seeking to discover and retain reliable honest peers. This paper defines an abstract peer management architecture, relying on flexible ranking and clustering of peers. A concrete implementation of this architecture in the Scorex framework is also described.

1 Introduction

Blockchain systems have been increasingly adopted in recent years. For instance, the total market cap of Bitcoin [?], the most popular cryptocurrency and blockchain application, has grown to more than 130 billion US dollars¹ since its inception in 2009. And Bitcoin is not an isolated phenomenon. Fifteen other cryptocurrencies have market caps larger than 1 billion US dollars. With so much at stake, the security of these systems is critical. Consensus protocols, such as Bitcoin's Proof of Work (PoW) [?] and variants of Proof of Stake (PoS) [?] (e.g. Cardano's Ouroboros [?] and Ethereum's TODO [?]), ensure that the communicating peers agree on a single blockchain and that the blockchain remains valid as long as honest nodes detain the majority of computation power or stake. However, the network itself is vulnerable to attacks that may prevent the peers from communicating effectively. A recent study [?] found, for example, that Bitcoin's peer management is significantly vulnerable to eclipse attacks, where an attacker running several nodes attempts to monopolize all incoming and outgoing connections of an honest node. If the attacker succeeds, the mining power or stake of that honest node, is effectively removed from the network. Furthermore, the attacker may fork the chain and lead the honest node to waste mining power (and profits) on the forked chain. The possibility of such network layer attacks highlights the importance of resilient peer management strategies and motivates the work presented here.

Bitcoin's rigid and hard-coded peer management strategy [?] distributes known peers in two tables, one for new peers and the other for tried peers, and groups them in buckets so that peers belonging to the same subnet (and hence more likely to be controlled by an attacker) are more likely to belong to

Data obtained from http://www.worldcoinindex.com in 2017 on the 24th of November 2017.

the same bucket. A bitcoin node then tries to connect to peers from different buckets and from different tables.

The goal of the peer management architecture proposed here is to generalize Bitcoin's approach and make it more flexible, because there may be many other characteristics, besides IP address and being new or tried, that may be relevant for a node to consider when deciding whether to connect to a peer. The proposed architecture, described abstractly in Section ??, depends on a peer ranking algorithm (cf. Section ??) and a peer clustering algorithm (cf. Section ??); and procedures and protocols for manipulating peers (cf. Section ??), from peer banning to selection of peers for connection, may use the ranking and the clusters. Finally, an implementation of the proposed architecture in the Scalabased Scorex framework for blockchains is presented in Section ??.

2 Abstract Peer Management Systems

A peer is a pair (addr, port), where addr is its IP address and port is its port. The set of all peers is denoted \mathcal{P} . Any characteristic of a peer (e.g. number of past connections, date of last connection, reputation, delivery delays) is a feature and the finite set of all relevant features is denoted \mathcal{F} .

A peer management structure is a tuple (G, B, C), where G is the set of good peers, B is the set of banned peers, and C is the set of connected peers, satisfying the following conditions: $G \cap B = \emptyset$, $C \subseteq G$, $G \subseteq \mathcal{P}$, $B \subseteq \mathcal{P}$. A peer management system is a tuple $(M, \phi, \rho_{\phi}, \xi_{\phi}, \Pi)$ where:

- M is a peer management structure;

- $-\phi$ is the feature map such that $\phi(p, f)$ is the value of feature f for peer p;
- $\frac{-\rho_{\phi}: \forall S: 2^{\mathcal{P}}.S \to \mathbb{N}}{p \text{ in the set of peers}}$ is a peer ranking such that $\rho_{\phi}(S, p)$ is the rank of peer
- $-\xi_{\phi}: 2^{\mathcal{P}} \to 2^{2^{\mathcal{P}}}$ is a peer clusterizer using the feature map ϕ that takes a set of peers S and returns a set of clusters $\xi_{\phi}(S) \equiv \{S_1, \ldots, S_m\}$ such that $S_i \cap S_j = \emptyset$ when $i \neq j$ and $\bigcup_{1 \leq i \leq m} S_i = S$;
- Π is a set of *protocols* and *procedures* that take the current M and the current feature map ϕ and, possibly using ρ_{ϕ} and ξ_{ϕ} , return a modified peer management structure M' or a modified feature map ϕ' .

Example 1. Examples of peer rankings include: a function that sorts the set of peers S according to any combination of features in ϕ and returns the index of p in the sorted ordered set; or a function that returns the number of days since the last connection with p.

Example 2. Bitcoin's usage of separate tables for new peers, with whom there has never been a connection, and already tried peers, can be seen as a simple and rigid example of clusterizer. Bitcoin's bucket system, which partition peers according to their subnets, can also be seen as a clusterizer.

TODO: do you think we need to separate ingoing outgoing connections?

I'm using a dependent type notation here, in a sloppy way. This could be improved.

TODO: More examples not related to bitcoin, here.

3 A Concrete Peer Ranking

The peer ranking function ρ_{ϕ} is defined by:

$$\rho_{\phi}(S,p) \equiv \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{sort}(S, f, \mathrm{isIncreasing}(f))))) * ff \} \\ \texttt{Weight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{sort}(S, f, \mathrm{isIncreasing}(f)))) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{sort}(S, f, \mathrm{isIncreasing}(f)))) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{sort}(S, f, \mathrm{isIncreasing}(f)))) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{sort}(S, f, \mathrm{isIncreasing}(f)))) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{sort}(S, f, \mathrm{isIncreasing}(f)))) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{sort}(S, f, \mathrm{isIncreasing}(f)))) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{sort}(S, f, \mathrm{isIncreasing}(f)))) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{sort}(S, f, \mathrm{isIncreasing}(f)))) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(p, \mathrm{indexOf}(f)) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meight:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f)) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f) * ff \} \\ \texttt{Meigh:Int}(f) = \sum_{f \in F_r} (\mathrm{indexOf}(f) * ff \}$$

where:

- $-F_r$ is the set of features relevant for ranking, which has averageDeliveryDelay, numberOfPastConnections and elapsedTimeSinceLastConnection as elements
- indexOf(p, L) is the index of p in a list L.
- $\operatorname{sort}(S, f, b)$ is the list of elements in S sorted by their values on feature f. Sorting is increasing if $b = \operatorname{true}$, and decreasing otherwise.
- isIncreasing(f) is true iff the sorting of peers according to feature f should be increasing. isIncreasing is true for numberOfPastConnections and elapsedTimeSinceLastConnection and false for averageDeliveryDelay.
- For each feature f, f weight: Int is a configurable parameter indicating the importance of feature f for the ranking.

Peers having a higher rank are considered to be better than peers having a lower rank. By sorting peers decreasingly with respect to their average delivery delay, faster peers are prioritized. By sorting peers increasingly with respect to their number of past connections, reliable peers that have already been tried and used many times are preferred. By sorting peers increasingly with respected to their elapsed time since last connection, rotation of peers is encouraged.

4 A Concrete Clusterizer

The peer clusterizer ξ_{ϕ} is defined by:

$$\xi_\phi(S) \equiv \mathtt{k-means} \, \|_{\mathtt{numberOfClusters}} (\mathrm{map}(\lambda p. \mathrm{toVector}_{F_c}^\phi(p) * \mathtt{weightVector}, S))$$

where:

- k-means | is the parallel scalable k-means++ clustering algorithm [?]
- numberOfClusters is the configurable parameter specifying how many clusters should be generated. It defaults to maxConnections.
- to Vector $_{F_c}^{\phi}(p)$ transforms a peer p into its vector of feature values map $(\lambda f.\phi(p,f),F_c)$.
- $-F_c$ is the vector of features relevant for clustering, which has ipOctetOne, ipOctetTwo, ipOctetThree, ipOctetFour, numberOfPastConnections.
- weightVector is the vector of weights map(λf .\${f}Weight, F_c).

By having IP octets as features relevant for clustering, peers that belong to the same subnet will tend to be in the same cluster. Since the subnet masking convention considers the leftmost bits of an IP address to be more significant (i.e. TODO: There is a minor typing issue here that I need to fix. kmeans is returning clusters of vectors, but it should be returning clusters of peers corresponding to those vectors.

dividing an IP address into 4 dimensions in these 4 octets takes care of clustering in the same cluster/bucket peers that belong to the same subnet when IP allocation follows the old classfull network scheme. It probably works well for variable-length subnet masking as well. But, in this case, maybe something better could be done if we knew the routing prefix. But I think we don't. Or do we? More info: https://en.wikipedia.org/wiki/Classless_Inter-Domain, Routing Classless_Inter-Domain, Routing

the more leftmost bits are shared by two IP addresses, the more likely they are to belong to the same subnet), it should be ensured that ipOctetOneWeight >> ipOctetTwoWeight >> ipOctetThreeWeight >> ipOctetFourWeight, where >> means "significantly greater than".

By taking numberOfPastConnections into account, peers that differ in how known or new they are will tend to belong to different clusters.

5 Peer Management Protocols and Procedures

Peers are inserted and removed from the sets of good peers, banned peers and connected peers through protocols and procedures. The following subsections describe the concrete protocols and procedures that are implemented in Scorex.

5.1 Peer Discovery

A peer discovery protocol requests new peers from a source and inserts them into the set of good peers G. Generally, a source may be another peer, a trusted central server or an untrusted communication channel (e.g. IRC, Twitter, ...).

In Scorex, when the size of G is below minGoodPeers:Int, the peer requests new peers from $p^* = \arg\min_p \rho_{\phi}(G, p)$. p^* replies by sending a set of new peers N. G is then updated to $G' \equiv G \cup (N \setminus B)$.

TODO: have "referencedBy" column in the peers table

monitor and record åÄIJnumber of incoming connectionsåÄİ and åÄIJnumber of outgoing connectionsåÄİ in the feature table. Then it is up to the ranking and clustering functions to use this information as they wish.

TODO: request from a central server? download from an untrusted communication channel?

5.2 Peer Selection

The peer selection protocol is executed whenever the size of C is smaller than minConnections: Int. Its goal is to select a subset N of $n \equiv \max \text{Connections}: \text{Int} - |C|$ peers from $G \setminus C$ and then update C to $C' \equiv C \cup N$.

This is achieved through the following steps:

- 1. Partition G into the clusters $\{G_1, \ldots, G_m\} \equiv \xi_{\phi}(G)$.
- 2. Initialize C' with C.
- 3. For each cluster G_k , let $h(G_k) \equiv |C' \cup G_k|$.
- 4. Then, for i from 1 to n:
 - (a) Let $H = \{G_k \in \xi_{\phi}(G) | h(G_k) = \min_{g \in \xi_{\phi}(G)} h(g) \}$
 - (b) Randomly choose a cluster G^* from H with uniform probability.
 - (c) Randomly choose an integer r in the [0,99] integer interval and then choose p^* from G^* with one of the following probability distributions (for all $p \in G^*$):
 - $-P(p) \equiv 1/|G^*|$, if r < peerExplorationProbability:Int.
 - $-P(p) \equiv (\rho_{\phi}(G^*, p))/(\sum_{q \in G^*} \rho_{\phi}(G^*, q))G^*$, otherwise.
 - (d) Let $C' \equiv C' \cup \{p^*\}$.

will be

Finally, the occasional selection of peers with uniform probability, irrespectively of their rank, gives a chance of connection even to new peers that might have a low rank.

As the peer selection procedure chooses to connect to a set of peers that belong to different clusters, it will tend to choose a mix of old and new peers that belong to different subnets.

5.3 Feeler Connections

Bitcoin does the following: // * Choose a random address from new and attempt to connect to it if we can connect // successfully it is added to tried. // * Start attempting feeler connections only after node finishes making outbound // connections.

5.4 Handshake

During a handshake protocol a peer tries to get information on the availability of another peer and its capabilities. For example, if another peer is working in SPV mode, so does not have full blocks, they should not being asked from it. Or if the peer is interested in encrypted traffic exchange only, it would like to refuse. Encryption key is also to be formed during the handshaking.

key to be agreed on during handshaking (via elliptic curve Diffie-Hellman key exchange); then symmetric encryption

there are couple of stan-

dards how to use symmetric crypto with the key agreed on; ECIES-KEM /

5.5 Good Peer Eviction

Removes peers from **GoodPeers** when its number of peers exceeds maxGoodPeers:

TODO: use ranking and cluster

5.6 Banned Peer Eviction

Removes peers from **BannedPeers** when its number of peers exceeds maxBannedPeers:Int. TODO: use ranking and cluster

5.7 Peer Banning

Moves a peer from GoodPeers to BannedPeers.

5.8 Peer Rotation

Replaces a peer in **ConnectedPeers** by another peer from **GoodPeers**. Speeds up recovery from eclipse attacks.

6 Information Exchange Protocols

7 Encryption and Authentication

ECIES, AES, MAC

8 Configuration Advice

In order to increase the peer-to-peer protocol's resilience against eclipse attacks by a group of malicious peers that pretend to be good until the moment when they decide to attack, maxGoodPeers: Int should be set sufficiently high, in order to keep the proportion of malicious peers in **GoodPeers** low compared to the proportion of honest peers. Then the probability that *all* connections are made with malicious peers, as required for an eclipse attack to succeed, is low as well.

As long as we do not want to ever forgive banned peers, maxBannedPeers:Int should be set as high as possible without exceeding the available memory.

Would we ever want to for give banned peers?

I'm assuming that we want to load BannedPeers entirely into memory to be able to check quickly whether an incoming connection request is from a banned peer. Is this assumption correct?

9 Implementation in Scorex

In order to allow a high degree of flexibility, the proposed peer management architecture has many configurable parameters. Throughout the paper, configurable parameters and their types are formatted as "anExampleParameter:ItsType".

TODO: Parallel implementation of clustering

The peer management system maintains persistent and mutually disjoint tables of GoodPeers² and BannedPeers. There are at most maxGoodPeers:Int peers in GoodPeers and at most maxBannedPeers:Int in BannedPeers.

Example 3. TODO: Show an example table

The peer management system maintains a non-persistent table of **ConnectedPeers** to which the peer is currently connected. There are at most maxConnections:Int peers in **ConnectedPeers**.

10 Conclusion

Future work: reputation system

Future work: Anomaly detection: Consider a peer anomalous if it is the only element of a cluster?

References

² In Bitcoin, presumably good peers are split in two tables: **New** and **Tried**. Here a new peer is simply a good peer with zero past connections.