# Dynamic Intelligent Aggregation Services

Love and Strife in large-scale decentralized systems

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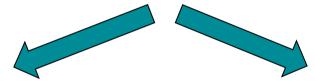
Systems Engineering Section



### PhD Research

#### **Distributed Computing**

Multi-level Reconfigurable Self-organization in **Overlay Services** 



**Dynamic Aggregation Services** 

Tree Self-organization Services











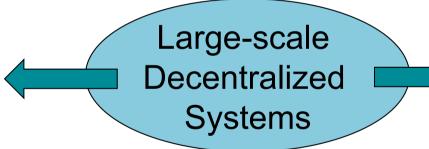
### Love and Strife



"These (elements) never cease changing place continually, now being all united by Love into one, now each borne apart by the hatred engendered of Strife, until they are brought together in the unity of the all, and become subject to it."



Continuous Information Change



Accurate Information Aggregation

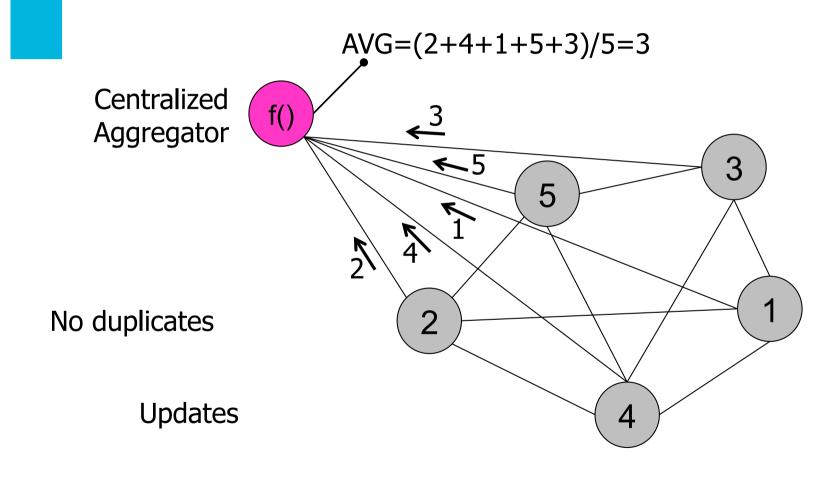
Local



Global

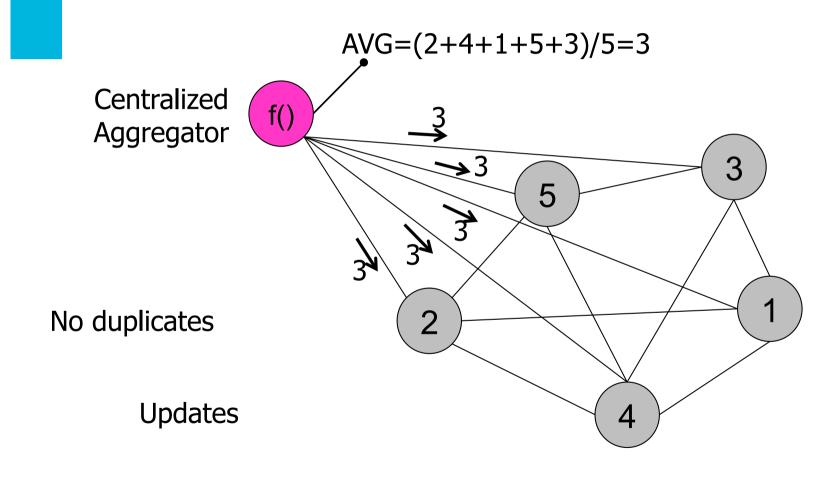


### The Aggregation Problem



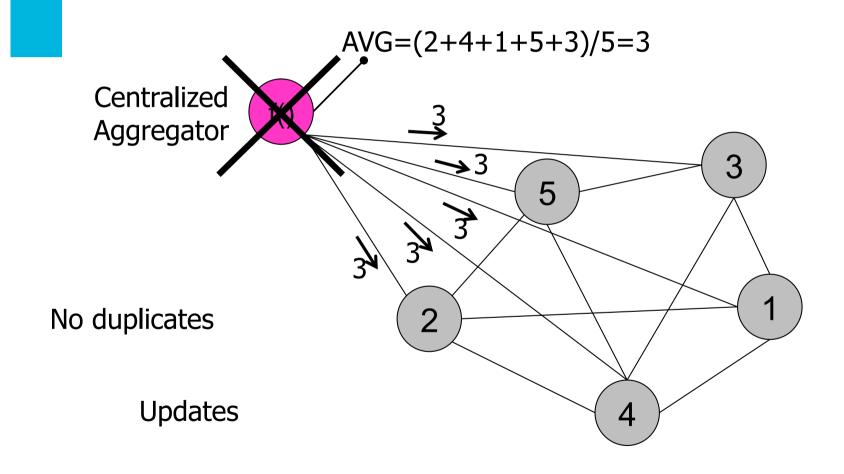


### The Aggregation Problem





### The Aggregation Problem





## Decentralized Aggregation

#### **Gossip-based Aggregation**

Aggregation-function dependent

Inaccuracies: Duplicate & outdated values

**Tree-based Aggregation** 

**Synopsis Diffusion** 

**Static Values** 



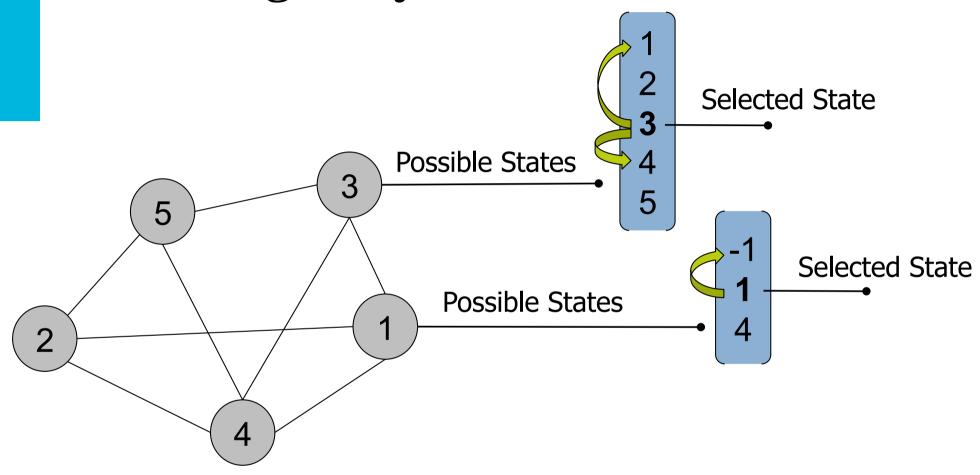
Routing-dependent

### Decentralized Aggregation (Cont.)

DIAS – Dynamic Intelligent Aggregation Service



### Modeling of Dynamics



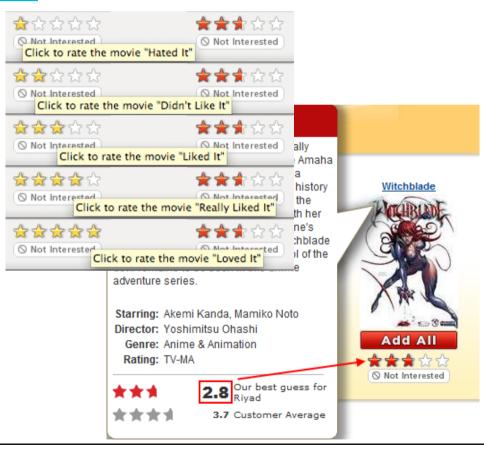
# of Possible Aggregates=# of possible states<sup># of Nodes</sup>=3<sup>10=</sup>59049!

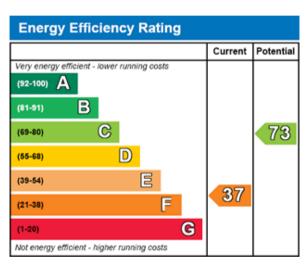


### Applications

#### **Recommender Systems**

#### **Smart Energy Systems**



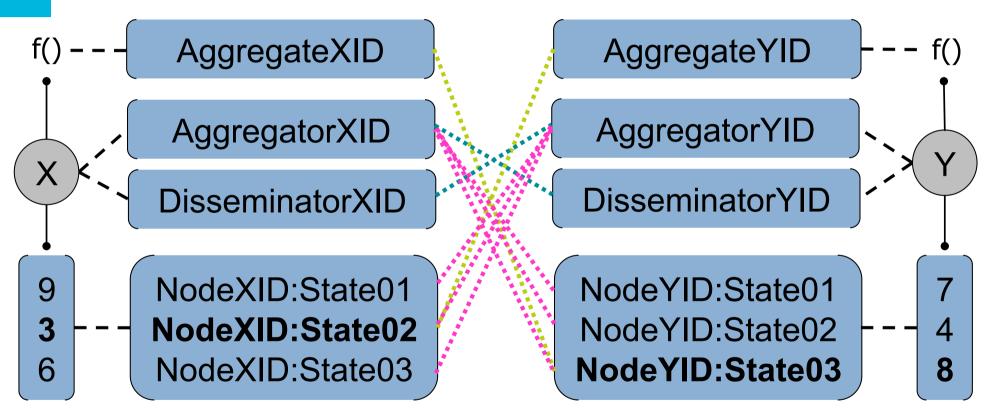






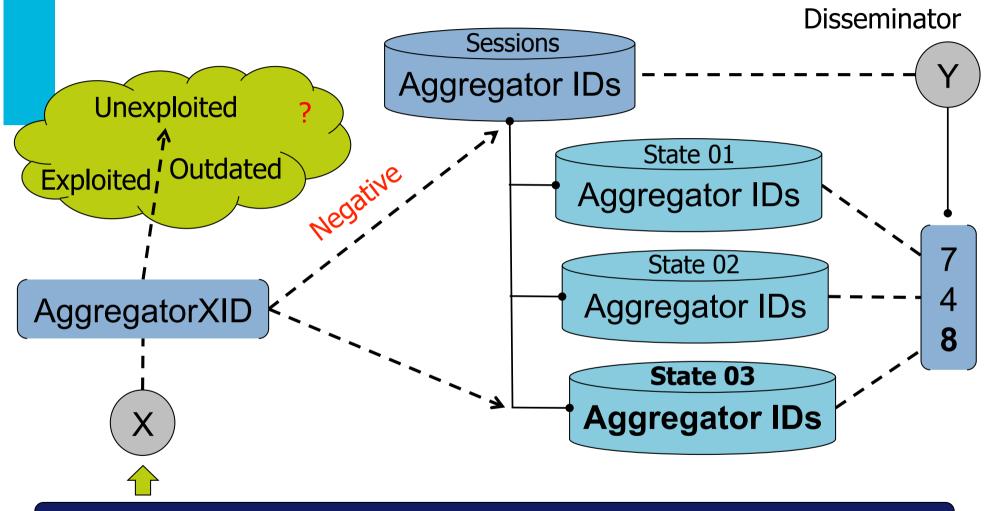
## Mutual Aggregation Memberships

Node: Aggregator and Disseminator

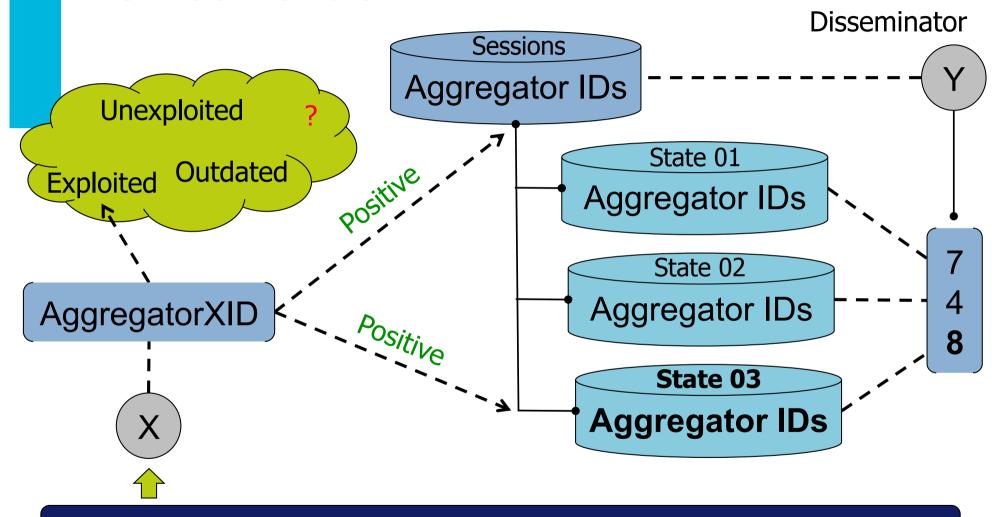


**Aggregation Session** 

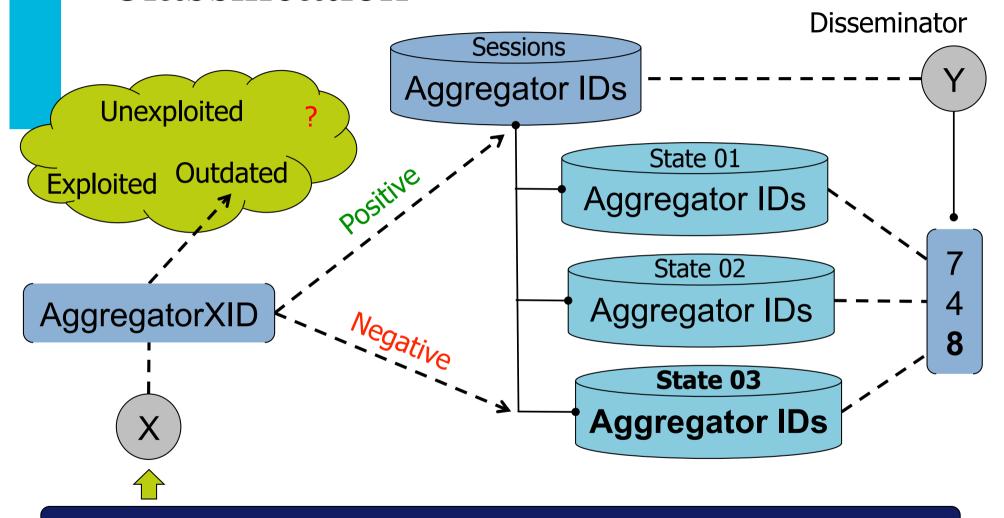














## Aggregation Memberships

Can we **explicitly** store aggregation memberships in a decentrality system?

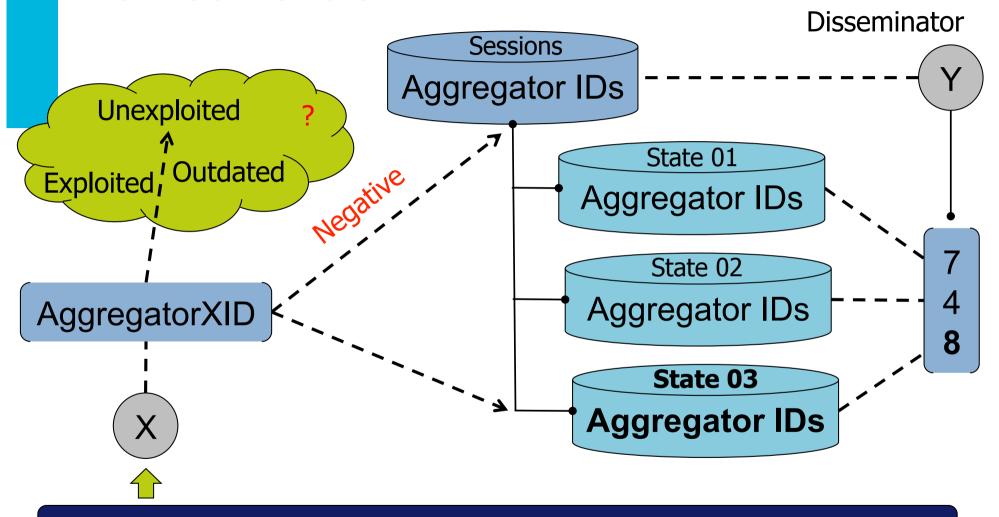
Can we overcome this problem?

Yes, with the Love and Strife of bloom filters

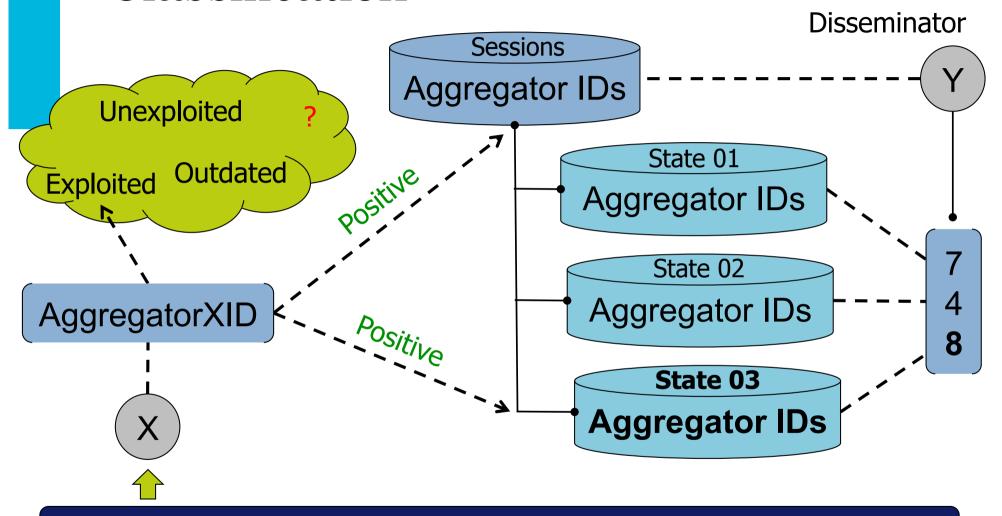
Probabilistic data structure

Large space savings at a cost of false positives

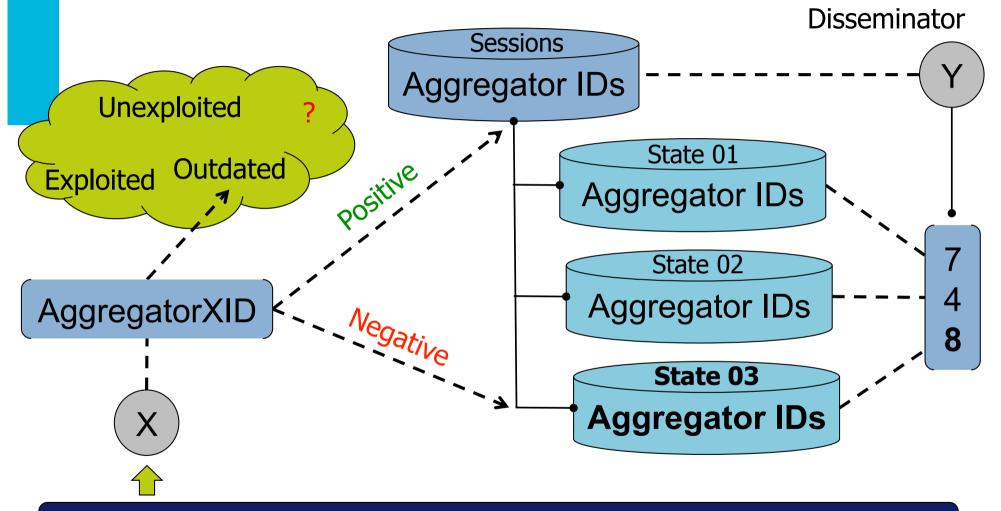














#### Bloom Filters

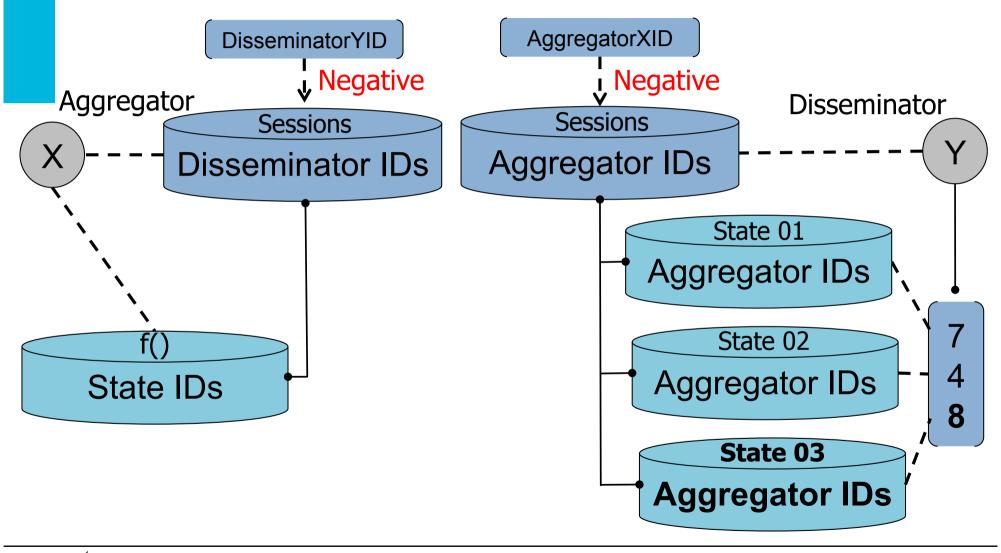
Inconsistent aggregation sessions are possible!

**Goal**: Minimization of aggregation inaccuracies due to false positives

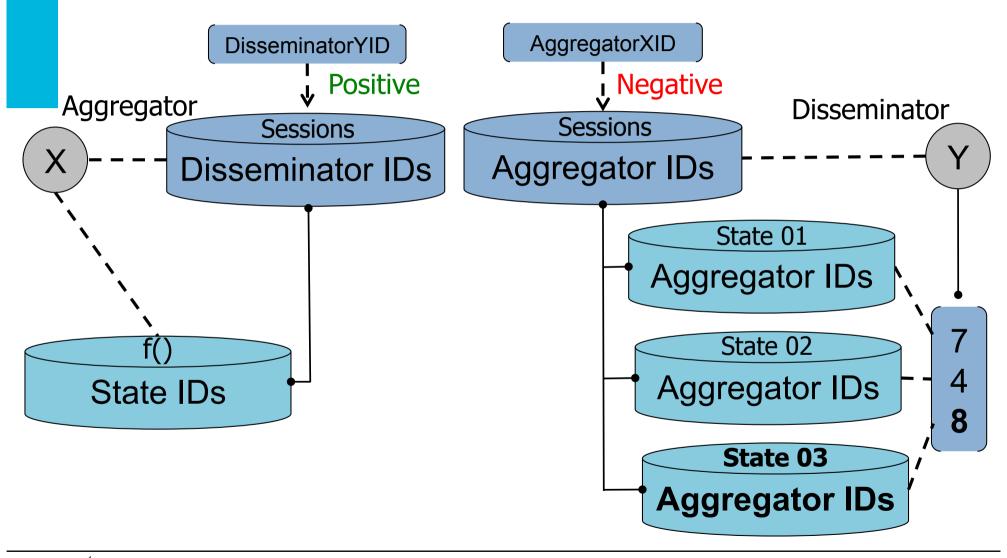
#### How?

Mutual membership checks!

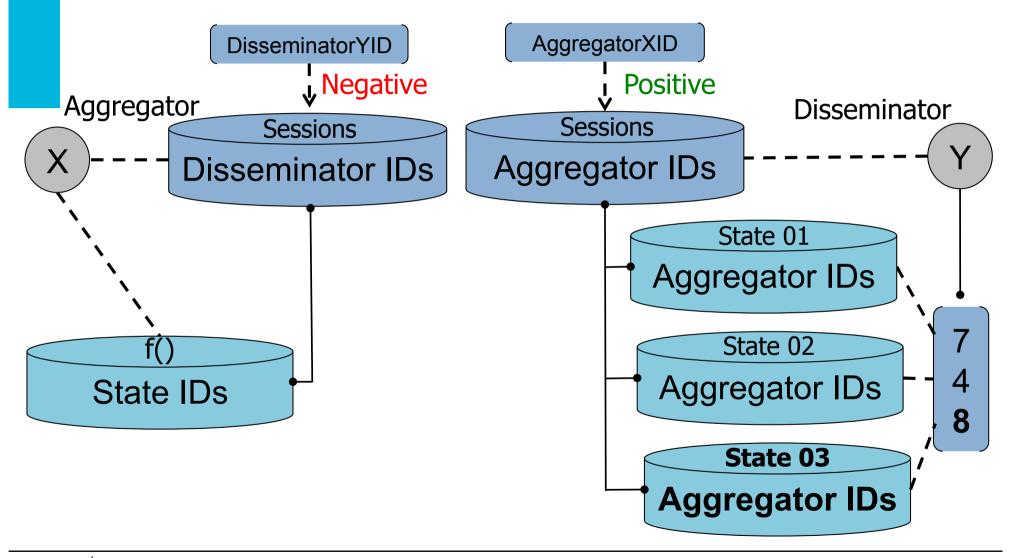




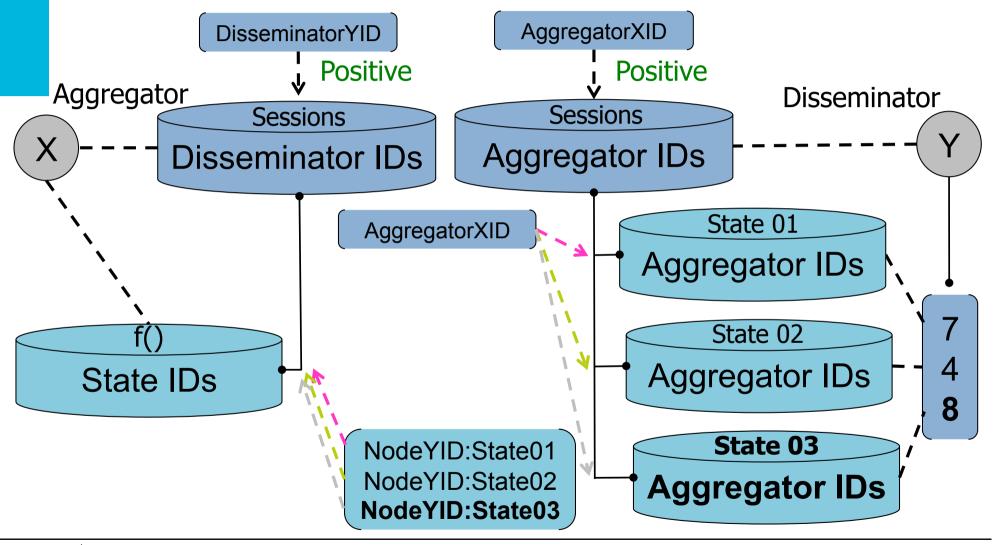














### Evaluation

DIAS prototype in Protopeer

XSiena Bloom Filter implementation

Node sampling: Gossiping (Peer Sampling Service)

Large-scale network: 1500 nodes

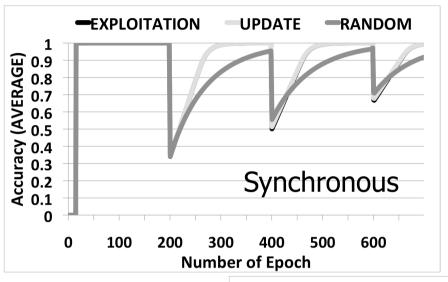
5 possible states/node

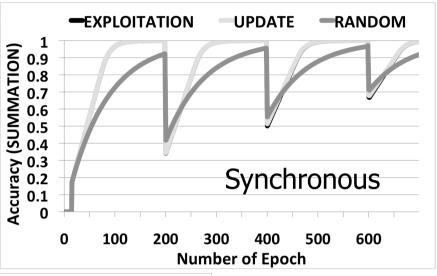
Synchronous/asynchronous state changes

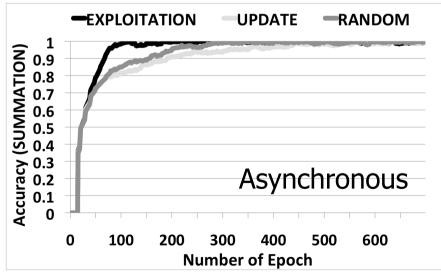
Aggregation strategies: EXPOITATION, UPDATE, RANDOM



### Evaluation







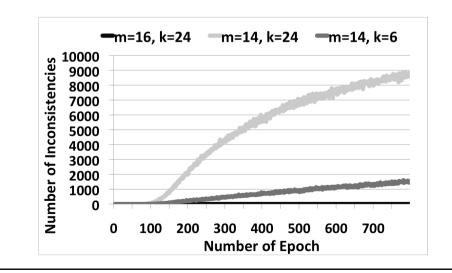


## Evaluation (cont.)

No influence in accuracy under false positive

Why?

Inconsistencies are detected by **mutual membership checks** 





### Conclusions

Generic-Multiple aggregation functions

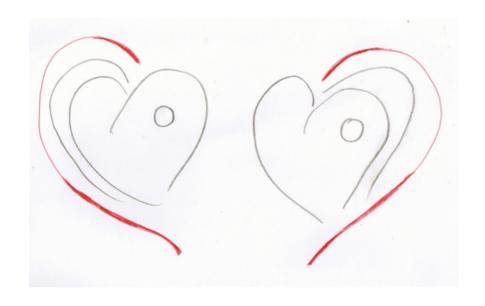
Dynamically changing values

Inaccuracies are minimized: Detection of duplicate and outdated values

Mutual memberships checks: False positive tolerance



## Questions?



#### **More information**

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