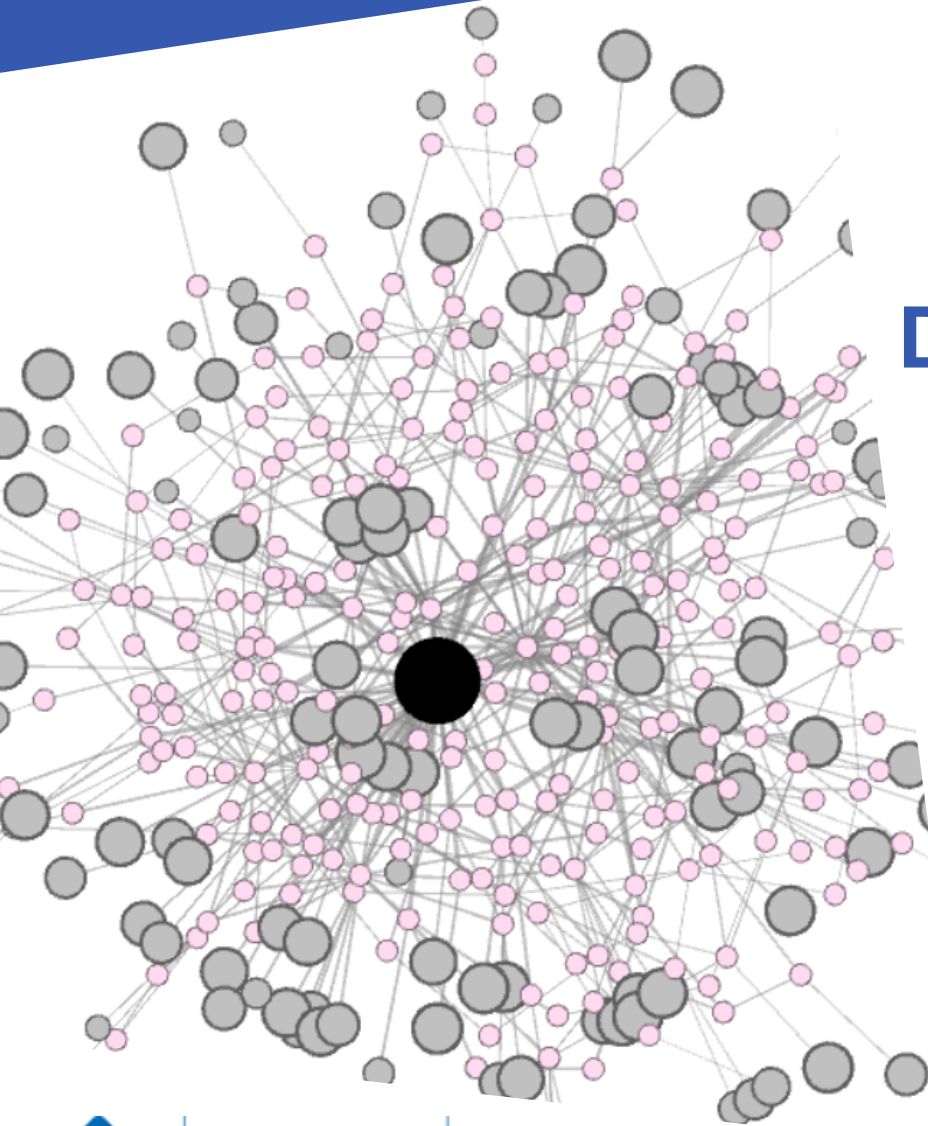




Comparing Centrality Indices for Network Usage Optimization of Data Placement Policies in Fog Devices

Isaac Lera, Carlos Guerrero,
Carlos Juiz

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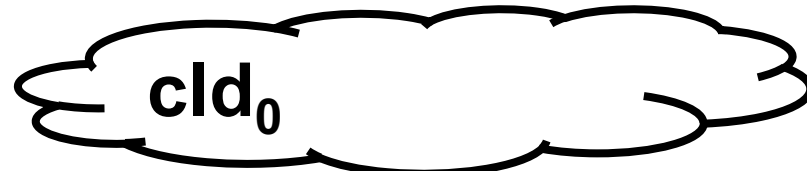
April 23rd, 2018
FMEC 2018, Barcelona

Introduction

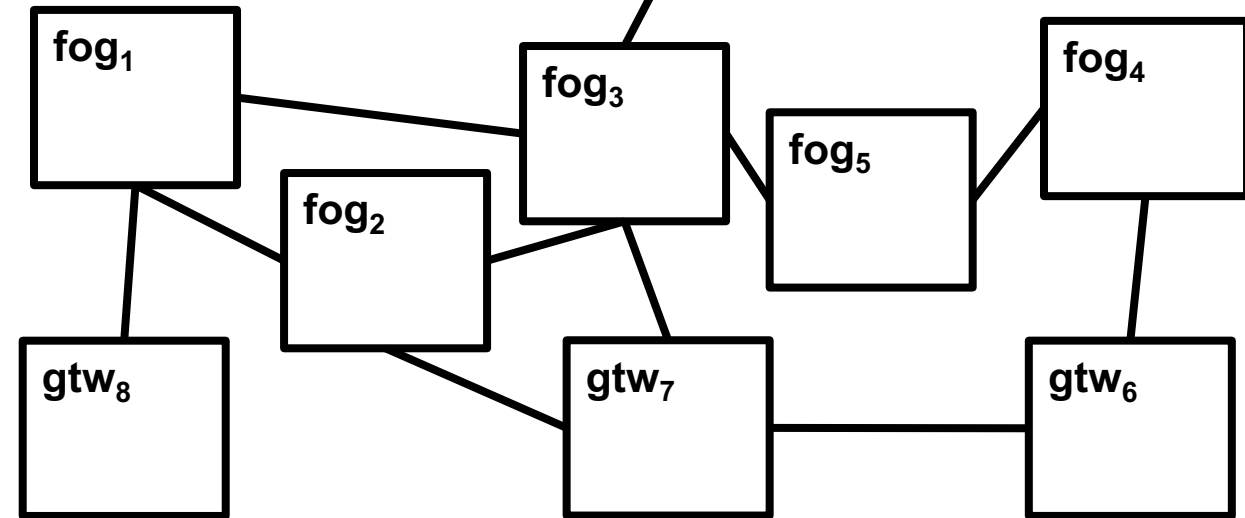
- Domain → Fog Computing
- Objective → Minimize network usage
- Decision variable → Data placement in fog devices
- Requirements → simple and scalable for large systems
- Tools → Complex Networks Metrics
- Solution → Centrality indices

Fog Architecture Model

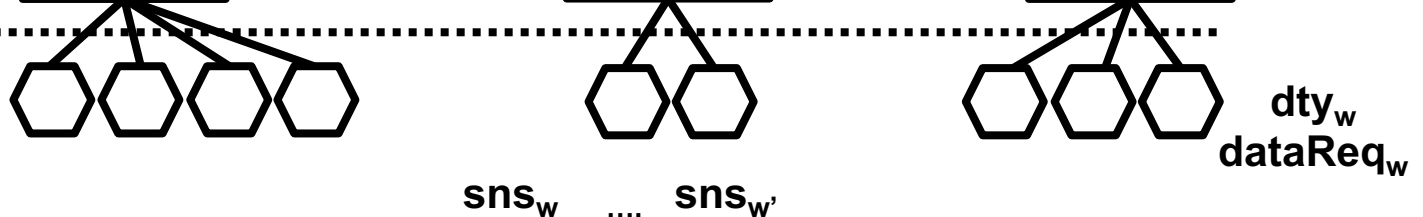
Cloud layer



Fog layer



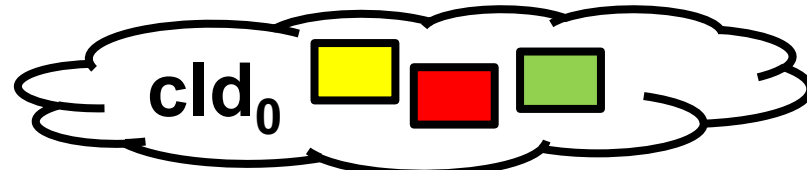
IoT
sensors



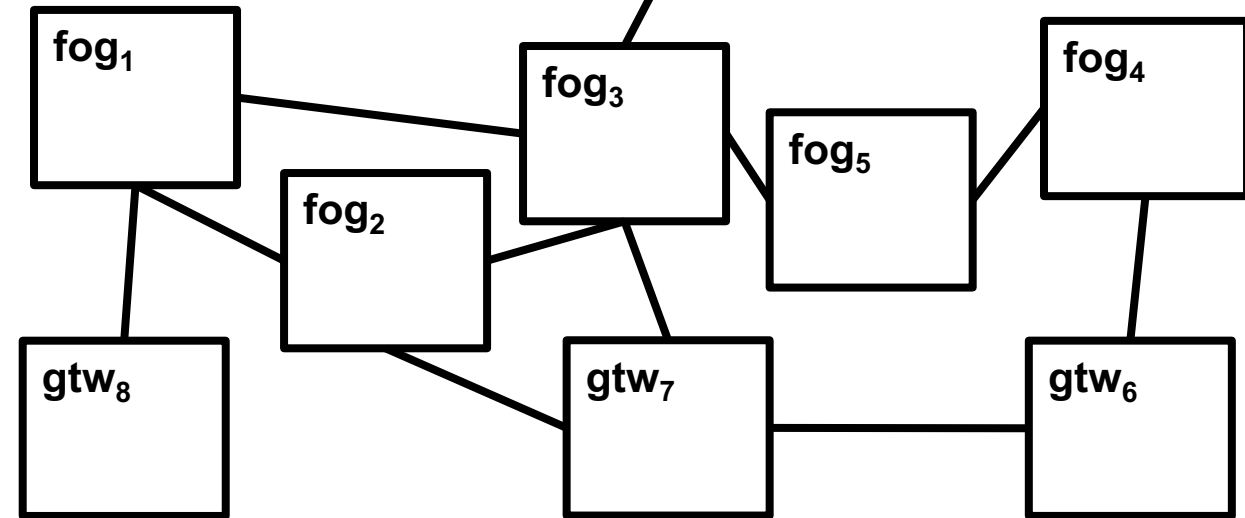
- Service placement
- Data Placement

Fog Architecture Model

Cloud layer



Fog layer



IoT
sensors



sns_w $sns_{w'}$

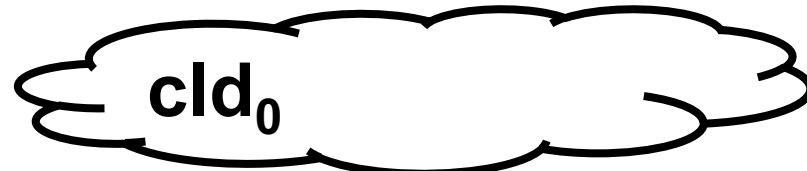


dt_{y_w}
 $dataReq_w$

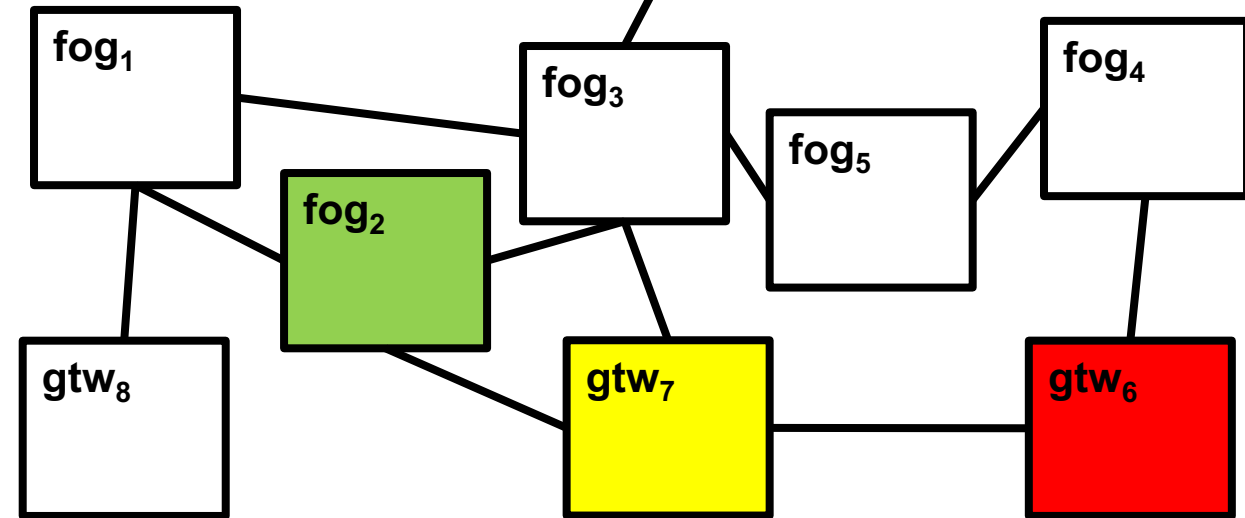
- Service placement
- Data Placement

Fog Architecture Model

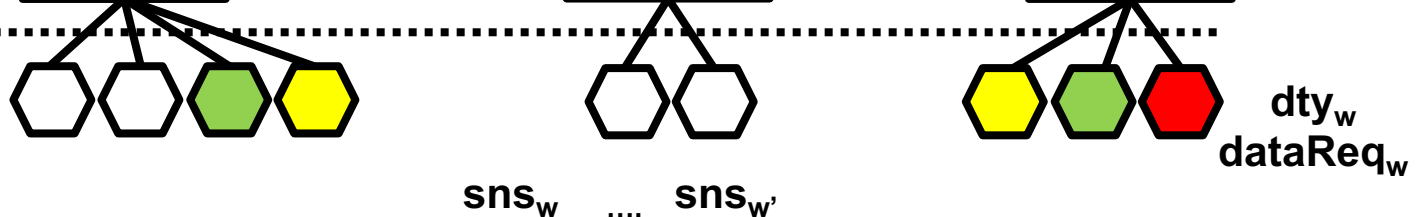
Cloud layer



Fog layer



IoT
sensors



- Service placement
- Data Placement

Data Placement Problem

- Solution

- Data placement matrix
- |Data types| x |Fog devices|

$$P = \begin{cases} p_{xu} = 1 & \text{if } dty_x \text{ is stored in } fog_u \\ p_{xu} = 0 & \text{otherwise} \end{cases}$$

- Constrains

- Data type stored in only 1 device
- Provisioned storage smaller than fog storage

$$\sum_{u=1}^{|U|} p_{xu} = 1, \forall x = (1, \dots, |X|)$$

$$\sum_{x=1}^{|X|} p_{xu} \times strReq_x \leq strCap_u, \forall u = (1, \dots, |U|)$$

- Objective

- Number of data packets transmitted

$$\min \sum_x^{|DTY|} \sum_u^{|FOG|} p_{xu} \times \sum_w^{|SNS|} dist(sns_w, fog_u)$$

Data Placement Algorithm

- (a) Model the network topology as a complex weighted network;
- (b) Weight the edges with the summation of the hop distances to the sensors of a given data source type
- (c) Calculate the centrality index of each node
- (d) Select the node with the highest centrality index
 - (d1) if not enough free storage capacity select the next highest centrality until a device with enough storage is found

Repeat the process for each data type, ordered by the data generation rate

Algorithm 2 Pseudo-code of the data allocation algorithm.

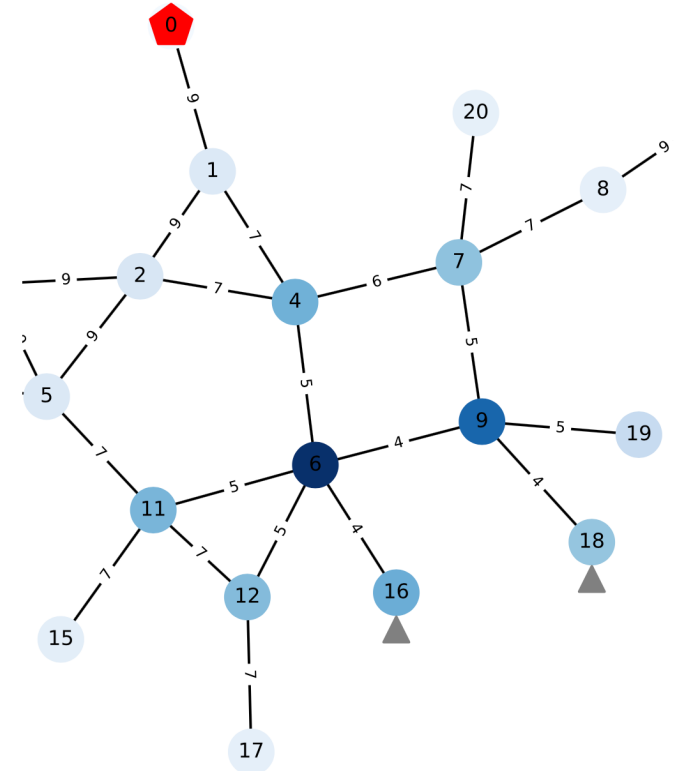
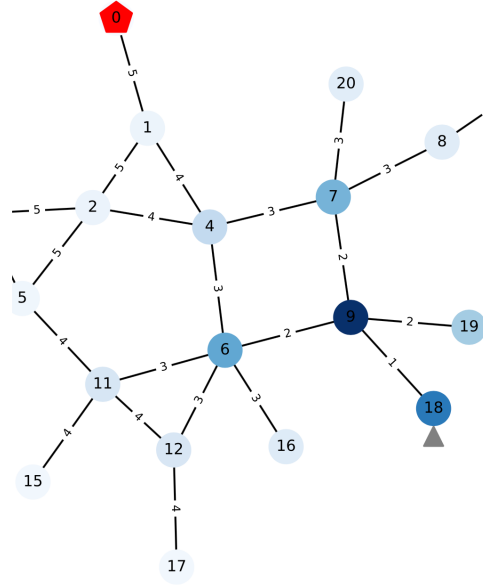
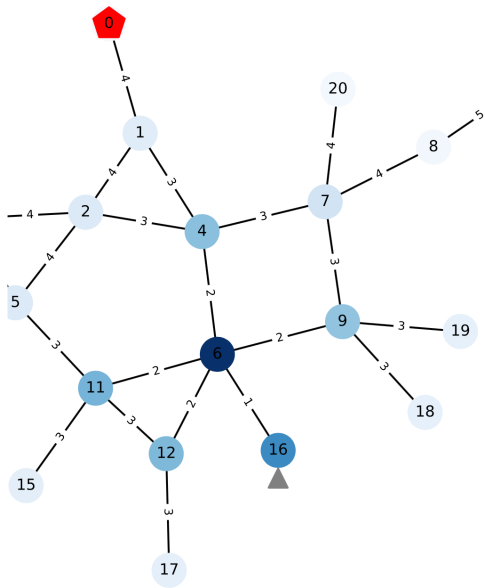
Procedure: DataAllocation()

Input: network topology, set of devices

Output: data type allocation (matrix P)

```
1: CN = EdgeWeightCalculation()
2: orderedDty = sort(|DTY|, "by generation rate")
3: for all  $dt_{y_x} \in \text{orderedDty}$  do
4:   nodesCentrality = centralityAlgorithm(CN[ $dt_{y_x}$ ])
5:   orderedNodes = sort(nodesCentrality, "by centrality")
6:   allocated = false
7:   while not allocated do
8:      $fog_u = \text{orderedNodes.next}()$ 
9:     if  $dt_{y_x}.dataReq_x < (fog_u.dataCap_u - fog_u.dataProv_u)$  then
10:      allocated = true
11:       $fog_u.dataProv_u += dt_{y_x}.dataReq_x$ 
12:       $p_{dt_{y_x}, fog_u} = 1$ 
13:     end if
14:   end while
15: end for
```

Data Placement Algorithm



Algorithm 1 Pseudo-code for the edge weight calculation.

Function EdgeWeightCalculation()

Input: network topology, set of sensors and set of data types

Output: complex weighted networks for each data type

```

1: for all  $dt_{y_x} \in |DTY|$  do
2:   totalWeightedGraph = 0
3:   for all  $sns_w \in dt_{y_x}$  do
4:      $snsWeightedGraph = calculateEdgeWeight(sns_w)$ 
5:     totalWeightedGraph += snsWeightedGraph
6:   end for
7:    $setCN[dt_{y_x}] = totalWeightedGraph$ 
8: end for
9: return setCN
    
```

Experimental evaluation

- Four network topologies:
 - Barabasi-Albert, Random Euclidean, Lobster, Grid
- Three centrality indices:
 - Eigenvector, Betweenness, Current flow betweenness
- Four experiment sizes:
 - Number of data types: 50, 100, 200, 300
- Simulator: YAFS (<https://github.com/acsicuib/YAFS>)

Experimental evaluation

Device capacity = 1.0

DT size = 1.0

Data rate = $[1/10..1/50]$

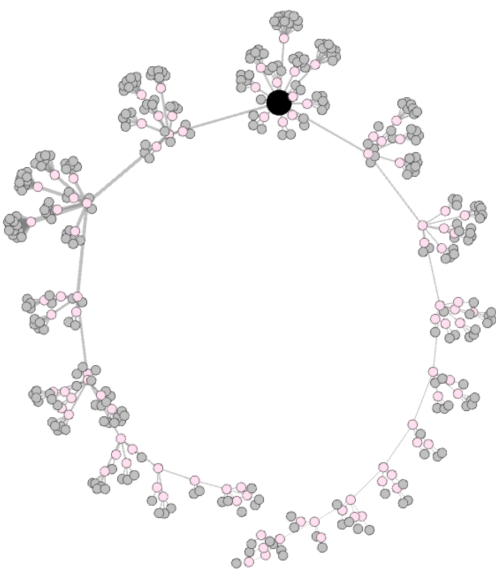
Proximity-based dist.
of the sensors

Link latency = 1.0

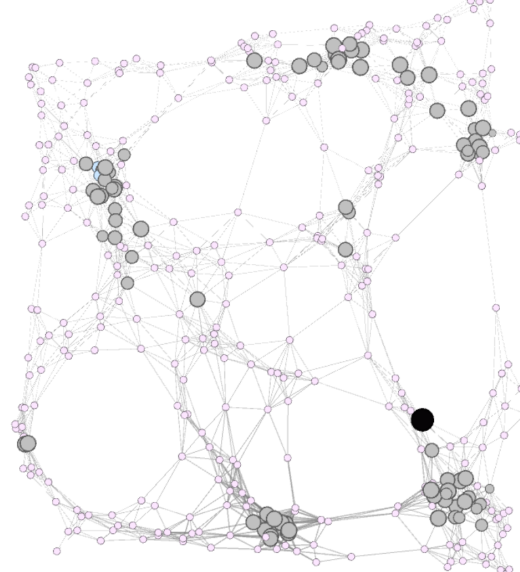
Packet size = 500 bytes

400 devices

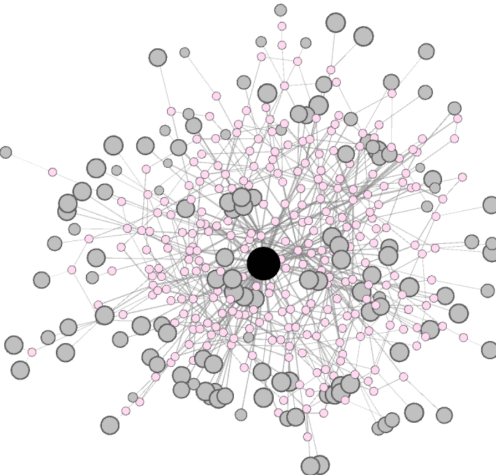
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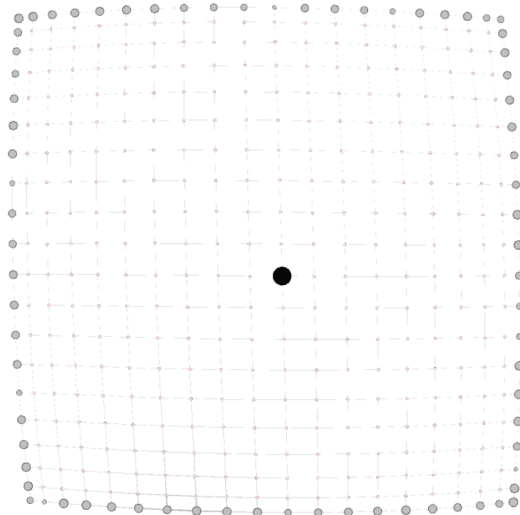
(a) Lobster topology



(b) Random Euclidean topology

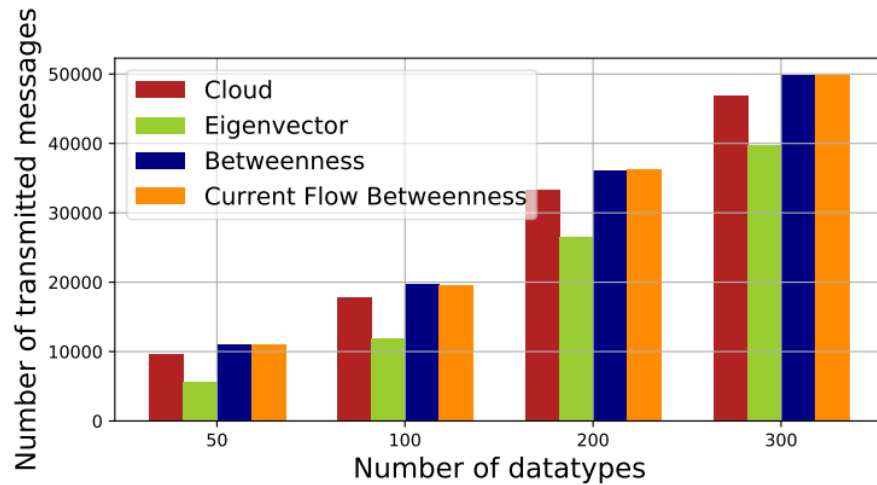


(c) Barabasi-Albert topology

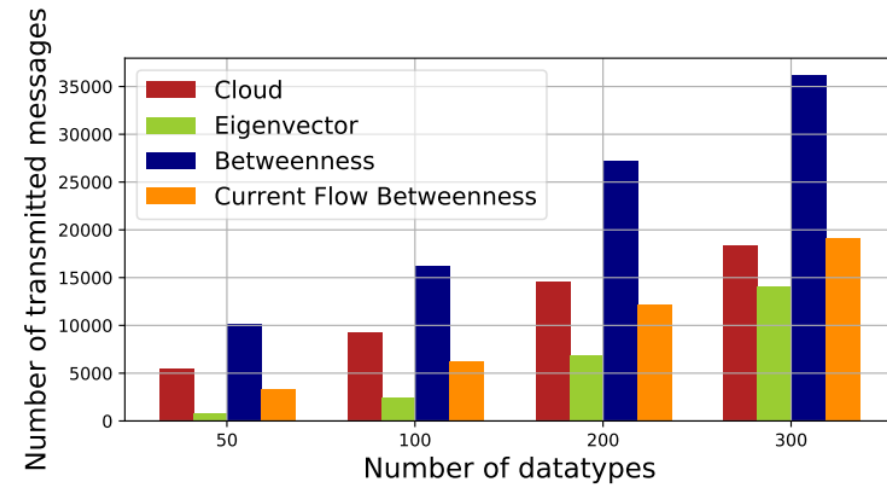


(d) Grid topology

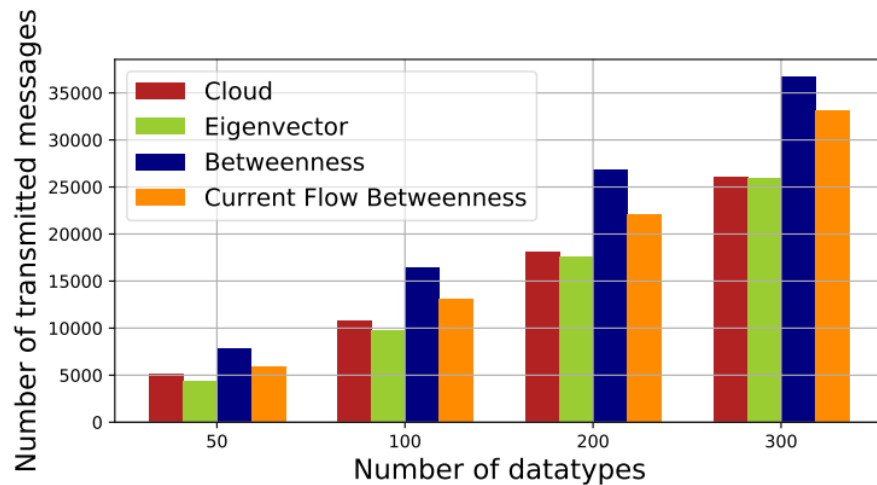
Results



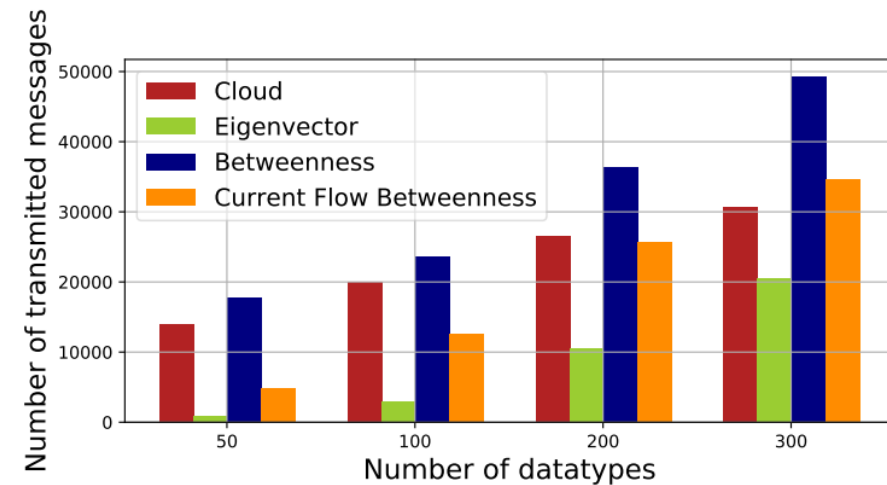
(a) Lobster topology



(b) Random Euclidean topology



(c) Barabasi-Albert topology



(d) Grid topology

Discussion

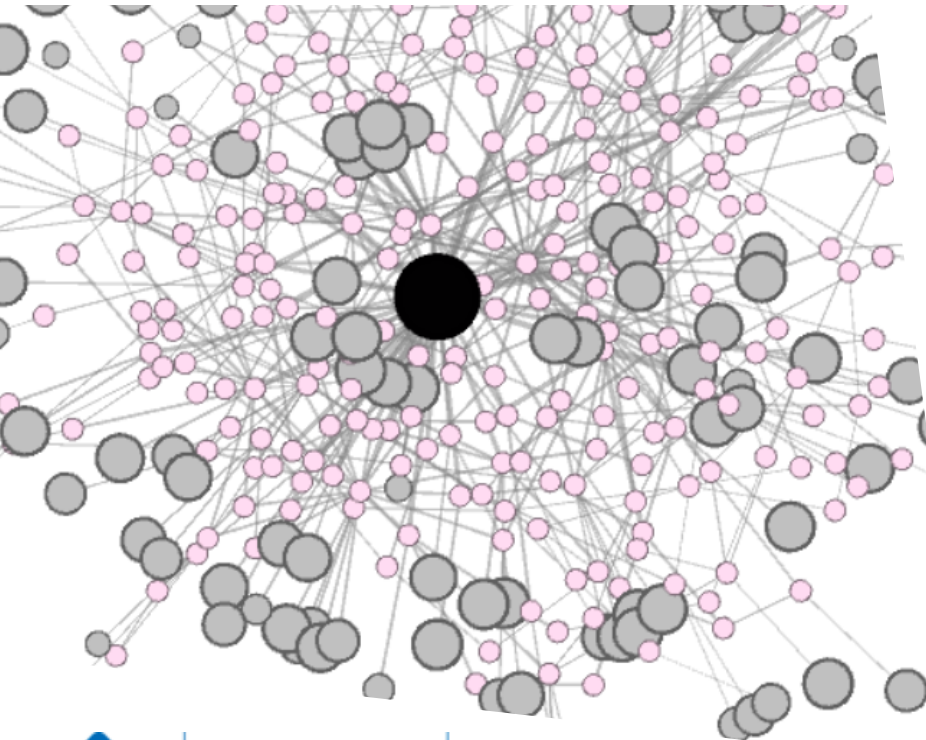
- Best centrality index: Eigenvector
- Worst centrality index: Betweenness
- Best topologies: Grid and Random Euclidean
- Increase of data types → Decrease of our approach improvements
 - Increase of data conflicts in the data nodes

Conclusion

- Data placement approach for fog architecture
 - Store the data as closer as possible to the data generators
 - Based on centrality indices
 - Improvement of number of packet transmissions
- Future works
 - Relation between topology features and results
 - Characterization of real fog architectures
 - Include a more sophisticated storage domain
 - Replicas, chunks, ...

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Thank you!!!! Q & A

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