REPORT(good)

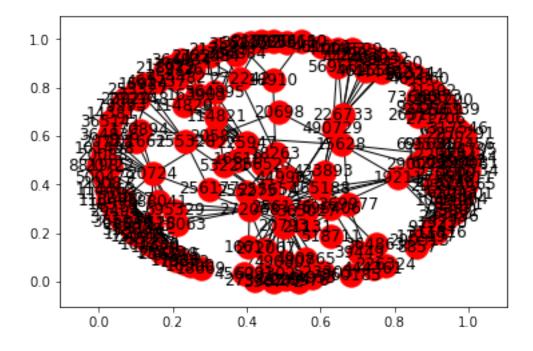
December 21, 2017

In [68]: import json

import networkx as nx

Please, give me an ID of the Author: 256176

Please, tell me, what distance of neighbors do you want?: 3



```
It was pleasure to meat you
Out[69]: <module 'Modules' from '/Users/Dario/Desktop/Modules.py'>
In [62]: dictAuthor = Modules_2.Author(data)
         dictPubl = Modules_2.Publ(data)
         dictConf = Modules_2.Conf(data)
In [63]: Gall = nx.Graph()
         for k,v in dictAuthor.items():
             Gall.add_node(k, id = v[0])
In [64]: for k,v in dictPubl.items():
             for i in itertools.combinations(v,2):
                 Gall.add_edge(i[0],i[1], weight = Modules_2.JaccardDistance(dictAuthor[i[0]],di
In [65]: print(nx.info(Gall))
Name:
Type: Graph
Number of nodes: 904664
Number of edges: 3679473
Average degree:
                  8.1345
In [24]: n = int(input())
         subgraph = Gall.subgraph(dictConf[n])
5262
```

Enter the number of part you want (2a, 2b, 3a, 3b) or type 'save me', if you want to quit: save

1 Homework 4: Group 19

2 Algorithmic Methods of Data Mining

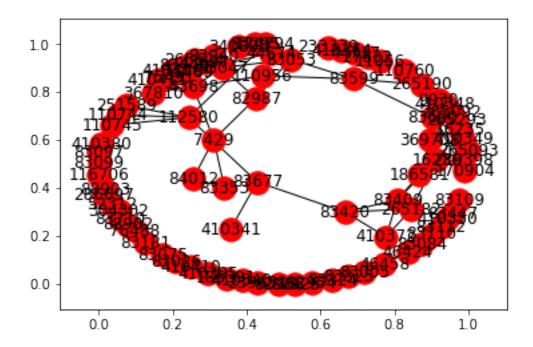
- 2.0.1 Through this PDF, we're going to explain all the results we obtained in the project.
- 2.0.2 First of all we created the Graph composed by:
 - Number of nodes: 904.664Number of edges: 3.679.473

- 2.0.3 At point 2.a, given a conference in input, we had to return the subgraph induced by the set of authors who published at the input conference at least once.
- 2.0.4 The number of the conference has to be given by input, but for showing the subgraph we've used as example the conference number 5262 with:

Number of nodes: 85Number of edges: 138

2.0.5 obtaining the following subgraph:

In [25]:



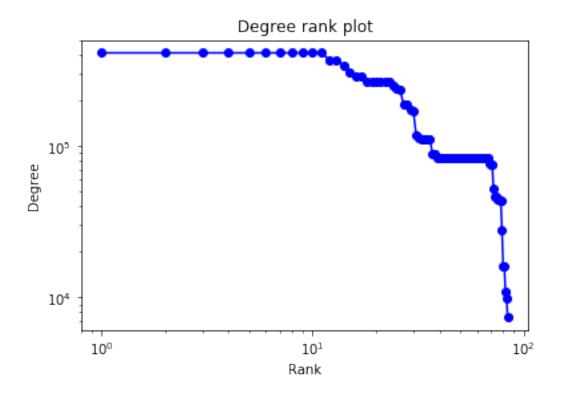
- 2.0.6 On this subgraph we computed some centralities measures (degree centrality, closeness centrality, betweeness centrality) and plotted them.
- 2.0.7 Before we show the results, we'll explain what these measures does and how they are defined.

3 Degree Centrality:

3.0.1 This measure is defined as the number of ties that a node has. If the graph were directed, we define two separate measures of degree centrality, namely *indegree* and *outdegree*.

* Indegree* is a count of the number of ties directed to the node and *outdegree* is the number of ties that the node directs to others. In this case, our graph is not directed.

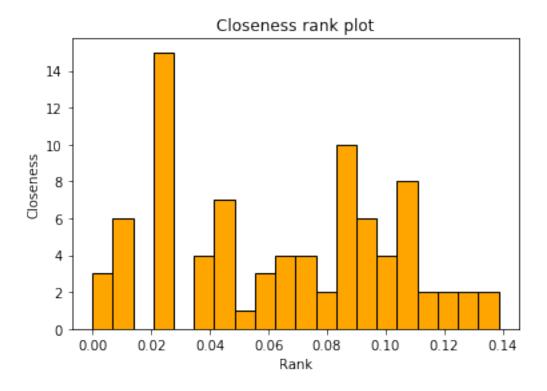
In [32]: Modules.degree(subgraph)



4 Closeness Centrality:

4.0.1 In a connected graph, the normalized closeness centrality of a node is the average length of the shortest path between the node and all other nodes in the graph. Thus the more central a node is, the closer it is to all other nodes.

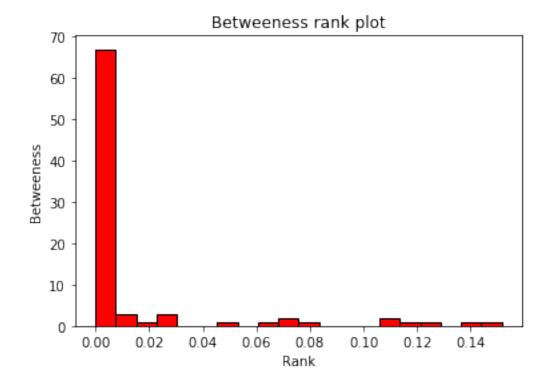
In [33]: Modules.closeness(subgraph)



5 Betweeness Centrality:

5.0.1 Betweeness Centrality represents the degree of which nodes stand between each other. It means that a node with higher betweenness would have more control over the network, because more information will pass through that node.

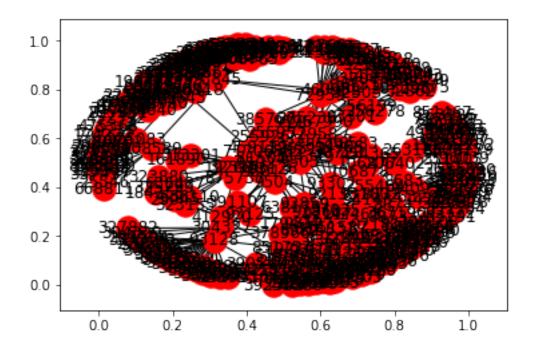
In [37]: Modules.betweeness(subgraph)



5.0.2 After statistics visualization, we created the subgraph induced by the nodes that have *hop distance* at most equal to an integer *d* with an input author.

5.0.3 The result is shown below.

```
In [77]: Modules.hop_distance(Gall)
Please, give me an ID of the Author: 93126
Please, tell me, what distance of neighbors do you want?: 2
```



5.0.4 In the exercise 3.a we had to implement the Dijkstra algorithm which calculate a generalized version of the Erdos number. In this case Erdos is Aris and our output will be the distance between an input author and Aris.

In [72]: Modules.dijkstra(Gall,256176,256177)

Out[72]: 0.9565217391304348