

VK analysis

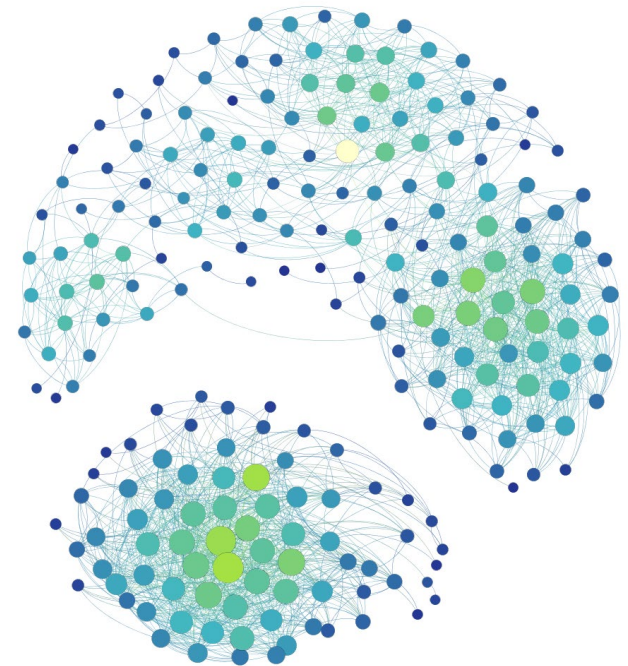
Network visualization

Konstantinov D.N.

M05-015a

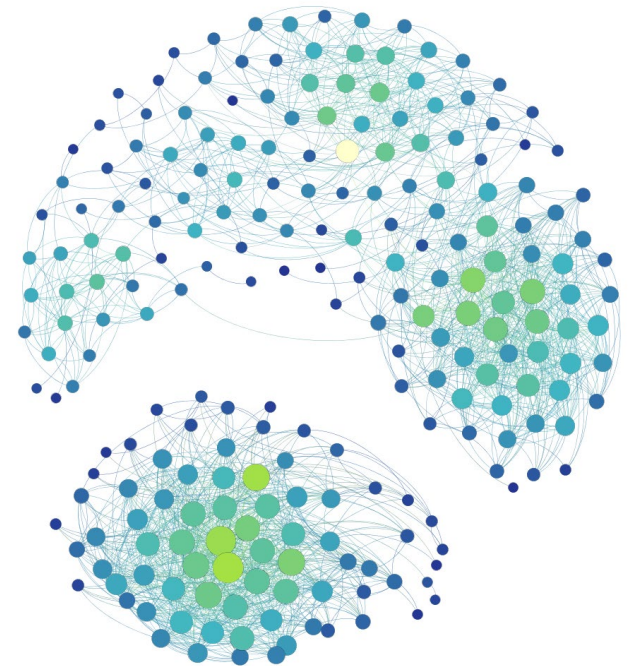
Contents

- Part 1. Network Summary
- Part 2. Structural Analysis
- Part 3. Community Detection



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- Part 1. **Network Summary**
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Network Summary

Statistics:

- Nodes - 227 (30 deleted)
- Edges - 1960
- Average node degree - 17.27
- Average clustering coefficient – 0.55
- Components - 2

Largest:

- Nodes – 151
- Edges - 1093
- Diameter – 8
- Average path length - 3

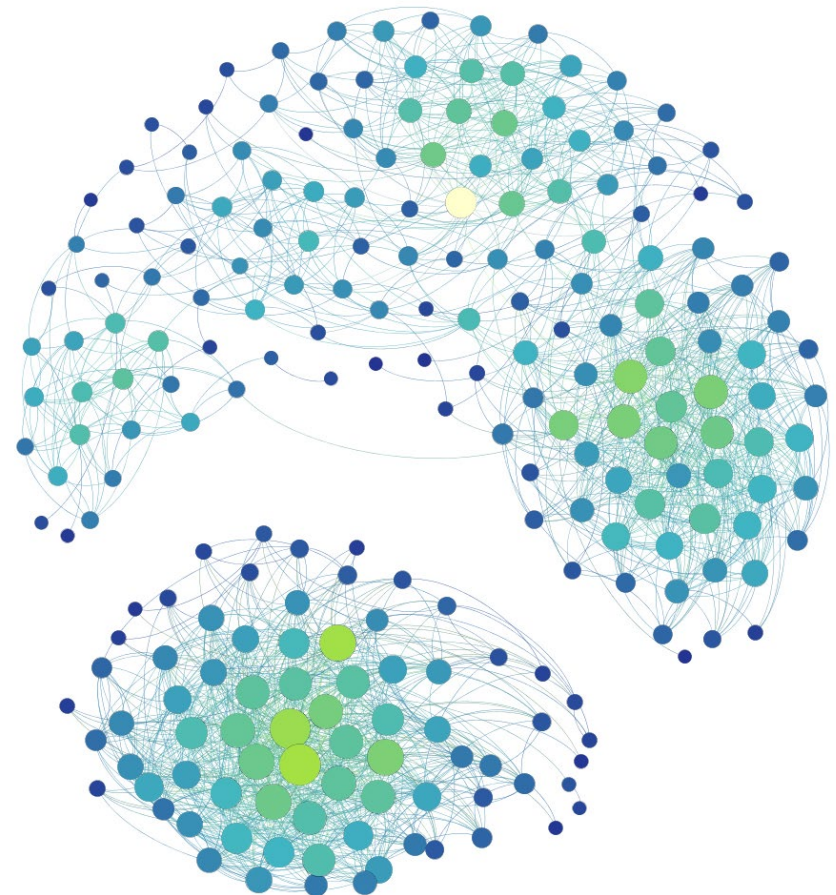
Smallest:

- Nodes – 76
- Edges - 867
- Diameter – 4
- Average path length – 1.8

Attributes of nodes:

- name
- sex
- city

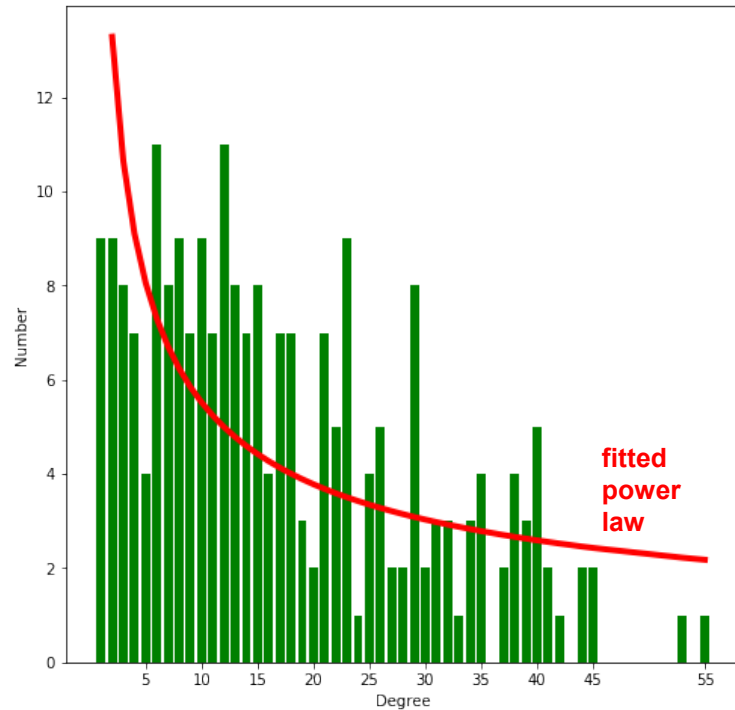
VK graph



vk.api
preprocessed;
sorted by degree

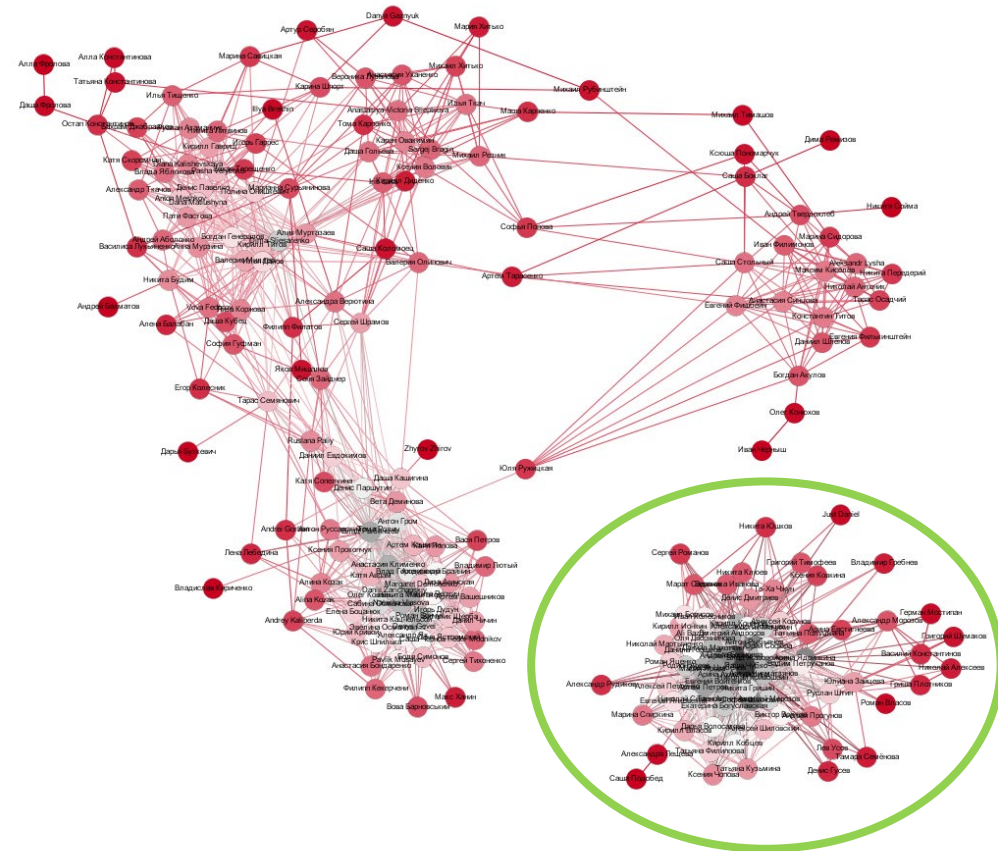
Network Summary

Degree distribution



- min degree – 1
- max degree – 55
- not similar to power law

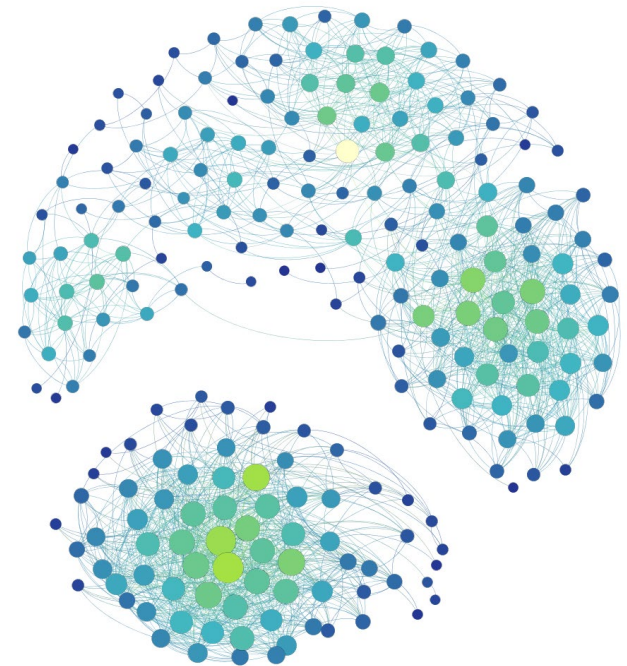
VK graph



MIPT DAFE

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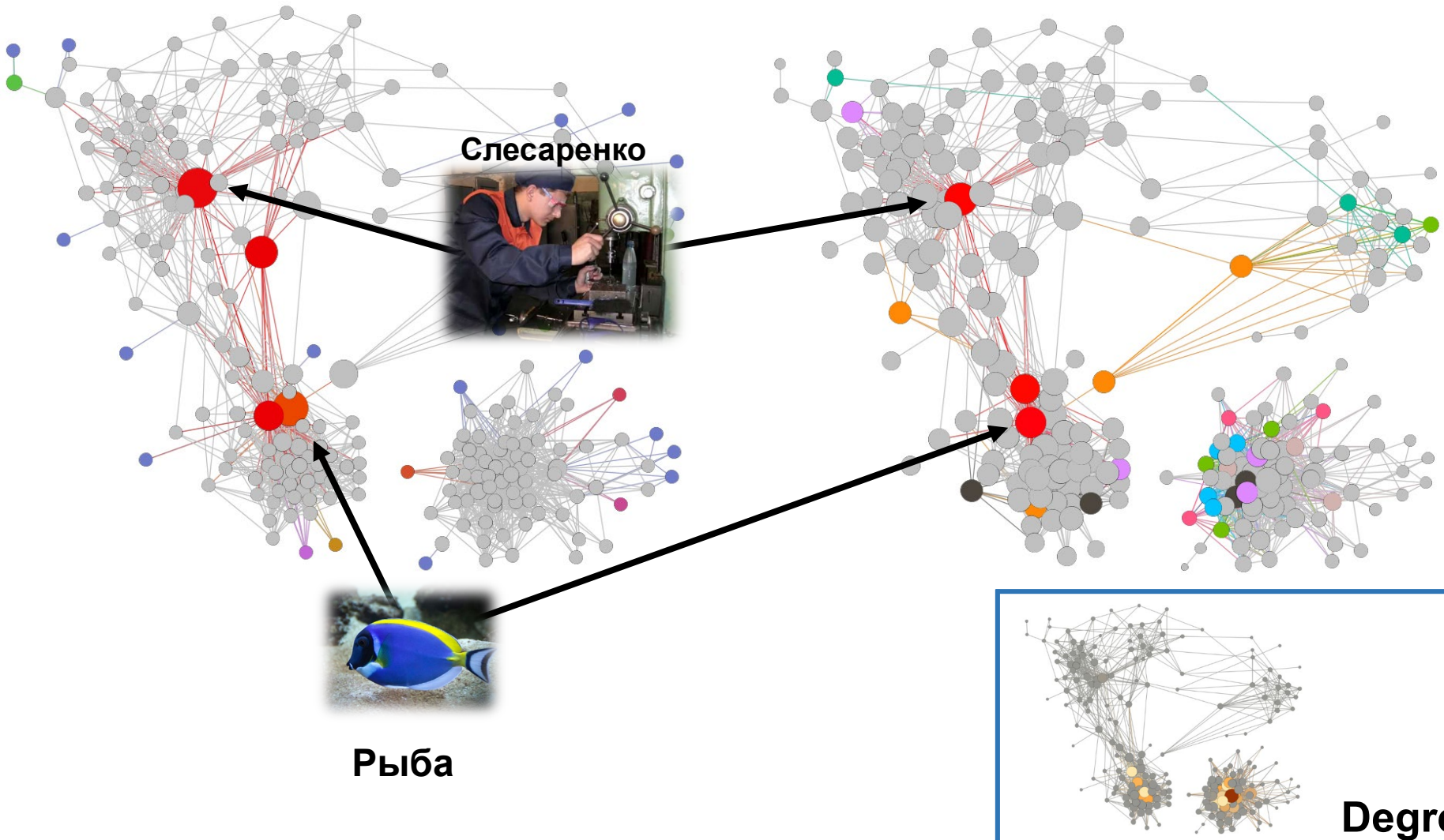
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Degree/Closeness/Betweenness centralities

Betweenness

Closeness

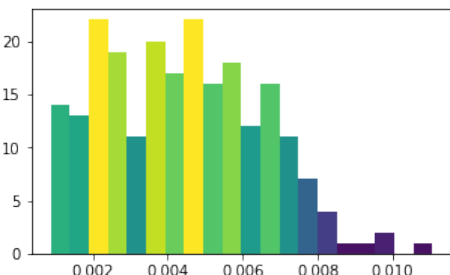


Structural Analysis

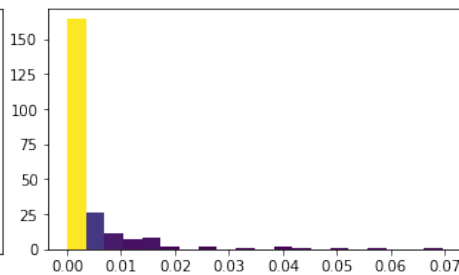
Top nodes interpretation

Degree centrality	Closeness centrality	Betweenness centrality	Pagerank
Паша Чубко	Дима Слесаренко	Дима Слесаренко	Дима Слесаренко
Ольга Борисова	Влад Рыбинцев	Антон Гром	Паша Чубко
Арина Ядринкина	Денис Паршутин	Сергей Шрамов	Арина Ядринкина
Антон Рыбьянов	Александра Верютина	Влад Рыбинцев	Ольга Борисова
Андрей Волков	Антон Гром	Валерия Олинович	Влад Рыбинцев

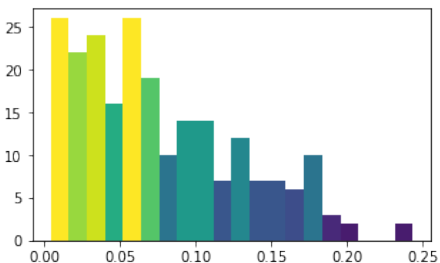
Pagerank



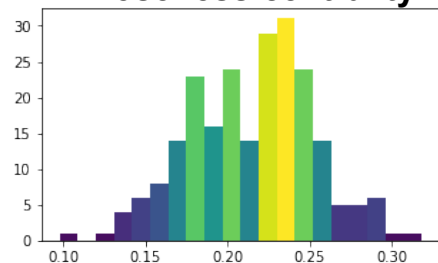
Betweenness centrality



Degree centrality

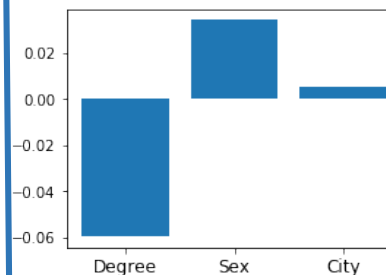


Closeness centrality



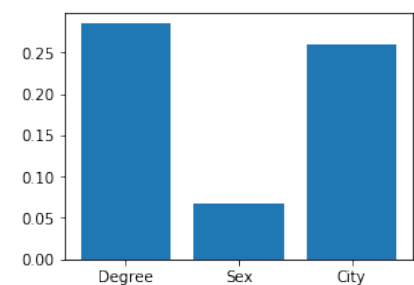
Assortative Mixing

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Disassortative by degree

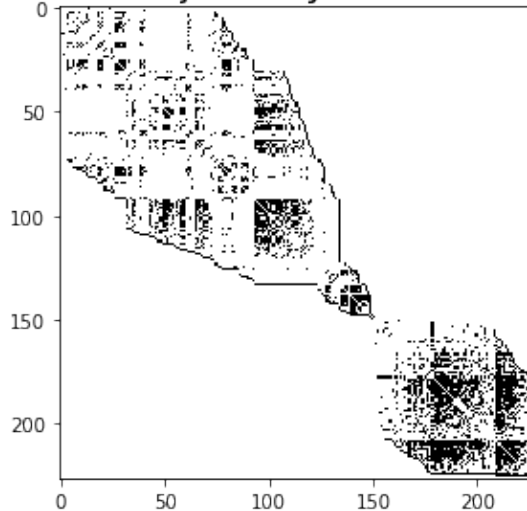
Large component



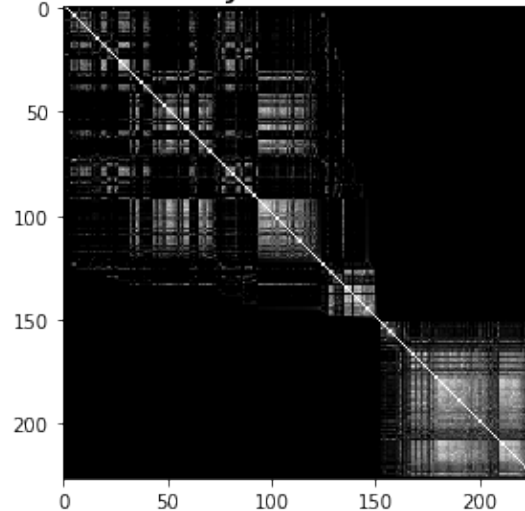
Assortative by degree and city

Node structural similarity

Adjacency Matrix

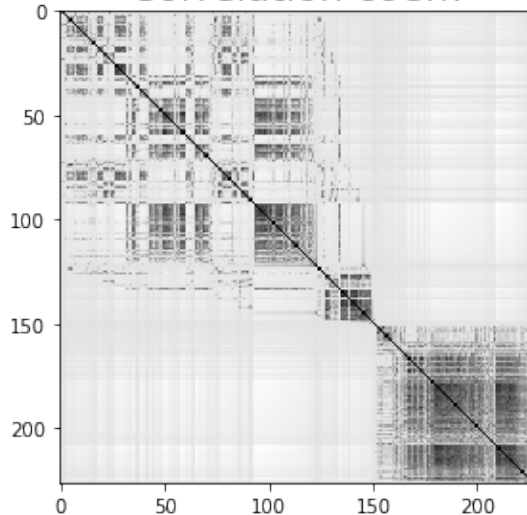


Jaccard

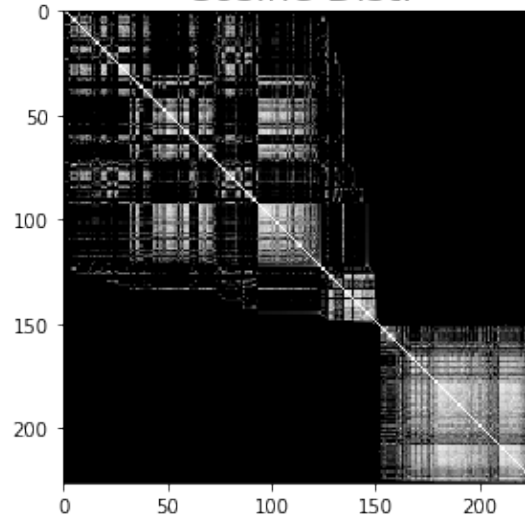


- Used the reverse Cuthill-McKee heuristic
- 3 metrics reveal ~3 clusters

Correlation coeff.



Cosine Dist.



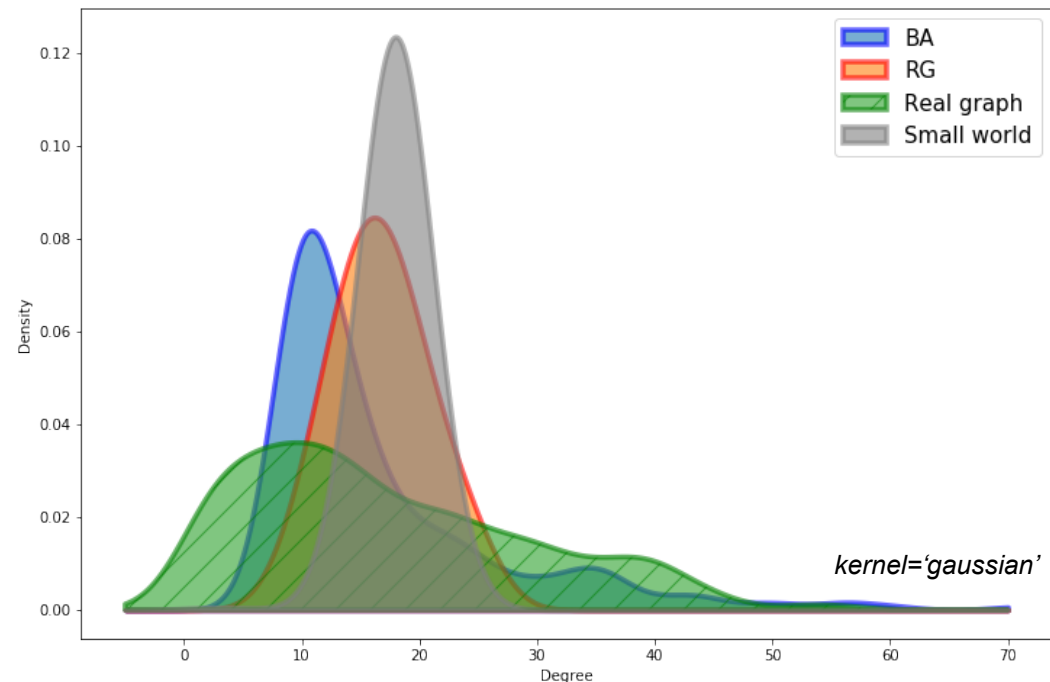
Closest random graph model similar to our network

Assumptions:

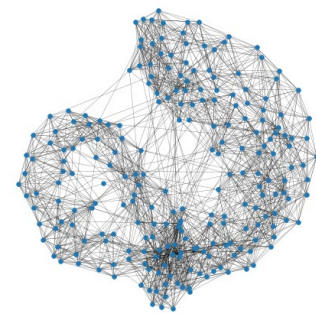
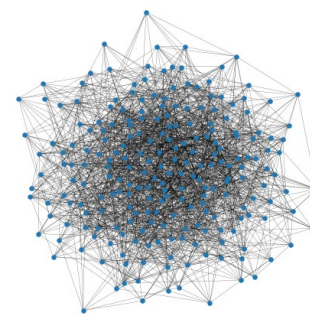
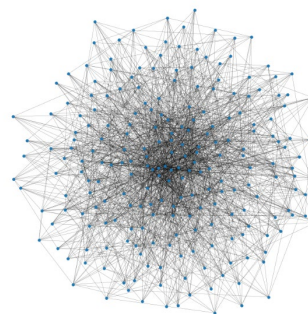
- same number of nodes and edges
- $p \approx \langle k \rangle / n$ for random graph (RG)
- $m = \langle k \rangle / 2$ for Barabase-Albert (BA)
- probability parameter for small world preferential attachment is optimized

	BA	RG	SW	real
nodes	227	227	227	227
edges	1962	1898	2041	1960
$\langle C \rangle$	0.16	0.08	0.55	0.55
$\langle L \rangle$	2.16	2.19	2.62	3; 1.8
$\langle k \rangle$	17.3	16.7	17.9	17.3
D	3	3	5	8; 4

Probability density function



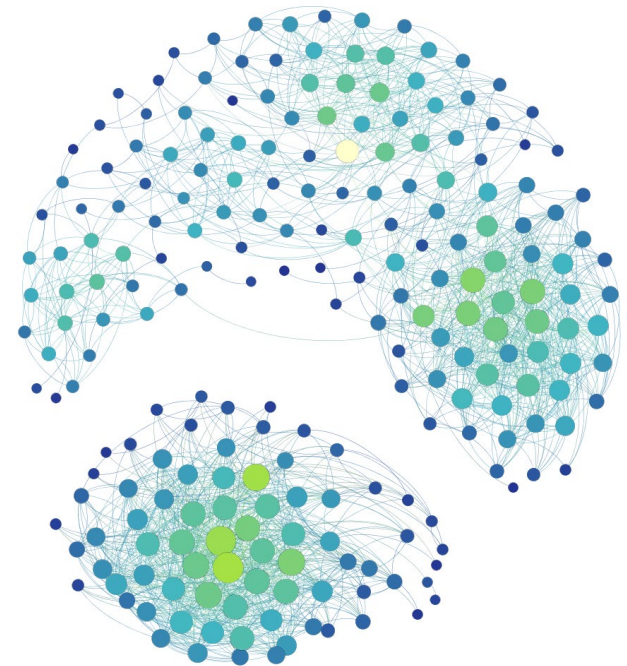
Barabási–Albert Random graph “Small world”



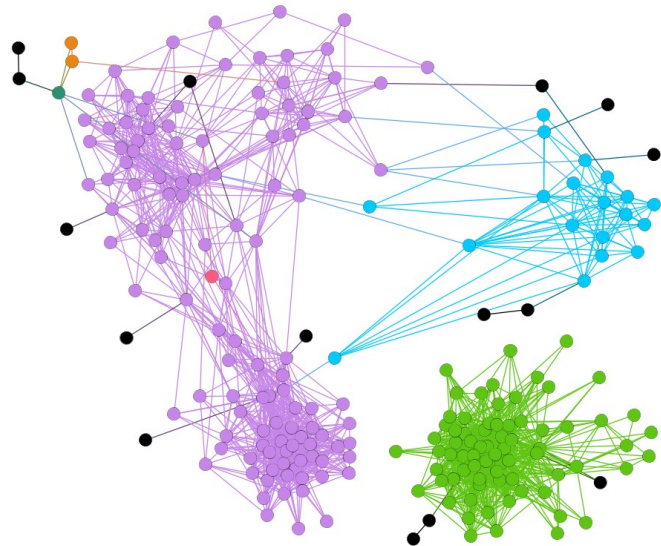
Small world has the best metrics, but still bad...

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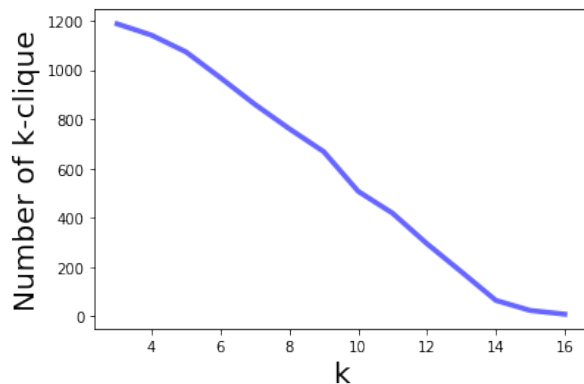
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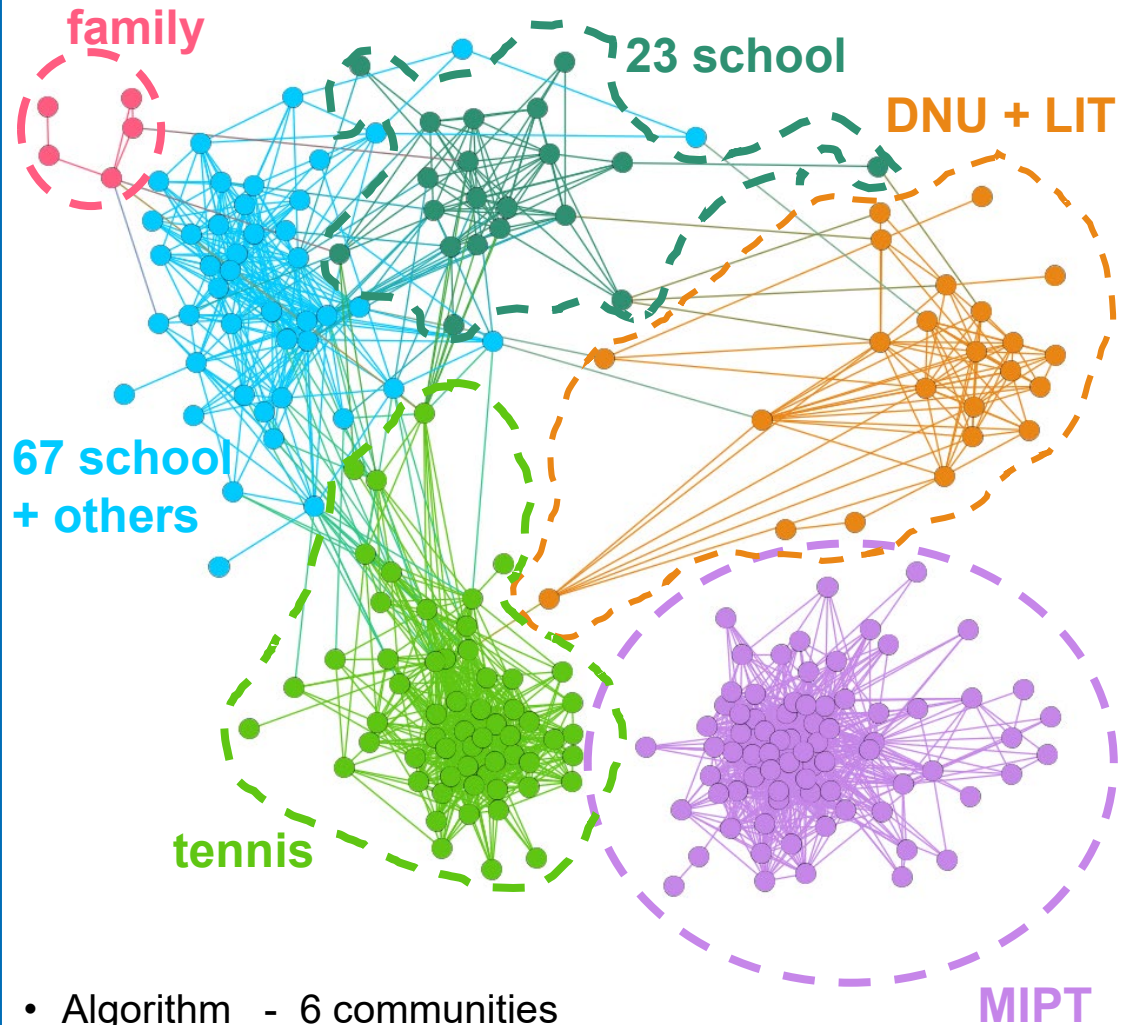
k-Clique search



#k-clique



Best results of various community detection algorithms



- Algorithm - 6 communities
- Real graph - 8 communities

igraph.community_label_propagation()
modularity = 0.35

Thank you for attention