NetPy '19: Introduction to Network Analysis in Python

Workshop instructor

Asst. Prof. Lovro Subelj, PhD, University of Ljubljana

Workshop schedule

Tuesday, 10th December 2019 at 3:30 PM (4 hours with breaks)

Workshop location

Lecture room 3 at UL FRI, Večna pot 113, Ljubljana, Slovenia

Materials and forum

- GitHub repository at http://github.com/lovre/netpy
- WWW repository at http://lovro.lpt.fri.uni-lj.si/netpy
- Piazza forum at http://piazza.com/fri.uni-lj.si/winter2020/netpy
- Slovenian dictionary at http://lovro.lpt.fri.uni-lj.si/netpy/dict.pdf

High-level description

The workshop is primarily aimed at Python programmers, either academics, professionals or students, that wish to learn the basics of modern network science and practical analyses of complex real networks, such as social, information and biological networks. Familiarity with the basics of probability theory, statistics and linear algebra is strongly encouraged. The workshop is based on masters level course Network Analysis offered at University of Ljubljana, Faculty of Computer and Information Science.

Recommended prerequisites

It is recommended that attendees bring a laptop with working installation of Python, Network and CDlib packages, and necessary dependencies. Alternatively, you can work with any other network analysis package such as Igraph, graph-tool or SNAP.py. Finally, for the purposes of visualization of smaller networks, it is recommended to have working installation of some network analysis software such as Gephi or Visione.

Tentative syllabus

- Challenge: Warmup Guimera's four knights challenge (10 min)
- 1. From classical graph theory to **modern network science** (30 min)
- 2. Large-scale structure of real networks and graph models (50 min)
- 3. Measures of **node importance** and link analysis **algorithms** (50 min)
- 4. Network community, core-periphery and other structures (50 min)
- 5. Network-based mining, visualization and some applications (50 min)
- **Hands-on**: Abstraction, centrality, communities, mining, visualization etc.

Networks data

All networks are available in Pajek, edge list in LNA formats.

- Simple toy example network (5 nodes)
- Zachary's karate club network (34 nodes)
- Davis's southern women network (32 nodes)
- Lusseau's bottlenose dolphins network (62 nodes)
- Game of Thrones character appearance network (107 nodes)
- Human diseasome network by common symptoms (117 nodes)
- Conflicts and alliances between world nations (180 nodes)
- Game of Thrones character kills network (284 nodes)
- Ljubljana public bus transport network (416 nodes)
- US airplane traffic transport network (1,323 nodes)
- Java software class dependency network (1,516 nodes)
- Ingredients network by common compounds (1,525 nodes)
- Map of Darknet from Tor network (7,178 nodes)
- IMDb actors collaboration network (17,577 nodes)
- Human protein-protein interaction network (19,634 nodes)
- WikiLeaks cable reference network (52,416 nodes)
- Internet map of autonomous systems (75,885 nodes)
- Amazon product copurchase network (262,111 nodes)
- Paper citation network of APS (438,943 nodes)
- Small part of Google web graph (875,713 nodes)
- Road/highway network of Texas (1,379,917 nodes)

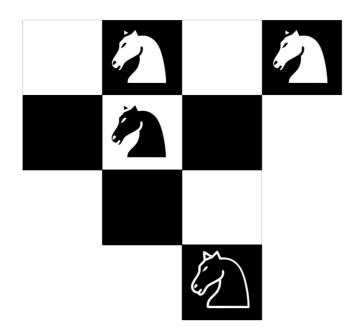
Let's start with Guimera's four knights challenge

Tentative duration

5+5 min

Challenge description

To be **revealed in class** =)



1. Classical graph theory → modern network science

Tentative duration

20+10 min

Brief description

Introduction of networks and selected **motivational examples**. From **classical graph theory** to social network analysis and **modern network science**. Network perspectives in different **fields of science**.



Lecture slides

- Networks introduction and motivational examples
- Brief historical development of network science
- <u>Network perspectives through science</u> (tentative)

Book chapters

- Ch. 1: Introduction in Barabási, A.-L., Network Science (Cambridge University Press, 2016).
- Ch. 1-5: Introduction etc. in Newman, M.E.J., <u>Networks: An Introduction</u> (Oxford University Press, 2010).
- Ch. 1: <u>Overview</u> in Easley, D. & Kleinberg, J., <u>Networks, Crowds, and Markets</u> (Cambridge University Press, 2010).

Selected must-reads

- Barabási, A.-L., The network takeover, *Nat. Phys.* **8**(1), 14-16 (2012).
- Motter, A.E. & Yang, Y., The unfolding and control of network cascades, *Phys. Today* 70(1), 33-39 (2017).
- Cramer, C., Porter, M.A. et al., <u>Network Literacy: Essential Concepts and Core Ideas</u> (Creative Commons Licence, 2015).

Selected papers

- Newman, M.E.J., The physics of networks, *Phys. Today* **61**(11), 33-38 (2008).
- Cimini, G., Squartini, T. et al., <u>The statistical physics of real-world networks</u>, *Nat. Rev. Phys.* **1**(1), 58-71 (2019).
- Newman, M.E.J., Communities, modules and large-scale structure in networks, *Nat. Phys.* 8(1), 25-31 (2012).
- Vespignani, A., Modelling dynamical processes in complex socio-technical systems, *Nat. Phys.* **8**(1), 32-39 (2012).
- Hegeman, T. & Iosup, A., <u>Survey of graph analysis applications</u>, e-print arXiv:180700382v1, pp. 23 (2018).
- Hidalgo, C.A., Disconnected, fragmented, or united? A trans-disciplinary review of network science,
 Appl. Netw. Sci. 1, 6 (2016).

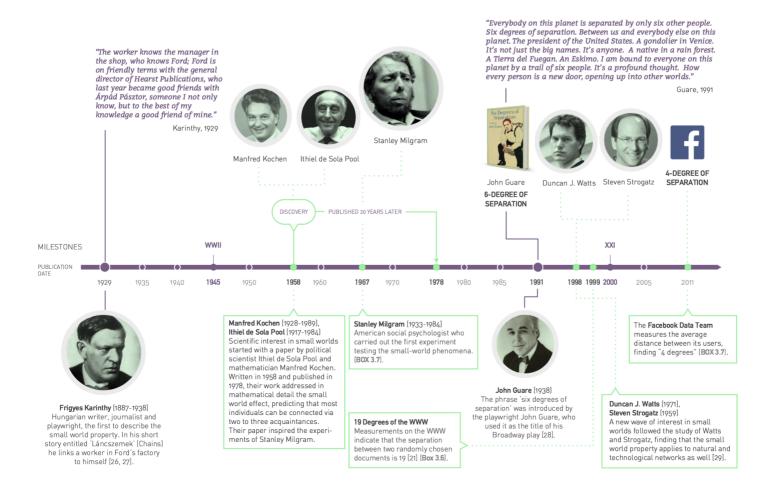
2. Large-scale network structure and graph models

Tentative duration

30+20 min

Brief description

Classical graph theory and modern network analysis. Random graphs, scale-free and small-world network models, and real network structure. Network representations, data formats and repositories.



Lecture slides

- Graph theory and modern network analysis
- Random graph models and network structure
- Scale-free and small-world network models
- Network representations, formats and data

Hands-on analysis

- Analysis of network structure and random graphs
- Starter script for network structure and random graphs

Networks data

- Zachary's karate club network in Pajek format
- <u>Davis's southern women network</u> in Pajek format
- Lusseau's bottlenose dolphins network in Pajek format
- <u>Ingredients network by common compounds</u> in Pajek format
- Map of Darknet from Tor network in Pajek format

- Human protein-protein interaction network in Pajek format
- Internet map of autonomous systems in Pajek format
- Amazon product copurchase network in Pajek format
- Paper citation network of APS in Pajek format
- Small part of Google web graph in Pajek format
- Road/highway network of Texas in Pajek format

Book chapters

- Ch. 2: <u>Graph theory</u>, Ch. 3.8-3.9: <u>Small worlds etc.</u> & Ch. 4-5: <u>Scale-free property etc.</u> in Barabási, A.-L., Network Science (Cambridge University Press, 2016).
- Ch. 6: Mathematics of networks & Ch. 12-15: Random graphs etc. in Newman, M.E.J., <u>Networks: An Introduction</u> (Oxford University Press, 2010).
- Ch. 2: <u>Graphs</u>, Ch. 18: <u>Power laws etc.</u> & Ch. 20: <u>Small-world phenomenon</u> in Easley, D. & Kleinberg,
 J., Networks, Crowds, and Markets (Cambridge University Press, 2010).

Selected must-reads

- Newman, M.E.J., Watts, D.J. & Strogatz, S.H., Random graph models of social networks, P. Natl. Acad. Sci. USA 99, 2566-2572 (2002).
- Ugander, J., Karrer, B. et al., <u>The anatomy of the Facebook social graph</u>, e-print arXiv:1111.4503v1, pp. 17 (2011).
- Backstrom, L., Boldi, P. et al., <u>Four degrees of separation</u>, In: *Proceedings of WebSci '12* (Evanston, IL, USA, 2012), pp. 45-54.

Selected papers

- Erdős, P. & Rényi, A., On random graphs I, Publ. Math. Debrecen 6, 290-297 (1959).
- Milgram, S., The small world problem, Psychol. Today 1(1), 60-67 (1967). Granovetter, M.S., The strength of weak ties, Am. J. Sociol. 78(6), 1360-1380 (1973).
- Watts, D.J. & Strogatz, S.H., Collective dynamics of 'small-world' networks, *Nature* 393(6684), 440-442 (1998).
- Barabási, A.-L. & Albert, R., Emergence of scaling in random networks, Science 286(5439), 509-512 (1999).
- Faloutsos, M., Faloutsos, P. & Faloutsos, C., On power-law relationships of the Internet topology, *Comput. Commun. Rev.* 29(4), 251-262 (1999).
- Albert, R., Jeong, H. & Barabási, A.-L., Error and attack tolerance of complex networks, *Nature* 406(6794), 378-382 (2000).
- Dorogovtsev, S.N. & Mendes, J.F.F., <u>Evolution of networks</u>, Adv. Phys. 51(4), 1079-1187 (2002).
- Clauset, A., Shalizi, C.R. & Newman, M.E.J., <u>Power-law distributions in empirical data</u>, *SIAM Rev.* 51, 661-703 (2009).
- De Domenico, M. & Arenas, A., <u>Modeling structure and resilience of the dark network</u>, *Phys. Rev. E* 95(2), 022313 (2017).

- Broido, A.D. & Clauset, A., Scale-free networks are rare, Nat. Commun. 10(1), 1017 (2019).
- Barabási, A.-L., <u>Love is all you need</u>, reply to e-print arXiv:1801.03400v1, pp. 6 (2018).
- Holme, P., Rare and everywhere, Nat. Commun. 10(1), 1016 (2019).

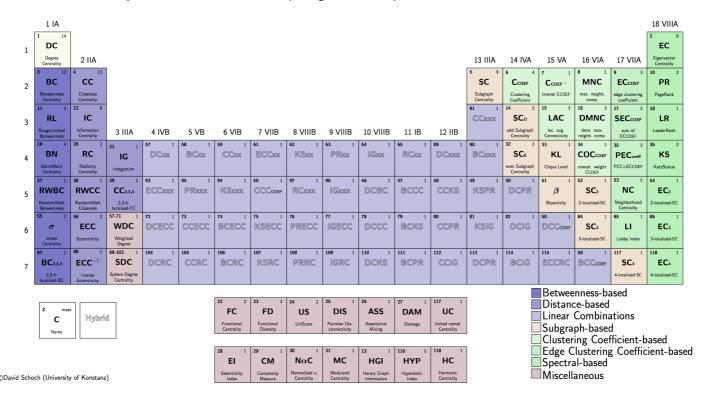
3. Measures of node importance and link analysis

Tentative duration

30+20 min

Brief description

Node importance and **measures of centrality**, i.e. clustering coefficients, spectral, closeness and betweenness centrality, and link analysis algorithms. **Link importance** and **measures of bridging**, i.e. betweenness centrality, embeddedness and topological overlap.



Lecture slides

- Node importance and measures of centrality
- <u>Link analysis for web page importance</u> (tentative)
- <u>Link importance and measures of bridging</u> (tentative)

Hands-on analysis

- Analysis of IMDb actors collaboration network
- Starter script for IMDb actors collaboration network

Networks data

• IMDb actors collaboration network in Pajek format

Book chapters

- Ch. 7: Measures and metrics in Newman, M.E.J., <u>Networks: An Introduction</u> (Oxford University Press, 2010).
- Ch. 14: <u>Link analysis and Web search</u> in Easley, D. & Kleinberg, J., <u>Networks, Crowds, and Markets</u> (Cambridge University Press, 2010).
- Ch. 14-15: Classical node centrality etc. in Estrada, E. & Knight, P.A., <u>A First Course in Network</u>
 Theory (Oxford University Press, 2015).

Selected must-reads

- Jeong, H., Mason, S.P. et al., Lethality and centrality in protein networks, *Nature* 411, 41-42 (2001).
- Jensen, P., Morini, M. et al., <u>Detecting global bridges in networks</u>, *J. Complex Netw.* 4(3), 319-329 (2015).
- Tong, H., Faloutsos, C. & Pan, J.-Y., Fast random walk with restart and its applications, In: *Proceedings of ICDM '06* (Washington, DC, USA, 2006), pp. 613-622.

Selected papers

- Freeman, L., A set of measures of centrality based on betweenness, *Sociometry* **40**(1), 35-41 (1977).
- Bonacich, P., Power and centrality: A family of measures, Am. J. Sociology 92(5), 1170-1182 (1987).
- Kleinberg, J., Authoritative sources in a hyperlinked environment, *J. ACM* **46**(5), 604-632 (1999).
- Franceschet, M. & Bozzo, E., <u>A theory on power in networks</u>, e-print arXiv:1510.08332v2, pp. 19 (2016).
- Everett, M.G. & Valente, T.W., Bridging, brokerage and betweenness, Soc. Networks 44, 202-208 (2016).
- Berkhin, P., A survey on PageRank computing, *Internet Math.* **2**(1), 73-120 (2005).

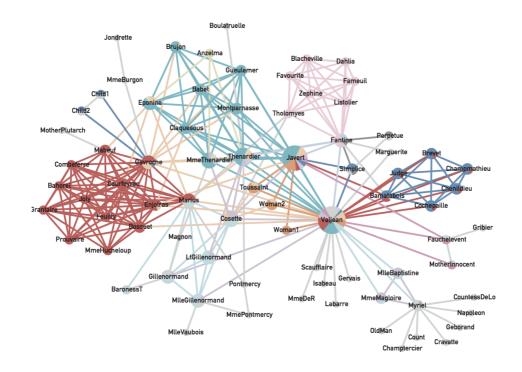
4. Clusters of nodes and network mesoscopic structure

Tentative duration

30+20 min

Brief description

Network **community**, **core-periphery** and other **mesoscopic** structures. Graph **partitioning**, community **detection**, **blockmodeling**, stochastic block models and core-periphery detection.



Lecture slides

- Node clustering and community structure
- Graph partitioning and community detection
- Blockmodeling and block models (tentative)
- <u>Core-periphery decomposition</u> (tentative)

Hands-on analysis

- Analysis of community and k-cores structure
- Starter script for community and k-cores detection

Networks data

- Zachary's karate club network in Pajek format
- Davis's southern women network in Pajek format
- Lusseau's bottlenose dolphins network in Pajek format
- Game of Thrones character appearance network in Pajek format
- Human diseasome network by common symptoms in Pajek format
- Conflicts and alliances between world nations in Pajek format
- US airplane traffic transport network in Pajek format
- Java software class dependency network in Pajek format
- IMDb actors collaboration network in Pajek format
- WikiLeaks cable reference network in Pajek format

Book chapters

• Ch. 9: Communities in Barabási, A.-L., Network Science (Cambridge University Press, 2016).

- Ch. 7.12-7.13: Homophily etc. & Ch. 11: Graph partitioning in Newman, M.E.J., <u>Networks: An Introduction</u> (Oxford University Press, 2010).
- Ch. 21: Communities in networks in Estrada, E. & Knight, P.A., <u>A First Course in Network Theory</u> (Oxford University Press, 2015).
- Ch. 3: <u>Strong and weak ties</u> in Easley, D. & Kleinberg, J., <u>Networks, Crowds, and Markets</u> (Cambridge University Press, 2010).

Selected must-reads

- Hric, D., Darst, R.K. & Fortunato, S., <u>Community detection in networks: Structural communities versus</u> ground truth, *Phys. Rev. E* 90(6), 062805 (2014).
- Fortunato, S. & Hric, D., Community detection in networks: A user guide, Phys. Rep. 659, 1-44 (2016).
- Schaub, M.T., Delvenne, J.-C. et al., <u>The many facets of community detection in complex networks</u>, *Appl. Netw. Sci.* 2, 4 (2017).
- Rossetti, G., Milli, L., & Cazabet, R., <u>CDlib: A python library to extract, compare and evaluate communities from complex networks</u>, *Appl. Netw. Sci.* 4(1), 1–26 (2019).

Selected papers

- Granovetter, M.S., The strength of weak ties, Am. J. Sociol. 78(6), 1360-1380 (1973).
- Girvan, M. & Newman, M.E.J., <u>Community structure in social and biological networks</u>, *P. Natl. Acad. Sci. USA* 99(12), 7821-7826 (2002).
- Fortunato, S., Community detection in graphs, Phys. Rep. 486(3-5), 75-174 (2010).
- Leskovec, J., Lang, K.J. et al., <u>Community structure in large networks</u>, *Internet Math.* **6**(1), 29–123 (2009).
- Borgatti, S.P. & Everett, M.G., Models of core/periphery structures, Soc. Networks 21(4), 375–395 (2000).
- Holme, P., Core-periphery organization of complex networks, Phys. Rev. E 72(4), 46111 (2005).
- Newman, M.E.J. & Leicht, E.A., <u>Mixture models and exploratory analysis in networks</u>, *P. Natl. Acad. Sci. USA* 104(23), 9564 (2007).
- Raghavan, U.N., Albert, R. & Kumara, S., <u>Near linear time algorithm to detect community structures in large-scale networks</u>, *Phys. Rev. E* 76(3), 036106 (2007).
- Rosvall, M. & Bergstrom, C.T., Maps of random walks on complex networks reveal community structure, P. Natl. Acad. Sci. USA 105(4), 1118-1123 (2008).
- Blondel, V.D., Guillaume, J.-L. et al., <u>Fast unfolding of communities in large networks</u>, *J. Stat. Mech.*, P10008 (2008).
- Traag, V.A., Waltman, L. & Van Eck, N.J., <u>From Louvain to Leiden: Guaranteeing well-connected communities</u>, *Sci. Rep.* **9**, 5233 (2019).
- Peixoto, T.P., <u>Bayesian stochastic blockmodeling</u>, e-print *arXiv:170510225v7*, pp. 44 (2018).

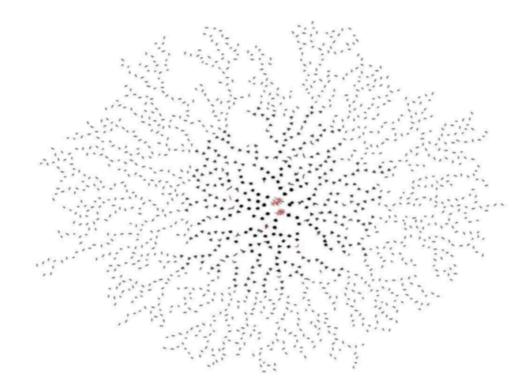
5. Network-based mining, visualization and applications

Tentative duration

30+20 min

Brief description

Network-based node clustering, classification and regression. Force-directed node layout and network visualization. Selected applications of network analysis (i.e. automobile insurance fraud and historical development of science).



Lecture slides

- Network-based node classification etc.
- Network layout and visualization (tentative)
- Fraud detection and automobile insurance
- Historical development of scientific knowledge

Hands-on analysis

• Demo script for network-based classification

Networks data

- Zachary's karate club network in Pajek format
- Java software class dependency network in Pajek format

Selected must-reads

- Grover, A. & Leskovec, J., node2vec, In: Proceedings of KDD '16 (San Francisco, CA, USA, 2016), pp. 855-864.
- Zanin, M., Papo, D. et al., <u>Combining complex networks and data mining: Why and how, Phys. Rep.</u>
 635, 1-44 (2016).
- Ma, K.-L. & Muelder, C.W., Large-scale graph visualization and analytics, Computer 46(7), 39-46 (2013).

Selected papers

- Getoor, L. & Diehl, C.P., Link mining: A survey, SIGKDD Explor. 7(2), 3–12 (2005).
- Getoor, L., Friedman, N. et al., Learning probabilistic models of link structure, J. Mach. Learn. Res. 3, 679–707 (2002).
- Neville, J. & Jensen, D., Iterative classification in relational data, In: *Proceedings of SRL '00* (Austin, TX, USA, 2000), pp. 13–20.
- Macskassy, S.A. & Provost, F., Classification in networked data: A toolkit and a univariate case study,
 J. Mach. Learn. Res. 8, 935-983 (2007).
- Bhagat, S., Cormode, G. & Muthukrishnan, S., Node classification in social networks, e-print arXiv:1101.3291v1, pp. 37 (2011).
- Šubelj, L., Exploratory and predictive tasks of network community detection, In: *Proceedings of NetSci* '15 (Zaragoza, Spain, 2015), p. 1.
- Hric, D., Peixoto, T.P. & Fortunato, S., <u>Network structure</u>, <u>metadata and the prediction of missing nodes</u>, *Phys. Rev. X* 6(3), 031038 (2016).
- Perozzi, B., Al-Rfou, R. & Skiena, S., <u>DeepWalk</u>, In: *Proceedings of KDD '14* (New York, NY, USA, 2014), pp. 701-710.
- Figueiredo, D.R., Ribeiro, L.F.R. & Saverese, P.H.P., <u>struc2vec</u>, In: *Proceedings of KDD '17* (Halifax, Canada, 2017), pp. 1–9.
- Peel, L., <u>Graph-based semi-supervised learning for relational networks</u>, In: *Proceedings of SDM '17*(Houston, TX, USA, 2017), pp. 1-11.
- Eades, P., A heuristic for graph drawing, Congressus Numerantium 42, 146-160 (1984).
- Kamada, T. & Kawai, S., An algorithm for drawing general undirected graphs, *Inform. Process. Lett.* **31**(1), 7-15 (1989).
- Fruchterman, T.M.J. & Reingold, E.M., Graph drawing by force-directed placement, *Softw: Pract. Exper.* **21**(11), 1129-1164 (1991).
- Kobourov, S.G., <u>Spring embedders and force directed graph drawing algorithms</u>, e-print arXiv:1201.3011v1, pp. 23 (2012).
- Gibson, H., Faith, J. & Vickers, P., A survey of two-dimensional graph layout techniques for information visualisation, *Infor. Visual.* **12**(3-4), 324-357 (2013).
- Šubelj, L., Furlan, Š. & Bajec, M., <u>An expert system for detecting automobile insurance fraud using social network analysis</u>, *Expert Syst. Appl.* 38(1), 1039-1052 (2011).

• Šubelj, L., Waltman, L. et al., Intermediacy of publications, R. Soc. Open Sci., pp. 19 (2019).