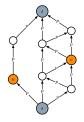
applications science

introduction to network science in Python (NetPy)

Lovro Šubelj University of Ljubljana 14th Dec 2021

science historiography

- algorithmic historiography tracks evolution of field
- relying on citations between scientific publications

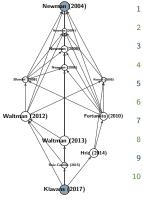


(input) selected source & target publications s & t
 (method) each citation is relevant/active with probability p
(output) importance is probability of path from s to t through u

science modularity

(target) Newman & Girvan (2004), Finding and evaluating community..., Phys. Rev. E 69(2), 026113.

(source) Klavans & Boyack (2017), Which type of citation analysis generates..., JASIST 68(4), 984-998.



- Waltman & Van Eck (2013), A smart local moving algorithm for largescale modularity-based community detection, *EPJB* **86**, 471.
- Waltman & Van Eck (2012), A new methodology for constructing a publication-level classification system..., ASIST 63(12), 2378-2392.
 Hric et al. (2014), Community detection in networks: Structural community detection in networks.
 - munities versus ground truth, *Phys. Rev. E* **90**(6), 062805.
 - Fortunato (2010), Community detection in graphs, *Phys. Rep.* **486**(3-5), 75-174.
- Newman (2006), Modularity and community structure in networks, PNAS 103(23), 8577-8582.
- Ruiz-Castillo & Waltman (2015), Field-normalized citation impact indicators using algorithmically.... J. Informetr. 9(1), 102-117.
- Blonds et al. (2008), Fast unfolding of communities in large networks,
- J. Stat. Mech., P10008.

 Newman (2006). Finding community structure in networks using the
- eigenvectors of matrices, *Phys. Rev. E* **74**(3), 036104.

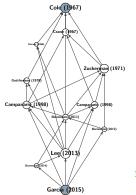
 Newman (2004). Fast algorithm for detecting community structure in
- Newman (2004), Fast algorithm for detecting community structure in networks, *Phys. Rev. E* 69(6), 066133.
 - Rosvall & Bergstrom (2008), Maps of random walks on complex networks reveal community structure, *PNAS* **105**(4), 1118-1123.

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science peer review

(target) Cole & Cole (1967), Scientific output and recognition, Am. Sociol. Rev. 32(3), 377-390.

(source) Garcia et al. (2015), The author-editor game, Scientometrics 104(1), 361-380.



- 1 Lee et al. (2013), Bias in peer review, JASIST 64(1), 2-17.
- Zuckerman & Merton (1971), Patterns of evaluation in science: Institutionalisation, structure and functions..., Minerva 9(1), 66-100.
- 3 Campanario (1998), Peer review for journals as it stands today: Part 1, Sci. Commun. 19(3), 181-211.
- 4 Crane (1967), The gatekeepers of science: Some factors affecting the selection of articles for scientific journals, Am. Sociol. 2(4), 195-201.
- Campanario (1998), Peer review for journals as it stands today: Part
 2. Sci. Commun. 19(4), 277-306.
- Gottfredson (1978), Evaluating psychological research reports: Dimensions, reliability, and correlates. Am Psychol. 33(10), 920, 934
- sions, reliability, and correlates..., Am. Psychol. 33(10), 920-934.

 Bornmann (2011), Scientific peer review, Annu. Rev. Inform. Sci.
- 45(1), 197-245.

 Bornmann (2012), The Hawthorne effect in journal peer review, *Sci*-
- entometrics 91(3), 857-862.
- 9 Bornmann (2014), Do we still need peer review? An argument for change, JASIST 65(1), 209-213.
- 10 Merton (1968), The Matthew effect in science, Science 159(3810), 56-63.

science small-world

(target) Watts & Strogatz (1998), Collective dynamics of 'small-world' networks, Nature 393(6684), 440-442.

(source) Backstrom et al. (2012), Four degrees of separation, In: Proceedings of the WebSci '12, pp. 45-54.

- Newman (2003), The structure and function of complex networks, SIAM Rev. 45(2), 167-256.
- Albert & Barabási (2002), Statistical mechanics of complex networks, Rev. Mod. Phys. 74(1), 47-97.
- 3 Li et al. (2005), Towards a theory of scale-free graphs: Definition, properties, and implications, *Internet Math.* 2(4), 431-523.
- 4 Leskovec et al. (2007), Graph evolution: Densification and shrinking diameters, ACM Trans. Knowl. Discov. Data 1(1), 1-41.
- 5 Liben-Nowell et al. (2005), Geographic routing in social networks, P. Natl. Acad. Sci. USA 102(33), 11623-11628.
- 6 Strogatz (2001), Exploring complex networks, Nature 410(6825), 268-276.
- 7 Boldi et al. (2011), Layered label propagation: A multiresolution coordinate-free ordering for compressing social networks, In: *Proceedings of the WWW '11*, pp. 587-596.
- 8 Dorogovtsev (2002), Evolution of networks, Adv. Phys. 51(4), 1079-1187.
- 9 Ye et al. (2010), Distance distribution and average shortest path length estimation in real-world networks, In: Proceedings of the ADMA '10, pp. 322-333.
- Lattanzi et al. (2011), Milgram-routing in social networks, In: Proceedings of the WWW '11, pp. 725-734.

science scale-free

```
(target) Barabási & Albert (1999), Emergence of scaling in
random networks, Science 286(5439), 509-512.
(source) Liu et al. (2011), Controllability of
complex networks, Nature 473(7346), 167-173.
```

- Albert & Barabási (2002), Statistical mechanics of complex networks, Rev. Mod. Phys. 74(1), 47-97.
- Strogatz (2001), Exploring complex networks, Nature 410(6825), 268-276.
- Boguñá et al. (2004), Cut-offs and finite size effects in scale-free networks, Eur. Phys. J. B 38(2), 205-209.
- 4 Nishikawa et al. (2003), Heterogeneity in oscillator networks: Are smaller worlds easier to synchronize?, Phys. Rev. Lett. 91(1), 014101.
- 5 Kim & Motter (2009), Slave nodes and the controllability of metabolic networks, New J. Phys. 11, 113047.
- 6 Newman (2003), The structure and function of complex networks, SIAM Rev. 45(2), 167-256.
- 7 Sorrentino et al. (2007), Controllability of complex networks via pinning, Phys. Rev. E 75(4), 046103.
- 8 Dorogovtsev (2002), Evolution of networks, Adv. Phys. 51(4), 1079-1187.
- 9 Pastor-Satorras et al. (2001), Dynamical and correlation properties of the Internet, *Phys. Rev. Lett.* 87(25), 258701.
- Yu et al. (2009), On pinning synchronization of complex dynamical networks, Automatica 45(2), 429-435.

in-house version of Scopus database at CWTS

science deep learning

```
(target) LeCun et al. (2015), Deep learning, Nature 521(7553), 436-444.
(SOURCE) Silver et al. (2017), Mastering the game of Go without human knowledge, Nature 550(7676), 354-359.
```

- Silver et al. (2016), Mastering the game of Go with deep neural networks and tree search, Nature 529(7587), 484-489.
- Jouppi et al. (2017), In-datacenter performance analysis of a tensor processing unit, In: Proceedings of the ISCA '17, pp. 1-12.
- Reagen et al. (2016), Minerva: Enabling low-power, highly-accurate deep neural network accelerators, In: Proceedings of the ISCA '16, pp. 267-278.
- 4 Chen et al. (2016), DianNao family: Energy-efficient hardware accelerators for machine learning, Commun. ACM 59(11), 105-112.
- 5 Chen et al. (2016), Eyeriss: A spatial architecture for energy-efficient dataflow for convolutional neural networks, In: *Proceedings of the ISCA '16*, pp. 127-138.
- Moravčík et al. (2017), DeepStack: Expert-level artificial intelligence in heads-up no-limit poker, Science 356(6337), 508-513.
- 7 Albericio et al. (2016), Cnylutin: Ineffectual-neuron-free deep neural network computing, In: Proceedings of the ISCA '16, pp. 1-13.
- 8 Han et al. (2016), EIE: Efficient inference engine on compressed deep neural network, In: Proceedings of the ISCA '16, pp. 243-254.
- 9 Shafiee et al. (2016), ISAAC: A convolutional neural network accelerator with in-situ analog arithmetic in procedure, of the ISCA 116, pp. 14-26
- in crossbars, In: *Proceedings of the ISCA '16*, pp. 14-26.

 10 Adolf et al. (2016), Fathom: Reference workloads for modern deep learning methods. In: *Proceedings*
- Adolf et al. (2016), Fathom: Reference workloads for modern deep learning methods, In: Proceedings of the IISWC '16, pp. 1-10.

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science references



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Intermediacy of publications.

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