intermediacy of publications

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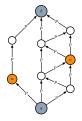
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introduction & motivation

algorithmic historiography for evolution of field (Garfield et al., 2003) relying on citations between scientific publications from WoS & Scopus

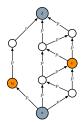


existing approaches include main paths (Hummon & Doreian, 1989) (longest/shortest paths) many irrelevant/miss relevant publications (intermediacy) important publications should only be well-connected

intermediacy measure

(input) selected source & target publications s & t (method) each citation is relevant/active with probability p (measure) importance of publication u called intermediacy $\phi_{\rm u}$

$$\phi_u = \Pr(X_{st}^u) = \Pr(X_{su}) \Pr(X_{ut})$$

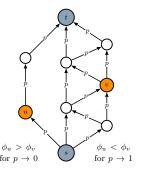


 X_{st} exists path from s to t & X_{st}^{u} exists path through u

intermediacy for $p \rightarrow 0$

for p o 0 intermediacy ϕ governed by ℓ (proof)

for
$$ho o 0$$
 if $\ell_u < \ell_v$ then $\phi_u > \phi_v$

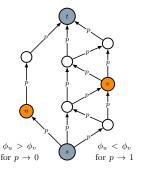


 ℓ_u is **length** of **shortest paths** from s to t through u

intermediacy for $p \rightarrow 1$

for $p \rightarrow 1$ intermediacy ϕ governed by σ (proof)

for
$$p \to 1$$
 if $\sigma_u < \sigma_v$ then $\phi_u < \phi_v$

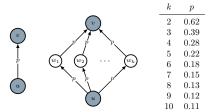


 σ_u is **number** of **independent paths** from s to t through u

intuition for p

for what p is **direct citation** equivalent to **k indirect citations**

$$Pr(X_{uv}) = p = 1 - (1 - p^2)^k$$

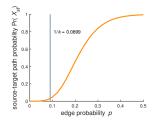


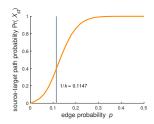
k is **number** of **independent paths** from u to v

phase transition

for what p source-target path $\Pr(X_{st}) > 0$ & intermediacy $\exists u : \phi_u > 0$

$$p \ge n/2m = 1/k$$



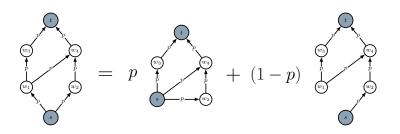


k is **average** number of **citations**/**references**

exact algorithm

decomposition algorithm by edge contraction & removal (Ball, 1979)

$$\Pr(X_{st} \mid G) = p \Pr(X_{st} \mid G/e) + (1-p) \Pr(X_{st} \mid G-e)$$

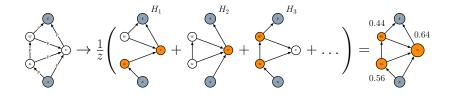


runs in exponential time since NP-hard even in DAG (Johnson, 1984)

approximate algorithm

simple Monte Carlo simulation algorithm by edge sampling

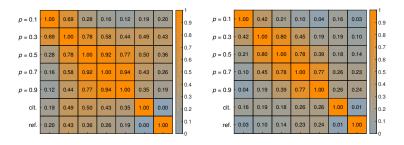
$$\phi_u = \Pr(X_{st}^u \mid G) = \frac{1}{z} \sum_{k=1}^z \mathrm{I}(X_{st}^u \mid H_k)$$



runs in quasi linear time using p-DFS over say 10^6 samples

intermediacy \neq centrality

correlation coefficient between intermediacies ϕ & citations/references

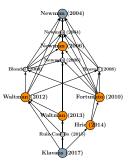


intermediacy ϕ uncorrelated with standard centrality measures

modularity example

(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.
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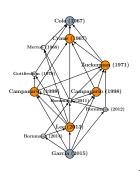
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- 6 Ruiz-Castillo & Waltman (2015), Field-normalized citation impact indicators using algorithmically..., J. Informetr. 9(1), 102-117.
- 7 Blondel et al. (2008), Fast unfolding of communities in large networks, J. Stat. Mech., P10008.
- Newman (2006), Finding community structure in networks using the eigenvectors of matrices, Phys. Rev. E 74(3), 036104.
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- Rosvall & Bergstrom (2008), Maps of random walks on complex networks reveal community structure, PNAS 105(4), 1118-1123.

peer review example

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

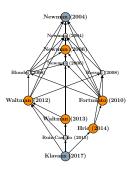
(SOURCE) Garcia et al. (2015), The author-editor game, Scientometrics 104(1), 361-380.



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- 5 Campanario (1998), Peer review for journals as it stands today: Part 2, Sci. Commun. 19(4), 277-306.
- 6 Gottfredson (1978), Evaluating psychological research reports: Dimensions, reliability, and correlates..., Am. Psychol. 33(10), 920-934.
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- 10 Merton (1968), The Matthew effect in science, Science 159(3810), 56-63.

conclusions & future work

(proposal) measure of importance of publications called intermediacy (theory) conceptually clear & provable behavior in extreme cases (practice) intermediacy shows promising results in case studies (future) applicability on general (un)directed networks?



(paper) soon on arXiv.org (code) soon on github.com

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