

network abstraction with backbones and skeletons: spanning trees vs convex skeletons

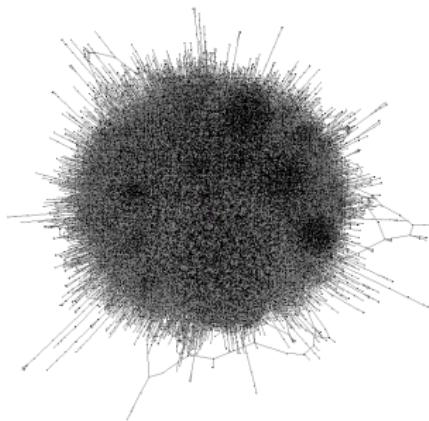
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network abstraction

many real networks are too **large/dense/complex/noisy**
for superlinear algorithms, GPU memory, clear visualizations, etc.

abstraction techniques try to simplify network
preserving network structure/dynamics as much as possible



roadmap

1. backbones/skeletons
2. spanning trees
3. convex skeletons
4. conclusions

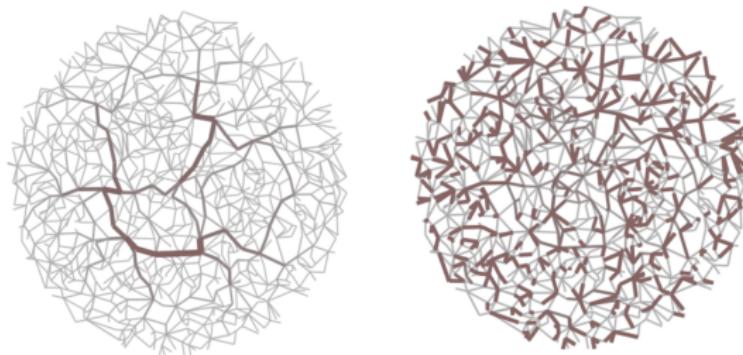
backbones and skeletons

network backbone keeps most important edges (e.g., information flow)

sparsification technique that removes as many edges as possible

network skeleton preserves overall structure (e.g., with simpler graph)

simplification technique that retains as many edges as possible



† Grady et al. (2012) *Nature Communications* 3, 864.

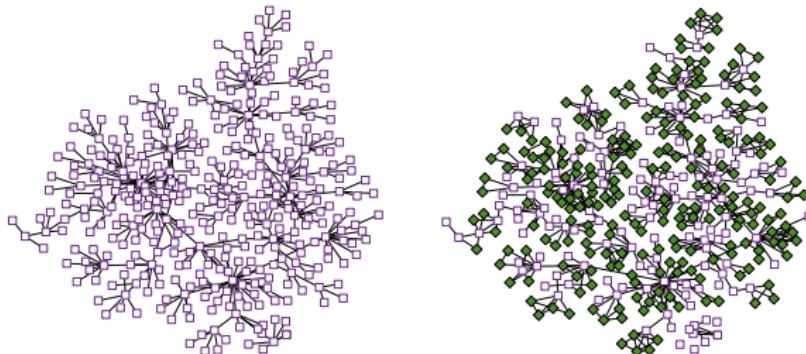
network skeletons

spanning tree is smallest connected graph on all nodes

network simplification technique that retains tree (of edges)

convex skeleton is tree connecting network cliques

network simplification technique that retains tree of cliques



network structure

simple undirected **unweighted network** (n nodes and m edges)

$$\langle k \rangle = \frac{1}{n} \sum_i k_i = \frac{2m}{n}$$

most large real networks are almost **connected and sparse**

$$\rho = \frac{m}{\binom{n}{2}} = \frac{\langle k \rangle}{n-1} \rightarrow 0 \text{ when } n \rightarrow \infty$$

small-world networks have high clustering and short distances

$$\langle C \rangle \gg \frac{\langle k \rangle}{n-1} \text{ and } \langle d \rangle \sim \log n$$

scale-free networks have heavy-tailed degree distribution

$$p_k \sim k^{-\gamma} \text{ for } \gamma > 1$$

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spanning trees

network abstraction with **spanning trees**

$$\langle k \rangle = \frac{1}{n} \sum_i k_i = \frac{2m}{n}$$

spanning trees preserve **connectivity and sparsity**

$$m = n - 1 \text{ and } \langle k \rangle = 2 - \frac{2}{n}$$

spanning trees **lack clustering** or longer cycles

$$\langle C \rangle = 0 \text{ by definition}$$

are spanning trees also **small-world and scale-free?**

$$\langle d \rangle \sim \log n \text{ and } p_k \sim k^{-\gamma}?$$

[†] $\langle d \rangle \sim \sqrt{n}$ in random trees (Re  yi and Szekeres, 1967)

spanning algorithms

Kruskal's algorithm

1. start with forest of trees each consisting of one node
2. merge trees until only one tree remains

Prim's algorithm

1. start with single tree consisting of seed node
2. add one new node at each step

breadth-first search

1. start with single tree consisting of seed node
2. add new neighbors at each step (in breadth-first order)

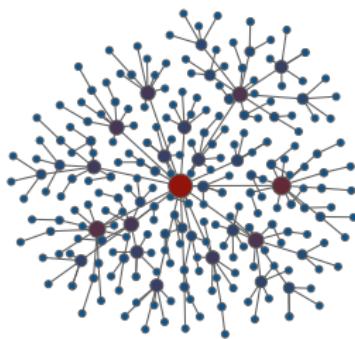
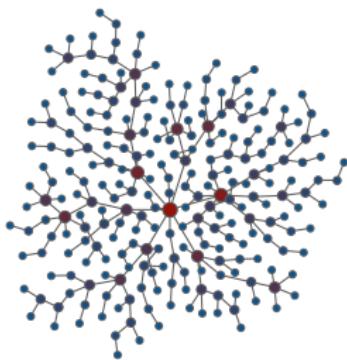
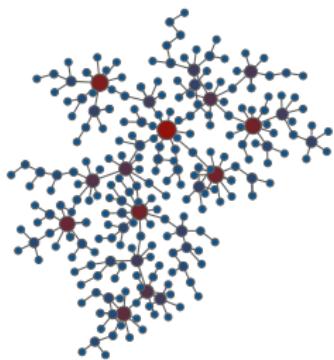
other algorithms

depth-first or beam search, Sollin's algorithm, etc.

wiring diagrams

examples of spanning trees of random graph

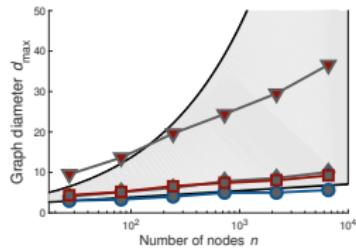
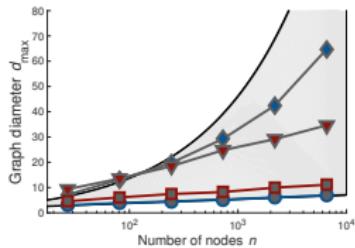
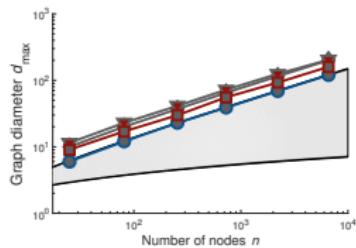
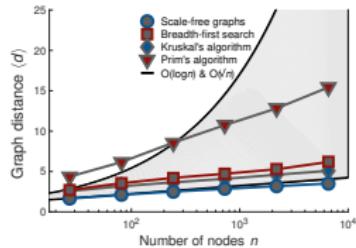
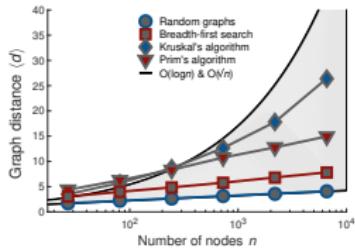
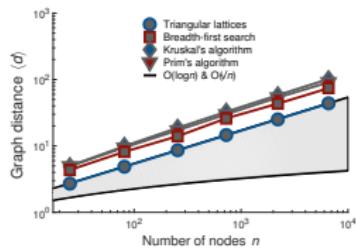
Kruskal's algorithm, Prim's algorithm and breadth-first search



synthetic graphs

breadth-first search preserves distances $\langle d \rangle$ in synthetic graphs

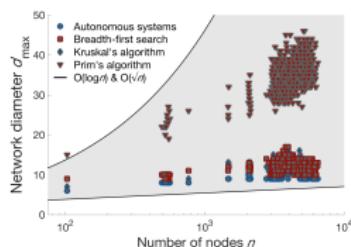
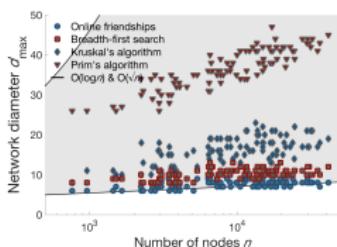
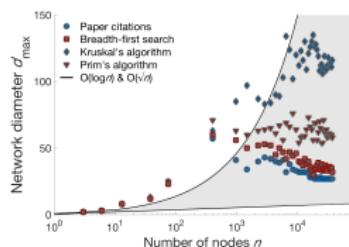
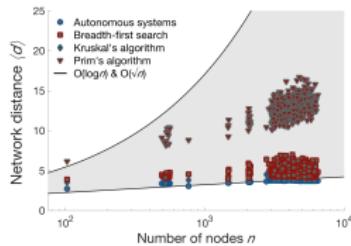
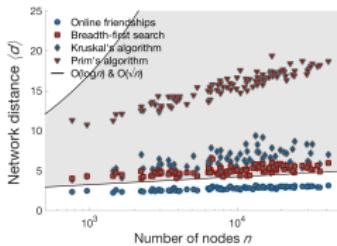
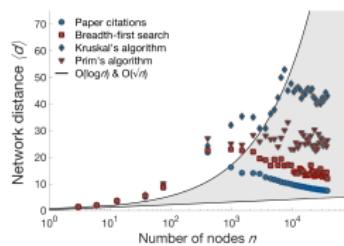
$\langle d \rangle \sim \sqrt{n}$ in lattices, $\langle d \rangle \sim \log n$ in random and $\langle d \rangle \sim \frac{\log n}{\log \log n}$ in scale-free graphs



small-world networks

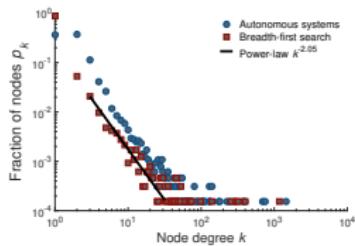
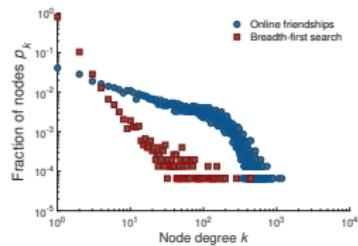
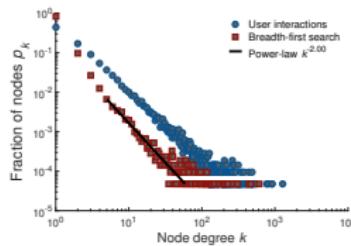
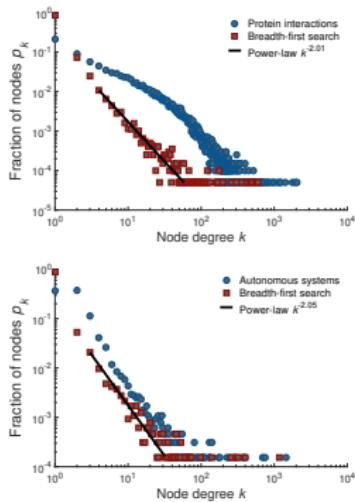
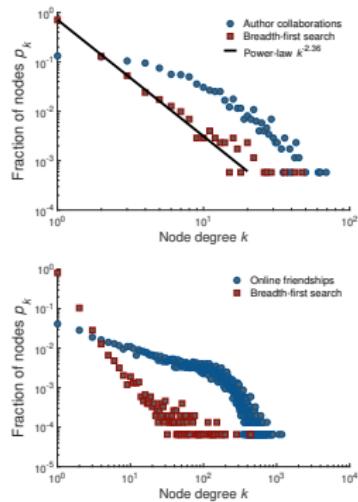
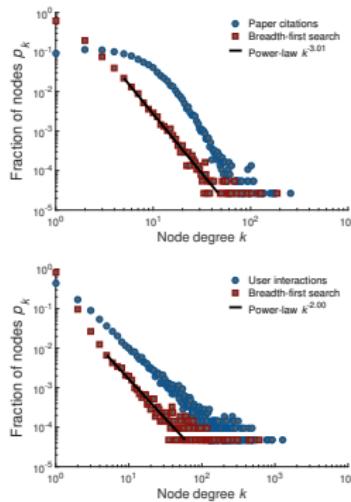
breadth-first search preserves short distances $\langle d \rangle$ in real networks

$\langle d \rangle \sim \log n$ in small-world and $\langle d \rangle \sim \log \log n$ in ultra small-world networks



scale-free networks

breadth-first search power-law $p_k \sim k^{-\gamma}$ as in real networks

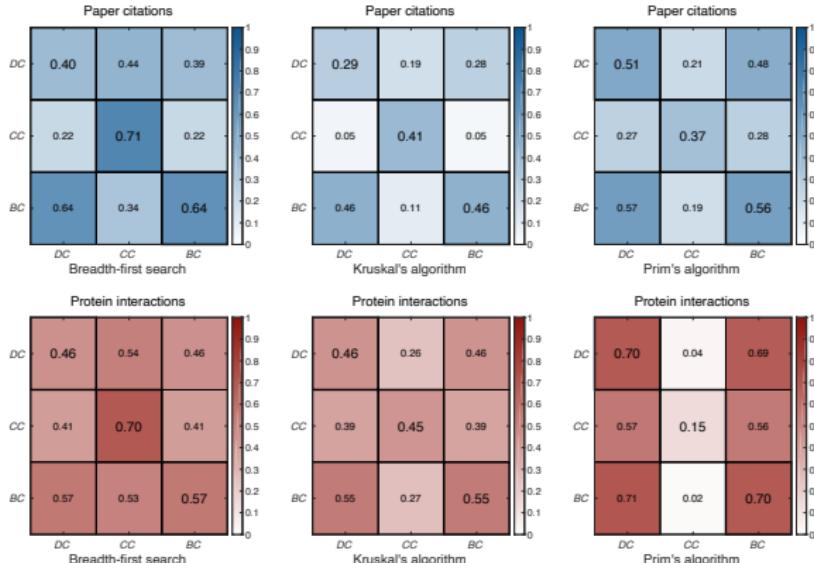


†

Clauset et al. (2009) *SIAM Review* 51(4), 661-703.

node position

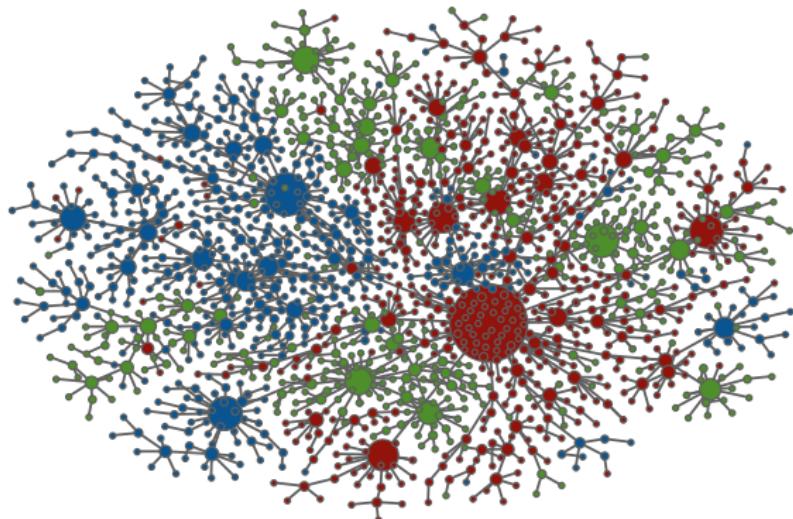
breadth-first search preserves closeness centrality of network nodes
correlations btw degree (DC), closeness (CC) and betweenness (BC) centrality



network visualization

breadth-first search tree of Slovenian scientists coauthorships

primary disciplines = **natural sciences**, **engineering**, **medical sciences**, etc.

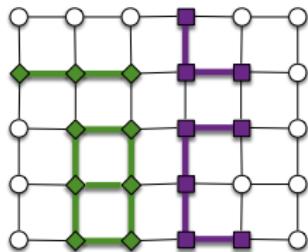
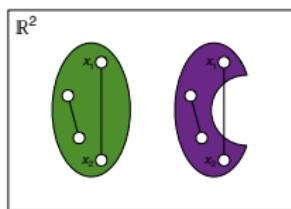
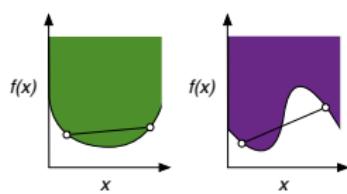


roadmap

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convexity in networks

convex/non-convex real functions, sets in \mathbb{R}^2 & subgraphs



convex graph \equiv every connected induced subgraph is convex

convexity estimated by simulating growth of convex subgraphs

convex network = tree of cliques (cliques connected tree-like)

† Marc & Šubelj (2018) *Network Science* 6(2), 176-203.

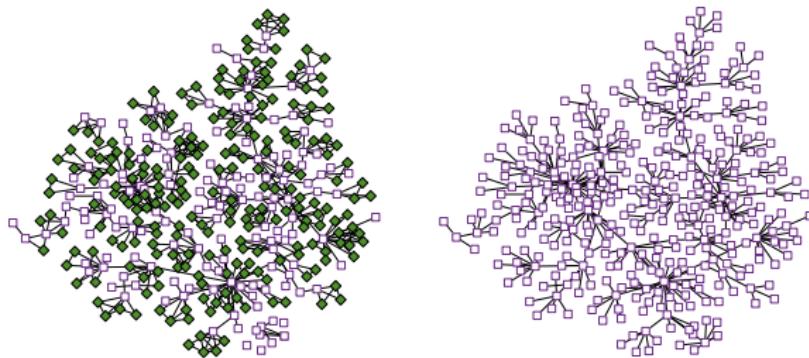
convex skeleton

convex skeleton = largest **high-convexity** subgraph of network

convex skeletons extracted by non-convex edge removal

convex skeleton \sim generalized spanning tree

tree of cliques \supseteq tree (of edges)



convex skeletons

network abstraction with **convex skeletons**

$$\langle k \rangle = \frac{1}{n} \sum_i k_i = \frac{2m}{n}$$

convex skeletons preserve **connectivity and sparsity**

$$m \geq n - 1 \text{ and } \langle k \rangle \geq 2 - \frac{2}{n}$$

convex skeletons preserve **clustering and cliques**

$$\langle C \rangle \gg 0 \text{ by construction}$$

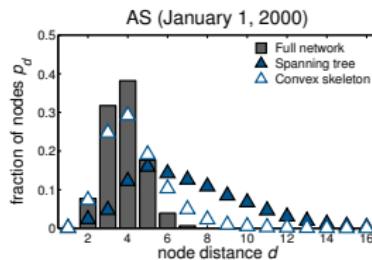
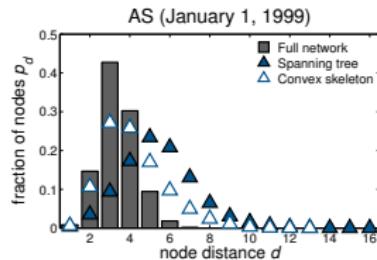
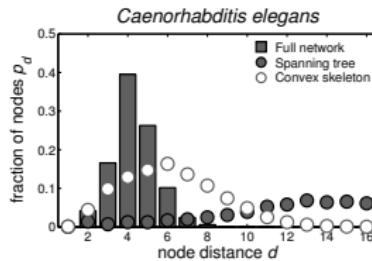
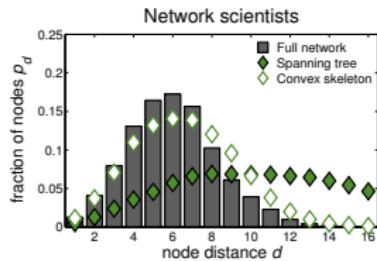
are convex skeletons also **small-world and scale-free**?

$$\langle d \rangle \sim \log n \text{ and } p_k \sim k^{-\gamma}?$$

† $\langle d \rangle \sim \sqrt{n}$ in random trees (Re  yi and Szekeres, 1967)

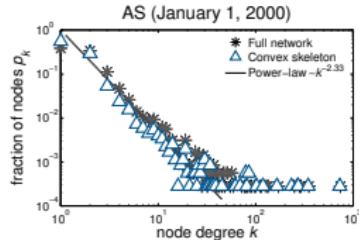
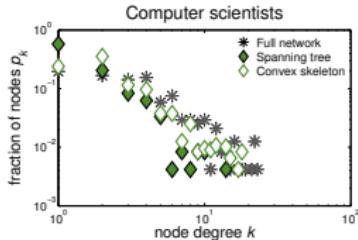
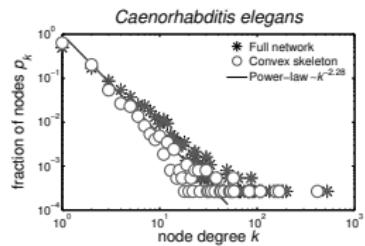
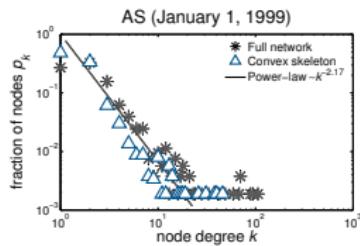
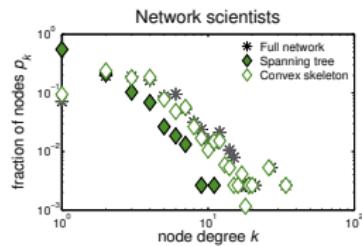
small-world networks

convex skeleton preserves distance distribution p_d of real networks



scale-free networks

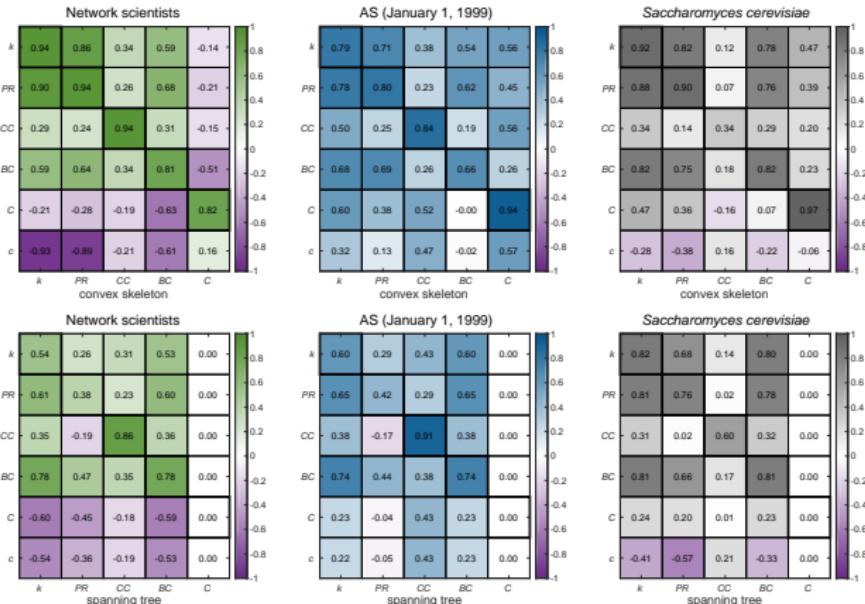
convex skeleton preserves degree distribution p_k of real networks



† Clauset et al. (2009) *SIAM Review* 51(4), 661-703.

node position

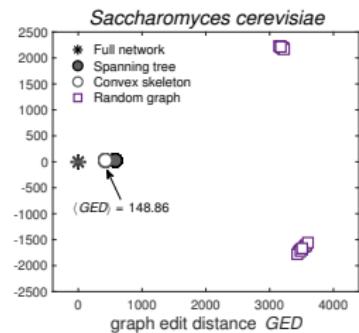
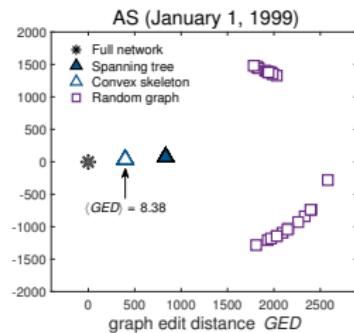
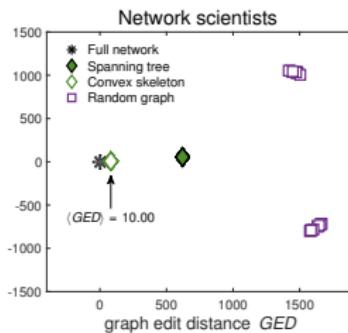
convex skeleton preserves position/centrality of network nodes
 correlations btw degree (k), closeness (CC), betweenness (BC), clustering (C), etc.



edit distances

convex skeleton provides good model of real networks

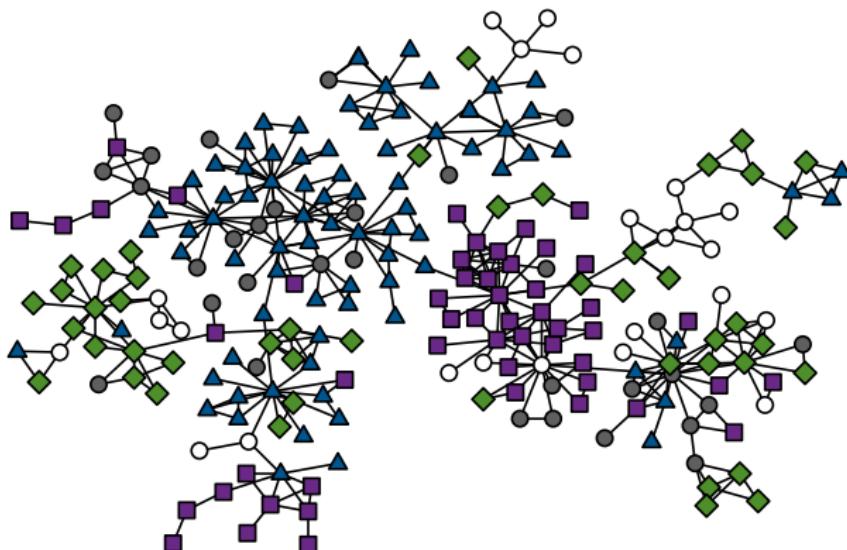
MDS maps of edit distances between networks and skeletons



network visualization

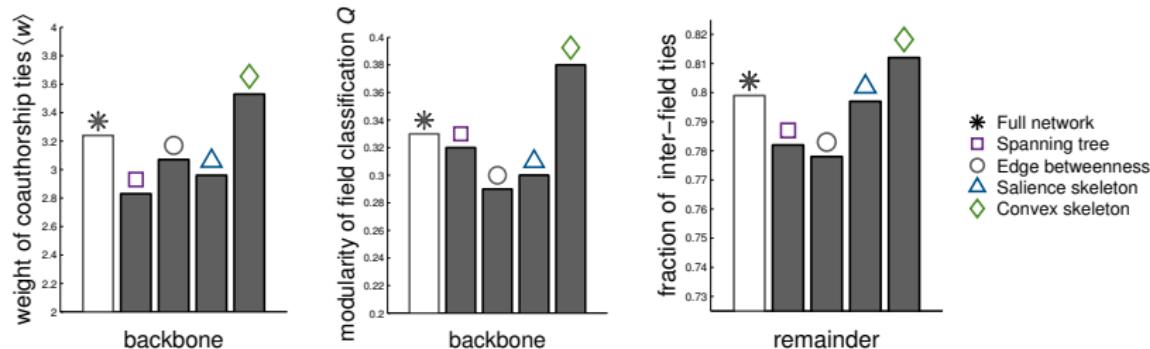
convex skeleton of Slovenian computer scientists coauthorships

research areas = **computer theory**, **intelligent systems**, **information technology**, etc.



network skeletons

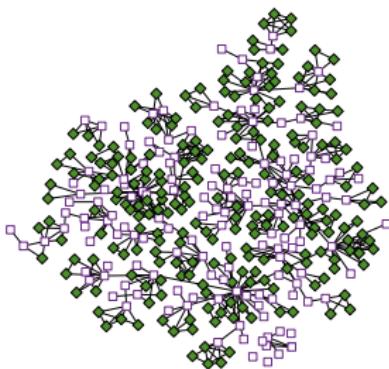
skeletons of Slovenian computer scientists coauthorships
convex skeleton strengthens desirable properties of scientists coauthorships



roadmap

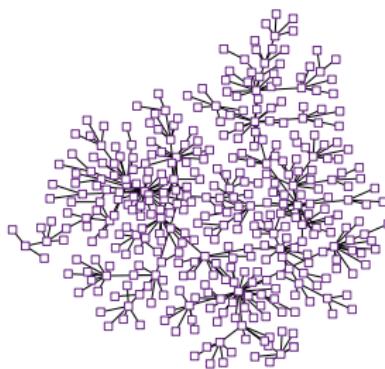
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conclusions



convex skeleton

tree with cliques



spanning tree

tree without cliques

quality → convex skeleton \gg spanning tree

complexity → spanning tree \ll convex skeleton

thank you!

Marc & Šubelj (2018) **Convexity in complex networks**. *Network Science* **6**(2), 176-203.

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Šubelj et al. (2019) **Convexity in scientific collaboration networks**. *Journal of Informetrics* **13**(1), 10-31.

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