

# convexity in complex networks

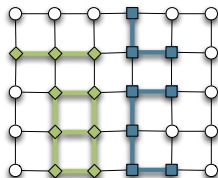
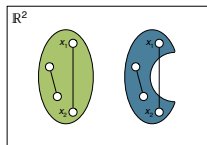
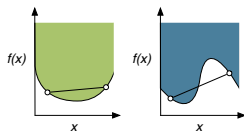
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LCN2 '17

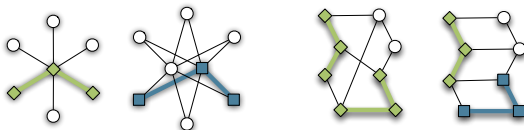
# definitions of convexity

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



disconnected  $\supseteq$  connected  $\supseteq$  induced  $\supseteq$  isometric  $\supseteq$  convex subgraphs

connected subgraphs induced on simple undirected graph  $\rightarrow$



## convexity **in** networks?

(**sna**)  $k$ -clubs/ $k$ -clans are convex  $k$ -cliques

(**cd**) community often defined as “convex” subgraph

- **subset**  $S$  is convex if it induces convex **subgraph**
- convex **hull**  $\mathcal{H}(S)$  is smallest convex subset including  $S$

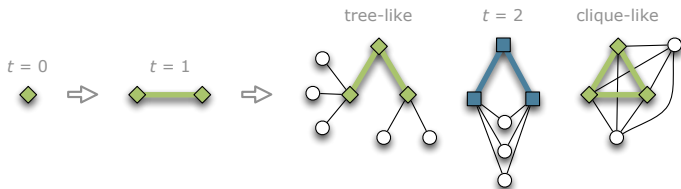
hull number =  $\min\{|S| : \mathcal{H}(S) \text{ includes } n \text{ nodes}\}$  (Everett & Seidman, 1985)

- ↑ hull number measures how **quickly** convex subsets can grow
- ↓ how **slowly** randomly grown convex subsets expand

# expansion of convex subsets

**grow** subset  $S$  by one node & **expand**  $S$  to convex hull  $\mathcal{H}(S)$

- $S = \{\text{random node } i\}$
- until  $S$  contains  $n$  nodes:
  1. select  $i \notin S$  by random edge
  2. expand  $S = \mathcal{H}(S \cup \{i\})$

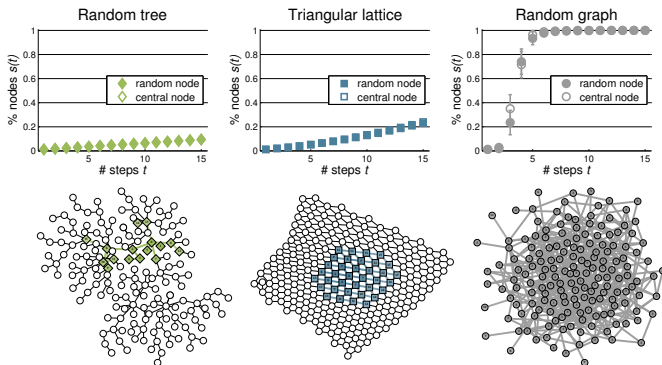


$S$  quantifies (locally) **tree-like**/**clique-like** structure of graphs

# convex expansion in graphs

$s(t)$  = average fraction of nodes in  $S$  after  $t$  expansion steps

$s(t) = (t + 1)/n$  in **convex** &  $s(t) \gg (t + 1)/n$  in **non-convex** graphs

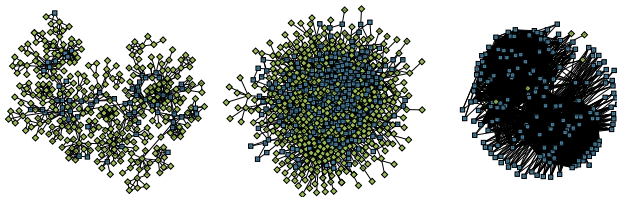
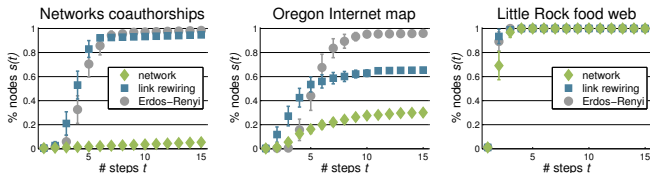


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# convex expansion in networks

$s(t)$  = average fraction of nodes in  $S$  after  $t$  expansion steps

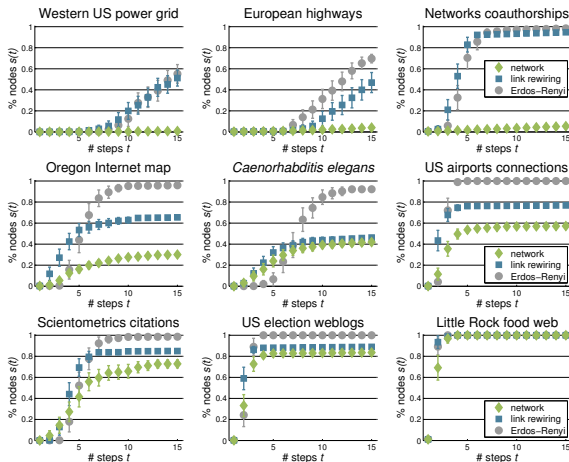
$s(t) = (t + 1)/n$  in **convex** &  $s(t) \gg (t + 1)/n$  in **non-convex** networks



$s(t)$  quantifies (locally) **tree-like**/**clique-like** structure of networks

# convex expansion in networks

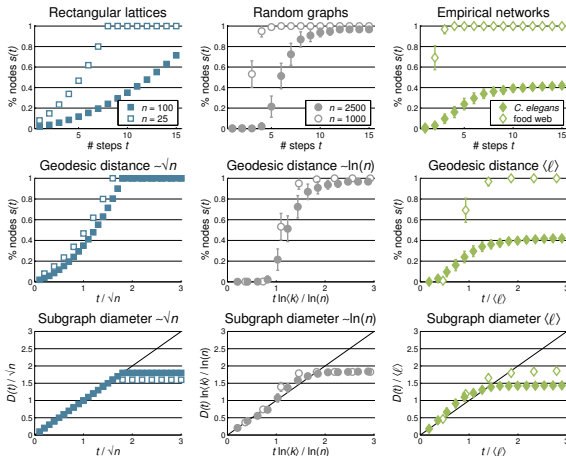
convex infrastructure and collaboration & non-convex food web



random graphs fail to reproduce convexity in empirical networks

# when/why sudden expansion?

(**why**) steps  $t \approx \text{diameter } D(t) > \text{distance } \langle \ell \rangle$  (**when**)

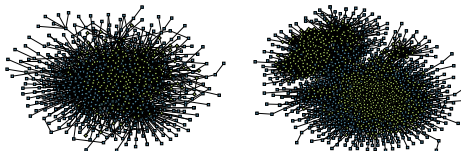
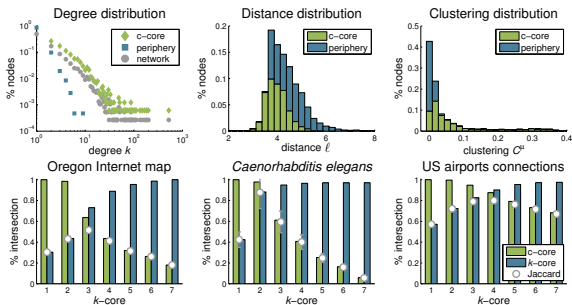


random graphs **convex** for  $< \mathcal{O}(\ln n)$  & **non-convex** for  $> \mathcal{O}(\ln^2 n)$



# when/why expansion settles?

(**when**)  $S$  extends to c-core (**why**) smallest convex subset  $\supseteq S$



core-periphery networks have **convex** periphery & **non-convex** c-core

# global measure $c$ -convexity

$$X_c = 1 - \sum_{t=1}^{n-1} \sqrt[n]{\max(\Delta s(t) - 1/n, 0)} \quad X_c \geq X_c^{\text{RW}} \geq X_c^{\text{ER}}$$

$X_c$  highlights **tree-like**/**clique-like** networks (cliques connected tree-like)

	$X_1$	$X_1^{\text{RW}}$	$X_1^{\text{ER}}$	$X_{1.1}$	$X_{1.1}^{\text{RW}}$	$X_{1.1}^{\text{ER}}$
Western US power grid*	0.95	0.32	0.24	0.91	0.10	0.01
European highways*	0.66	0.23	0.27	0.44	-0.02	0.06
Networks coauthorships	0.91	0.09	0.06	0.83	-0.05	-0.09
Oregon Internet map	0.68	0.36	0.06	0.53	0.20	-0.09
<i>Caenorhabditis elegans</i>	0.57	0.54	0.07	0.43	0.40	-0.13
US airports connections	0.43	0.24	0.00	0.30	0.16	-0.07
Scientometrics citations	0.24	0.16	0.02	0.04	0.00	-0.13
US election weblogs	0.17	0.12	0.00	0.06	0.04	-0.08
Little Rock food web	0.03	0.03	0.02	-0.06	-0.02	-0.02

$X_c$  measures **global** & **regional** (periphery) convexity in networks

# local measure of convexity

$$L_c = 1 + \max\{t \mid s(t) < (t + c + 1)/n\} \quad L_1 \leq L_1^{\text{ER}} \approx \ln n / \ln \langle k \rangle$$

$L_c$  highlights locally **tree-like**/**clique-like** networks & random graphs

	$L_t$	$L_t^{\text{ER}}$	$L_1$	$L_1^{\text{ER}}$	$\ln n / \ln \langle k \rangle$
Western US power grid	14	9	6	9	8.66
European highways	16	7	7	7	7.54
Networks coauthorships	17	4	7	4	3.77
Oregon Internet map	3	4	3	4	4.40
<i>Caenorhabditis elegans</i>	2	5	2	5	5.79
US airports connections	2	3	2	3	2.38
Scientometrics citations	3	4	3	4	4.30
US election weblogs	2	2	2	2	2.15
Little Rock food web	2	2	2	2	1.59

$L_c$  measures **local** & **absolute** (tree/clique) convexity in networks

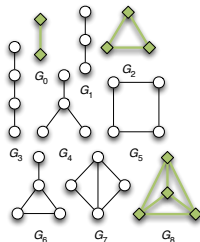
# probability of convex subgraphs

$P$  = probability that random  $G_{1-8}$  convex

$$P \leq P^{\text{ER}}$$

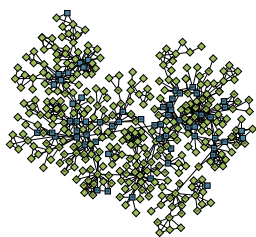
$P$  highlights locally **tree-like**/**clique-like** networks & random graphs

	$P$	$P^{\text{ER}}$	$\ln n / \ln \langle k \rangle$
Western US power grid	77.0%	99.4%	8.66
European highways	83.2%	97.6%	7.54
Networks coauthorships	53.3%	71.3%	3.77
Oregon Internet map	56.0%	86.4%	4.40
<i>Caenorhabditis elegans</i>	77.8%	97.6%	5.79
US airports connections	5.5%	12.9%	2.38
Scientometrics citations	30.5%	89.2%	4.30
US election weblogs	2.7%	6.0%	2.15
Little Rock food web	2.2%	0.3%	1.59



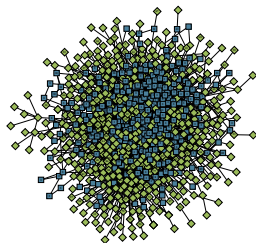
$P$  measures **local** (up to **4 nodes**) convexity in networks

# types of network convexity



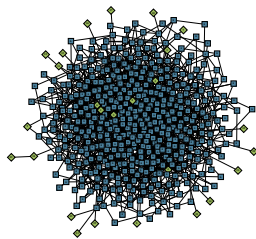
**global** convexity

tree/clique-like  
networks



**regional** convexity

core-periphery  
networks etc.



**local** convexity

random graphs  
 $< \ln n / \ln \langle k \rangle$

**c-convexity**  $\neq$  standard measures & **c-core**  $\neq$   $k$ -cores  
robustness, navigation, optimization, abstraction, comparison etc.

to be continued...

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# convex skeletons of networks

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# corrected measure of convexity

$$X_s = s - \sum_{t=1}^{sn-1} \sqrt[n]{\max(s\Delta s(t) - 1/n, 0)} \quad s = \text{fraction of nodes in LCC}$$

$X_s$  highlights **tree-like**/**clique-like** networks & synthetic graphs

	$n$	$\langle k \rangle$	$X_s$		$n$	$\langle k \rangle$	$X_s$
Jazz musicians	198	27.70	0.12	Random graphs	2500	10.00	0.00
Network scientists	379	4.82	0.85		1000	10.00	0.01
Computer scientists	239	4.75	0.64		225	10.00	0.03
<i>Plasmodium falciparum</i>	1158	4.15	0.43	Triangular lattice	225	5.48	0.23
<i>Saccharomyces cerevisiae</i>	1458	2.67	0.68	Rectangular lattice	225	3.73	0.13
<i>Caenorhabditis elegans</i>	3747	4.14	0.56	Core-periphery graph	3747	4.48	0.39
AS (January 1, 1998)	3213	3.50	0.66	Convex graphs	2500	5.97	1.00
AS (January 1, 1999)	531	4.58	0.49		1000	5.97	1.00
AS (January 1, 2000)	3570	3.94	0.59		225	6.01	1.00
Little Rock Lake	183	26.60	0.02	convex graphs are random trees of cliques			
Florida Bay (wet)	128	32.42	0.03				
Florida Bay (dry)	128	32.91	0.03				

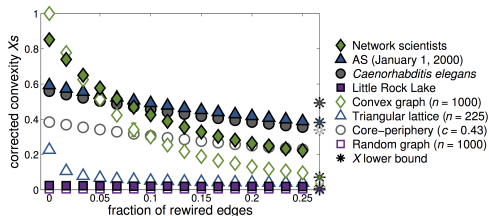
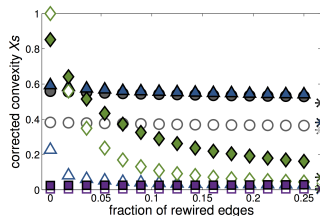
$X_s$  measures **global** & **regional** convexity in (disconnected) networks



# convexity under randomization

$$X \geq s_1 \quad s_1 = \text{fraction of pendant nodes}$$

$X_s$  under **degree-preserving**/full randomization by edge rewiring

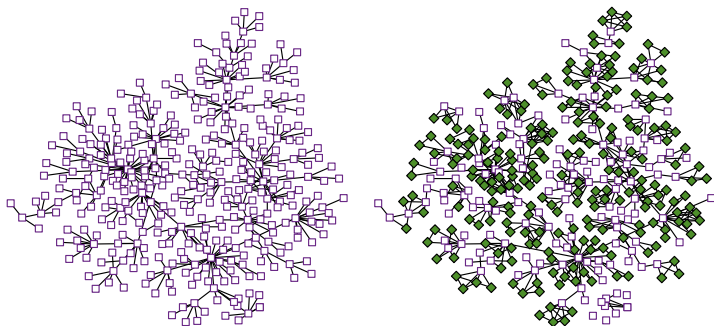


$X_s$  very **sensitive** to **random perturbations** of network structure

# convex skeletons of networks

convex skeleton = largest high- $X_s$  subnetwork (every  $S$  is convex)

spanning tree & convex skeleton of network scientists coauthorships

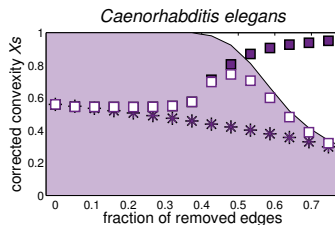
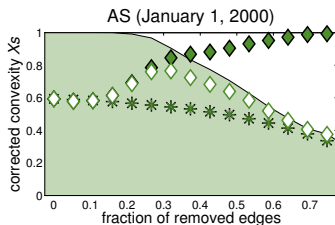


convex skeleton is tree of cliques extracted by edge removal

# extraction of convex skeletons

$$c_i = \sum_{j \in \Gamma_i} p_j - \sum_{j \in \Gamma_i} 1 - p_j \quad p_i = \text{probability that } i \in \text{c-core}$$

Xs under **removal** of edges  $\{i, j\}$  based on **c-centrality**  $c_i + c_j$



c-centrality  $c_i + c_j$  for **core-periphery** & clustering  $\Delta C_i + \Delta C_j$  for **others**

# statistics of convex skeletons

$$\langle C \rangle = \frac{1}{n} \sum_i \frac{2t_i}{k_i(k_i - 1)} \quad \langle \sigma \rangle = \frac{2}{n(n-1)} \sum_{i < j} \sigma_{ij} \quad X_s = \dots$$

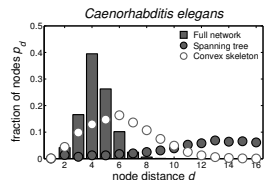
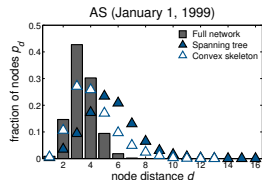
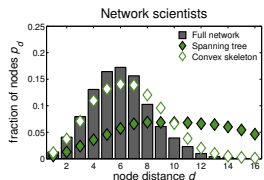
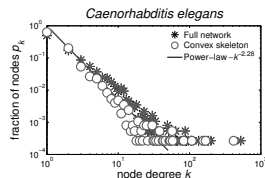
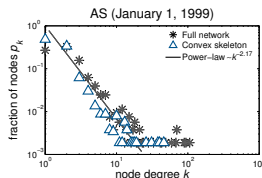
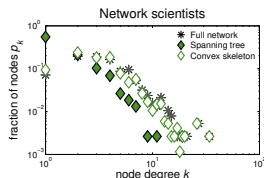
statistics of **convex skeletons** & **spanning trees** of networks

	clustering $\langle C \rangle$			geodesics $\langle \sigma \rangle$			convexity $X_s$		
	N	CS	ST	N	CS	ST	N	CS	ST
Jazz musicians	0.62	0.81	0.00	9.71	1.97	1.00	0.12	0.84	1.00
Network scientists	0.74	0.75	0.00	2.66	1.47	1.00	0.85	0.95	1.00
Computer scientists	0.48	0.54	0.00	4.08	1.42	1.00	0.64	0.95	1.00
<i>Plasmodium falciparum</i>	0.02	0.07	0.00	3.71	1.77	1.00	0.43	0.95	1.00
<i>Saccharomyces cerevisiae</i>	0.07	0.10	0.00	2.58	1.19	1.00	0.68	0.88	1.00
<i>Caenorhabditis elegans</i>	0.06	0.12	0.00	6.79	3.03	1.00	0.56	0.85	1.00
AS (January 1, 1998)	0.18	0.21	0.00	3.87	2.32	1.00	0.66	0.91	1.00
AS (January 1, 1999)	0.18	0.27	0.00	3.54	2.05	1.00	0.49	0.95	1.00
AS (January 1, 2000)	0.20	0.25	0.00	4.81	3.07	1.00	0.59	0.90	1.00
Little Rock Lake	0.32	0.69	0.00	22.13	4.32	1.00	0.02	0.82	1.00
Florida Bay (wet)	0.33	0.79	0.00	9.17	1.37	1.00	0.03	0.92	1.00
Florida Bay (dry)	0.33	0.82	0.00	9.37	1.65	1.00	0.03	0.93	1.00

**convex skeleton** is generalization of **spanning tree** retaining **clustering**

# distributions of convex skeletons

node distributions of **convex skeletons** & **spanning trees** of networks

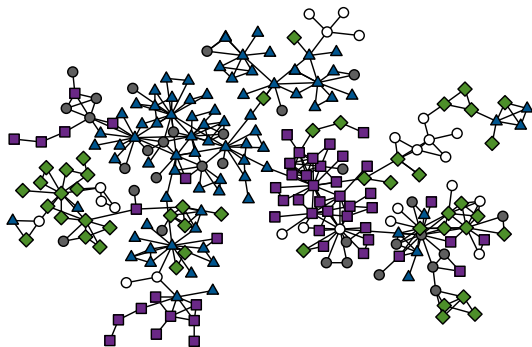


**convex skeletons** retain distributions in contrast to **spanning trees**

# convex skeletons of coauthorships

convex skeleton  $\sim$  network abstraction technique

**convex skeleton** of Slovenian **computer scientists** coauthorships

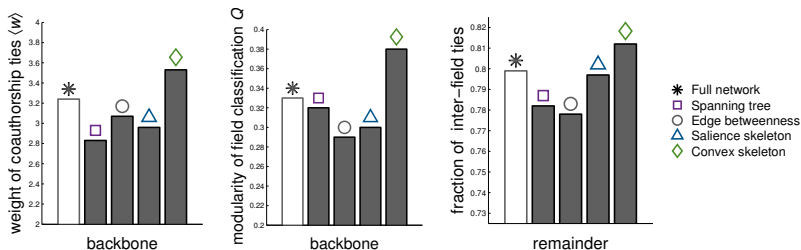


computer theory (◆), information systems (■), intelligent systems (▲),  
programming technologies (○) & other (●)

# network backbones of coauthorships

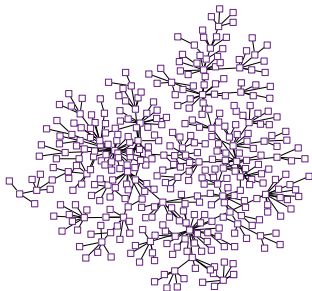
convex skeleton  $\gg$  high-betweenness & high-salience skeletons

properties of **backbones** of Slovenian **computer scientists** coauthorships



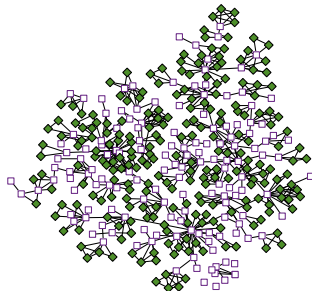
**convex skeletons** increase properties in contrast to **other backbones**

# convex skeletons of networks



spanning **tree**

tree w/o cliques



**convex** skeleton

tree w/ cliques

**convex skeleton**  $\gg$  backbones & **c-centrality**  $\neq$  centralities

abstraction, sampling, sparsification, modeling, dynamics etc.



thank you!

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