

A Basis for Systematic Analysis and Generation of Network Topologies

Priya Mahadevan

UC San Diego

Dmitri Krioukov (CAIDA), Kevin Fall (Intel Research),

Amin Vahdat (UC San Diego)

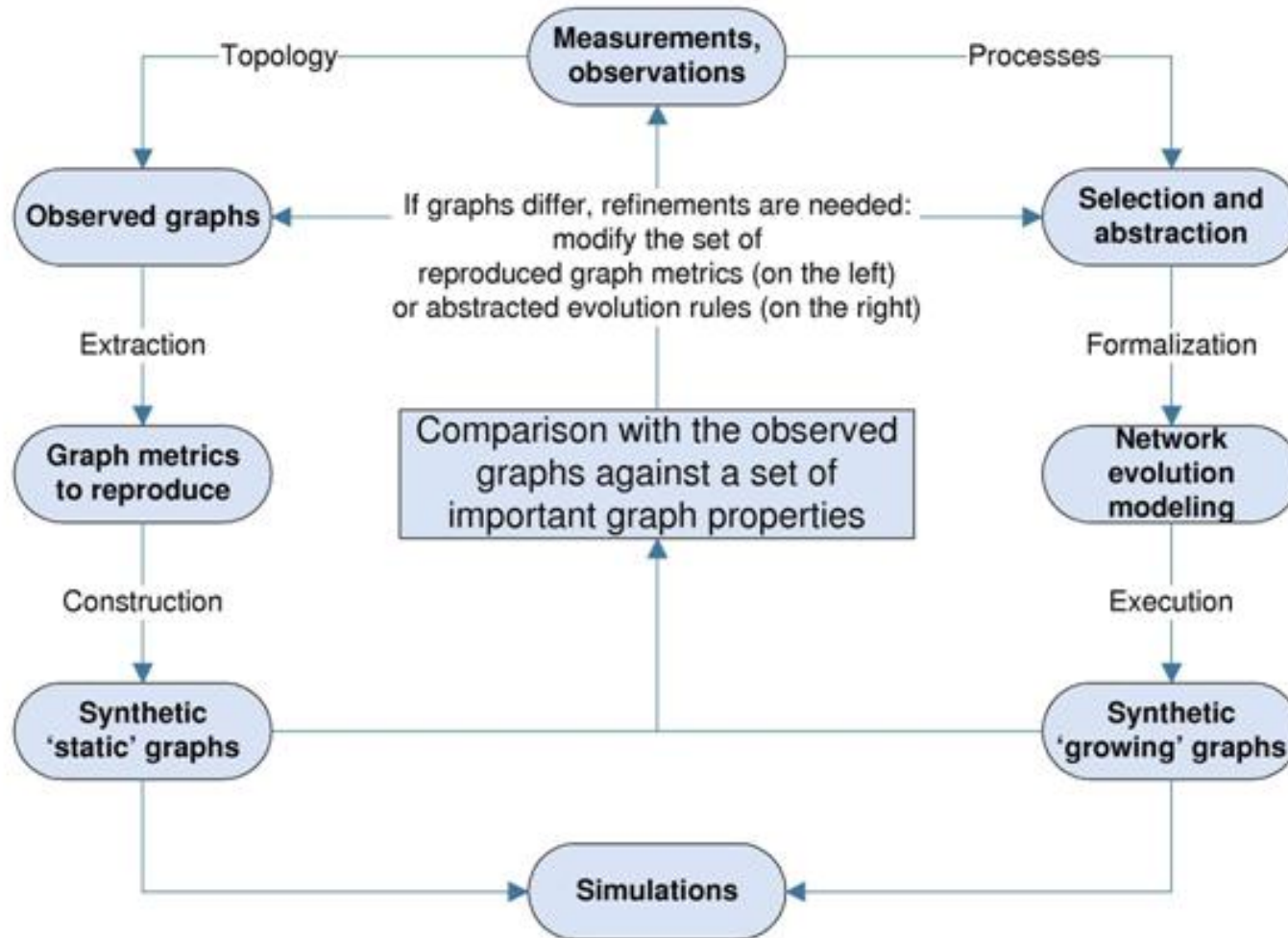
Importance of Network Topology

- Performance of protocols and applications
 - Routing, overlay networks

For example: new routing protocol might offer X -time smaller routing tables for today but scale Y -time worse, with $Y \gg X$

- Robustness of the network
- Traffic engineering
- Network management
- Spread of worms, etc.

Methodologies of Topology Research



Topology Evaluation Metrics

- Distance distribution
- Betweenness
- Clustering
- Assortativity coefficient / likelihood
- Spectrum

Problem in generating graphs?

- No known techniques to produce graphs with a given form of distance distribution, betweenness, etc.
- What if a new important metric is discovered?

Our Approach

Enumerable set of properties P_d , $d = 0, 1, \dots$ that satisfy:

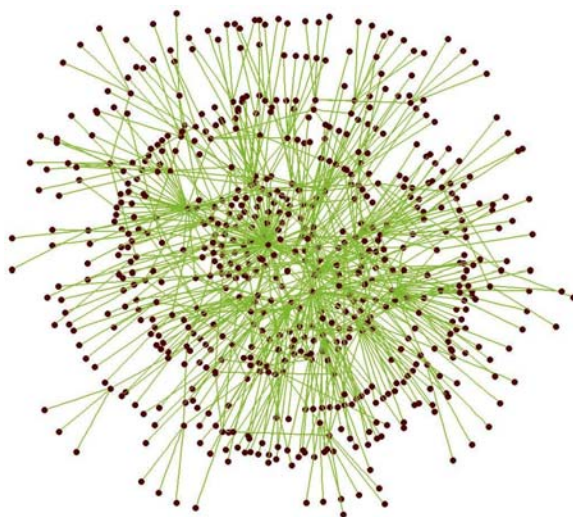
- *Constructibility*: construct graphs having these properties
- *Inclusion*: property P_d subsumes P_j where $j = 0, \dots, d-1$
- *Convergence*: As d increases, the set of graphs having property P_d converges to the original graph G

Connectivity is the most basic property of network topologies

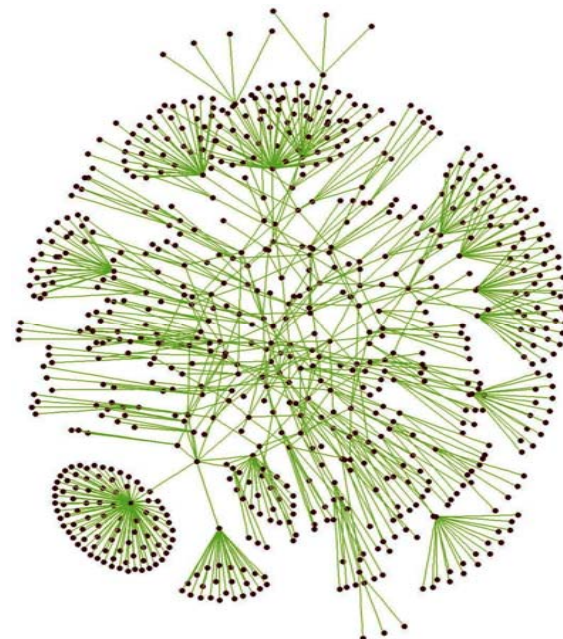
- We consider degree correlations among increasingly larger set of connected nodes
- $P_0, P_1, P_2, \dots, P_n$ correspond to degree correlations among connected nodes of size $0, 1, 2, \dots, n$



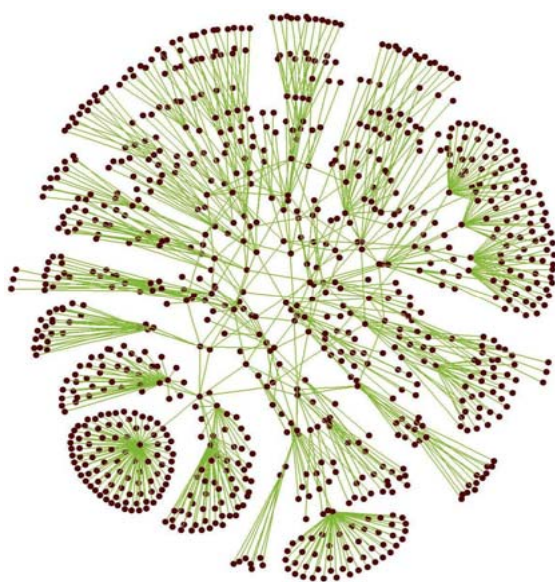
Graph reproducing P_0



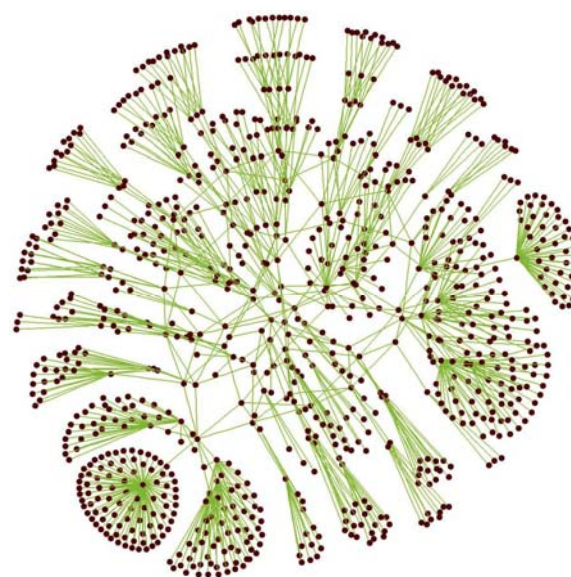
Graph reproducing P_1



Graph reproducing P_2



Graph reproducing P_3







Original graph (HOT)

Outline

- **Background**
- Methodology
 - Graph generation
- Validation
 - AS-level graph (skitter)
 - Router-level graph (HOT)
- Limitations
- Conclusions

Our dK -series

dK -series: degree correlations within non-isomorphic simple connected subgraphs of size d

Tag	Name	Subgraphs of size:	Symbolics
$0K$	Average node degree	0	
$1K$	Node degree distribution	1	
$2K$	Joint degree distribution	2	
$3K$	Joint edge degree distribution	3	
...
nK	Full degree distribution	$n = \#$ of nodes	

More on dK -series

dK -series : degree correlations within non-isomorphic simple connected subgraphs of size d

How groups of d -nodes with different degrees interconnect

- $0K$ (average degree)
- $1K$ (degree distribution)
- $2K$



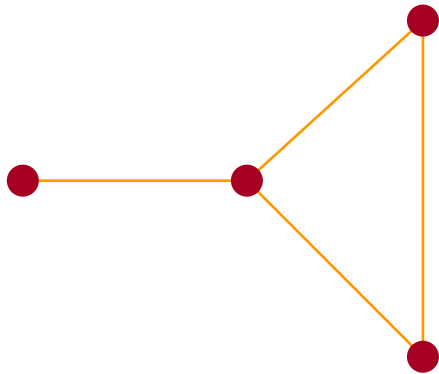
Connectivity between nodes having degrees k_1 and k_2

- $3K$

Connectivity among nodes having degrees k_1 , k_2 and k_3



Example

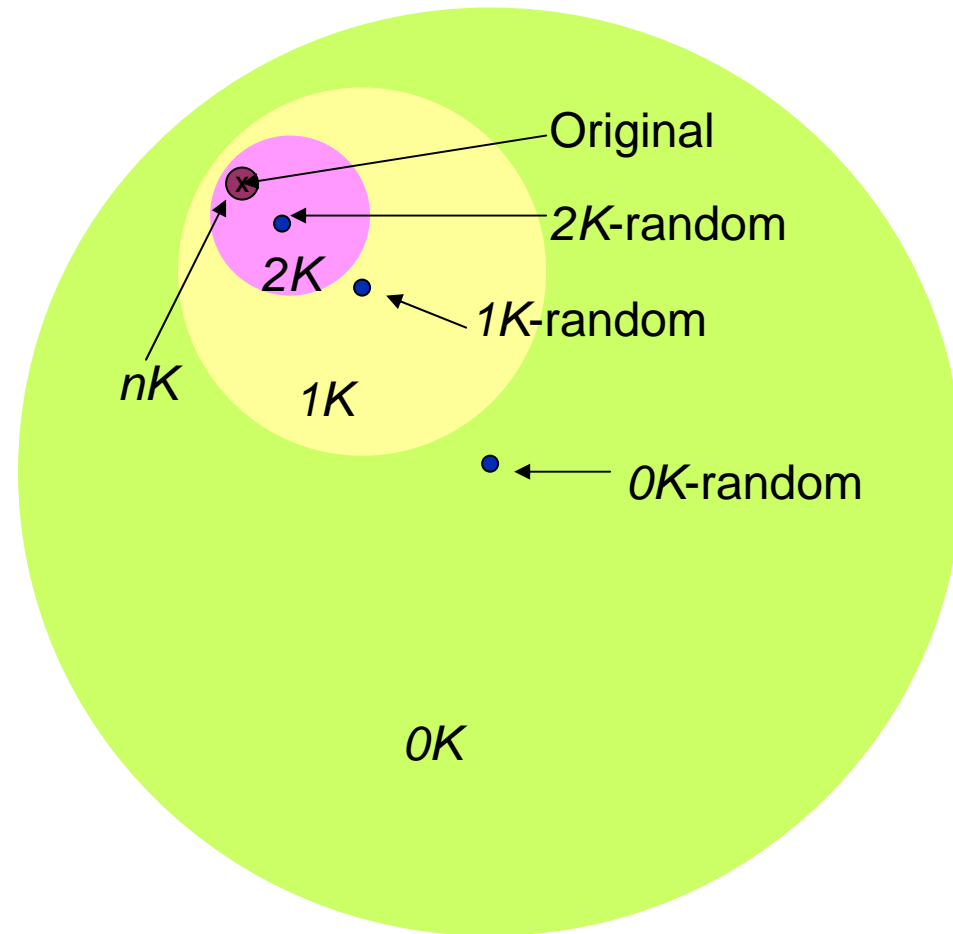


- $0K$: $avg\ deg = 2$
- $1K$: $N(1)=1, N(2)=2, N(3)=1$
- $2K$: $N(1,3)=1$
 $N(2,2)=1$
 $N(2,3)=2$
- $3K$: $N_{\Delta}(2,2,3)=1$
 $N_{\wedge}(1,3,2)=2$

Compute probability distributions for each

dK -graphs

- Graphs associated with each P_d
Original graph belongs to all subsets of dK -graphs ($d=0,..n$)
- As d increases, dK graphs converge to the original graph
- dK -random graph reproduces the specified property P_d , random with respect to all other properties
- What value of d is sufficient for practical purposes to reproduce Internet topologies?



Constructing dK -random graphs

- Stochastic approach
 - In theory, can be generalized to any d ; does not work well in practice
- Pseudograph (eg: PLRG)
 - In its original form, its only for $1K$
 - Extended it for $2K$, but not beyond
- Rewiring
 - Perturb the given graph (swap pairs of edges) such that property P_d is preserved



- Targeting Rewiring (Metropolis Dynamics)
 - Rewire graph such that it moves from reproducing P_{d-1} to P_d

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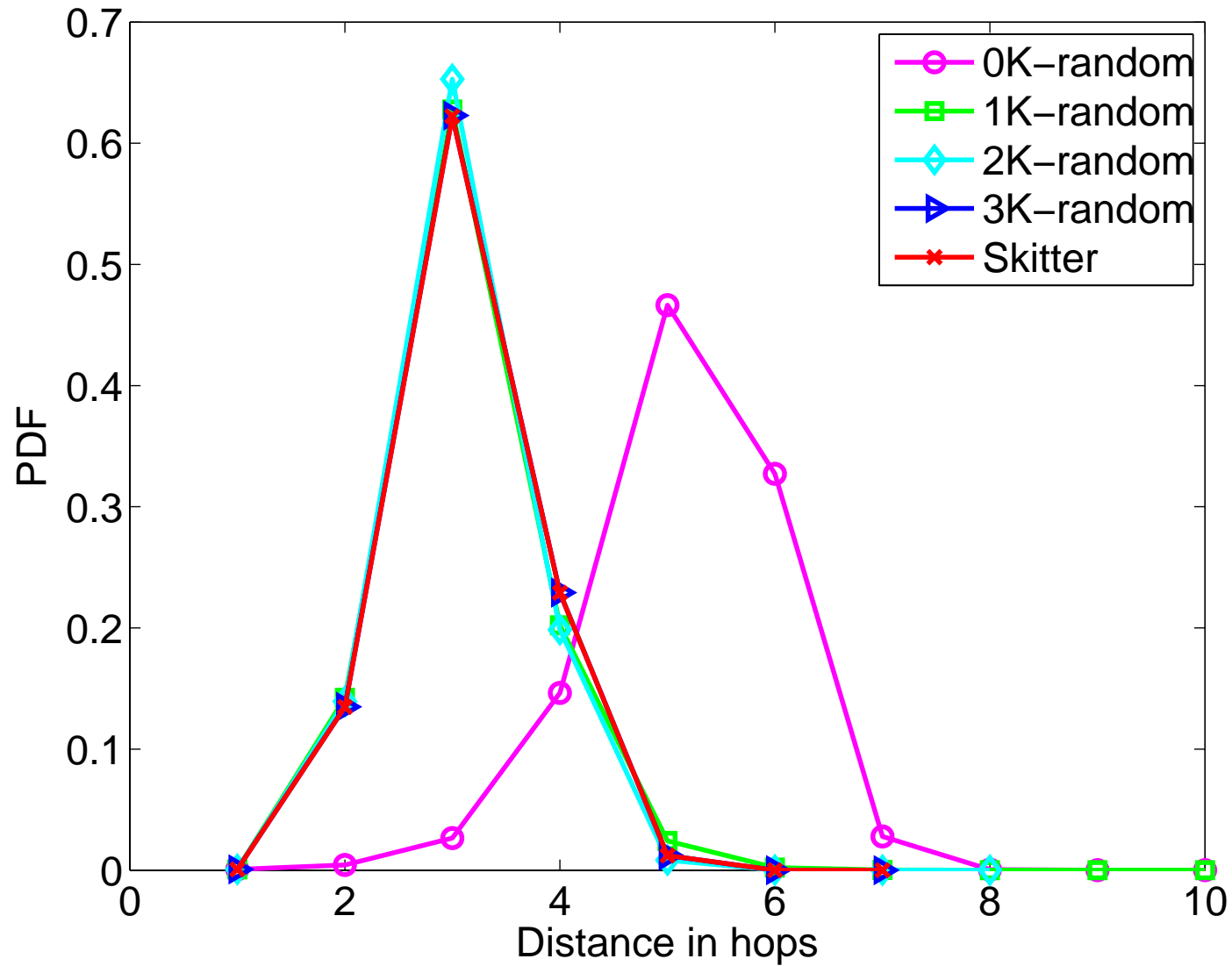
Validation

- We generated $0K$, $1K$, $2K$ and $3K$ -random graphs
- Compare our generated dK -random graphs with the original graph w.r.t. important topology metrics
- AS-level validation: skitter, BGP tables, WHOIS
- Router-level validation: HOT

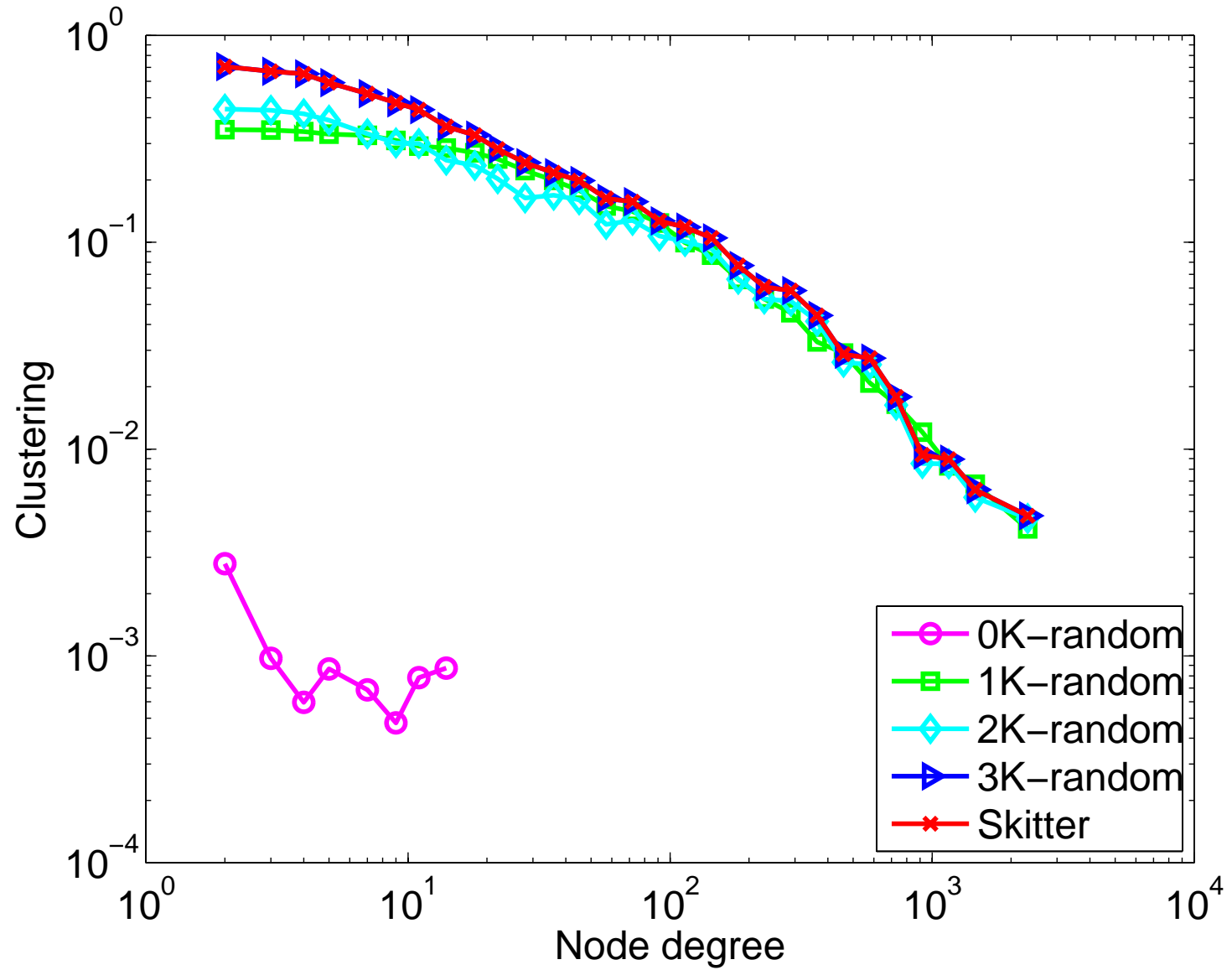
Scalar metrics comparison for skitter

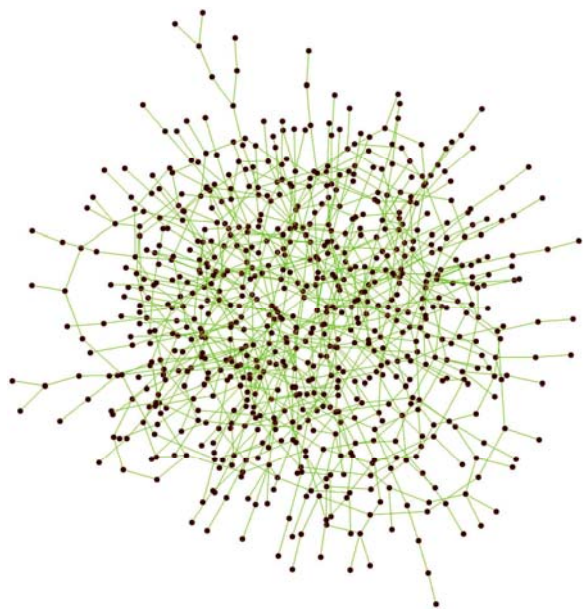
<i>Metric</i>	<i>0K</i>	<i>1K</i>	<i>2K</i>	<i>3K</i>	<i>skitter</i>
$\langle k \rangle$	6.31	6.34	6.29	6.29	6.29
r	0	-0.24	-0.24	-0.24	-0.24
$\langle C \rangle$	0.001	0.25	0.29	0.46	0.46
d	5.17	3.11	3.08	3.09	3.12
σ_d	0.27	0.4	0.35	0.35	0.37
λ_1	0.2	0.03	0.15	0.1	0.1
λ_{n-1}	1.8	1.97	1.85	1.9	1.9

Distance Distribution in skitter

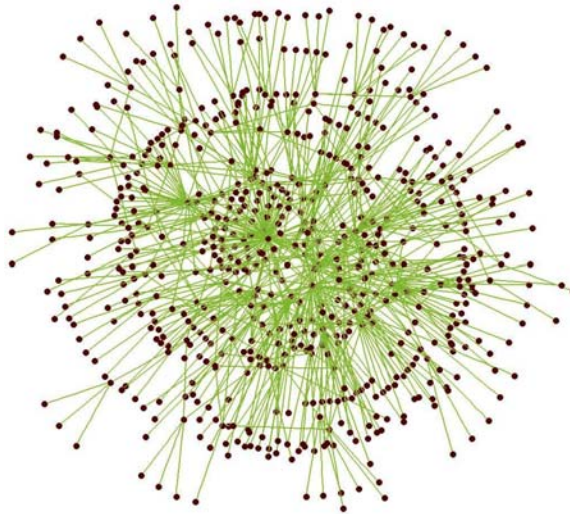


Clustering in skitter

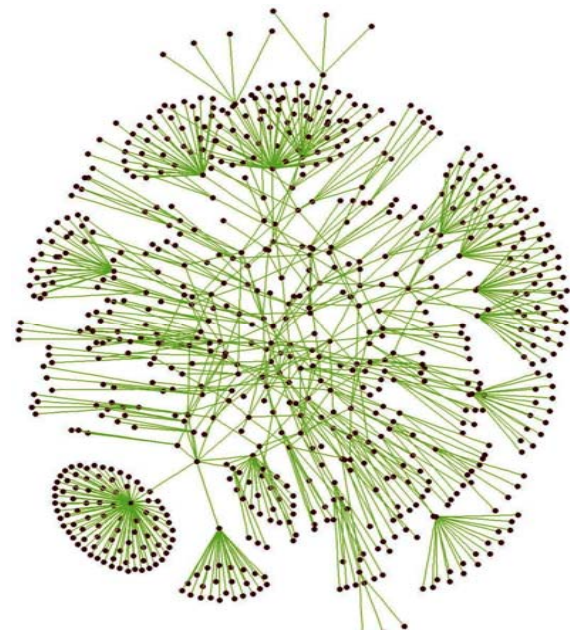




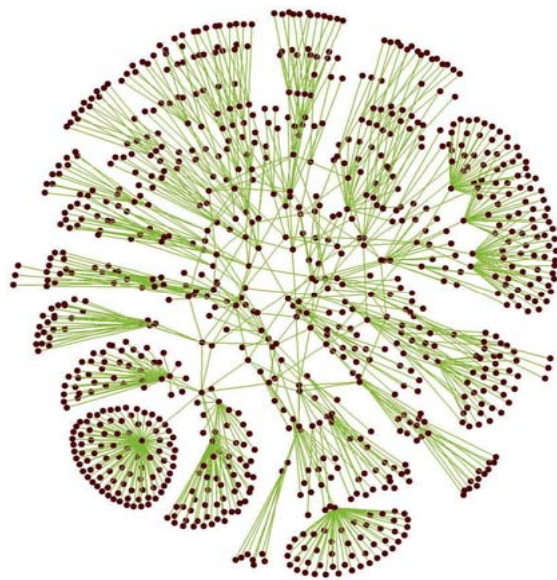
0K-random graph



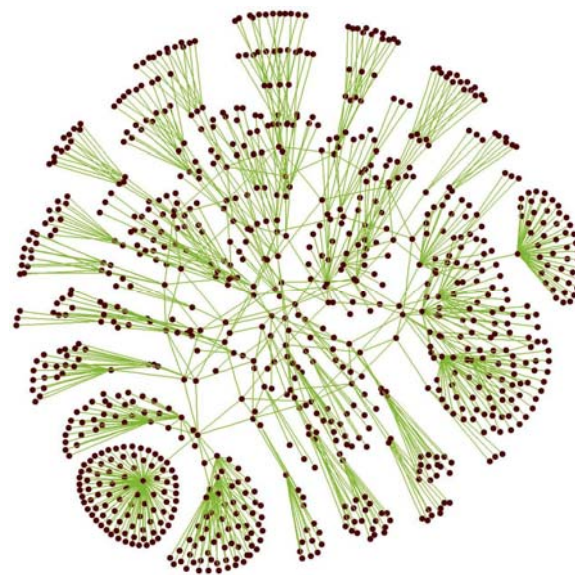
1K-random graph



2K-random graph



3K-random graph

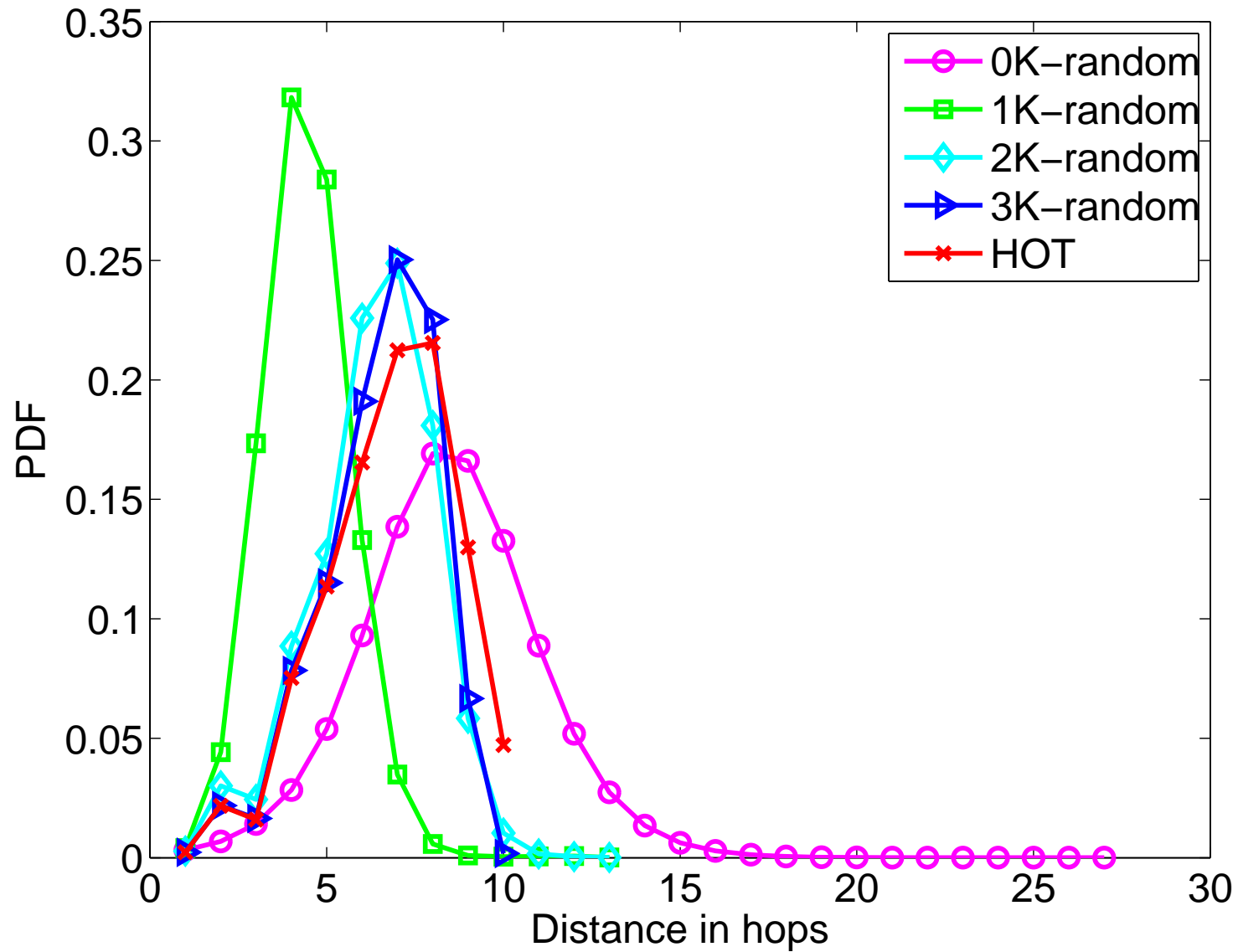


Original graph (HOT)

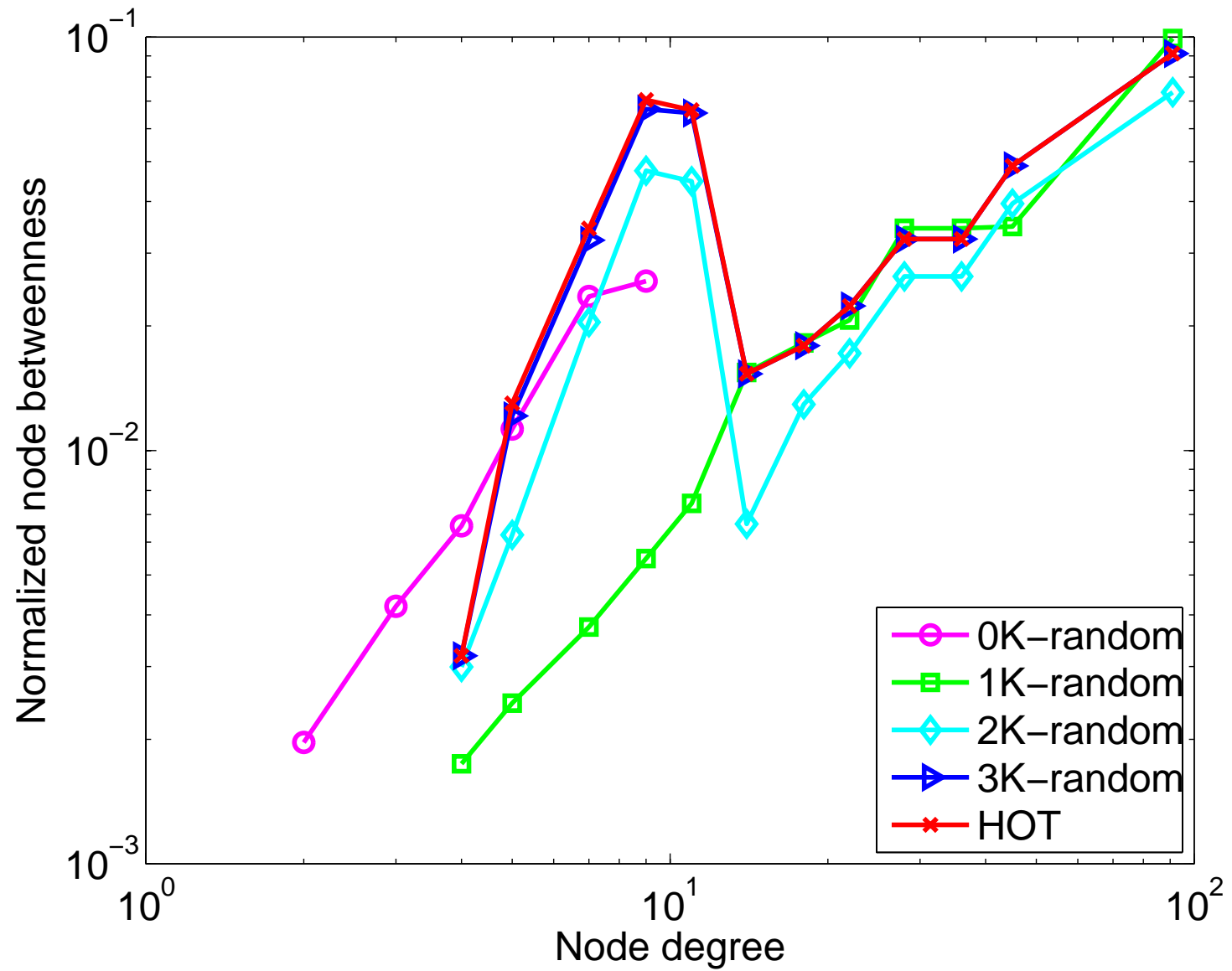
Scalar Metrics for HOT

<i>Metric</i>	<i>0K</i>	<i>1K</i>	<i>2K</i>	<i>3K</i>	<i>HOT</i>
$\langle k \rangle$	2.47	2.59	2.18	2.10	2.10
r	-0.05	-0.14	-0.23	-0.22	-0.22
$\langle C \rangle$	0.002	0.009	0.001	0	0
d	8.48	4.41	6.32	6.55	6.81
σ_d	1.23	0.72	0.71	0.84	0.57
λ_1	0.01	0.034	0.005	0.004	0.004
λ_{n-1}	1.989	1.967	1.996	1.997	1.997

Distance Distribution in HOT



Betweenness in HOT



Results summary

- For AS-level topologies:
 - dK -series convergence is fast
 - $2K$ -random graphs capture most metrics; need $3K$ for clustering
- For router-level topologies:
 - dK -series convergence is slower
 - $3K$ -random graphs reproduce most metrics
- Router-level topologies are less random; reflect careful design and planning

Discussions and Limitations

- For graphs that we considered, $d=3$ seems sufficient
- Not all graphs can be approximated using $3K$
- What if we have to reproduce a new metric?
 - Increase d
 - Extreme case of $d = n$, generated graphs must be isomorphic to the given graph
 - Computational complexity grows rapidly with d
- Cannot discover evolutionary growth of a network

Conclusions

- No need to capture individual metrics
- Our dK -series , $d = 0, 1, \dots, n$ specify degree correlations within non-isomorphic simple connected subgraphs of size d
- By increasing d , we capture more complex properties of a given graph
- $d=3$ is sufficient to reproduce important metrics of observed Internet graphs

More Information

1. A Basis for Systematic Analysis of Network Topologies,
Priya Mahadevan , Dmitri Krioukov, Kevin Fall, and Amin Vahdat. ACM SIGCOMM 2006.