

A general framework for analyzing long-range degree correlations in complex networks

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Keywords

Scale-Free Network

Degree-Degree Correlation

<https://arxiv.org/abs/1712.00910>

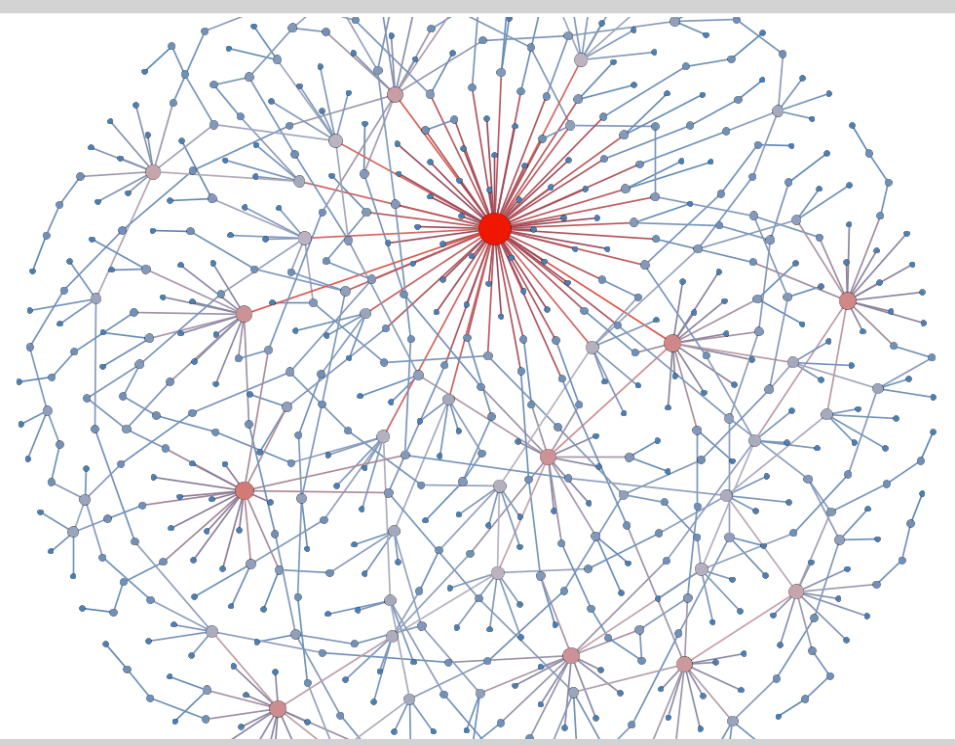
1. Introduction

Degree-degree correlations in real-world networks

Scale-Free Nature of degree distribution

$$P(k) \propto k^{-\gamma}$$

Degree $k_i = 3$

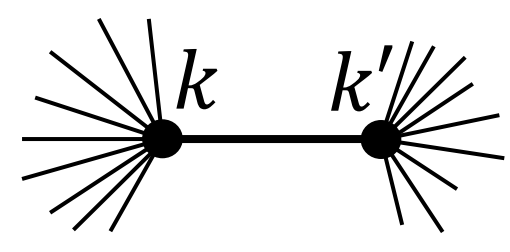


High degree nodes (**Hubs**) exist and play important roles.

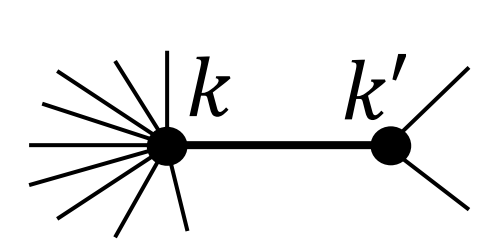
⇒ **Degree-degree correlations** (complexity peculiar to networks)

Previous research tends to focus on adjacent nodes

Nearest neighbor degree correlation (NNDC)



Assortative mixing
Social networks



Disassortative mixing
Biological or Technological networks

Assortativity [M. E. J. Newman (2002)]

$$r = \frac{4 \sum_{k,k'} k k' P(k, k') - [\sum_{k,k'} (k+k') P(k, k')]^2}{2 \sum_{k,k'} (k^2 + k'^2) P(k, k') - [\sum_{k,k'} (k+k') P(k, k')]^2}$$

Average degree of nearest neighbor nodes

$$k_{nn}(k) = \sum_{k'} k' P(k'|k)$$

More generally,

$P_{nn}(k, k')$: Probability that one end node of a randomly chosen edge has the degree k and the other end node has the degree k'

$P_{nn}(k'|k)$: Probability that a node adjacent to a randomly chosen node of degree k has the degree k'

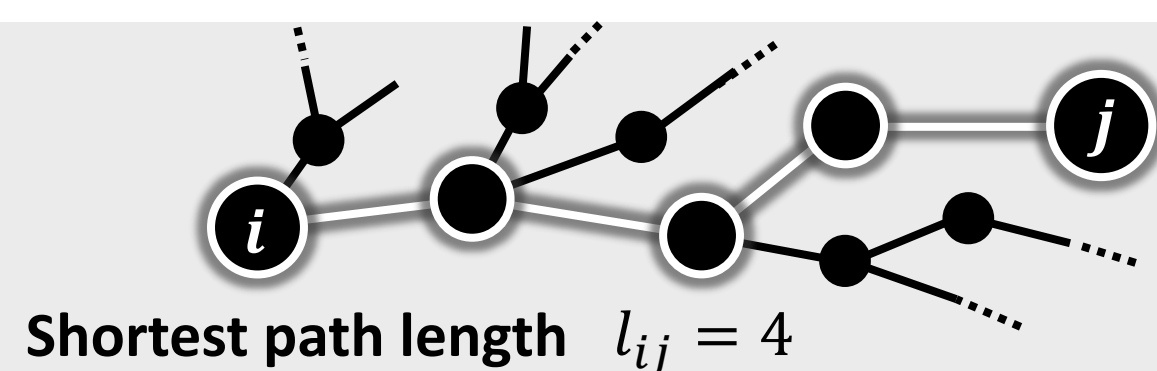
Impact of nearest neighbor degree correlations (NNDCs)

- Resilience for random or targeted attack [M. A. Serrano (2006), A. V. Goltsev (2008)]
- Diffusion of information or disease [A.-L. Barabási (2016), C. E. Gross (2006)]
- Synchronization of oscillator [C. E. La Rocca (2011), V. Avalos-Gaytan (2012)]
- Game theory [Z. Rong (2007)]

However

NNDC is not enough to characterize degree-degree correlations.

Degree correlations between nodes separated by more than one step (i.e., beyond nearest neighbors)



Long-Range Degree Correlation (LRDC)

NNDC is not enough to explain global properties

- Long-range hub repulsion in fractal networks [Y. Fujiki (2017)]
- Reconstruction of networks by NNDC [C. Orsini (2015)]

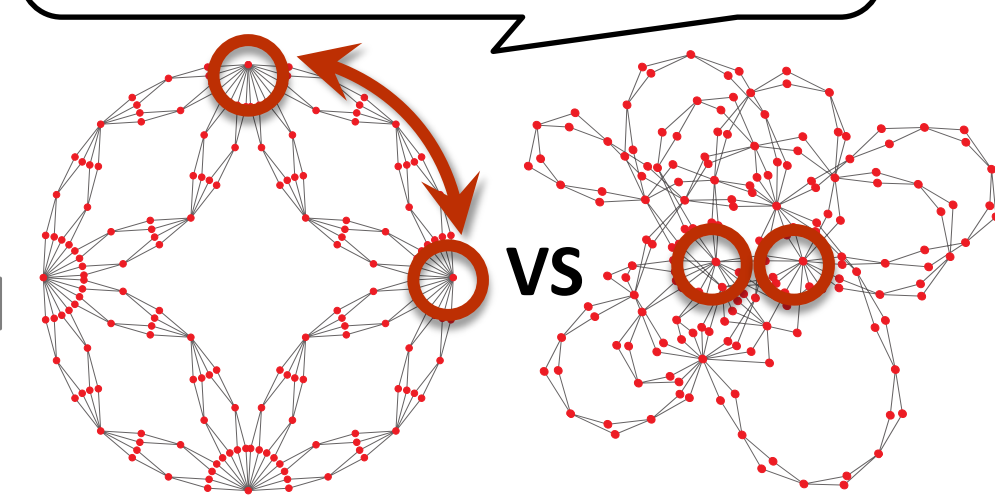
Path lengths between hubs affects on dynamics

- Jamming threshold in communication networks [B. Tadić (2004)]
- Epidemic threshold of the SIS model [M. Boguna (2013)]

Previous proposals for formulating LRDCs

- Long-range assortativity [M. Mayo (2015), A. Arcagni (2017)]
- Fluctuations of the degree along shortest paths [D. Rybski (2010)]
- Two-walks degree assortativity [A. Allen-Perkins (2017)]

☺ How far?
☹ Adjacent or not?



Specific aspects of LRDCs

General argument of LRDCs

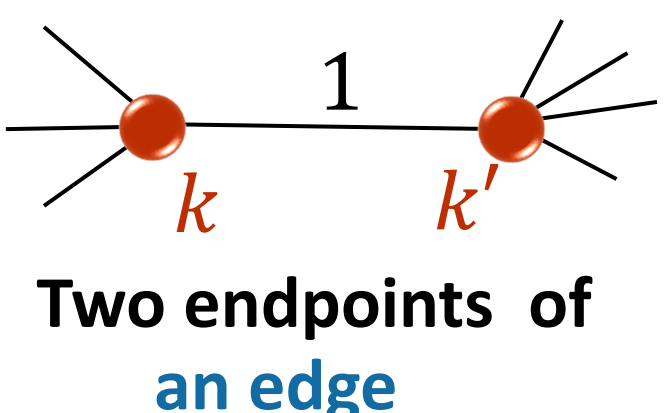
2. Objective

To provide a general description of long-range degree correlations in complex networks

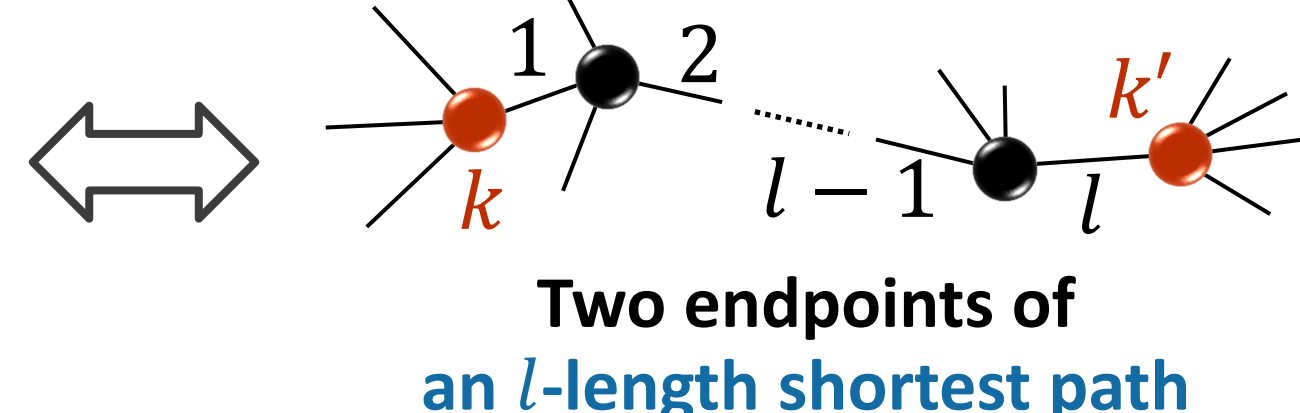
3. Description of LRDC

Five probability distributions characterizing LRDC

NNDC described by $P_{nn}(k, k')$, $P_{nn}(k'|k)$



Two endpoints of an edge



Two endpoints of an l-length shortest path

Joint distribution

$P(k, k', l)$: Probability that one node of a randomly chosen node pair has the degree k , the other node has the degree k' , and the path length between two nodes is l

Conditional distributions

$P(k, k'|l)$: Probability that one node of a randomly chosen node pair separated by l from each other has the degree k and the other node has the degree k'

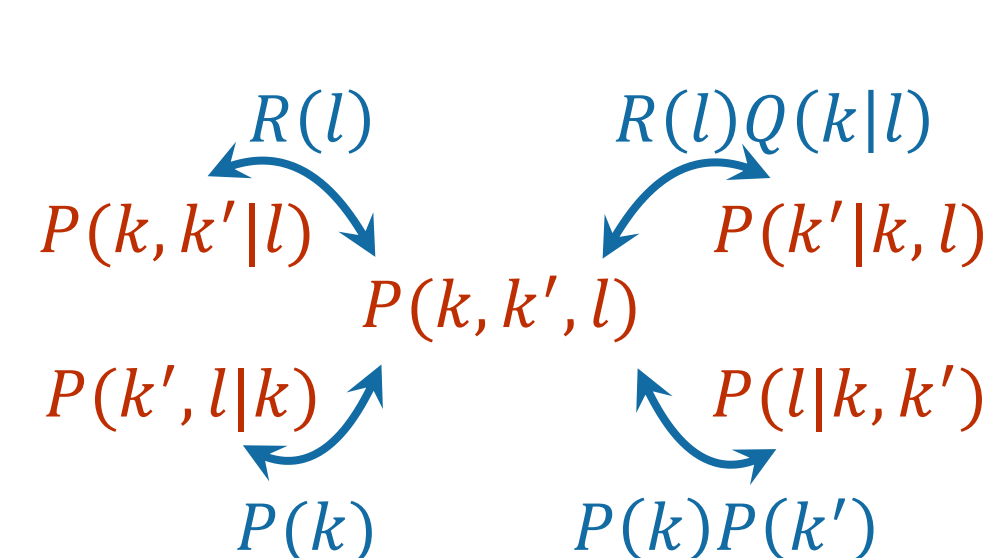
$P(k', l|k)$: Probability that a randomly chosen node has the degree k' and is separated by l from a node of degree k

$P(k'|k, l)$: Probability that a node separated by l from a randomly chosen node of degree k has the degree k'

$P(l|k, k')$: Probability that the path length between randomly chosen two nodes of degrees k and k' is l

Bayes' theorem

$$P(A \cap B) = P(A|B)P(B)$$



$P(k)$: Degree distribution
 $R(l)$: Shortest path length distribution

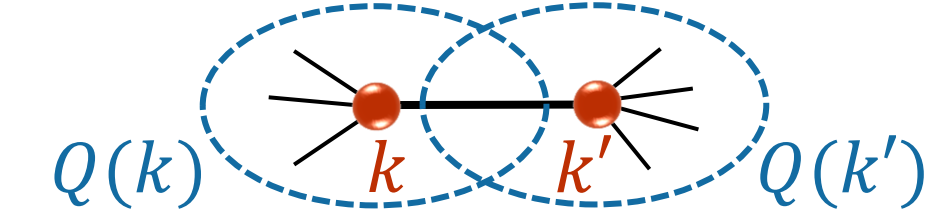
If one of the five probability distributions is given, we can calculate other distributions.

4. Long-range uncorrelated networks (LRUN)

Judgement of existence of LRDCs ← uncorrelated networks

Nearest neighbor uncorrelated networks

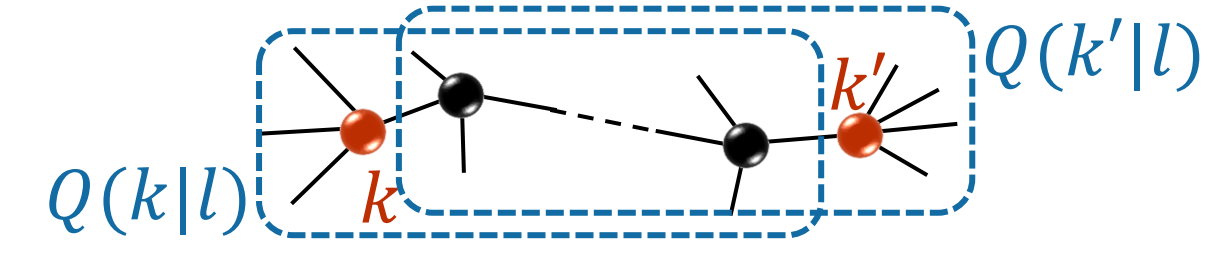
$$P_{nn}(k, k') = Q(k)Q(k')$$



$Q(k|l) = \sum_{k'} P(k, k'|l)$
Probability that one node of a randomly chosen node pair separated by l has the degree k

Long-range uncorrelated network (LRUN)

$$P(k, k'|l) = Q(k|l)Q(k'|l) \text{ for } \forall l$$



4.1. Infinite tree-like networks satisfy the condition of LRUN

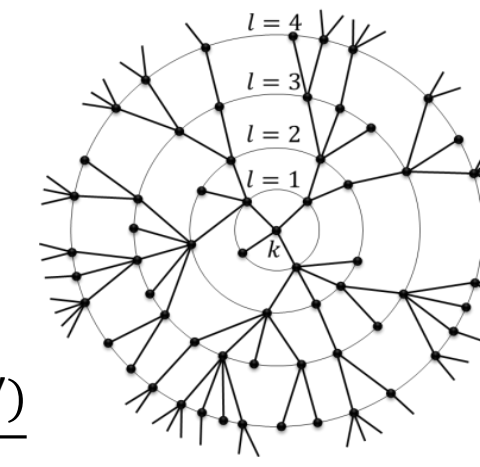
$$P(k, k', l) = 0$$

$$P(k', l|k) = 0$$

$$P(l|k, k') = 0$$

$$P(k, k'|l) = \frac{k k' P(k) P(k')}{\langle k \rangle^2}$$

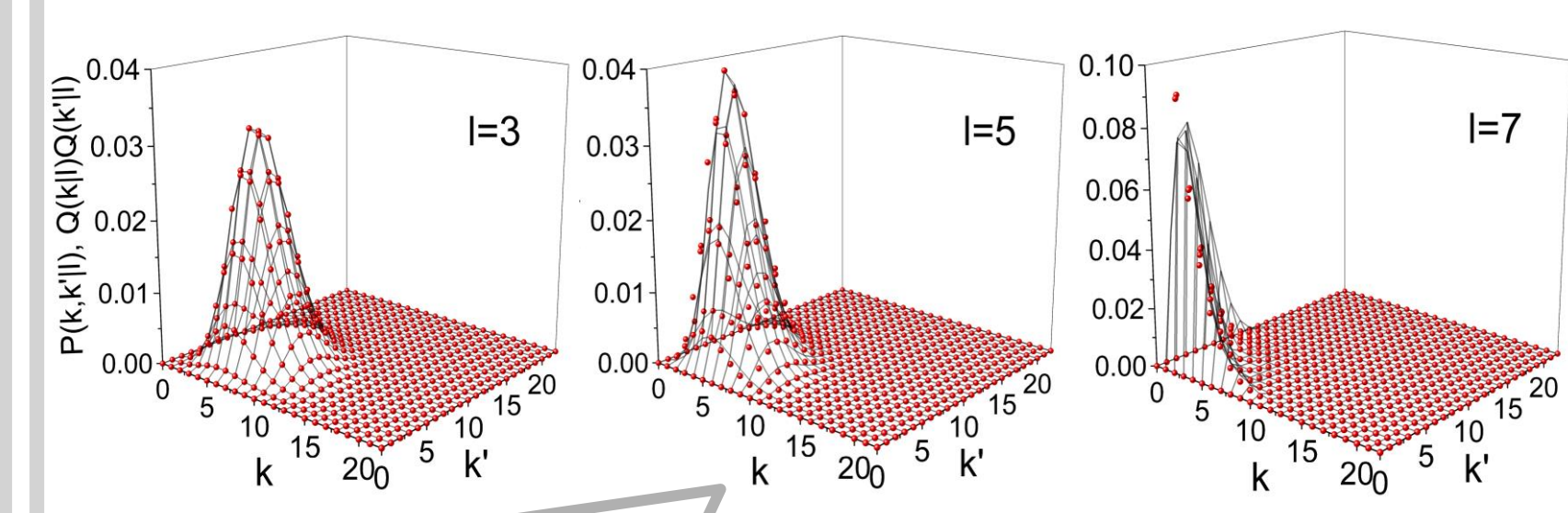
$$P(k'|k, l) = \sum_{k''} P(k''|k, l-1) P_{nn}(k'|k'') = \frac{k' P(k')}{\langle k \rangle}$$



4.2. Can finite networks satisfy that?

It is not easy to answer this question rigorously...

Most plausible candidate for LRUNs = random networks (configuration model)



Confirmation of $P(k, k'|l) = Q(k|l)Q(k'|l)$ for Erdős-Rényi random graphs

$$\begin{cases} P(k, k'|l) \neq Q(k|l)Q(k'|l) & \text{for } l \gg \langle l \rangle & \text{(finite-size effect)} \\ P(k, k'|l) = Q(k|l)Q(k'|l) & \text{for } l \ll \langle l \rangle & \text{(local tree-like)} \end{cases}$$

$\#$: $P(k, k'|l)$
 \bullet : $Q(k|l)Q(k'|l)$
 $N = 1,000$
 $\langle k \rangle = 5.0$
 $\langle l \rangle = 4.47$

From a practical viewpoint

Baseline for comparison

Random networks (P_0) with the same degree sequence

Calculation of P_0 using the mean-field and local-tree approximations

[S. Melnik and J. P. Gleeson, arXiv:1604.05521 (2016)]

$$P_0(l|k, k') = \rho_l^{k k'} - \rho_{l-1}^{k k'}$$

Recursion formula:

$$1 - \bar{q}_{l+1}^{k'} = G_1(1 - \bar{q}_l^{k'}) - \frac{k'(1 - \bar{q}_l^{k'})^{k'-1}}{N \langle k \rangle}$$

$\rho_l^{k k'}$: Probability that the distance between randomly chosen two nodes of degrees k and k' is equal to or less than l

$\bar{q}_l^{k'}$: Probability that an adjacent node of a randomly chosen node i lies within l from a k' -degree node j under the condition that i is separated more than l from j

Initial state: $\rho_0^{k k'} = \frac{\delta_{k k'}}{N P(k)}$, $\bar{q}_0^{k'} = \frac{k'}{N \langle k \rangle}$

$$1 - \rho_l^{k k'} = [1 - \rho_0^{k k'}][1 - \bar{q}_{l-1}^{k'}]^k$$

5. Applying to real-world networks

Measure to characterize LRDCs

Extensions of measures for NNDC

$$P_{nn}(k, k') = P(k, k'|l=1)$$

$$P_{nn}(k'|k) = P(k'|k, l=1)$$

➢ Long-range assortativity: [M. Mayo (2015)]

$$r_l = \frac{4 \sum_{k,k'} k k' P(k, k'|l) - [\sum_{k,k'} (k+k') P(k, k'|l)]^2}{2 \sum_{k,k'} (k^2 + k'^2) P(k, k'|l) - [\sum_{k,k'} (k+k') P(k, k'|l)]^2}$$

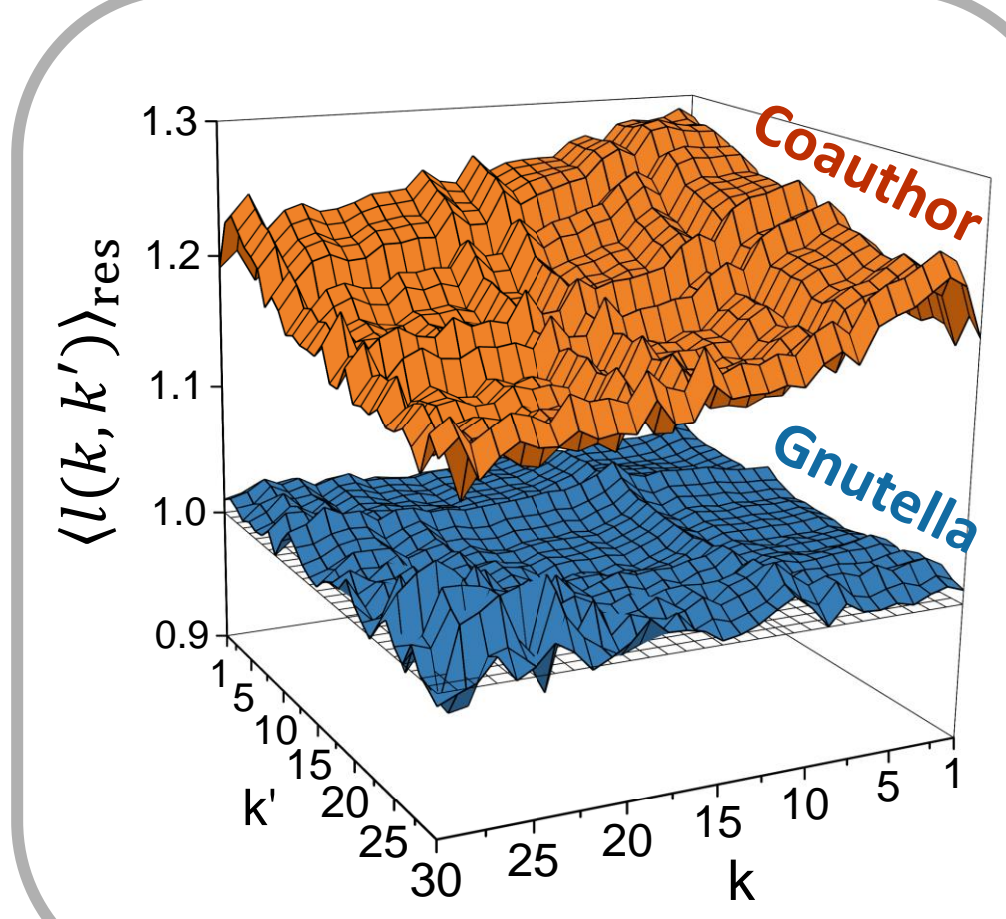
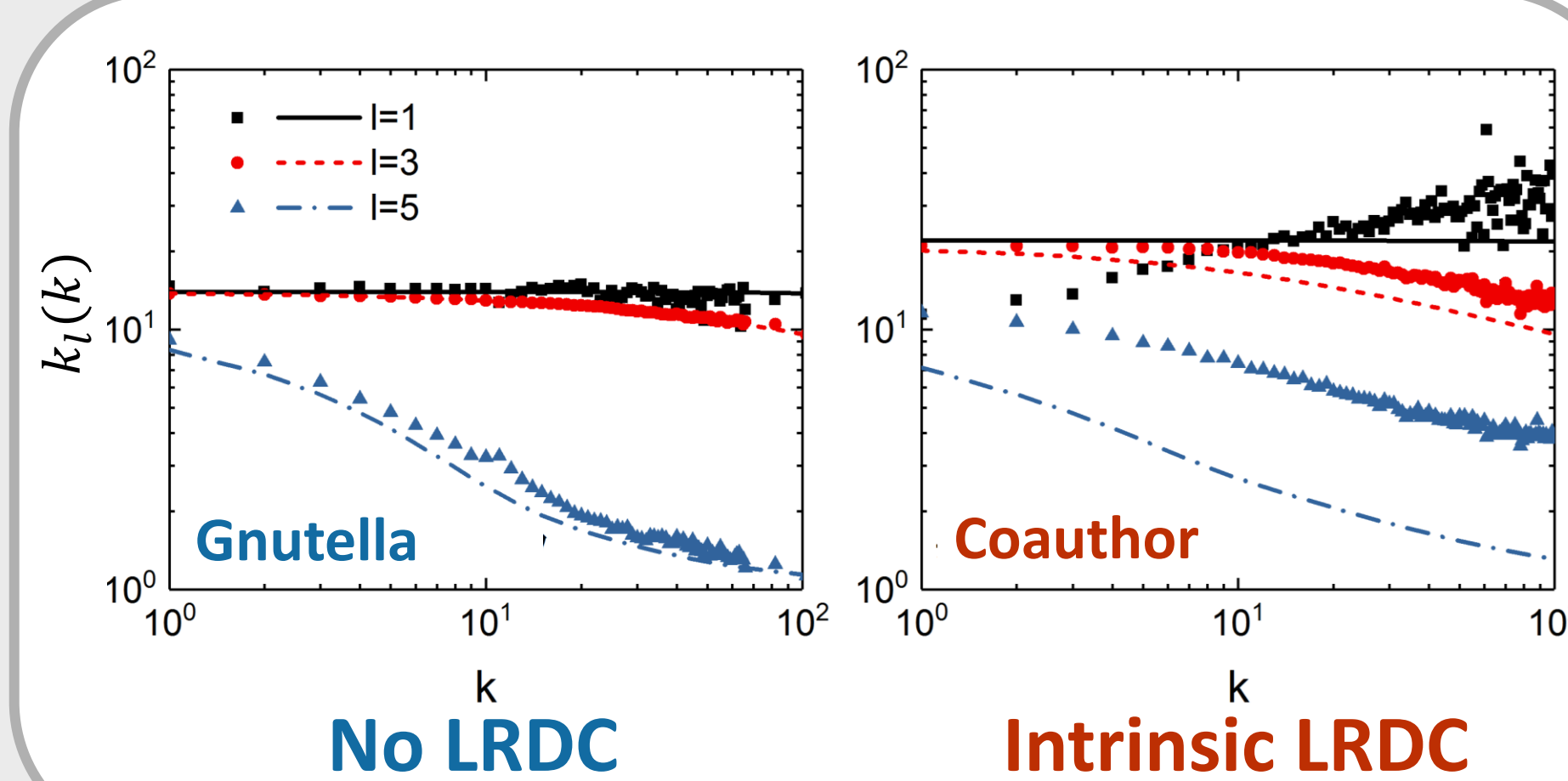
➢ Average degree of l -th neighbor nodes:

$$k_l(k) = \sum_{k'} k' P(k'|k, l)$$

Other possible indices

➢ Rescaled average shortest-path length:

$$\langle l(k, k') \rangle_{\text{res}} \equiv \frac{\sum_l l P(l|k, k')}{\sum_l P_0(l|k, k')}$$



Real-World Networks

Gnutella peer-to-peer network

$N = 10,876$ $\langle k \rangle = 7.4$
 $E = 39,994$ $\langle l \rangle = 4.6$

Coauthorship network

$N = 23,133$ $\langle k \rangle = 8.1$
 $E = 93,439$ $\langle l \rangle = 5.4$

Conclusions

We provided a general framework for analyzing LRDCs.

- To fully describe LRDCs, we introduced fundamental five distributions $P(k, k', l)$, $P(k, k'|l)$, $P(l|k, k')$, $P(k'|k, l)$, and $P(k', l|k)$. If one of them is given, we can calculate others using Bayes' theorem.
- We adopted random networks as a baseline to judge the existence of LRDCs, instead of LRUNs defined by $P(k, k'|l) = Q(k|l)Q(k'|l)$, and analytically calculated the probability distributions (P_0) for random networks within the mean-field approximation.
- The utility of our argument was demonstrated by applying it to real-world networks.

One can introduce new measures in our framework.