Lecture 11 Betweenness Centrality



Say, (1) EVERY pair of nodes exchanges a message With equal Drobability / cenit time (2) Messages travel on shortest path "Indomnation" through node /per cuit time \Rightarrow

~ # of shortest parts through the node

Xi ~ 5 nost/9st total # of shortest $\begin{array}{c} \# of shortest paths \\ \vdots & \text{drow s to t through i} \\ \text{if } g_{st} = 0 \implies n_{st}^{s} = 0 \end{array}$ pathas Nom Stot Remark-2: Convention. $\frac{n_{jt}}{3} = 0$ S= t Not excluded

Wide rauge in the value of betweenen centra. P. g.,

Central node $X_c = n^2 - (n-1)$ total # of # of peripheral n usdes $attue (n_{it}^{S} = g_{it})$

HODE

star

Leaf

From I to others n-1Opposite direct. n-1l to itself 1

$$\frac{Lower bound}{Lower bound} = \frac{n^2 - n + 1}{2n - 1} \approx \frac{1}{2}n$$

Example: FR: 7.5×108 }~103~n Nest: 8.9×105 }~103~n Lowest +USA 1971 di-em: The French Connection Fernando Rey With GENE Hackman





Scaling of Betweenness Centrality in Weighted Complex Networks



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Characterization of weighted complex networks

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To account for possible distinct functional roles played by different nodes and links in complex networks, we introduce and analyze a class of weighted scale-free networks. The weight of a node is assigned as a random number, based on which the weights of links are defined. We utilize the concept of *betweenness* to characterize the weighted networks and obtain the scaling laws governing the betweenness as functions both of the weight and of the degree. The scaling results may be useful for identifying influential nodes in terms of physical functions in complex networks.

FIG. 4. Algebraic scaling between $B_K(k)$ and k for a weighted scale-free network with m=2 and N=10000. The value of the scaling exponent is $\alpha \approx 1.5$. A similar scaling relation is obtained for the corresponding nonweighted network, where $\alpha \approx 1.6$, as shown in the inset.



Synchronization in Complex Networks

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Heterogeneity in Oscillator Networks: Are Smaller Worlds Easier to Synchronize?

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Small-world and scale-free networks are known to be more easily synchronized than regular lattices, which is usually attributed to the smaller network distance between oscillators. Surprisingly, we find that networks with a homogeneous distribution of connectivity are more synchronizable than heterogeneous ones, even though the average network distance is larger. We present numerical computations and analytical estimates on synchronizability of the network in terms of its heterogeneity parameters. Our results suggest that some degree of homogeneity is expected in naturally evolved structures, such as neural networks, where synchronizability is desirable.

Semi-random scale-free network model

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Smaller $\gamma \longleftrightarrow$ More heterogeneous degree distribution



Growing scale-free network model

Smaller $\alpha \longleftrightarrow$ More heterogeneous degree distribution



























Role of Betweenness Centrality (Load) in Network Synchronization

