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Networks and Graphs

Graphs :

use graph theory to analyse Networks



undirected edge



directed edge

graph can be stored as adjacency matrix or adjacency list

Characteristics: degree of a Node = # of edges connected to that node

path = sequena of edges that joins 2 Nodes

complete = an edge between every pair of Nodes

Network Characteristic

distribution of edges / nodes degrees:

- Anomaly detection
- Ranking / Recommendation
- network flow

Centrality of a node:

- discover clusters

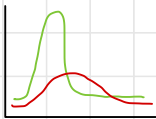
Network Analysis

state of a Network / Graph is Stochastic

Random Graph Model: distribution of the degree of Nodes

$$P(\deg(v) = k) = \binom{N-1}{k} p^k (1-p)^{N-1-k}$$

Power Law: Random Graph



Power Law



Metrics on Graph

1) diameter : $\text{diam}(G) = \max_{i,j} d_{ij}$

- 2) clustering coefficient $= c = \frac{\# \text{ of triangle}}{\# \text{ of triplets}}$
- 3) density: $N = \text{nodes}, M = \text{edges}$. Density $= \frac{2M}{N(N-1)}$

Metrics on Nodes

- 1) Degree Centrality: $C_{\text{deg}}(v) = \text{Deg}(v)$
- 2) Closeness Centrality: $C_{\text{close}}(v) = \frac{1}{\sum_u d(u,v)}$
- 3) Harmonic Centrality: $C_h(v) = \sum_{u \neq v} \frac{1}{d(u,v)}$
- 4) Betweenness Centrality: $C_b(v) = \sum_{s \neq v \neq t \in V} \frac{g_{st}(v)}{g_{st}}$

Ranking Aggregation

kendall τ distance: $d_\tau(w_1, w_2) = \frac{\# \text{ pairs ranked in different order}}{\# \text{ pairs}}$

Given m Rankings w_1, \dots, w_m , the aggregate ranking w^*

$$w^* = \arg \max_w \sum_{i=1}^m d_\tau(w, w_i)$$