

Graph Mining SD212

2. Random graphs

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Motivation

Random graph = random instance of a graph with some specific statistical properties

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Useful for:

- ▶ generating graphs “for free”
- ▶ testing algorithms (simulation)
- ▶ proving algorithms / providing performance guarantees (analysis)

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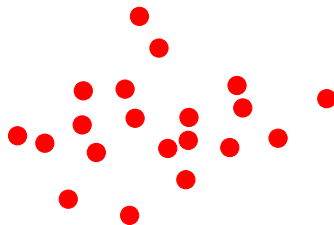
We focus on **undirected** graphs; the results naturally extend to directed graphs

Outline

1. Erdős-Rényi graphs
2. Preferential attachment
3. Configuration model
4. Stochastic block model

Erdős-Rényi graphs

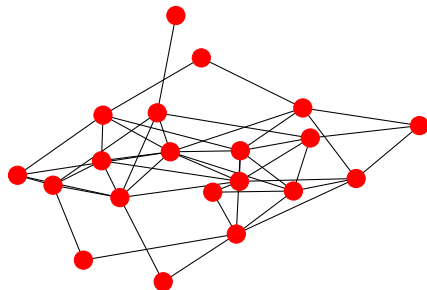
- ▶ n nodes
- ▶ $p \in (0, 1)$
- ▶ An edge with probability p between any distinct nodes u, v



$\sim B(n, p)$

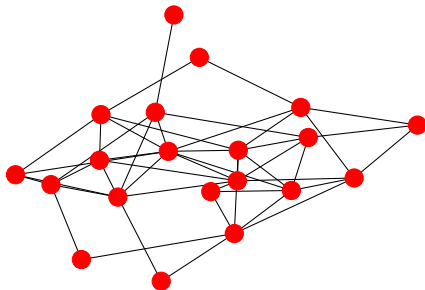
Erdős-Rényi graphs

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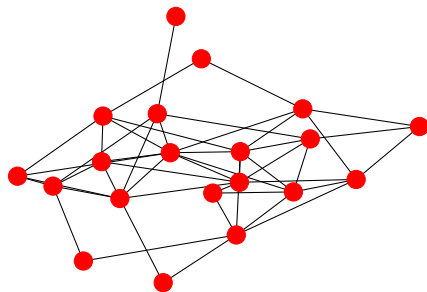
Erdős-Rényi graphs

- ▶ n nodes
- ▶ $p \in (0, 1)$
- ▶ An edge with probability p between any distinct nodes u, v
- ▶ Degree distribution



Erdős-Rényi graphs

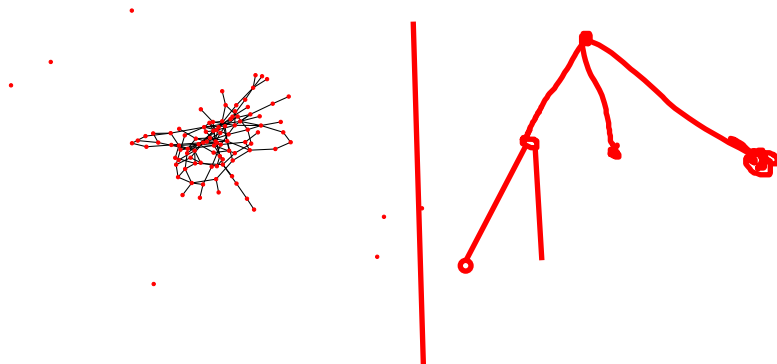
- ▶ n nodes
- ▶ $p \in (0, 1)$
- ▶ An edge with probability p between any distinct nodes u, v
- ▶ Degree distribution (all instances)
- ▶ Empirical degree distribution (one instance)



Large Erdős-Rényi graphs

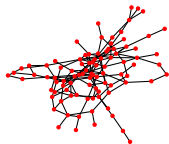
- ▶ $n \rightarrow +\infty$
- ▶ $p \rightarrow 0$
- ▶ $np \rightarrow \lambda$

Example with $n = 100$, $p = 0.03$ ($\lambda = 3$)



Large Erdős-Rényi graphs

- ▶ $n \rightarrow +\infty$
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$$-P(\lambda)$$

$$d_1, \dots, d_n \sim \text{i.i.d.}$$

$$p_k = \frac{1}{n} \sum_{i=1}^n \mathbb{I}_{\{d_i = k\}} \\ = E(\mathbb{I}_{\{d_i = k\}}) = e^{-\lambda} \frac{\lambda^k}{k!}$$

Galton-Watson tree

Recursive definition:

- ▶ A root
- ▶ The offspring of each node has a Poisson distribution with parameter λ $n \rightarrow +\infty$, $p \rightarrow 0$, $(n - n_{\text{new}})p = \lambda$
 $X \sim P(\lambda)$

$$\begin{aligned} G(t) &= E(t^*) \\ &= \sum \frac{\lambda^k}{k!} e^{-\lambda} t^k \\ &= e^{\lambda(t-1)} \\ G_k(t) &= E(k^{2k}) \end{aligned}$$

$$\begin{aligned} Z_{k+1} &= \sum_{i=2}^{\infty} Z_i X_i \\ G_{k+1}(t) &= E(t^{Z_{k+1}}) = E(t^{\sum_{i=2}^{\infty} Z_i X_i}) \\ &= \sum_{b \geq 0} P(Z_k = n) E(t^{\sum_{i=1}^n X_i}) = G_k(G(t)) \\ G_{k+1} &= G_k(G(E)) = G \circ \dots \circ G(t) \quad (k+1 \text{ 个 } G \text{ 复合}) \end{aligned}$$

$$E(Z_{k+1}) = \lambda E(Z_k)$$

Three regimes

$\lambda < 1$:



$$\frac{1}{1-\lambda}$$

$\lambda > 1$:



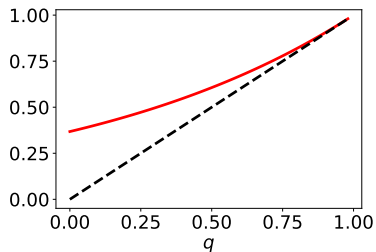
Extinction probability

Assume $\lambda \geq 1$ $q = \lim_{k \rightarrow +\infty} P(Z_k=0)$

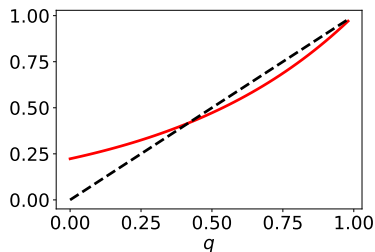
Fixed-point equation

$$q = e^{\lambda(q-1)}$$

$\lambda = 1$

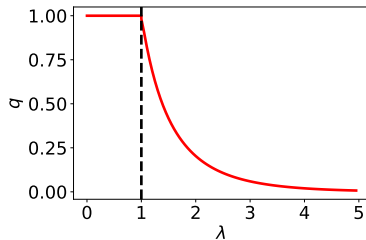


$\lambda > 1$



Extinction probability

$$q = e^{\lambda(q-1)}$$



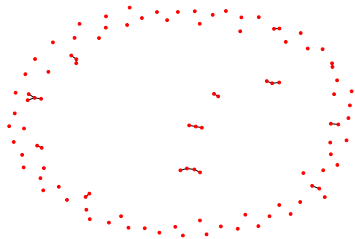
Back to Erdős-Rényi graphs

Three regimes:

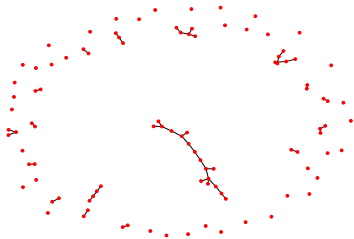
- ▶ **Subcritical** ($\lambda < 1$): finite tree \rightarrow many small components
- ▶ **Supercritical** ($\lambda > 1$): infinite tree with probability $p = 1 - q$
 \rightarrow one **giant component** containing a fraction p of the nodes (the others in small components)
- ▶ **Critical** ($\lambda = 1$): finite tree (but infinite expectation) \rightarrow one large component

Examples

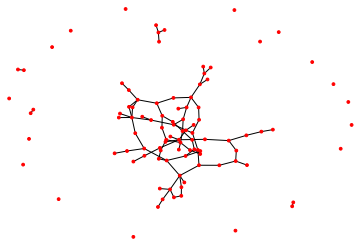
$\lambda = 0.5$



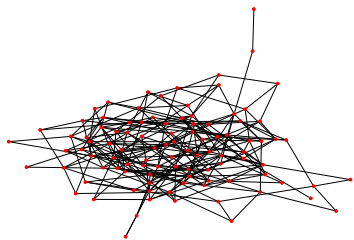
$\lambda = 1$



$\lambda = 2$



$\lambda = 5$



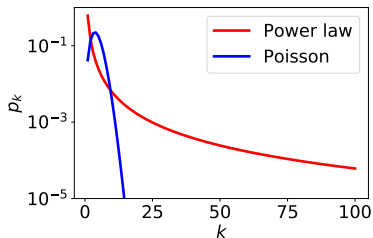
Outline

1. Erdős-Rényi graphs
2. **Preferential attachment**
3. Configuration model
4. Stochastic block model

Power law

- ▶ The typical degree distribution of real graphs is of the form

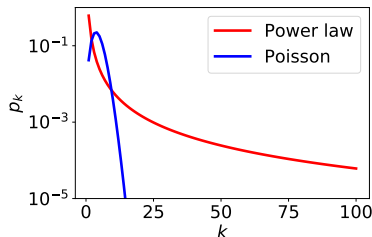
$$p_k \propto \frac{1}{k^\alpha} \quad \alpha > 1$$



Power law

- ▶ The typical degree distribution of real graphs is of the form

$$p_k \propto \frac{1}{k^\alpha} \quad \alpha > 1$$



- ▶ Explained by the “**rich get richer**” phenomenon
Barabasi & Albert 1999

Barabasi-Albert model

Explicit construction (with $d \geq 1$):

- ▶ Start from a clique of d nodes
- ▶ Add new nodes one at a time, each of degree d and with **preferential attachment**

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Configuration model

Given some sequence of integers d_1, \dots, d_n , can we generate a graph with this particular **degree sequence**?

Havel-Hakimi algorithm

<http://blog.51cto.com/sbp810050504/883904>

Random configuration

Number of self-loops

Outline

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4. **Stochastic block model**

Stochastic block model

Random graphs with some underlying **community** structure

- ▶ A partition of $\{1, \dots, n\}$ in k blocks
- ▶ A block connectivity matrix B

Summary

Random graphs are generated from **models**

- ▶ Erdős-Rényi graphs → No structure
- ▶ Preferential attachment → Power law
- ▶ Configuration model → Degree sequence
- ▶ Stochastic block model → Clusters

These may be combined to get more realistic (but complex) models