

# Graph Mining

## SD212

### 1. Node and edge sampling

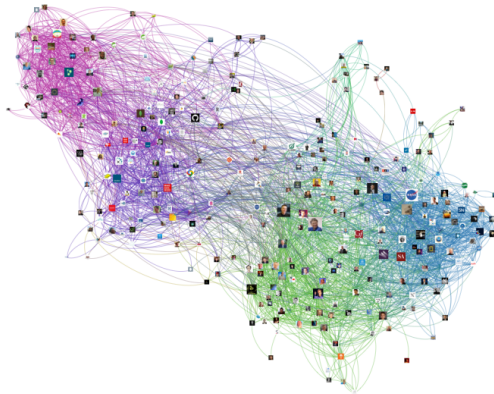
Thomas Bonald

2017 – 2018



# The friendship paradox

“Your friends have more friends than you on average!”



# Outline

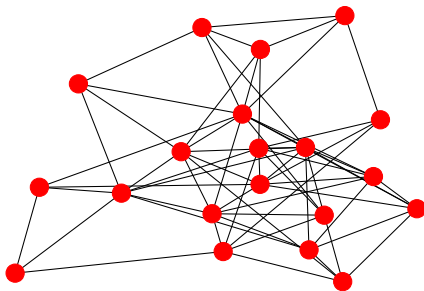
1. Undirected graphs
2. Directed graphs

# Setting

- ▶ Some **undirected** graph  $G = (V, E)$  of  $n$  nodes and  $m$  edges

$$V = \{1, \dots, n\} \quad E \subset \{\{u, v\}, u, v \in V\}$$

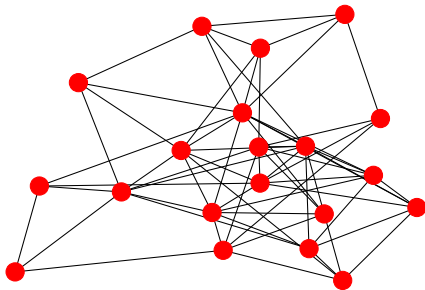
- ▶ No self-loops
- ▶  $d_u$ , degree of node  $u$



# Setting

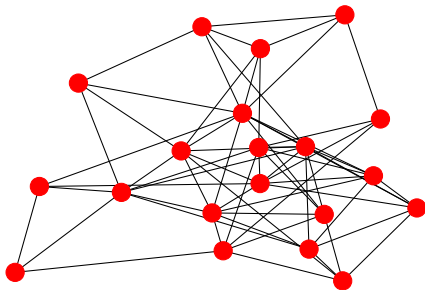
## Proposition

$$\sum_u d_u = 2m \qquad \bar{d} \equiv \frac{1}{n} \sum_u d_u = \frac{2m}{n}$$



## Random node

Sample a **node** uniformly at random



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- Degree distribution

$$p_k = \frac{1}{n} \sum_u 1_{\{d_u=k\}}$$

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Sample a **node** uniformly at random

- ▶ Degree distribution

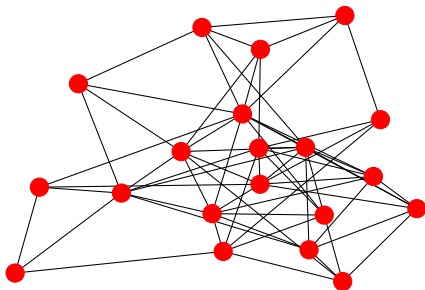
$$p_k = \frac{1}{n} \sum_u 1_{\{d_u=k\}}$$

- ▶ Expected degree



## Random edge

Sample an **edge** uniformly at random and one of the two ends of this edge uniformly at random



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- Degree distribution

$$\hat{p}_k \propto kp_k$$

This is the **size-biased** distribution

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## Example: power law distribution

- ▶ Typical degree distribution of real graphs

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

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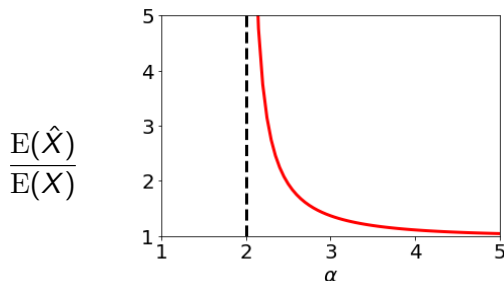
- ▶ Explained by the “**rich get richer**” phenomenon  
Barabasi & Albert 1999

## Example: power law distribution

- ▶ Typical degree distribution of real graphs

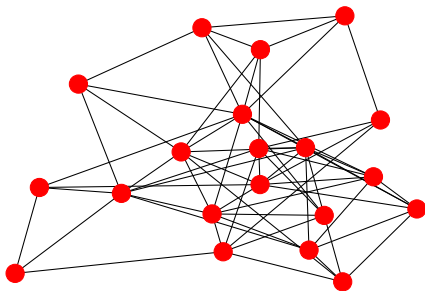
$$p_k \propto \frac{1}{k^\alpha} \quad \alpha > 1$$

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



## Random neighbor

Sample a **neighbor** uniformly at random (from a random node)



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Sample a **neighbor** uniformly at random (from a random node)

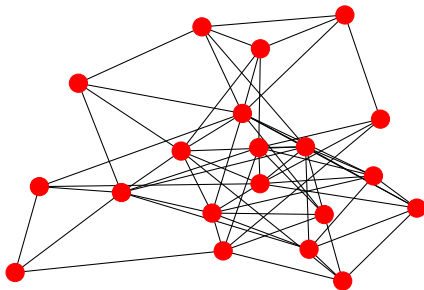
### Proposition

The expected degree is **larger** than  $\bar{d}$



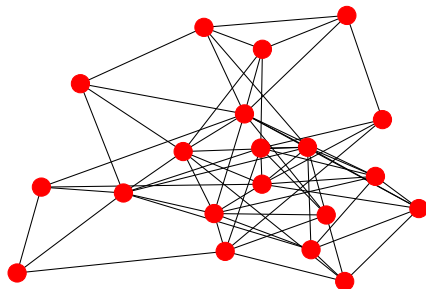
## Degree correlation

- Take the degrees of two **neighbors** (random edge)

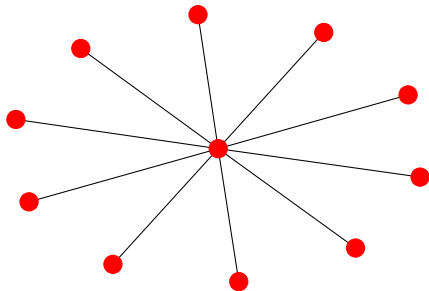


## Degree correlation

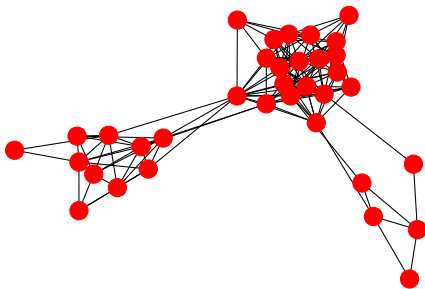
- ▶ Consider the degrees of two **neighbors** (random edge)
- ▶ Their correlation indicate the **assortativity** of the graph



## A disassortative graph



## An assortative graph



# Outline

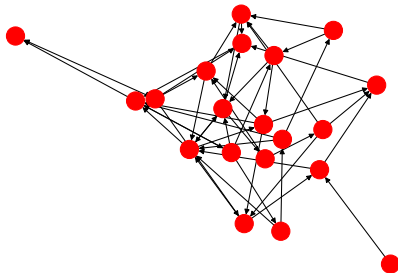
1. Undirected graphs
2. Directed graphs

# Setting

- ▶ Some **directed** graph  $G = (V, E)$  of  $n$  nodes and  $m$  edges

$$V = \{1, \dots, n\} \quad E \subset \{(u, v), u, v \in V\}$$

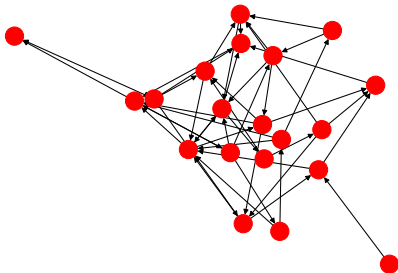
- ▶  $d_u^+ / d_u^-$ , out-degree/in-degree of node  $u$



# Setting

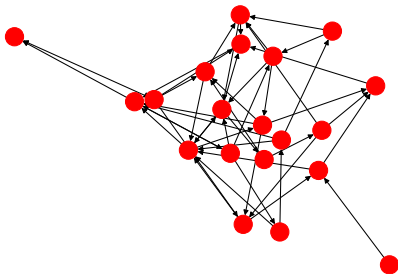
## Proposition

$$\sum_u d_u^+ = \sum_u d_u^- = m \quad \bar{d}^+ = \bar{d}^- = \frac{m}{n}$$



## Random node

Sample a **node** uniformly at random





## Random node

Sample a **node** uniformly at random

- Out/in degree distributions

$$p_k^+ = \frac{1}{n} \sum_u 1_{\{d_u^+ = k\}} \quad p_k^- = \frac{1}{n} \sum_u 1_{\{d_u^- = k\}}$$

## Random node

Sample a **node** uniformly at random

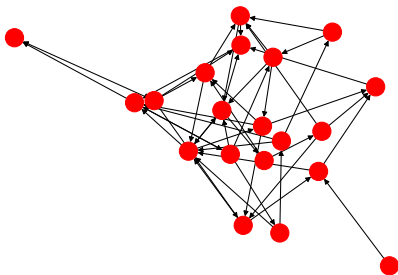
- ▶ Out/in degree distributions

$$p_k^+ = \frac{1}{n} \sum_u 1_{\{d_u^+ = k\}} \quad p_k^- = \frac{1}{n} \sum_u 1_{\{d_u^- = k\}}$$

- ▶ Expected out/in degrees

## Random edge

Sample an **edge** uniformly at random



## Random edge

Sample an **edge** uniformly at random

- ▶ Out-degree of the **origin** / in-degree of the **destination**

$$\hat{p}_k^+ \propto kp_k^+ \quad / \quad \hat{p}_k^- \propto kp_k^-$$

## Random edge

Sample an **edge** uniformly at random

- ▶ Out-degree of the **origin** / in-degree of the **destination**

$$\hat{p}_k^+ \propto kp_k^+ \quad / \quad \hat{p}_k^- \propto kp_k^-$$

- ▶ Expected out/in-degrees

## Random neighbor

Sample a **successor** / **predecessor** uniformly at random  
(from a random node)

# Summary

- ▶ Sampling from **nodes** / **edges** / **neighbors** is not the same!
- ▶ Something to think about

Are you **really** popular?

**Hint:** Look at your friends