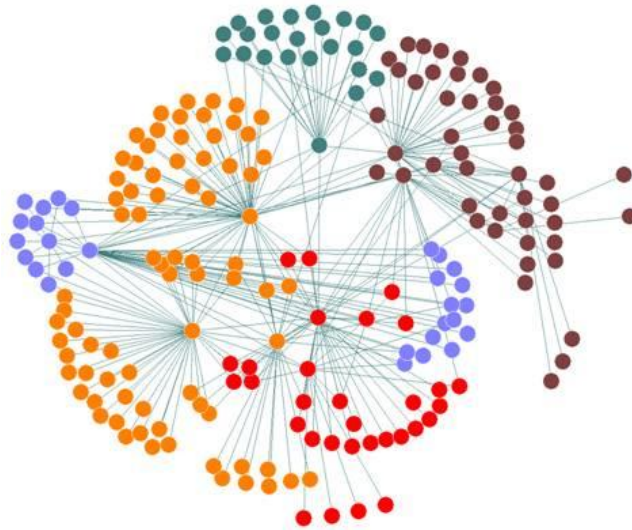




Algorithms and Applications in Social Networks



2023/2024, Semester A

Slava Novgorodov

Lesson #2

- Random network models
- Centrality measures

Random Graphs

Erdős–Rényi model

- Two variants of the model:
 - $G(n, m)$ – a graph is chosen uniformly from a set of graphs with n nodes and m edges
 - $G(n, p)$ – a graph is constructed on n nodes, with probability of edge equals to p
- We will focus on the second variant
- Expected number of edges and average degree:

$$\overline{m} = \frac{n(n-1)}{2} p \qquad \overline{k} = \frac{1}{n} \sum_i k_i = \frac{2\overline{m}}{n} = p(n-1)$$

Erdős–Rényi model

- Probability of node i having a degree k :

$$P(k_i = k) = \binom{n-1}{k} p^k (1-p)^{n-1-k}$$

- Binomial distribution, which becomes Poisson when $n \rightarrow \infty$

$$\lambda = pn$$

$$P(k_i = k) = \frac{\lambda^k e^{-\lambda}}{k!}$$

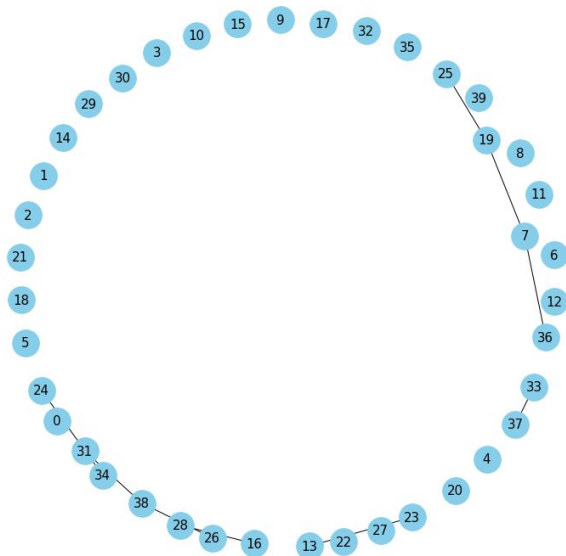
- Phase transition at p_c (critical point) = $1/n$

Erdős–Rényi model

- Example – 40 nodes, different p

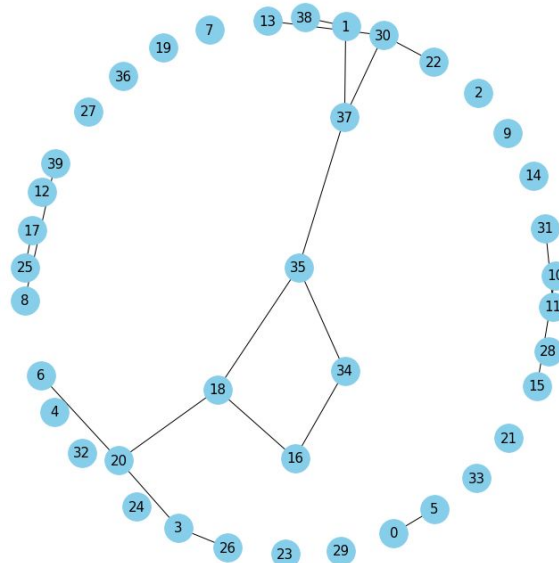
$p = 0.01$, 9 edges

Avg. degree = 0.45



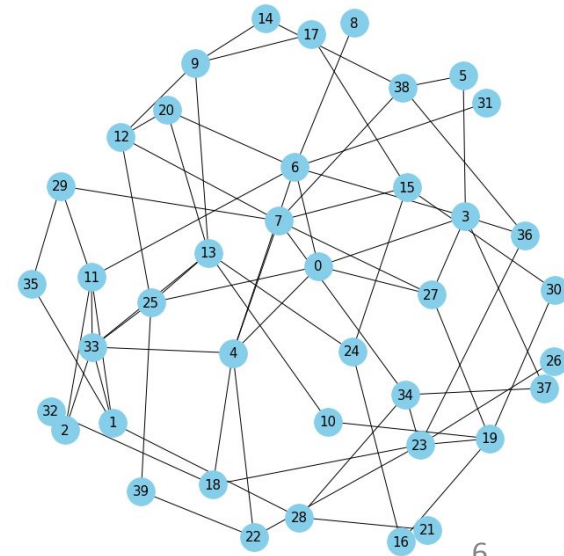
$p = 0.025$, 19 edges

Avg. degree = 0.95



$p = 0.1$, 69 edges

Avg. degree = 3.45



Erdős–Rényi model

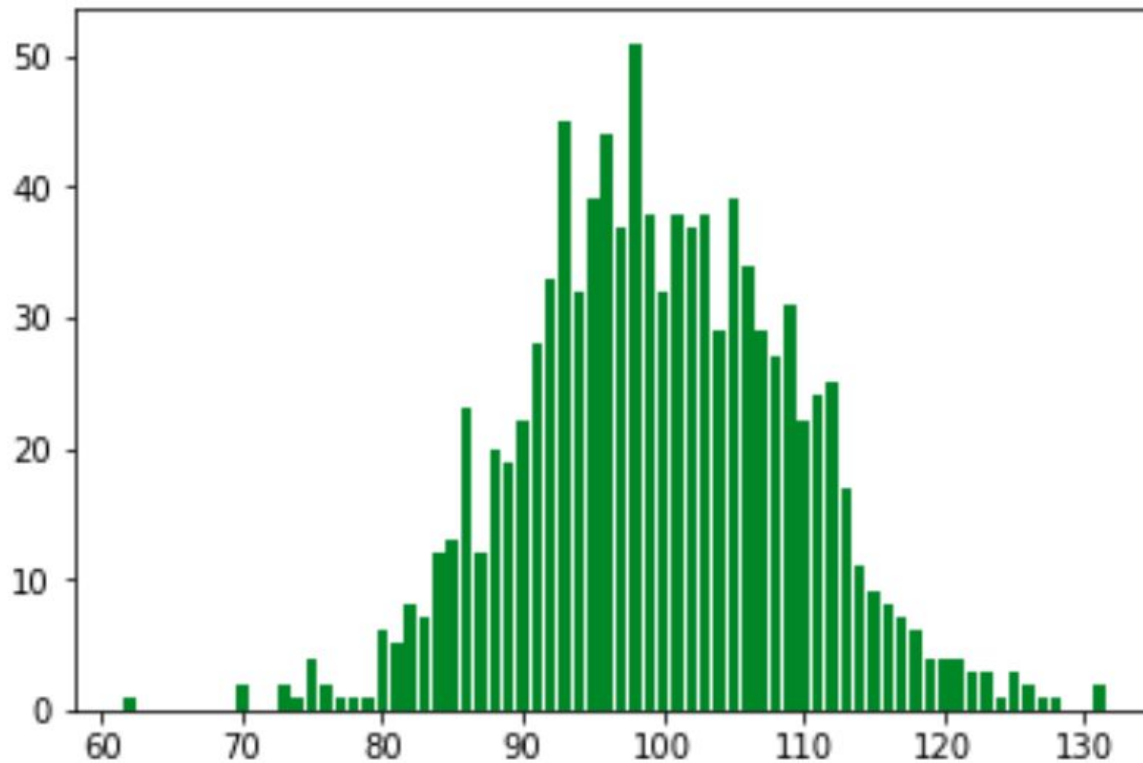
- Clustering coefficient = p

For a node with k neighbors:

$$\begin{aligned} \text{\#links between neighbors} / \text{\#max links between neighbors} &= \\ &= [p * (k(k-1)/2)] / [k(k-1)/2] = p \end{aligned}$$

Erdős–Rényi model

- Example – degree distribution for $G(1000, 0.1)$



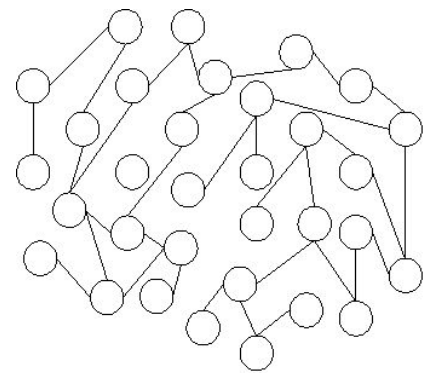
“Small-world” model

- Properties:
 - Small diameter (proportional to $\log N$)
 - High clustering coefficient
- A class of random graphs by Watts and Strogatz

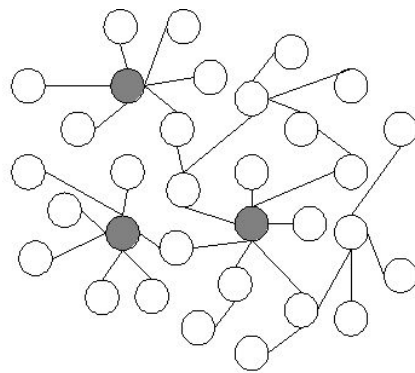
Scale-free networks

- A network whose degree distribution follows power law.

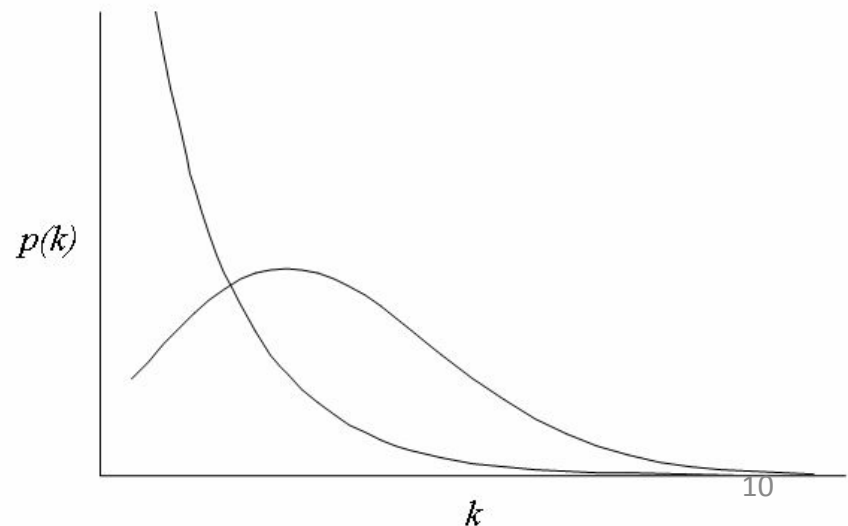
$$P(k) \sim k^{-\gamma}$$



(a) Random network

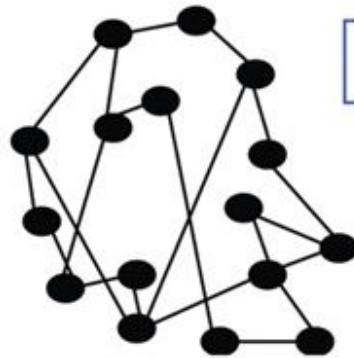


(b) Scale-free network

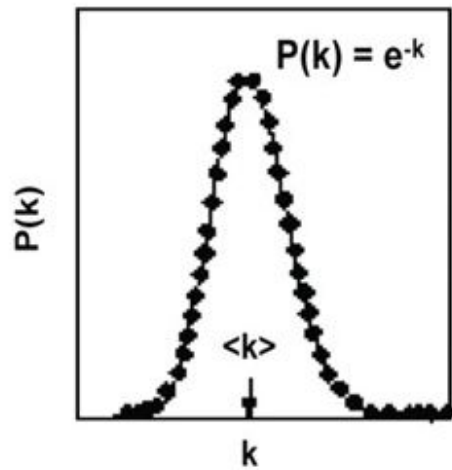


Random vs Scale-free networks

A Random Network

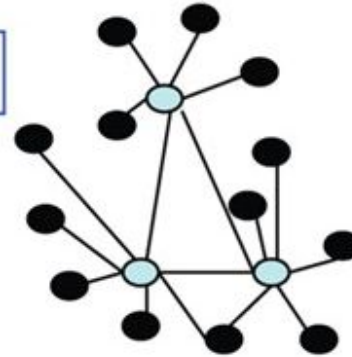


k =degree or #
nodal connections

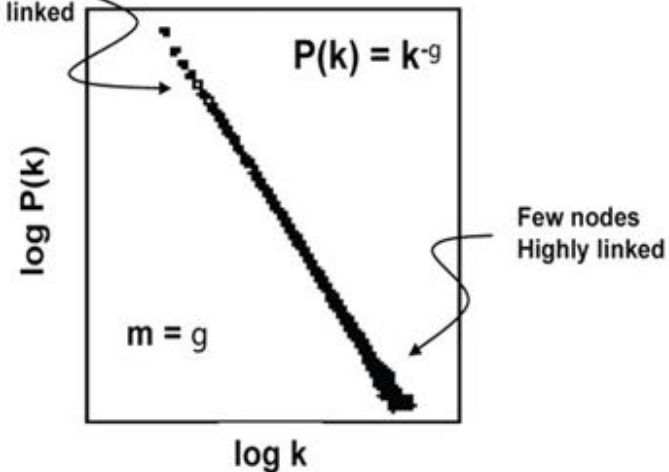


Poisson Distribution

Scale-free Network



Many nodes
Sparsely linked



Power Law Distribution

“Small-world” model

- Small-world examples:
 - Co-authors in the same domain
 - Colleagues
 - Classmates
- Non small-world examples:
 - “went-to-same-school” people

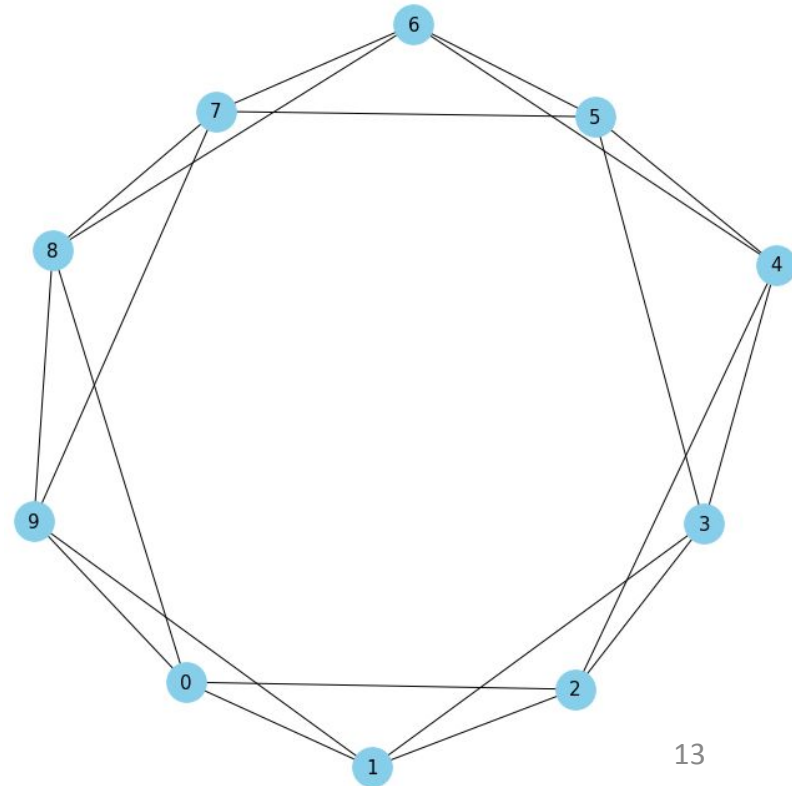
Watts-Strogatz model

- Input: **N** nodes, with average degree **K** and probability **p** of “recreating” the edge.

Step 1:

Create N nodes, connect each node to $K/2$ neighbors on the left and right (by IDs)

Result: High clustering coefficient, but also big diameter

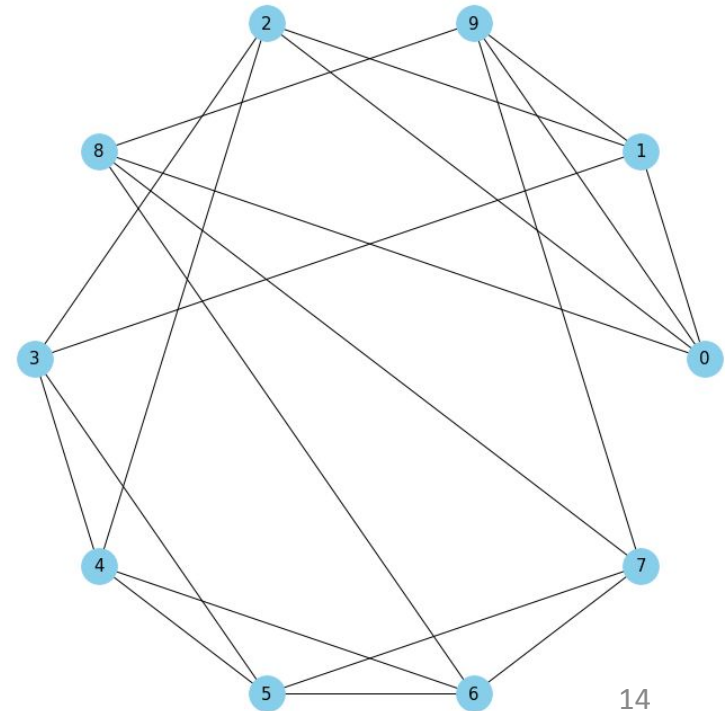


Watts-Strogatz model

Step 2:

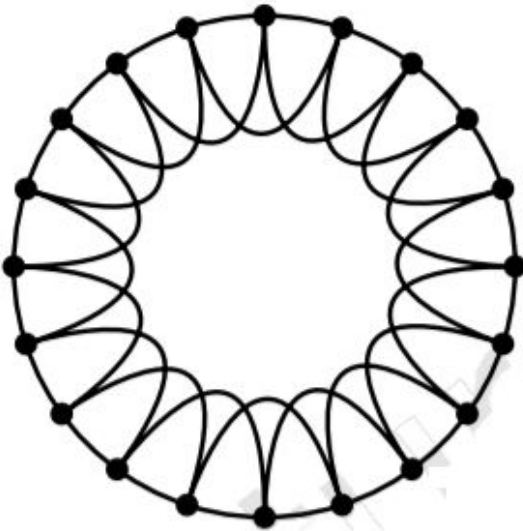
For each edge (i, j) , decide if it should be recreated with probability p

Result: High clustering coefficient,
and smaller diameter

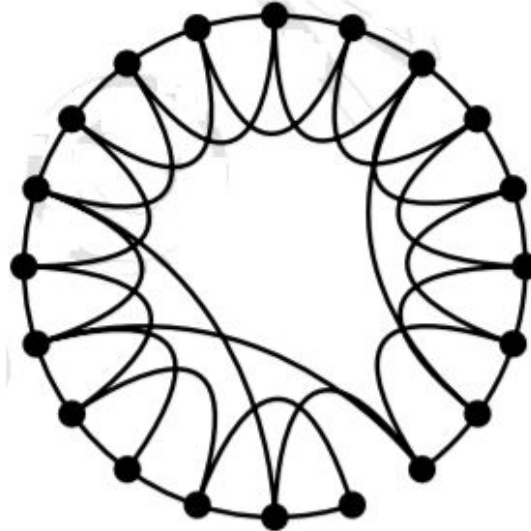


Watts-Strogatz model

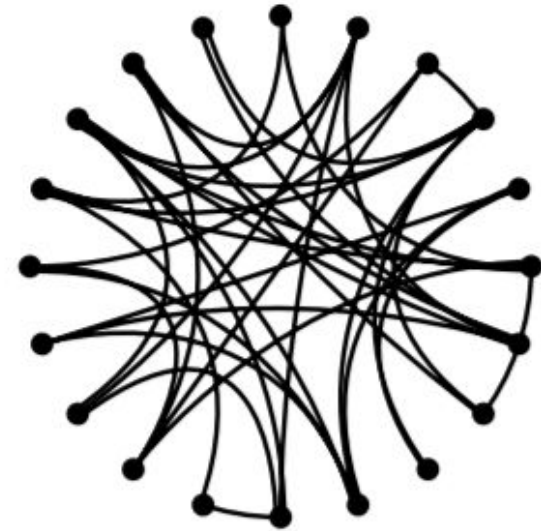
Regular



Small-world



Random



$p = 0$



$p = 1$

Increasing randomness

Summary

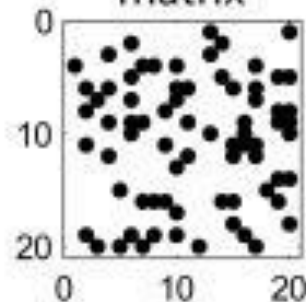
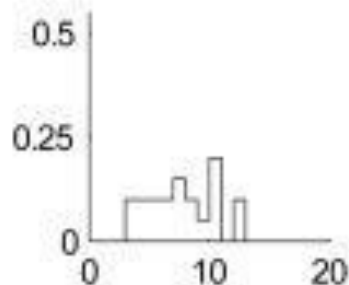
A.
structure

B.
degree distribution

C.
adjacency matrix

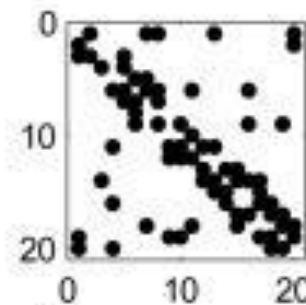
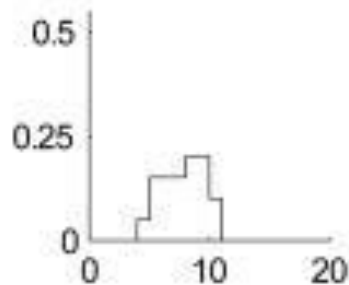
D.
Description

random



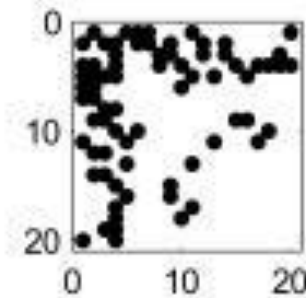
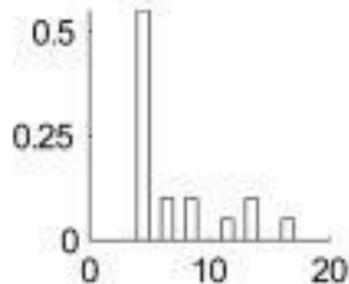
73 connections
among 20 nodes
assigned randomly

small-
world



High local clustering
and short average
path lengths.
'Hub-and-spoke'
architecture.

scale-
free



'Hub-and-spoke'
architecture is
maintained at
multiple spatial
scales.

Real World examples

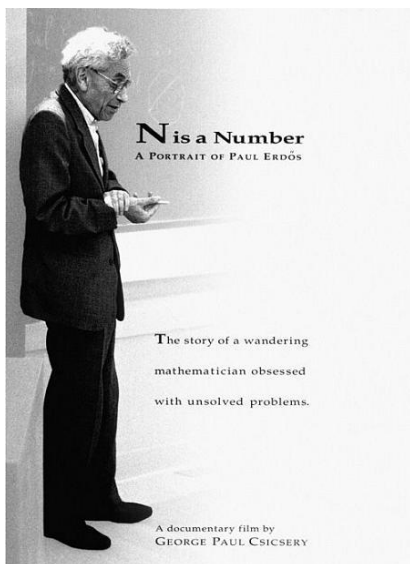
- Erdős number – collaboration distance to Erdős



- Kevin Bacon number:
 - Kevin Bacon himself has a Bacon number of 0.
 - Those actors who have worked directly with Kevin Bacon have a Bacon number of 1.
 - If the lowest Bacon number of any actor with whom X has appeared in any movie is N, X's Bacon number is N+1.

Erdős–Bacon number

- Paul Erdős has Erdős–Bacon number 3
 - Erdős number 0
 - Bacon number 3



Ronald Graham



Dave Johnson



Erdős–Bacon number

- Natalie Portman has Erdős–Bacon number 7
 - Erdős number 5
 - Bacon number 2



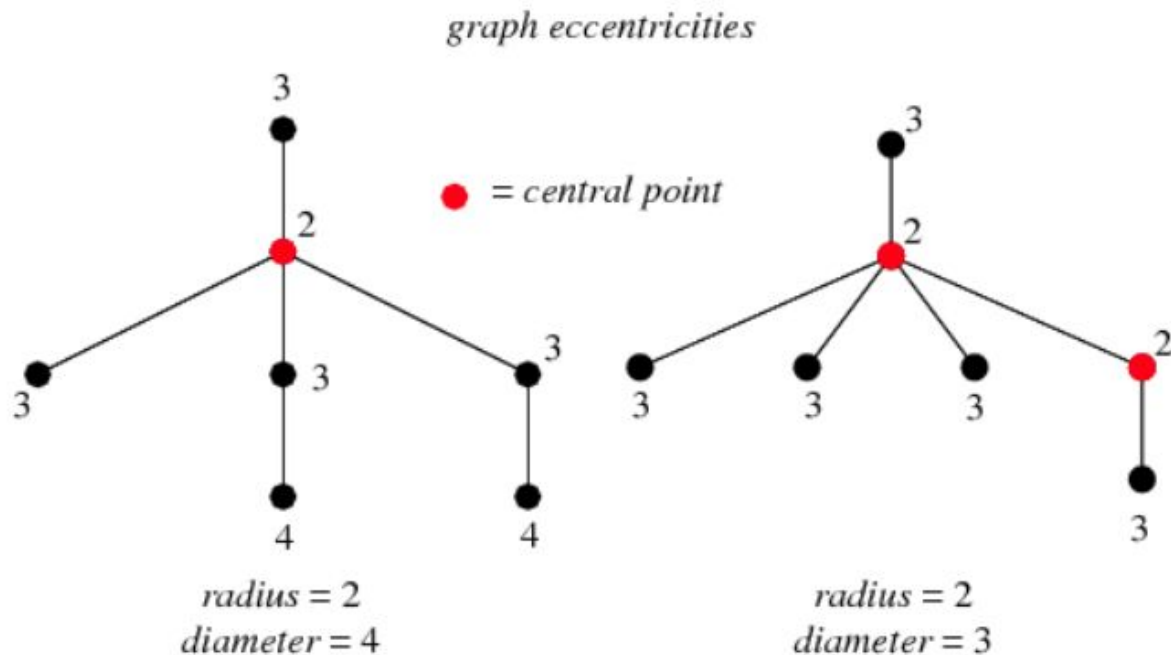
Centrality Measures

Centrality

- Identify the most important vertices in a graph
- Applications:
 - Most influential people
 - Key infrastructure nodes
 - Information spread points
- The measure we chose is often depends on the application

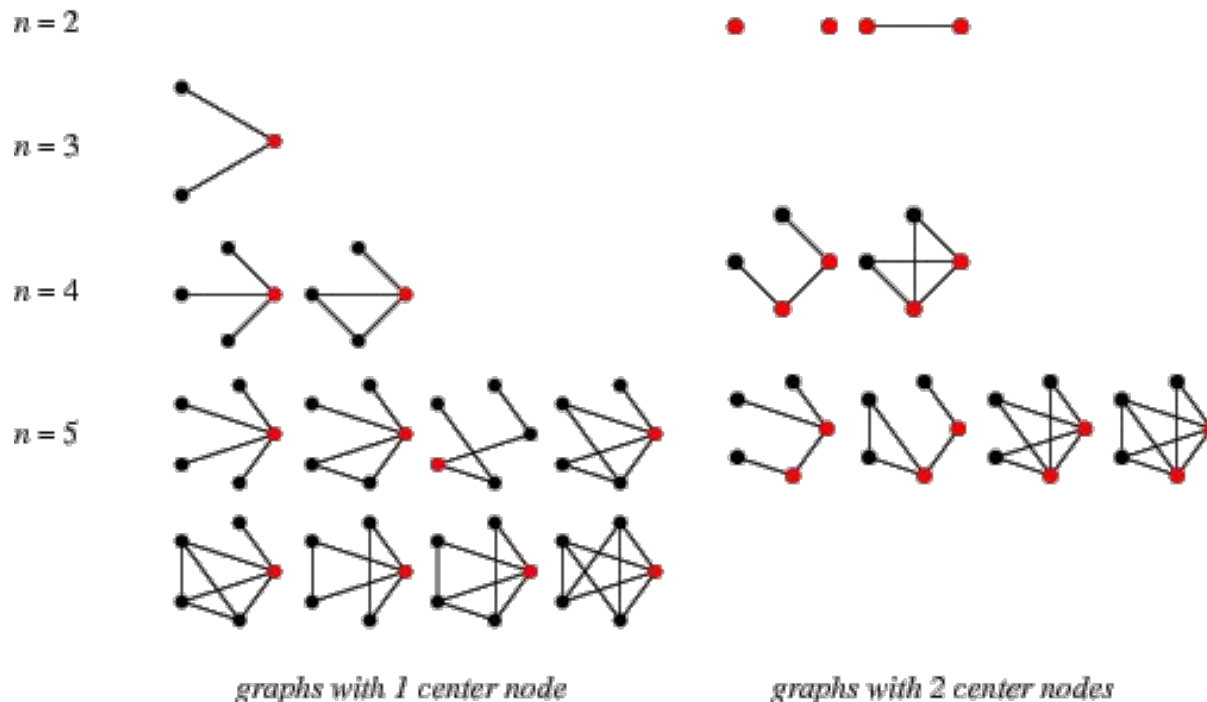
Preliminaries

- Eccentricity (of node v) – maximal distance between v and any other node.
- **Diameter** – *maximum* eccentricity in graph
- **Radius** – *minimum* eccentricity in graph



Preliminaries

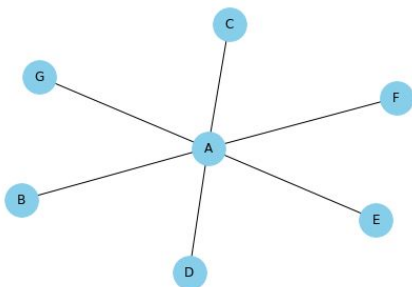
- **Central point** – node with eccentricity = radius
- **Graph center** – set of central points
- **Periphery** – set of nodes with eccentricity = diameter



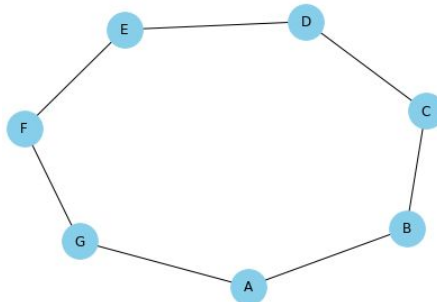
Types of Centrality

- There are many types of centrality measures:
 - Degree Centrality
 - Closeness Centrality
 - Betweenness Centrality
 - Eigenvector Centrality
- To demonstrate, we use 3 types of graphs:

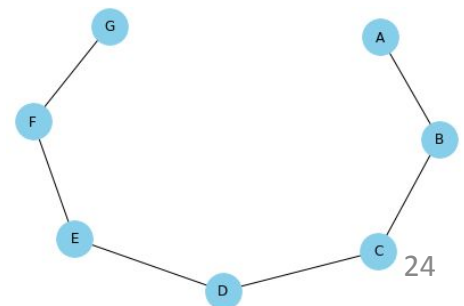
Star graph,



Circle graph,



Line graph



Things to measure

- Degree Centrality:
 - Connectedness
- Closeness Centrality:
 - Ease of reaching other nodes
- Betweenness Centrality:
 - Role as an intermediary, connector
- Eigenvector Centrality
 - “Whom you know...”

Degree Centrality

- How “connected” is a node?

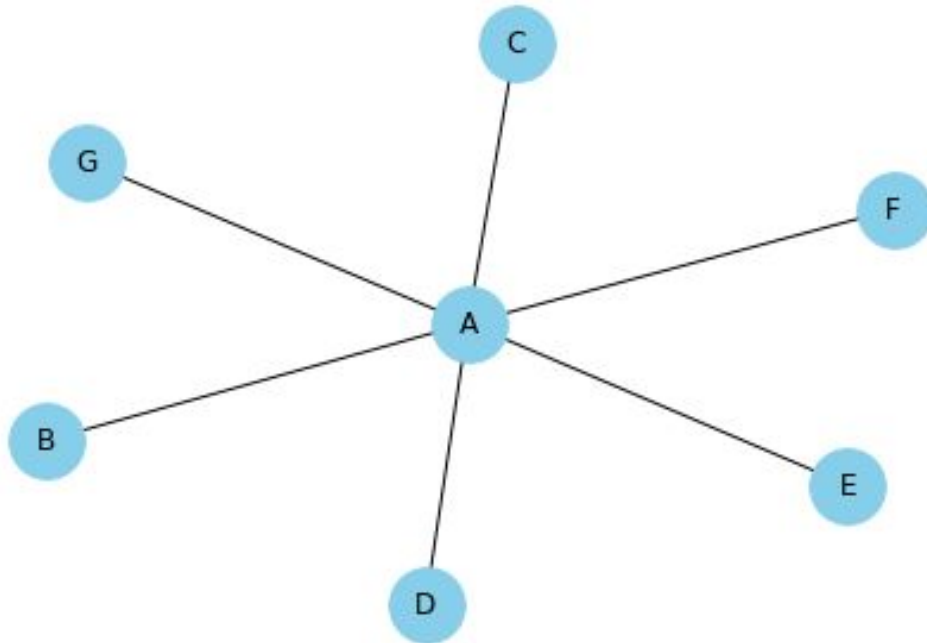
$$C_D(i) = k(i) = \sum_j A_{ij}$$

- Normalized: Divide by (n-1)

$$C_D^*(i) = \frac{1}{n-1} C_D(i)$$

- High centrality – direct contact with many others
- Low centrality – not active

Degree Centrality



$$C_D(A) = 6$$

$$C_D(B) = 1$$

$$C_D(C) = 1$$

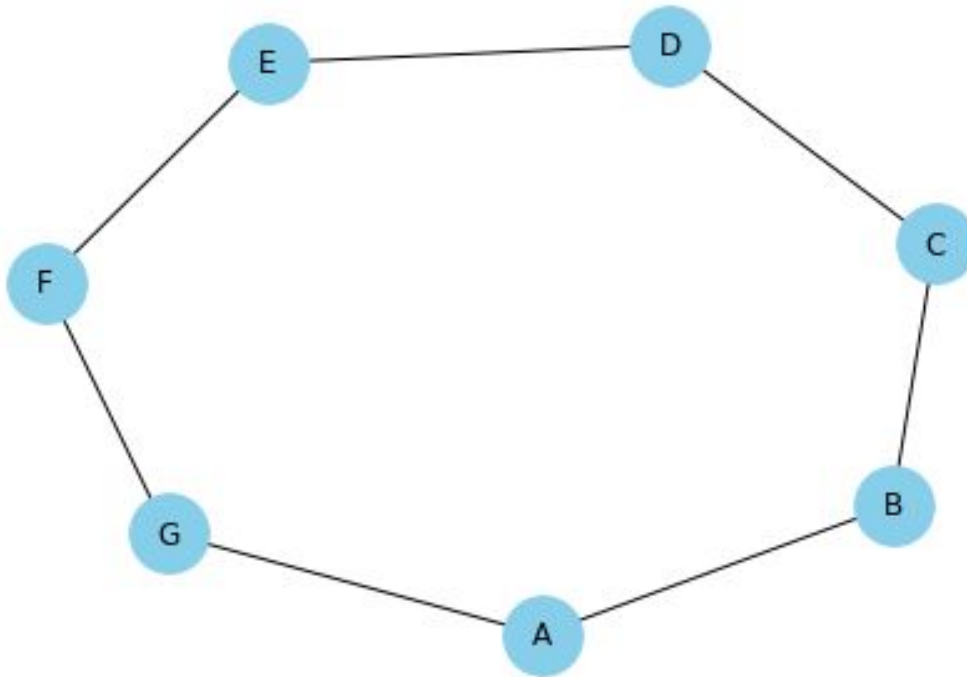
$$C_D(D) = 1$$

$$C_D(E) = 1$$

$$C_D(F) = 1$$

$$C_D(G) = 1$$

Degree Centrality



$$C_D(A) = 2$$

$$C_D(B) = 2$$

$$C_D(C) = 2$$

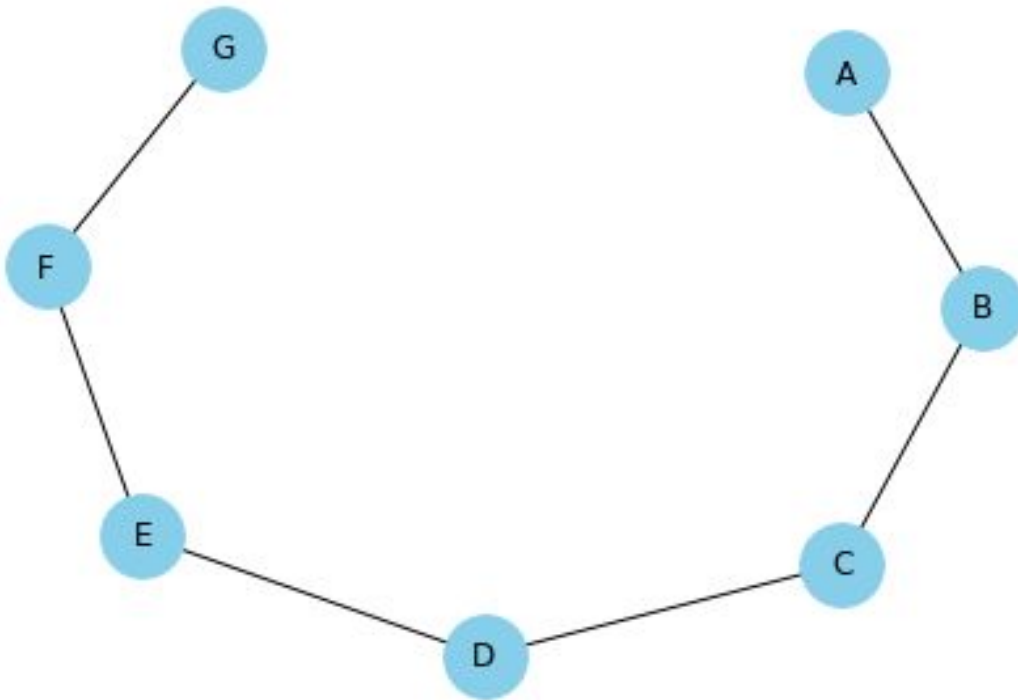
$$C_D(D) = 2$$

$$C_D(E) = 2$$

$$C_D(F) = 2$$

$$C_D(G) = 2$$

Degree Centrality



$$C_D(A) = 1$$

$$C_D(B) = 2$$

$$C_D(C) = 2$$

$$C_D(D) = 2$$

$$C_D(E) = 2$$

$$C_D(F) = 2$$

$$C_D(G) = 1$$

Closeness Centrality

- How close the node to other nodes in a graph

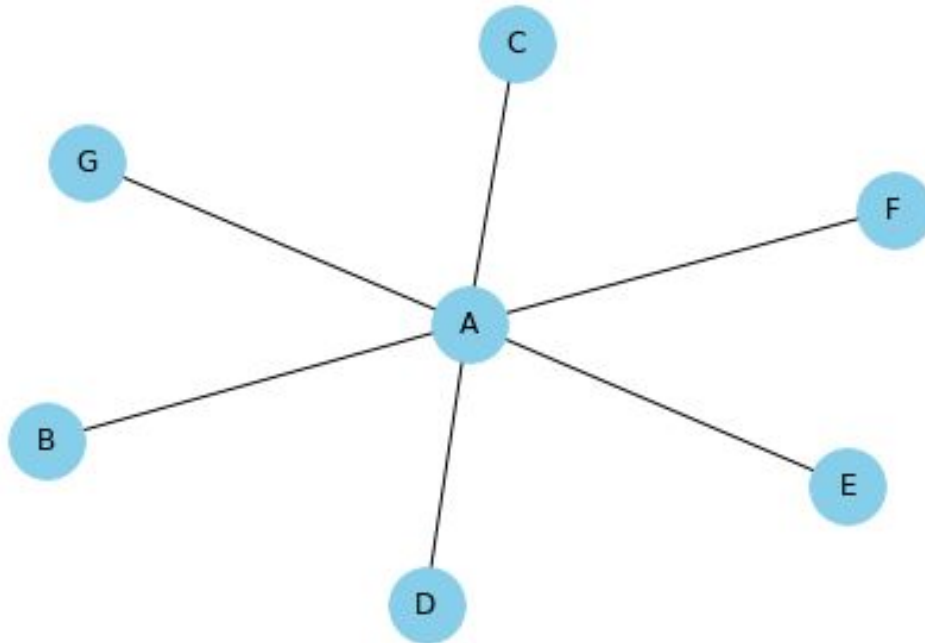
$$C_C(i) = \frac{1}{\sum_j d(i,j)}$$

- Normalized: Multiply by (n-1)

$$C_C^*(i) = (n - 1)C_C(i)$$

- High centrality – quick interaction with others, short communication path, low number of steps

Closeness Centrality



$$C_c(A) = 1/6$$

$$C_c(B) = 1/11$$

$$C_c(C) = 1/11$$

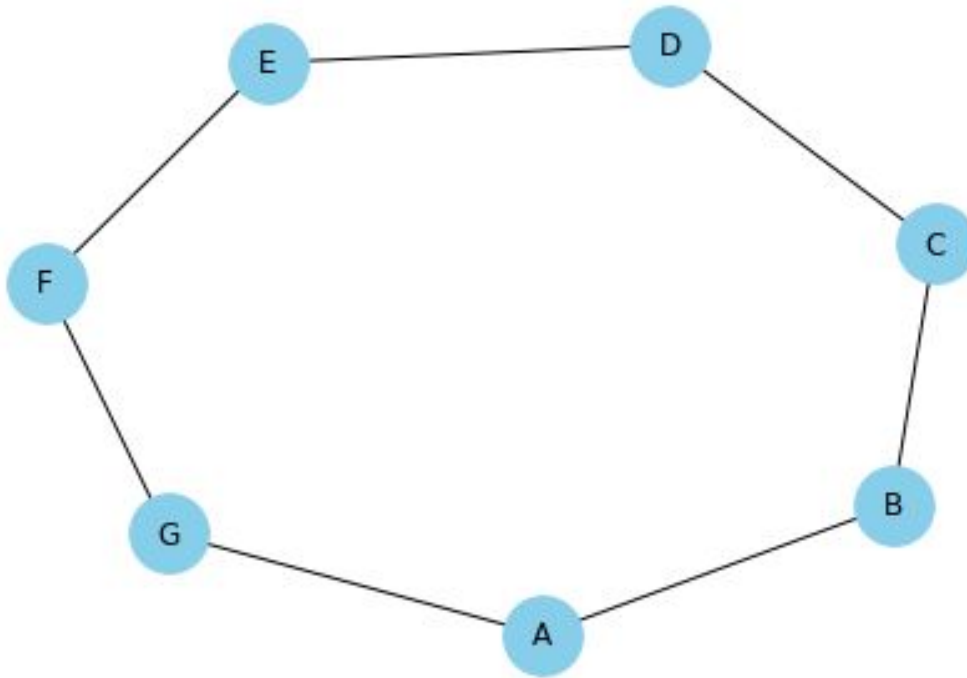
$$C_c(D) = 1/11$$

$$C_c(E) = 1/11$$

$$C_c(F) = 1/11$$

$$C_c(G) = 1/11$$

Closeness Centrality



$$C_c(A) = 1/12$$

$$C_c(B) = 1/12$$

$$C_c(C) = 1/12$$

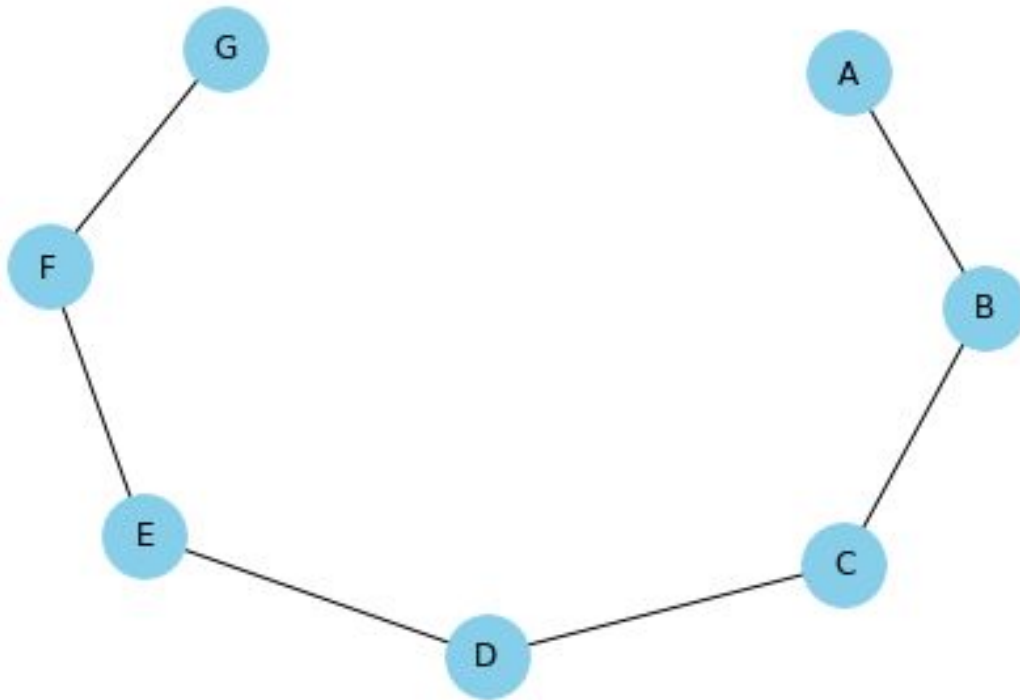
$$C_c(D) = 1/12$$

$$C_c(E) = 1/12$$

$$C_c(F) = 1/12$$

$$C_c(G) = 1/12$$

Closeness Centrality



$$C_c(A) = 1/21$$

$$C_c(B) = 1/16$$

$$C_c(C) = 1/13$$

$$C_c(D) = 1/12$$

$$C_c(E) = 1/13$$

$$C_c(F) = 1/16$$

$$C_c(G) = 1/21$$

Betweenness Centrality

- Number of shortest pathes going through **i**

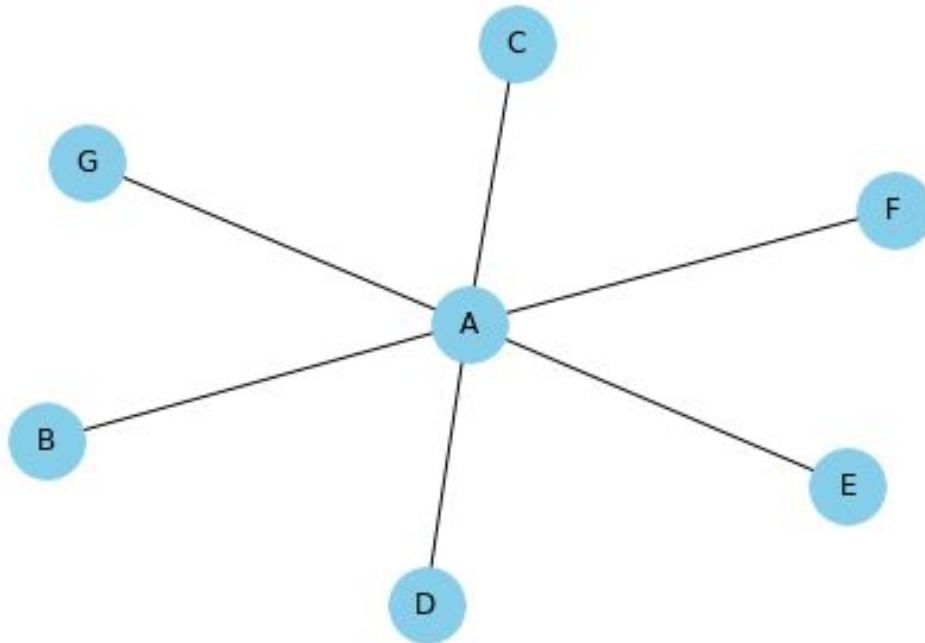
$$C_B(i) = \sum_{s \neq t \neq i} \frac{\sigma_{st}(i)}{\sigma_{st}}$$

- Normalized: Divide by $(n-1)(n-2)/2$

$$C_B^*(i) = \frac{2}{(n-1)(n-2)} C_B(i)$$

- High centrality – probability of communication between **s** and **t** going through **i**

Betweenness Centrality



$$C_B(A) = 15$$

$$C_B(B) = 0$$

$$C_B(C) = 0$$

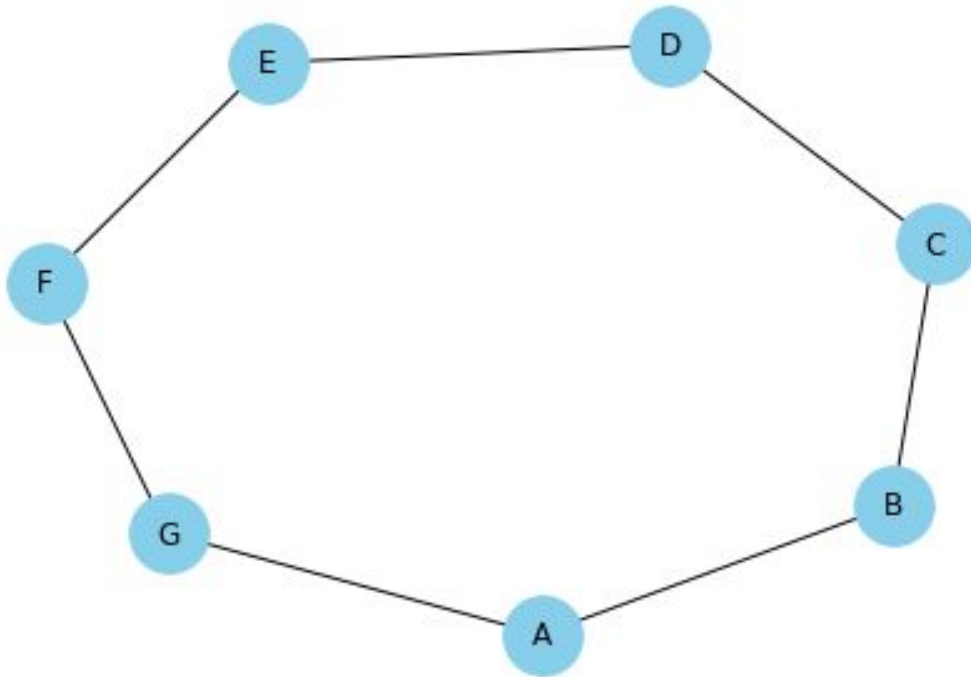
$$C_B(D) = 0$$

$$C_B(E) = 0$$

$$C_B(F) = 0$$

$$C_B(G) = 0$$

Betweenness Centrality



$$C_B(A) = 3$$

$$C_B(B) = 3$$

$$C_B(C) = 3$$

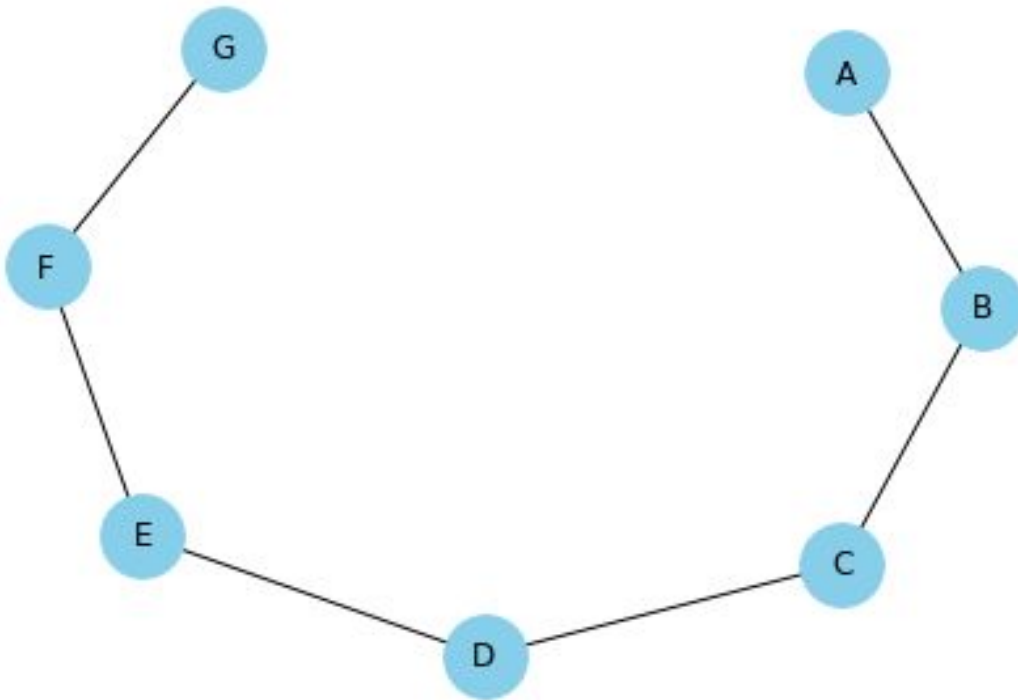
$$C_B(D) = 3$$

$$C_B(E) = 3$$

$$C_B(F) = 3$$

$$C_B(G) = 3$$

Betweenness Centrality



$$C_B(A) = 0$$

$$C_B(B) = 5$$

$$C_B(C) = 8$$

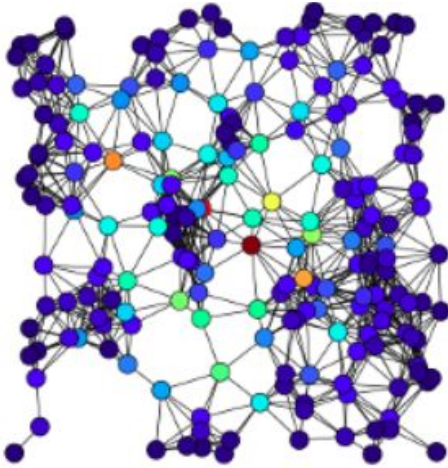
$$C_B(D) = 9$$

$$C_B(E) = 8$$

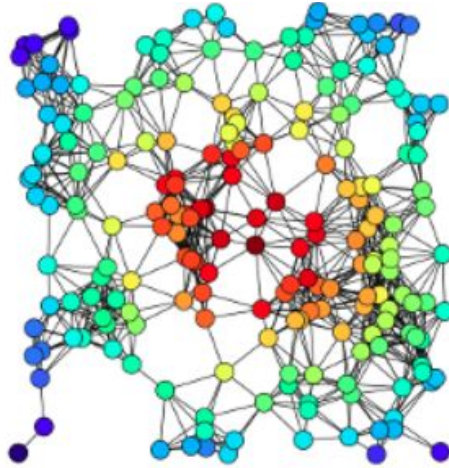
$$C_B(F) = 5$$

$$C_B(G) = 0$$

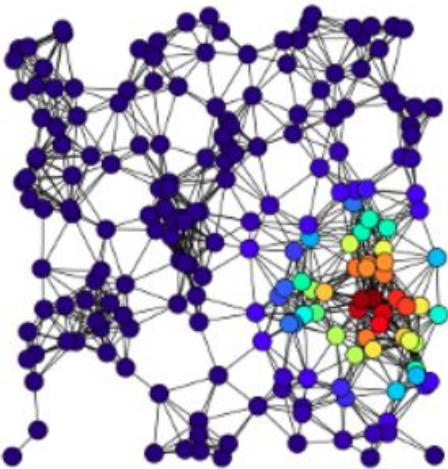
Centralities



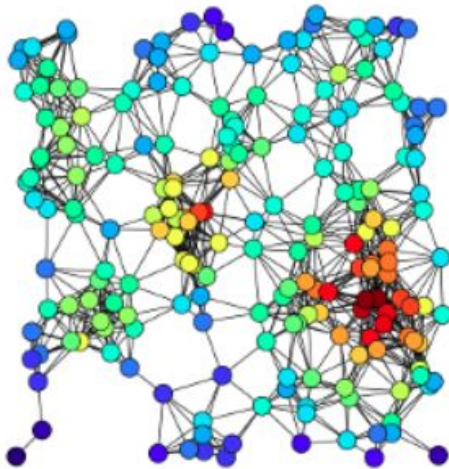
A



B



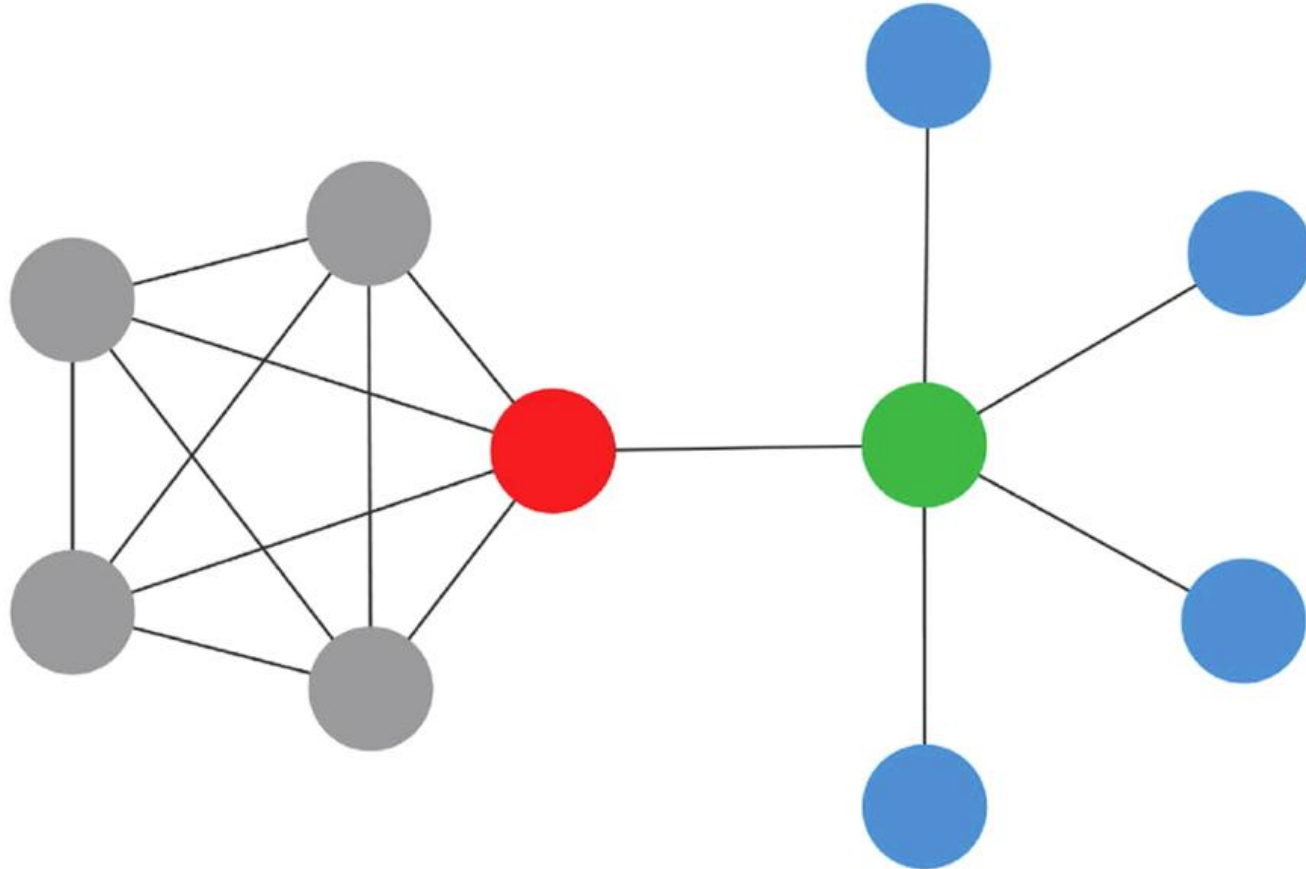
C



D

- A) Betweenness
- B) Closeness
- C) Eigenvector
- D) Degree

HW question example



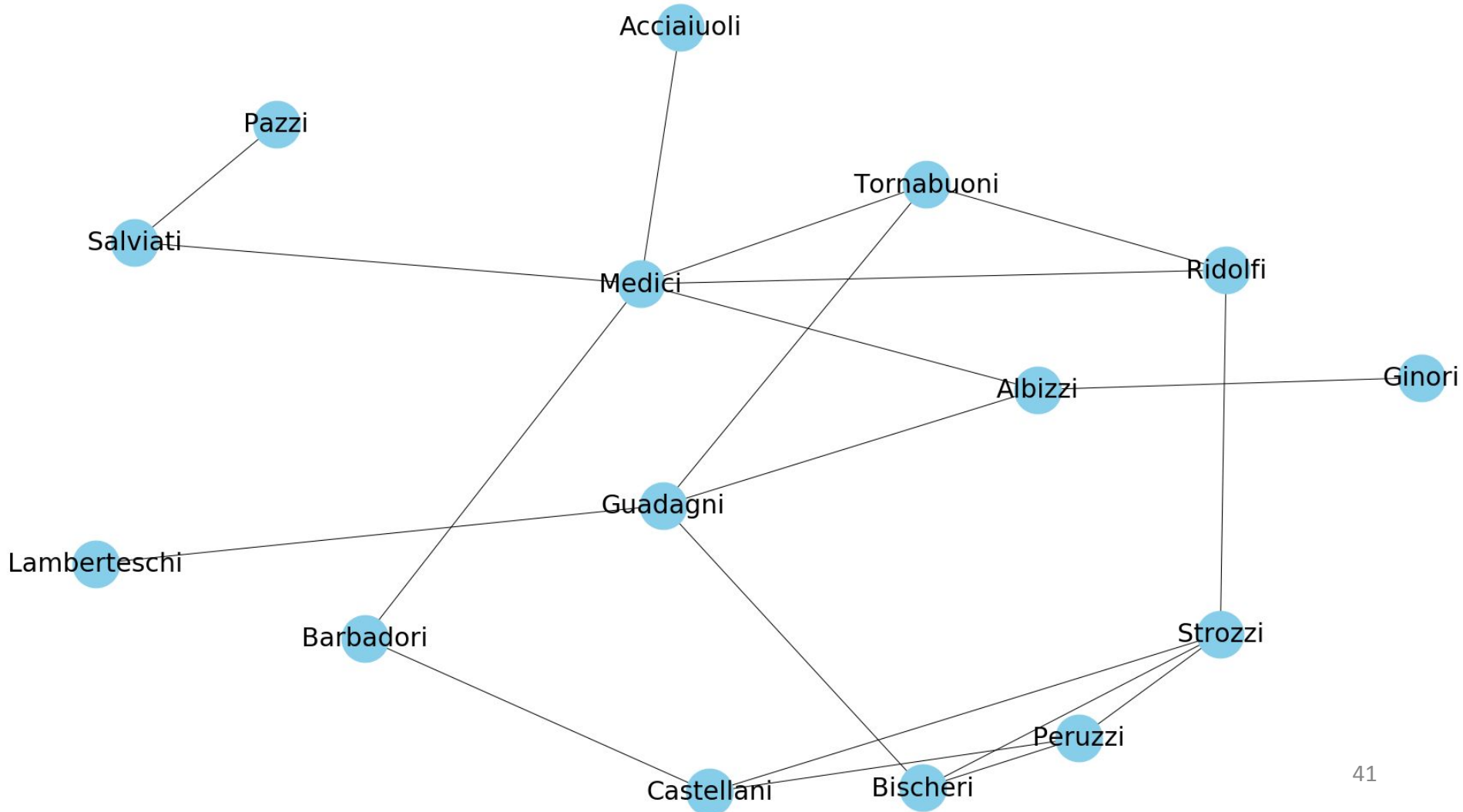
Compute and explain 3 types of centrality

Families of Florence

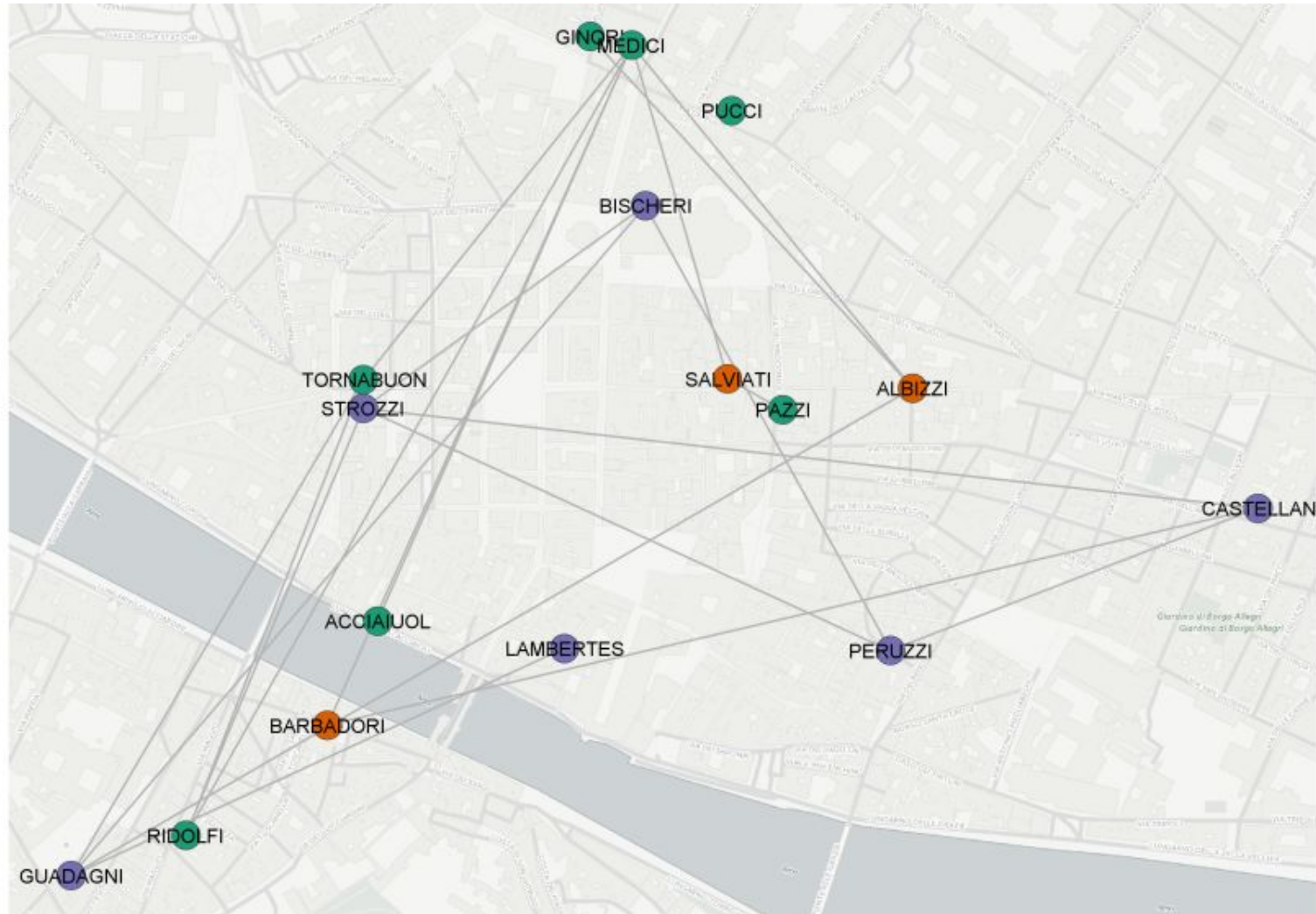
- Marriage and relationships of 16 families in Florence in middle ages
- Very interesting, “classic” network to analyze
- The rise of Medici family

https://www2.bc.edu/candace-jones/mb851/Mar12/PadgettAnsell_AJS_1993.pdf

Families of Florence



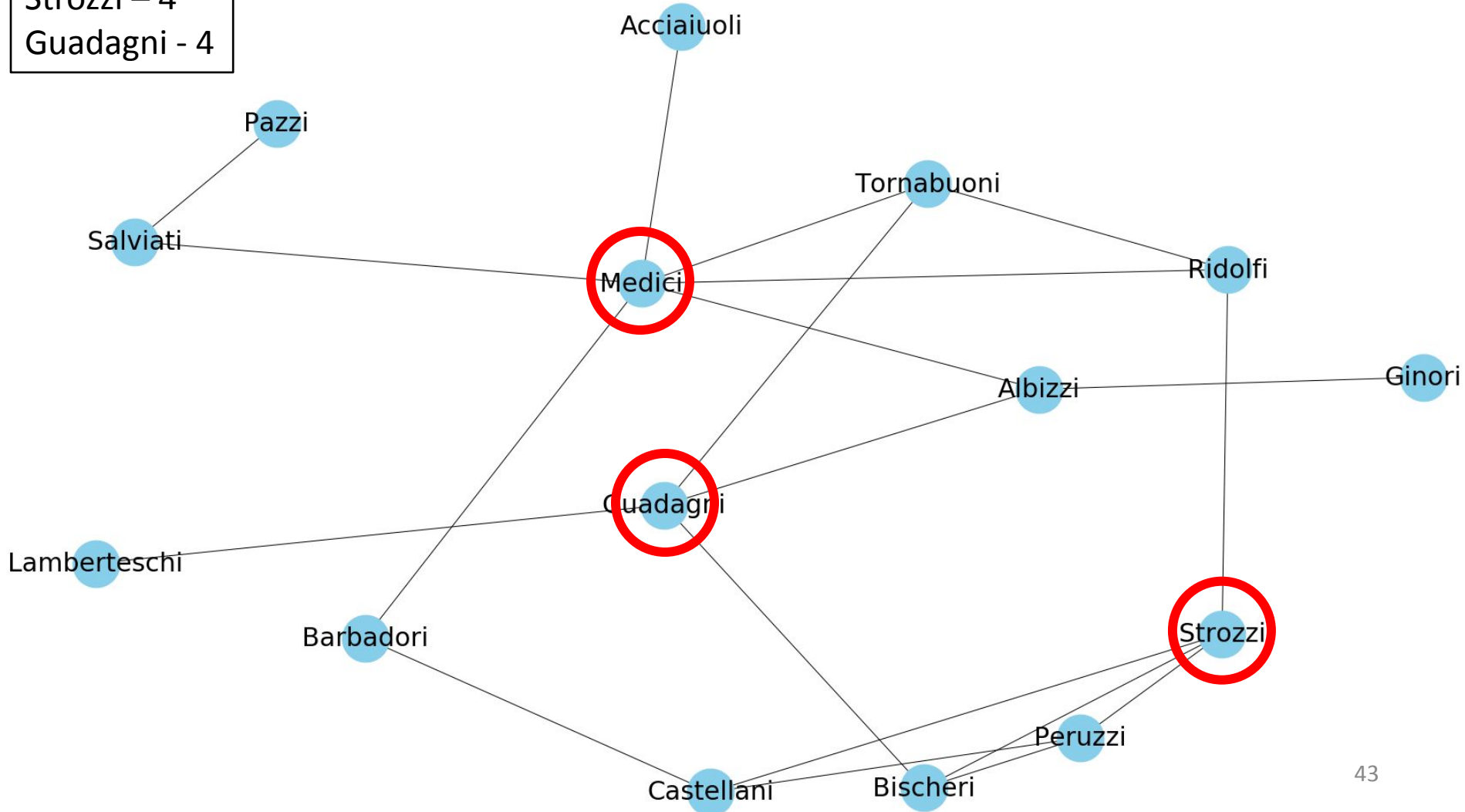
Families of Florence



From: <https://simonfink.wordpress.com/2016/05/11/the-medici-marriage-network-in-florence/>

Degree Centrality

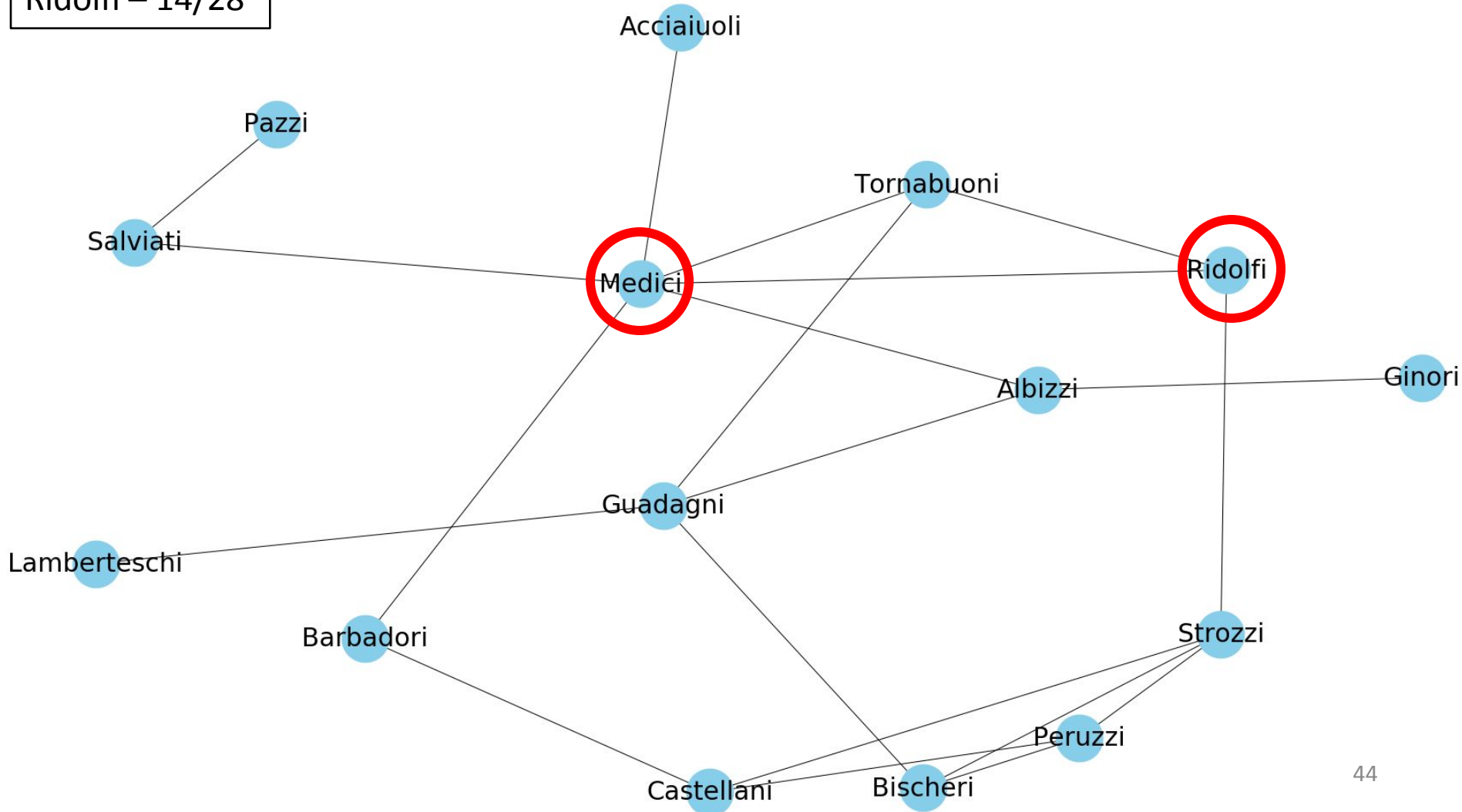
Medici – 6
Strozzi – 4
Guadagni - 4



Closeness Centrality

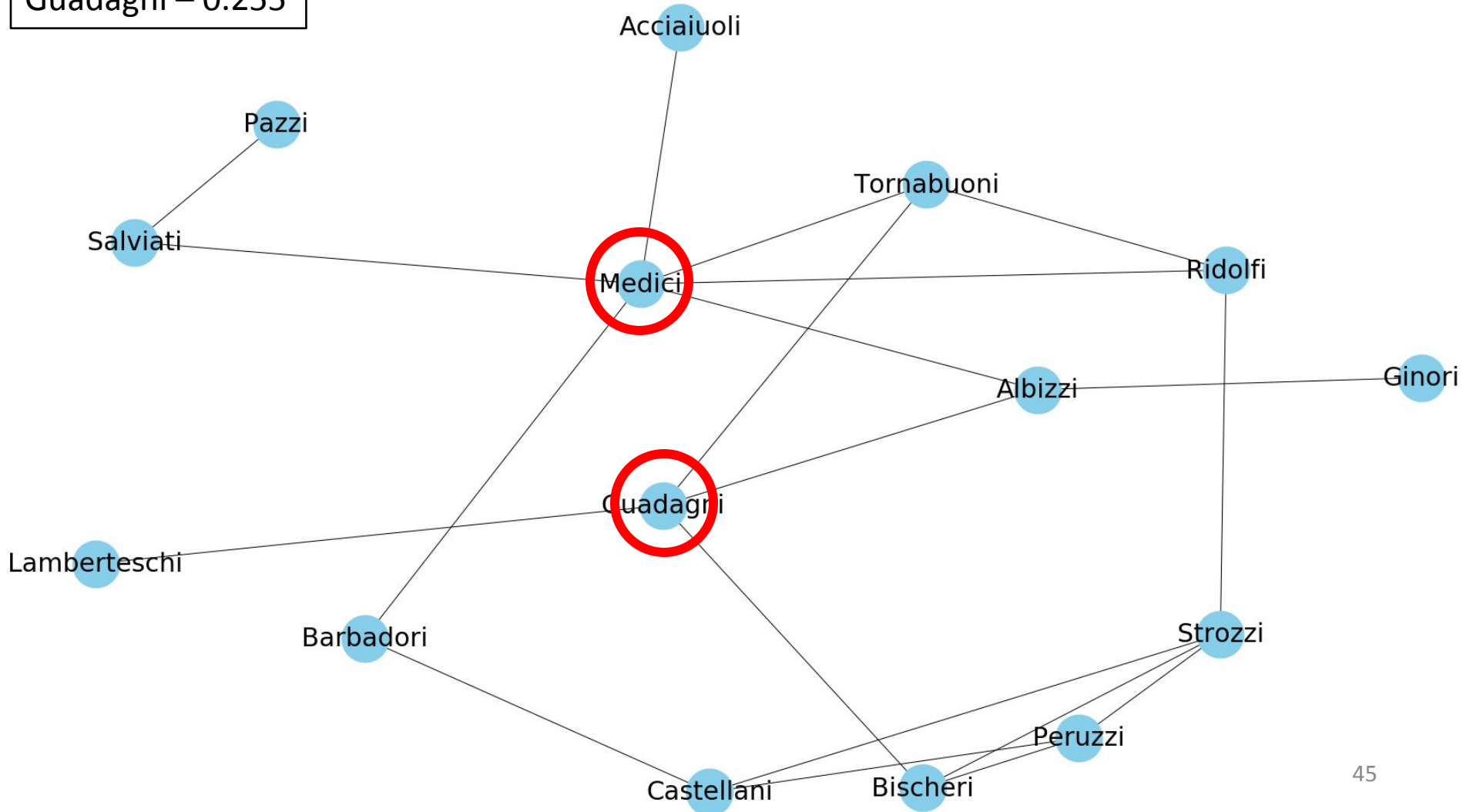
Medici – 14/25

Ridolfi – 14/28

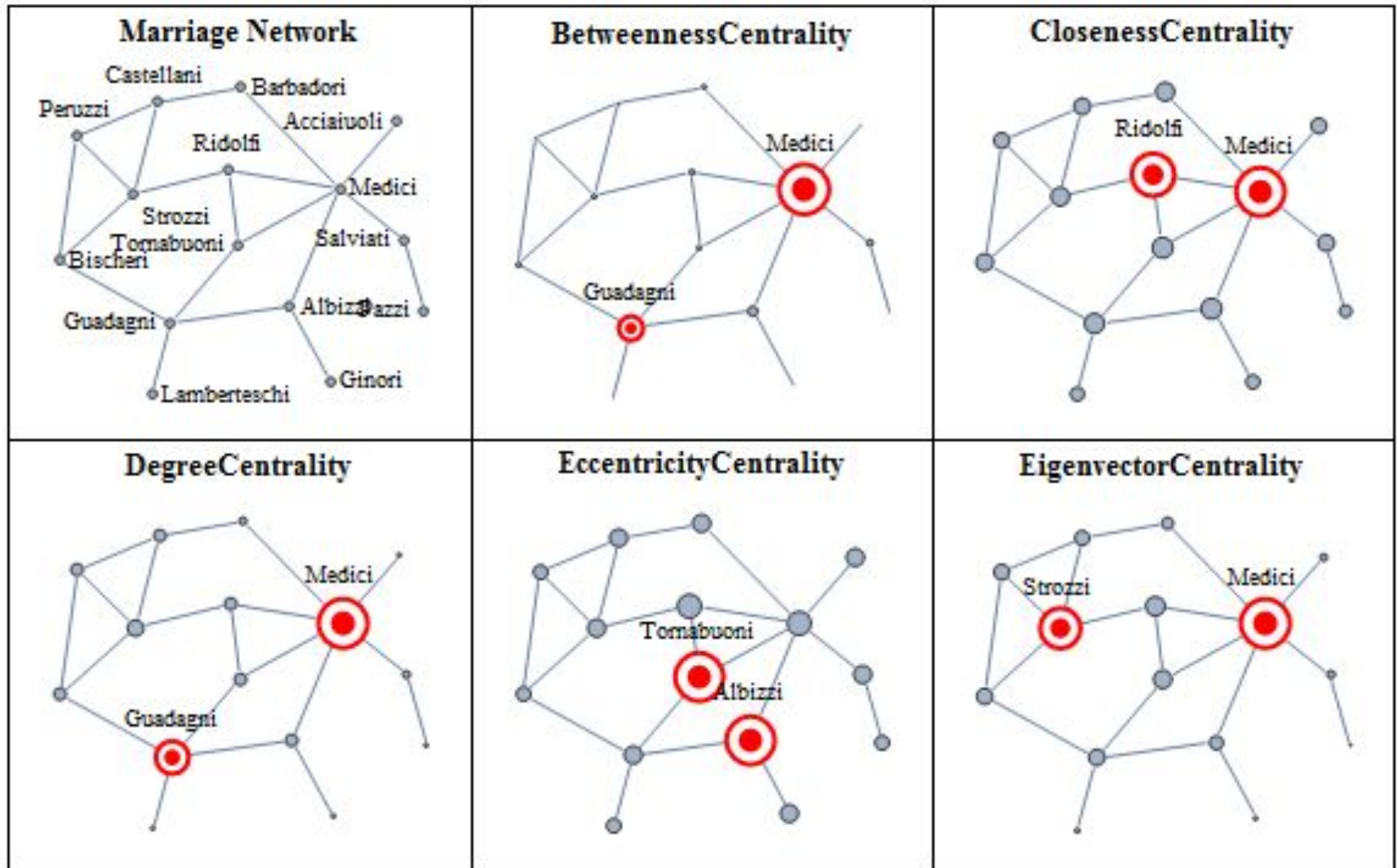


Betweenness Centrality

Medici – 0.522
Guadagni – 0.255

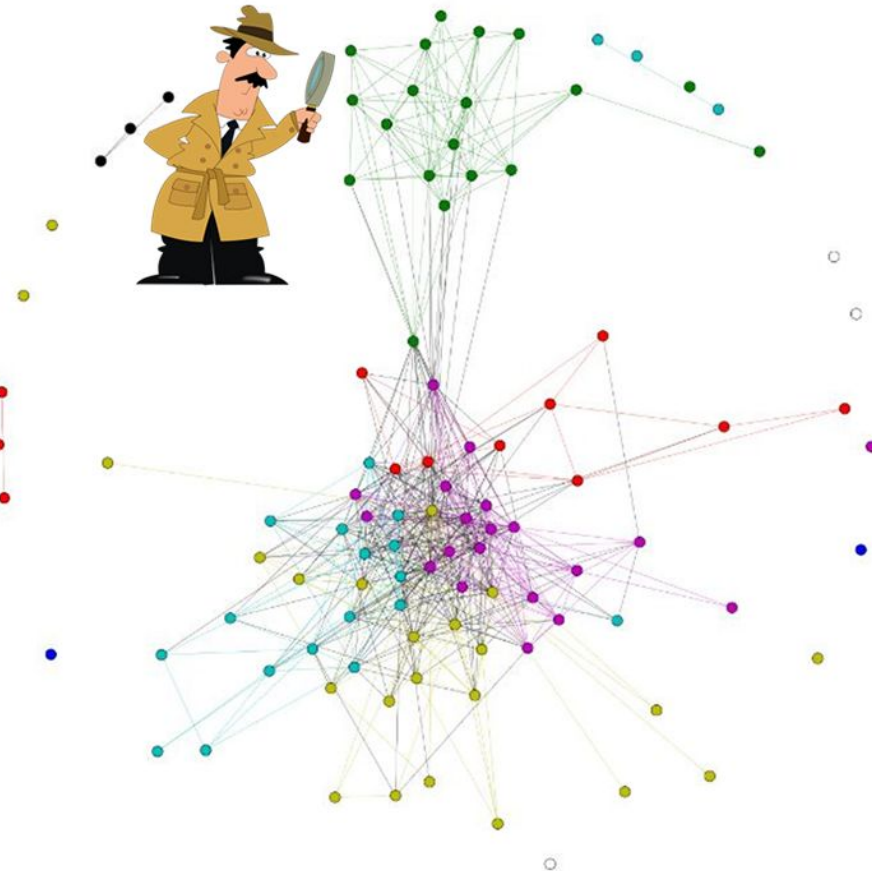


Families of Florence



Another usecase

- Sister found her “lost” brother by analyzing his (online)social network connections



Another usecase


- Sister found her “lost” brother by analyzing his (online)social network connections

| Degree centrality | Betweenness centrality | Closeness centrality |
|-------------------|------------------------|----------------------|
| A | A | A |
| B | E | E |
| C | F | B |
| D | G | C |
| E | B | D |

Another usecase

Happy End – brother was found through the connection E!

| Degree centrality | Betweenness centrality | Closeness centrality |
|-------------------|------------------------|----------------------|
| A | A | A |
| B | E | E |
| C | F | B |
| D | G | C |
| E | B | D |



Thank you!
Questions?