

DESCRIPTIVE NETWORK ANALYSIS

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BASIC CONCEPTS

STRUCTURE OF NETWORKS

Lecture 1

- A network is a collection of objects where some pairs of objects are connected by links

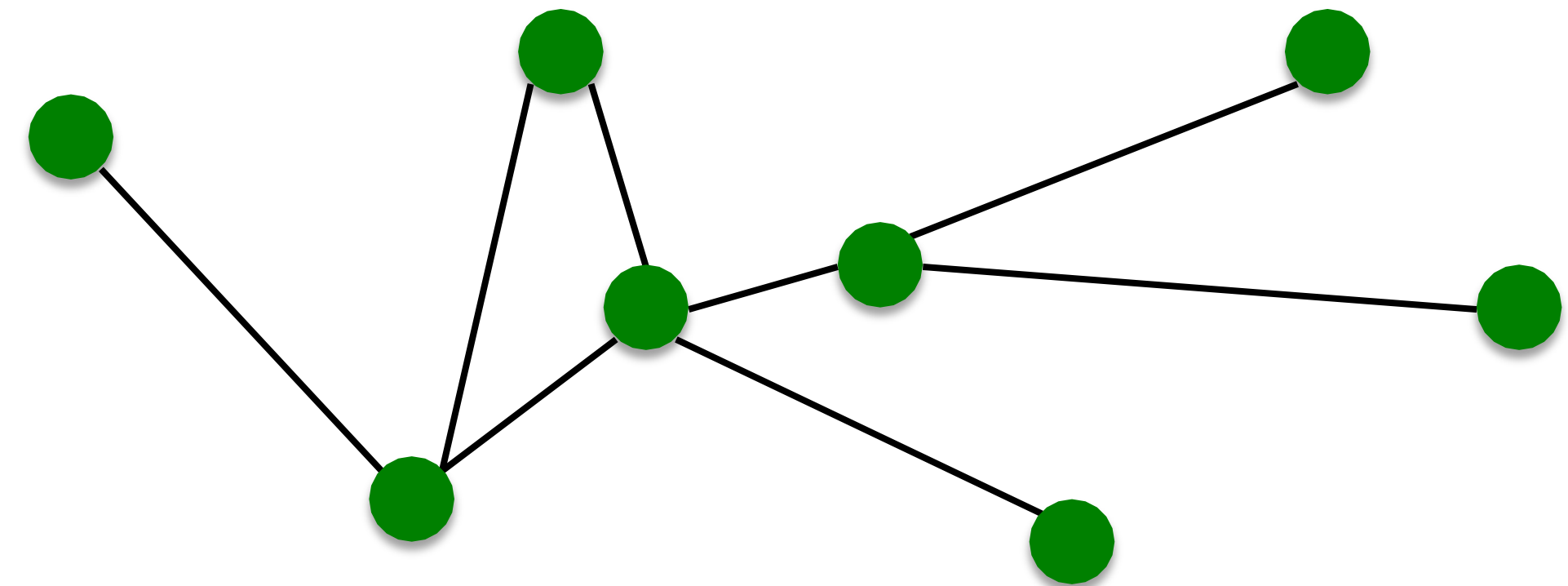


BASIC CONCEPTS

COMPONENTS OF A NETWORK

Lecture 1

- Objects: nodes, vertices N
- Interactions: links, edges E
- System: network, graph $G(N, E)$



A **graph** $G = (V, E)$ is an ordered pair of sets: a set of vertices V and a set edges E , where $n = |V|$, $m = |E|$

An **edge** $e_{ij} = (v_i, v_j)$ is pair of vertices (ordered pair for directed graph)

BASIC CONCEPTS

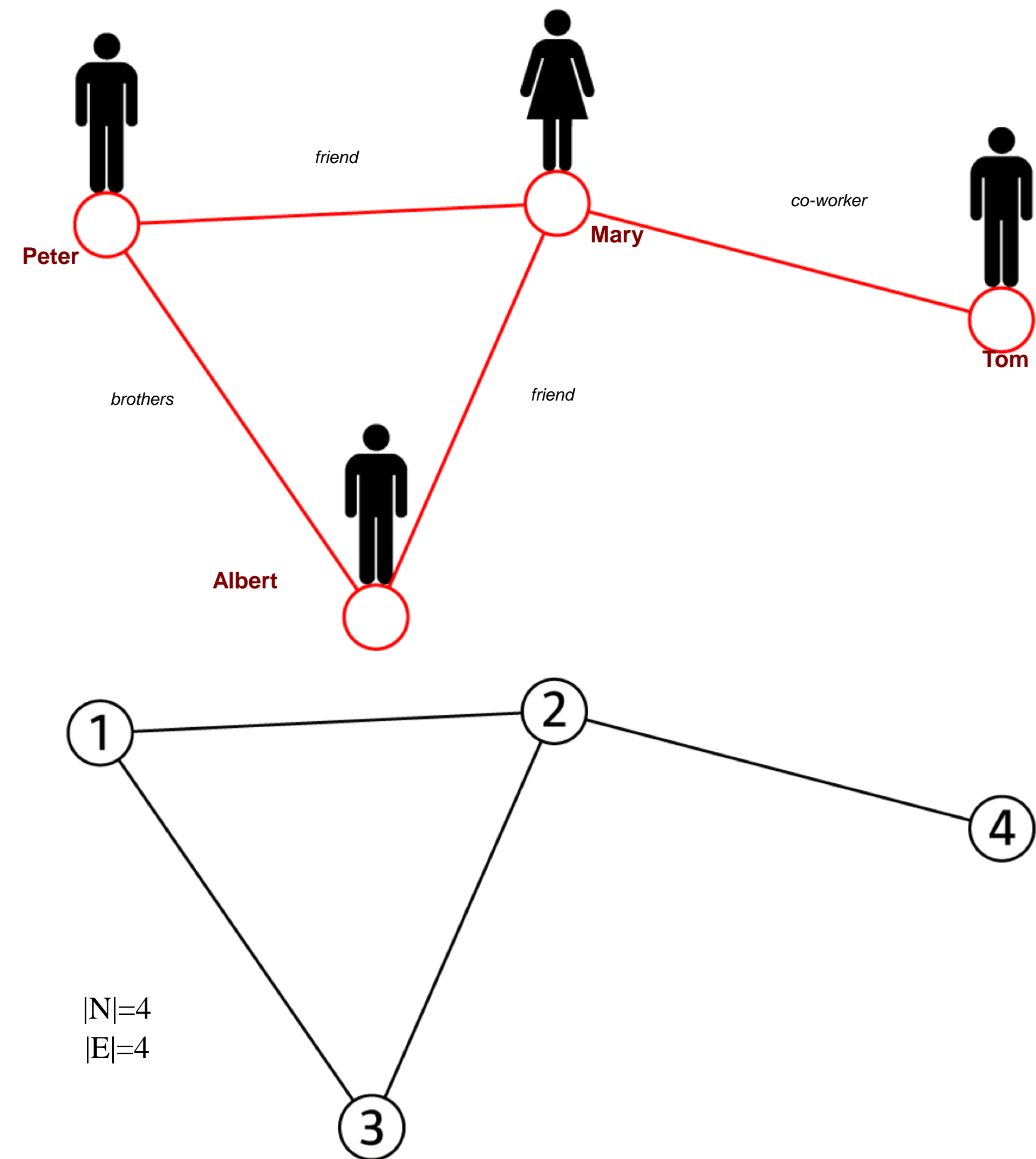
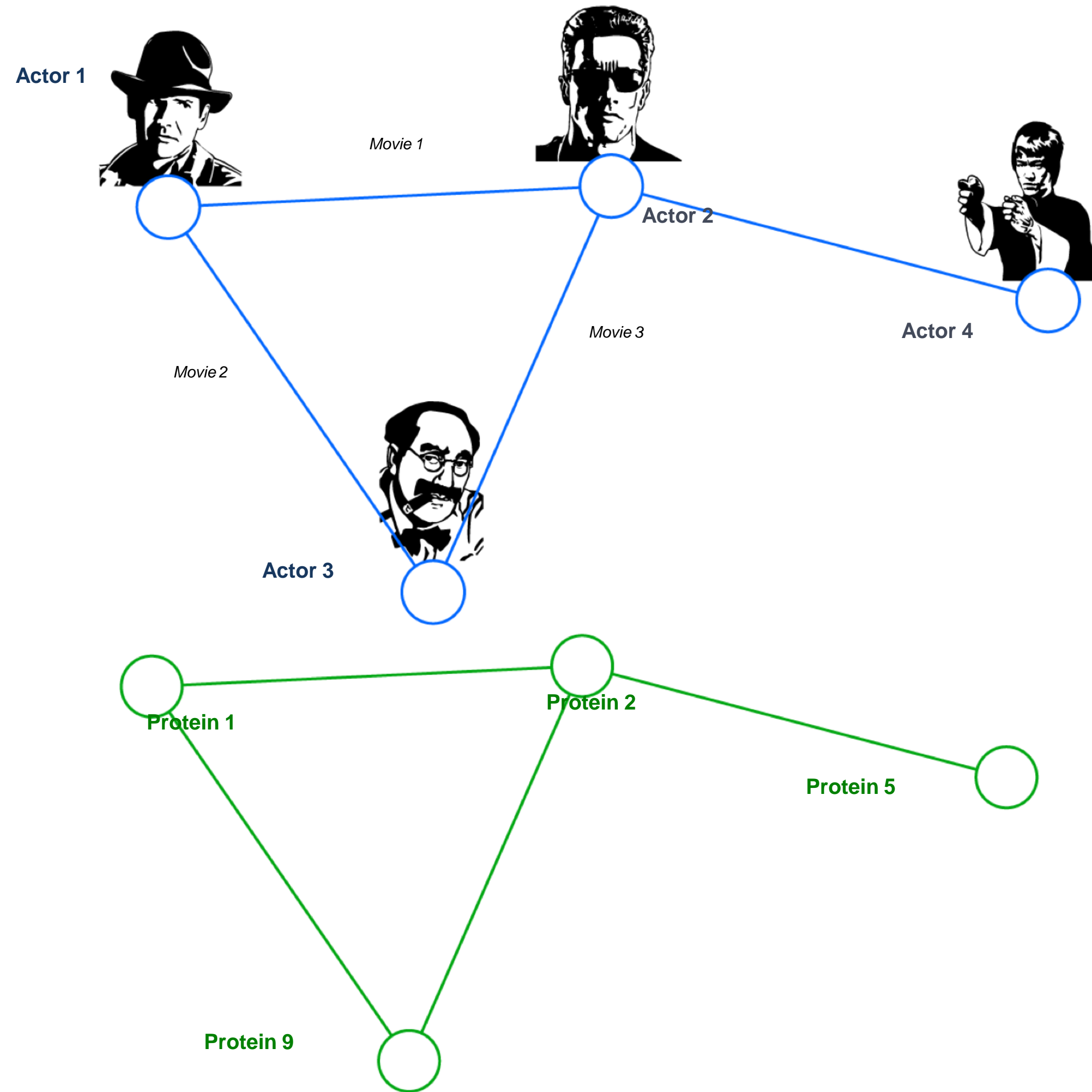
COMPONENTS OF A NETWORK

- Network often refers to real systems
Web, Social network, Metabolic network
Language: Network, node, link
- Graph is a mathematical representation of a network
Web graph, Social graph (a Facebook term)
Language: Graph, vertex, edge

BASIC CONCEPTS

COMPONENTS OF A NETWORK

Lecture 1



BASIC CONCEPTS

HOW TO DEFINE A NETWORK

Lecture 1

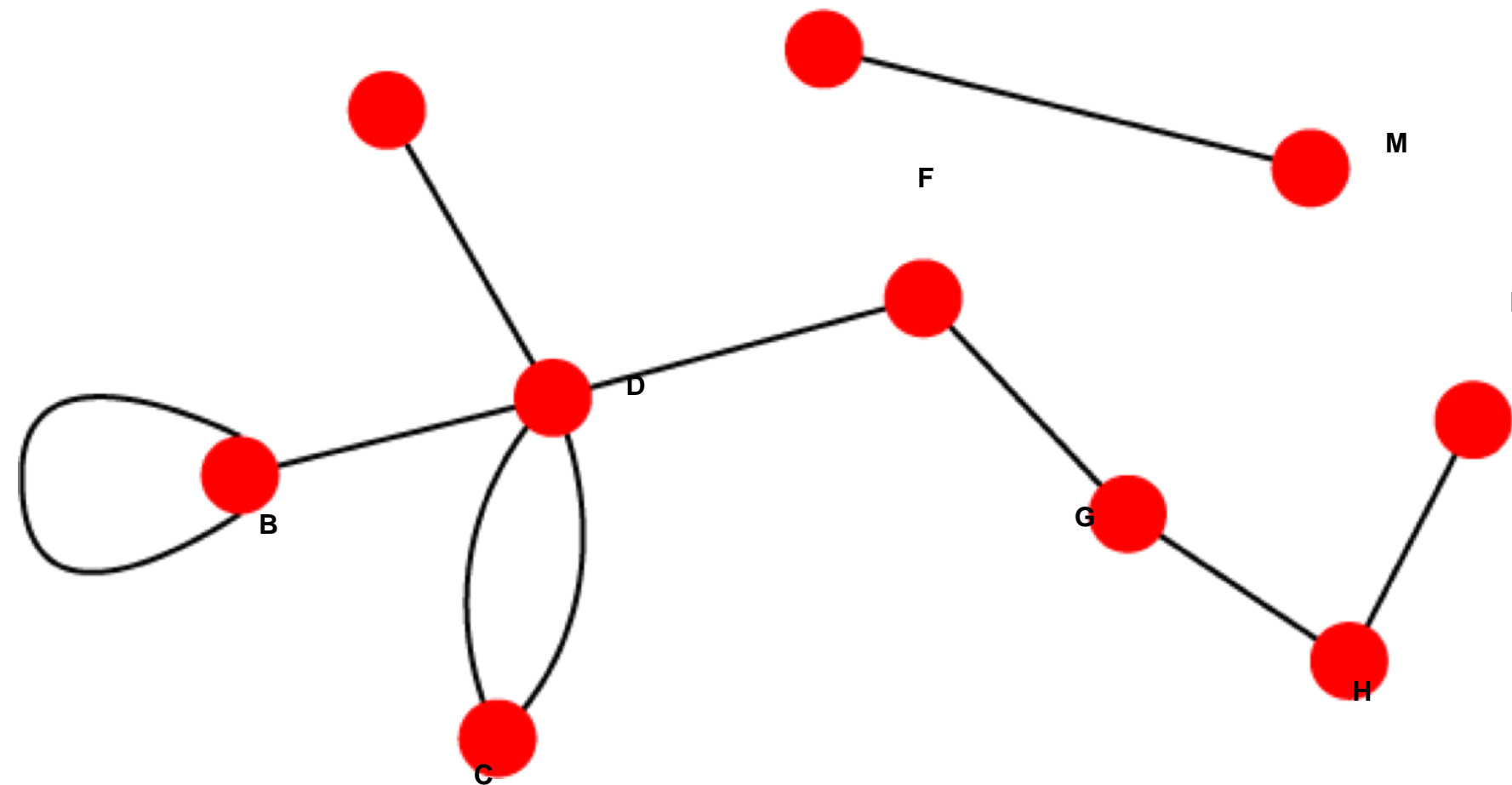
- How to build a graph:
 - What are nodes?
 - What are edges?
- Choice of the proper network representation of a given domain/problem determines our ability to use networks successfully:
 - In some cases there is a unique, unambiguous representation
 - In other cases, the representation is by no means unique
 - The way you assign links will determine the nature of the question you can study

BASIC CONCEPTS

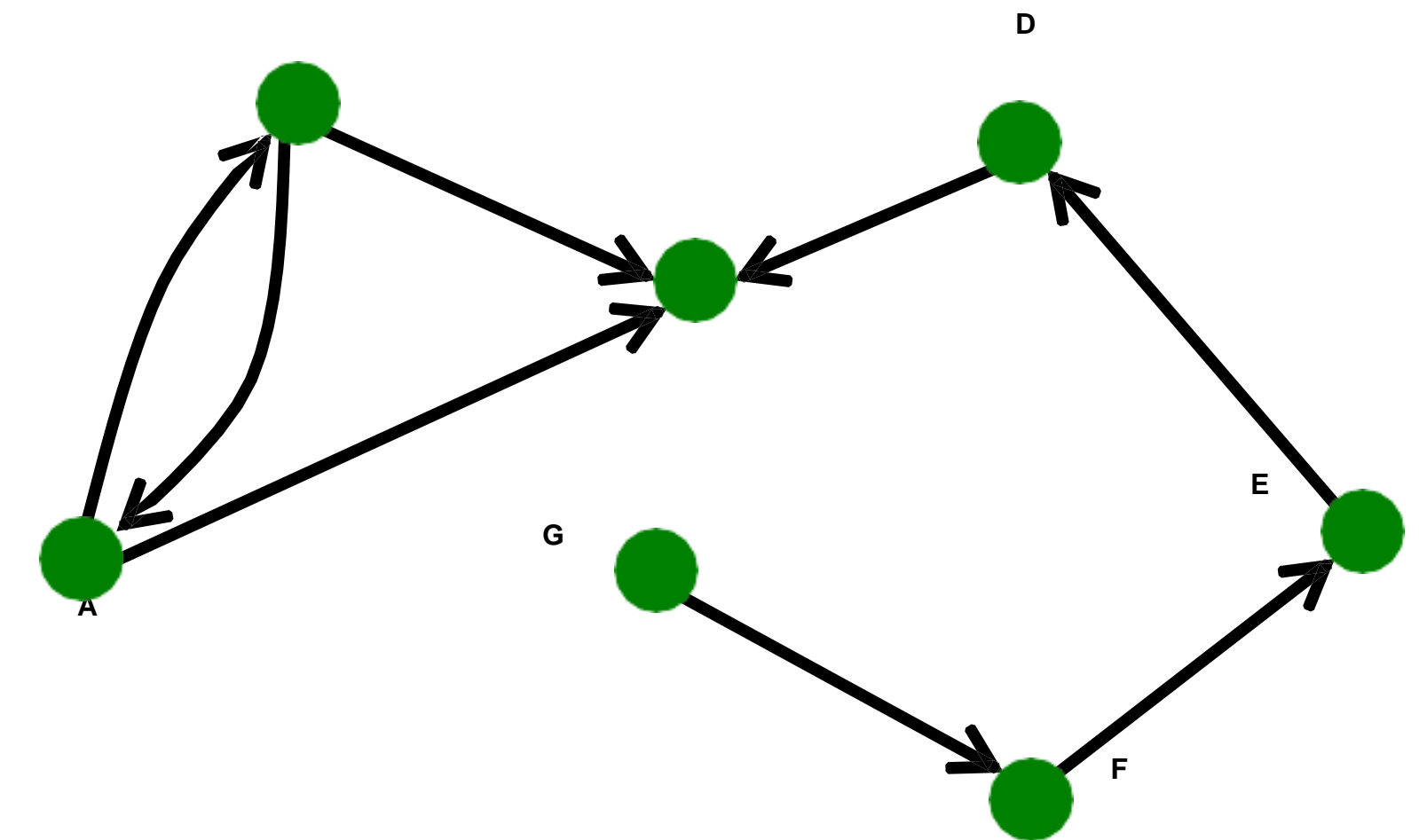
DIRECTED VS. UNDIRECTED GRAPHS

Lecture 1

- Undirected
Links: undirected (symmetrical, reciprocal)
- Examples:
Collaborations
Friendship on Facebook



- Directed
Links: directed (arcs)
- Examples:
Phone calls
Following on Twitter



Two nodes/vertices are *adjacent* if they share a common edge
An edge and a node on that edge are called *incident*.

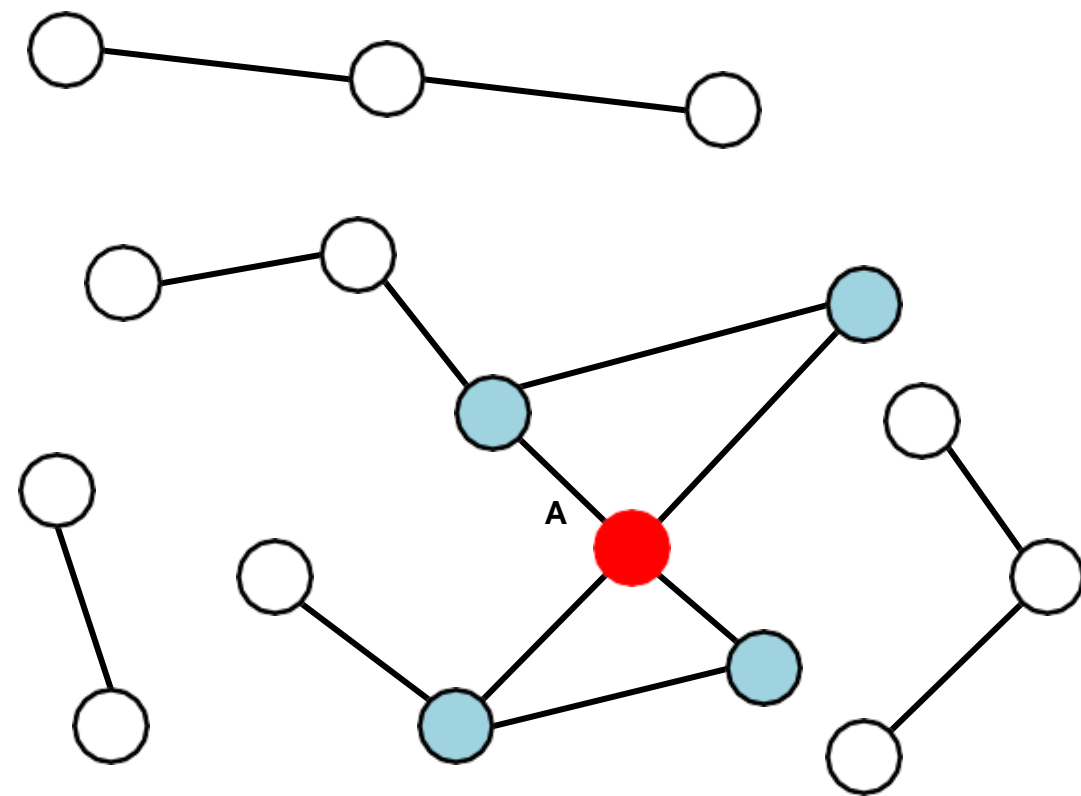
BASIC CONCEPTS

NODE DEGREES

Lecture 1

- Undirected
- Node degree, k_i : the number of edges adjacent to node i

$$k_A = 4 \quad \text{Avg. degree: } \bar{k} = \langle k \rangle = \frac{1}{N} \sum_{i=1}^N k_i = \frac{2E}{N}$$



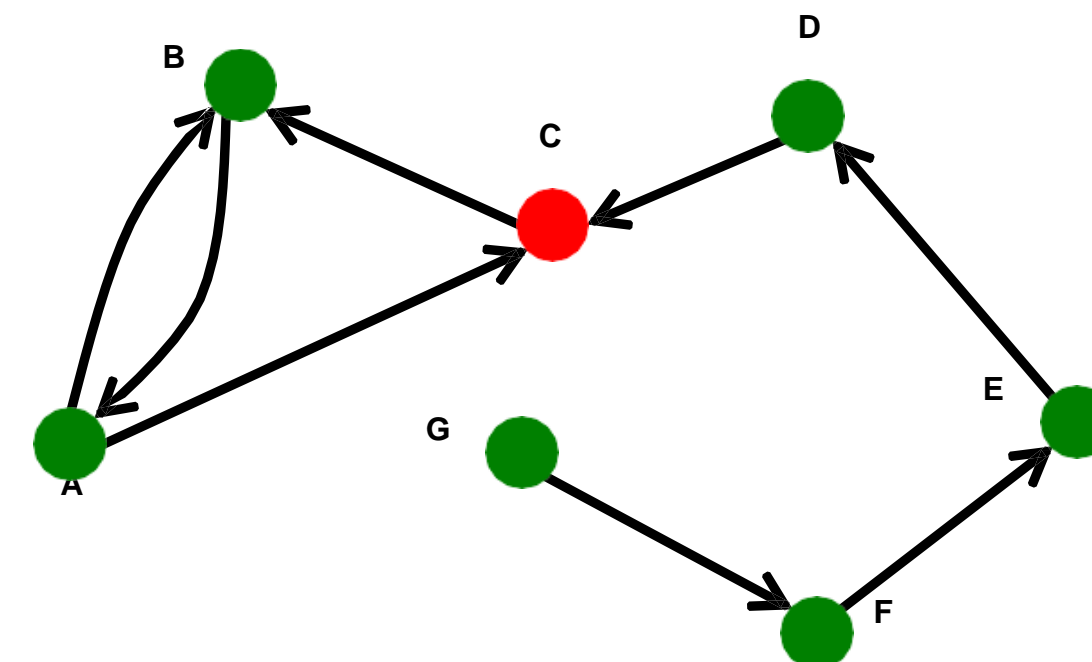
Source: Node with $k^{in} = 0$

Sink: Node with $k^{out} = 0$

- Directed
- In directed networks we define an in-degree and out-degree.
The (total) degree of a node is the sum of in- and out-degrees.

$$k^{in} = 2$$

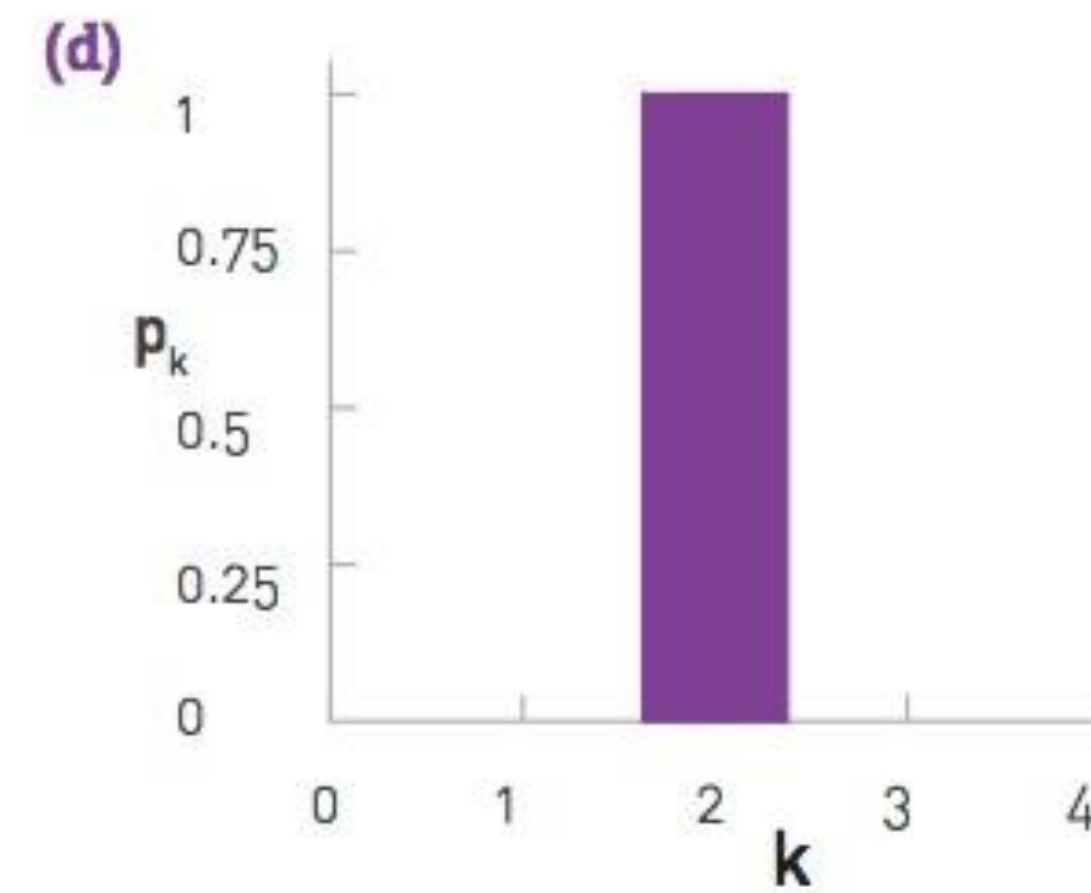
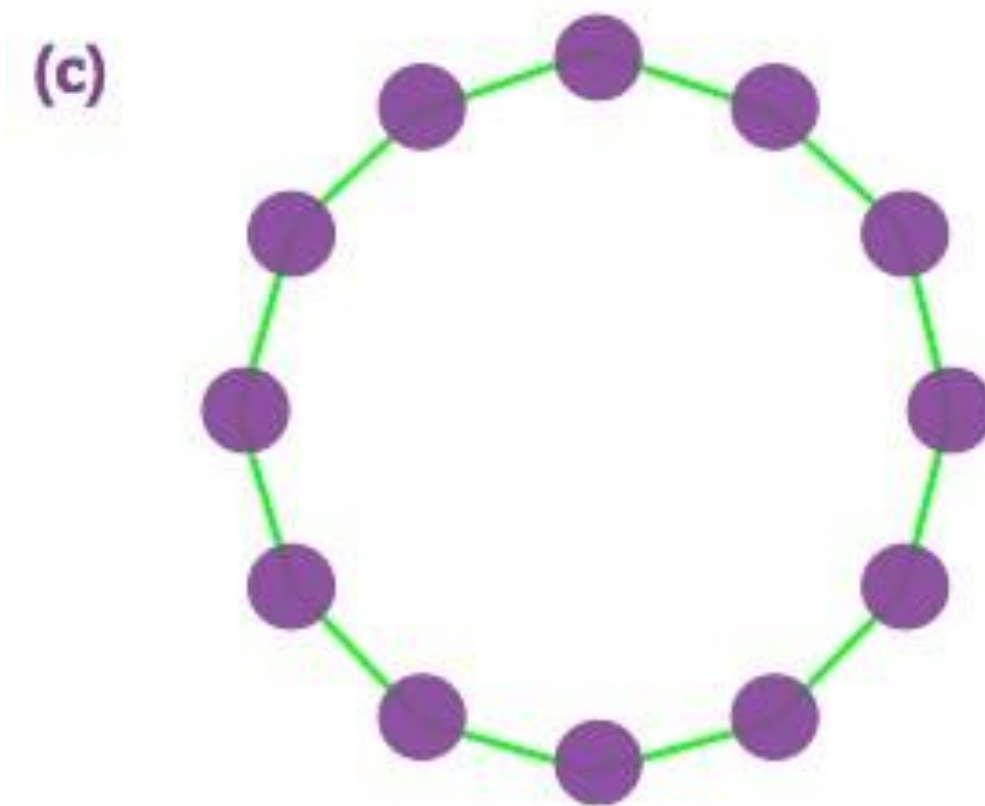
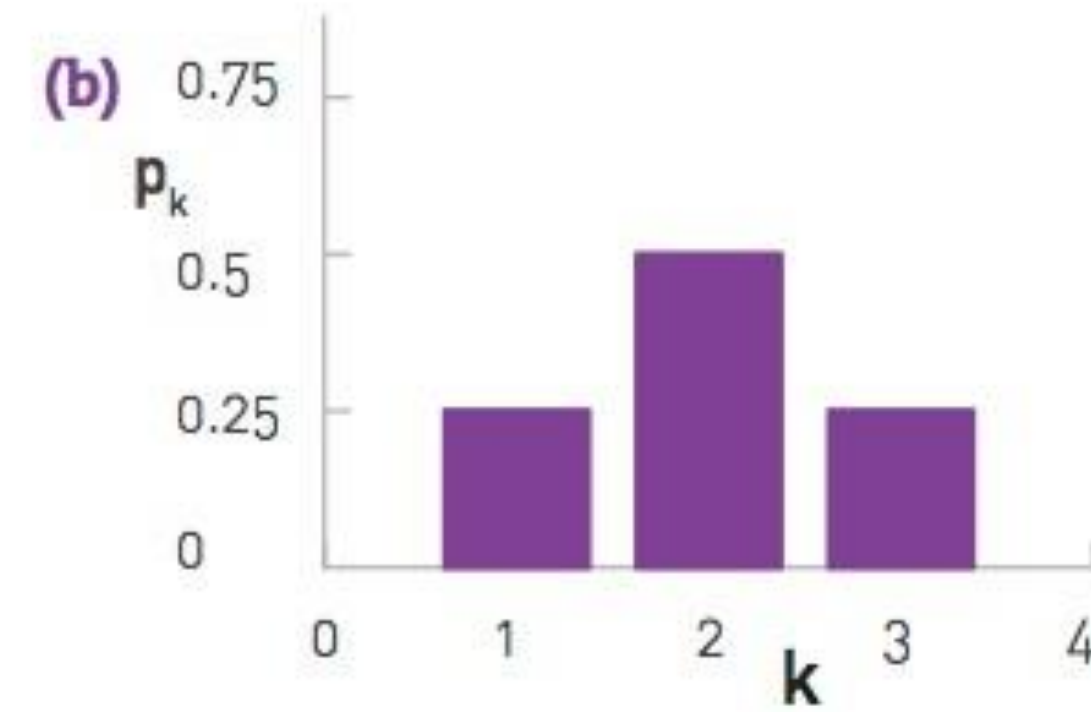
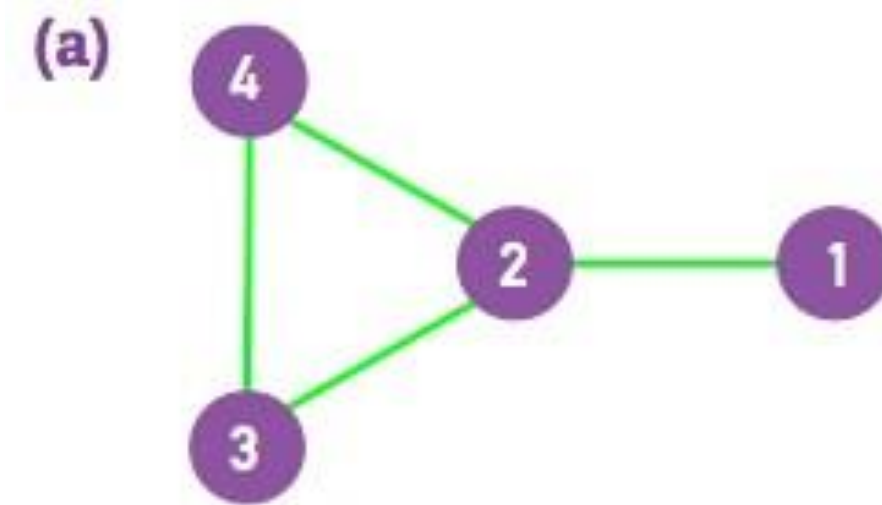
$$k_c^{out} = 1 \quad k_c = 3$$



BASIC CONCEPTS

DEGREE DISTRIBUTION

Lecture 1

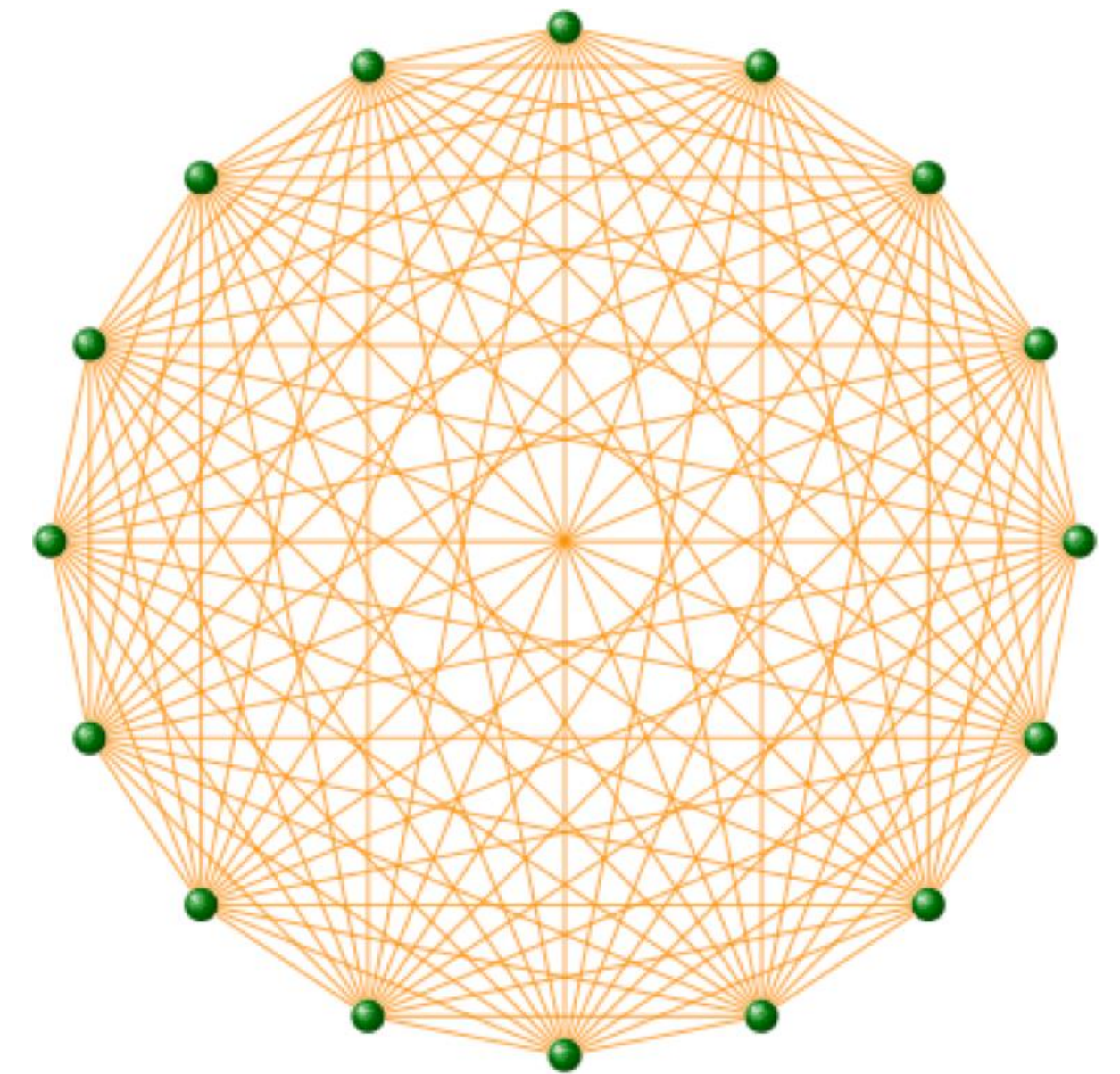


BASIC CONCEPTS

COMPLETE GRAPH

- The maximum number of edges in an undirected graph on N nodes is

$$E_{\max} = \binom{N}{2} = \frac{N(N-1)}{2}$$

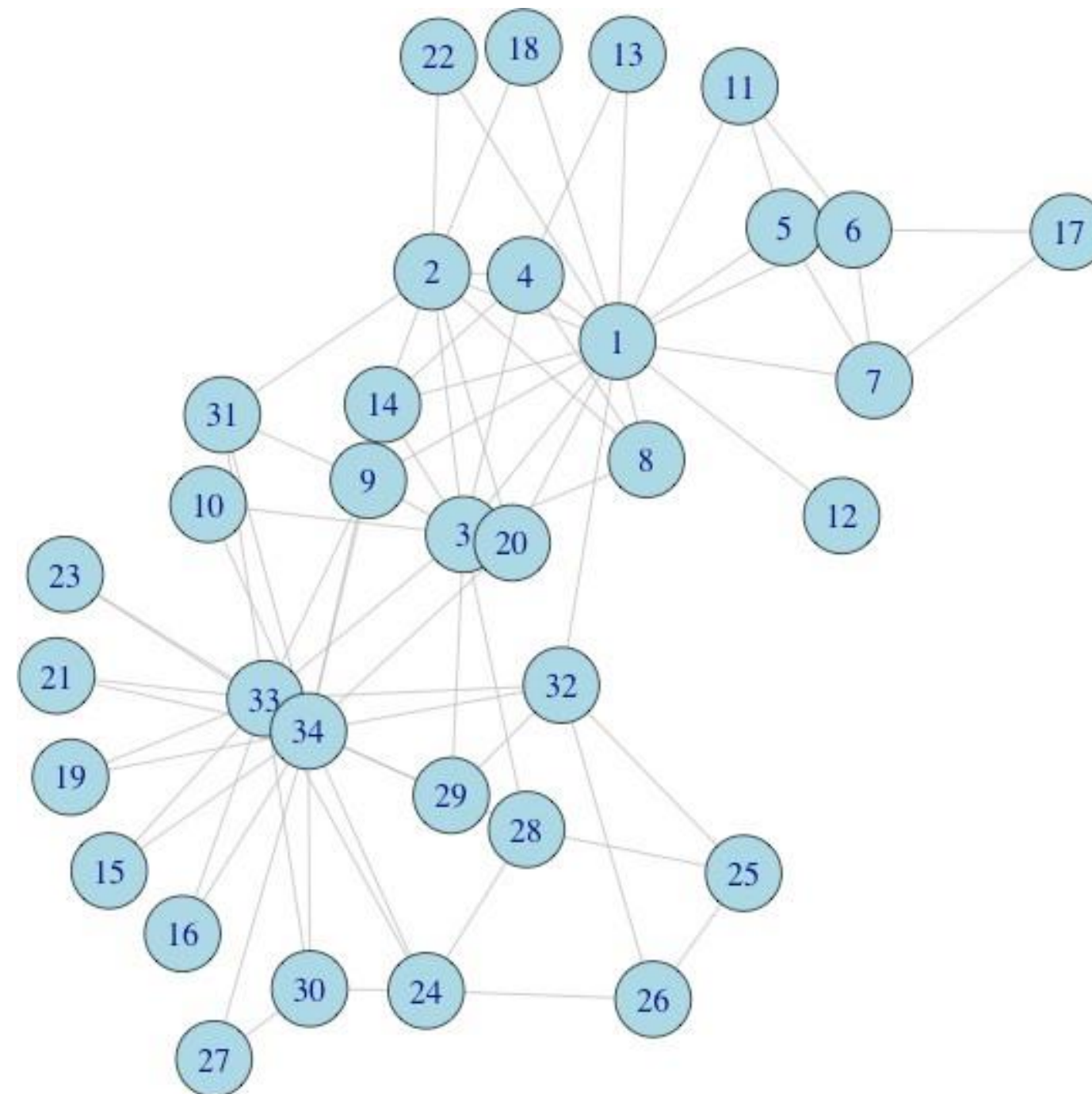


An undirected graph with the number of edges $E = E_{\max}$ is called a **complete graph**, and its average degree is $N-1$

BASIC CONCEPTS

DEGREE DISTRIBUTION

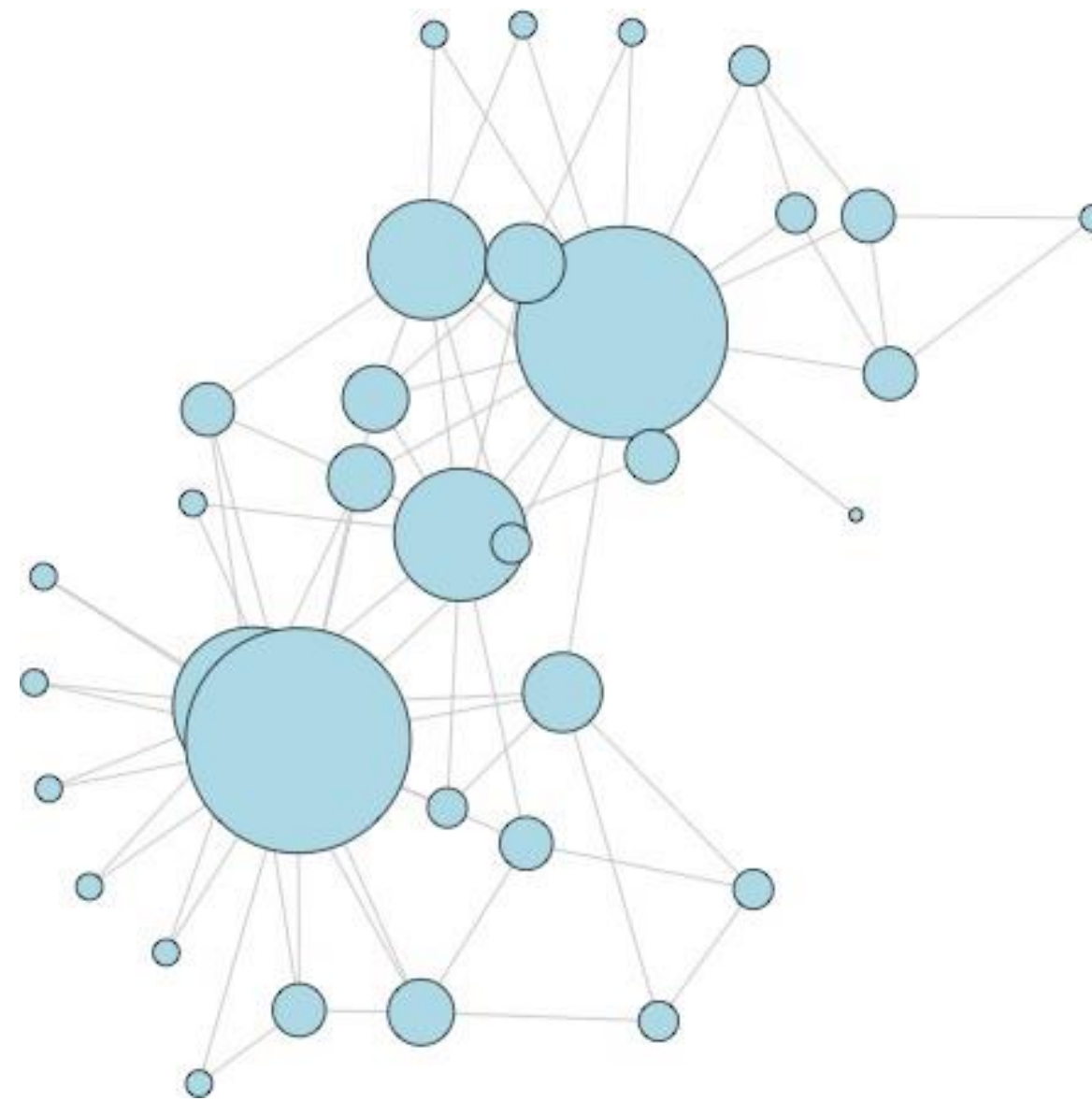
Lecture 1



BASIC CONCEPTS

DEGREE DISTRIBUTION

Lecture 1



BASIC CONCEPTS

AVERAGE DEGREES

Lecture 1

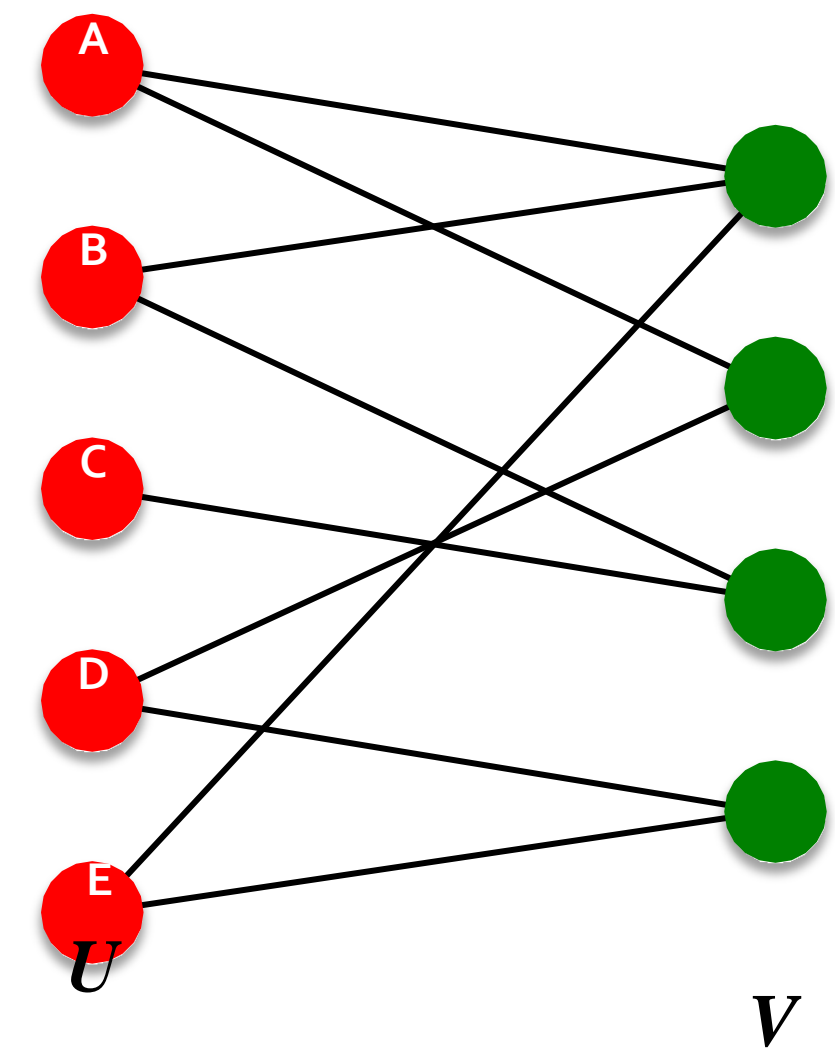
NETWORK	NODES	LINKS	DIRECTED UNDIRECTED	N	L	$\langle k \rangle$
Internet	Routers	Internet connections	Undirected	192,244	609,066	6.33
WWW	Webpages	Links	Directed	325,729	1,497,134	4.60
Power Grid	Power plants, transformers	Cables	Undirected	4,941	6,594	2.67
Mobile Phone Calls	Subscribers	Calls	Directed	36,595	91,826	2.51
Email	Email addresses	Emails	Directed	57,194	103,731	1.81
Science Collaboration	Scientists	Co-authorship	Undirected	23,133	93,439	8.08
Actor Network	Actors	Co-acting	Undirected	702,388	29,397,908	83.71
Citation Network	Paper	Citations	Directed	449,673	4,689,479	10.43
E. Coli Metabolism	Metabolites	Chemical reactions	Directed	1,039	5,802	5.58
Protein Interactions	Proteins	Binding interactions	Undirected	2,018	2,930	2.90

BASIC CONCEPTS

BIPARTITE GRAPH

Lecture 1

- Bipartite graph is a graph whose nodes can be divided into two disjoint sets U and V such that every link connects a node in U to one in V ; that is, U and V are independent sets
- Examples:
 - Authors-to-Papers (they authored)
 - Actors-to-Movies (they appeared in)
 - Users-to-Movies (they rated)
 - Recipes-to-Ingredients (they contain)
- “Folded” networks:
 - Author collaboration networks
 - Movie co-rating networks

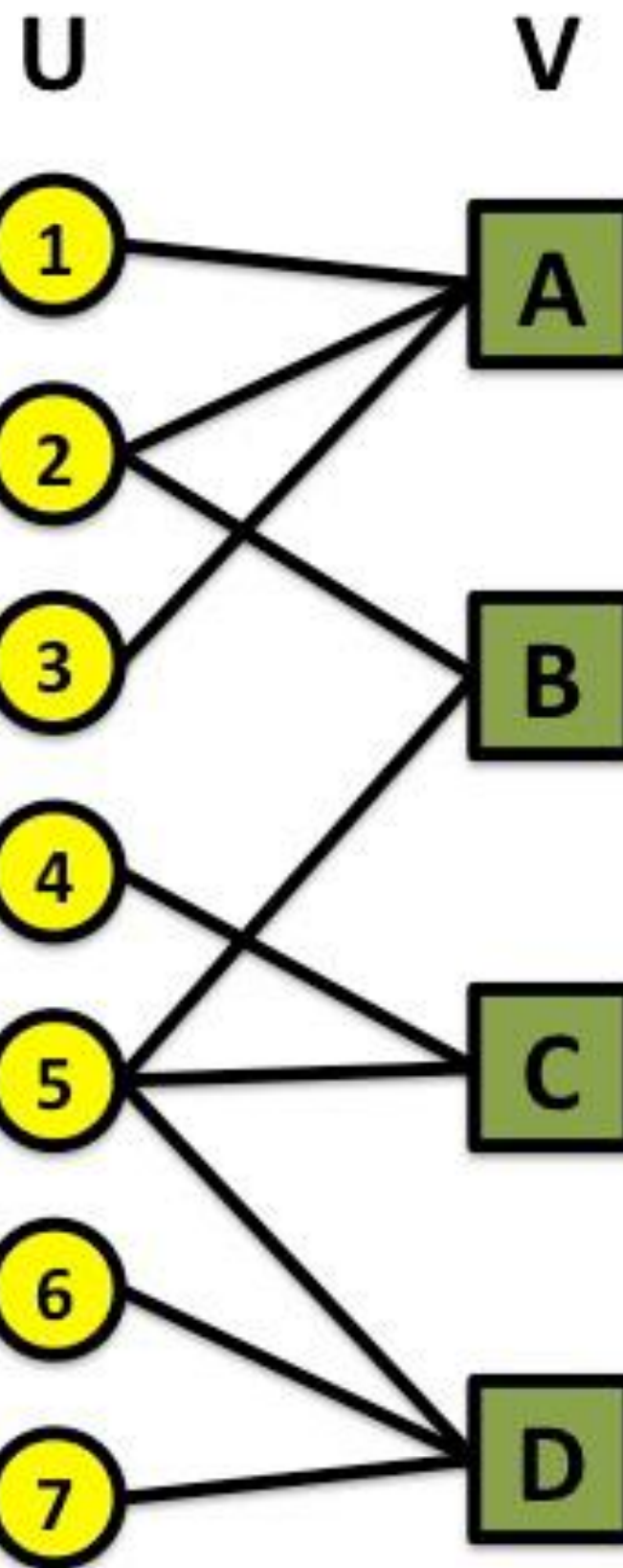
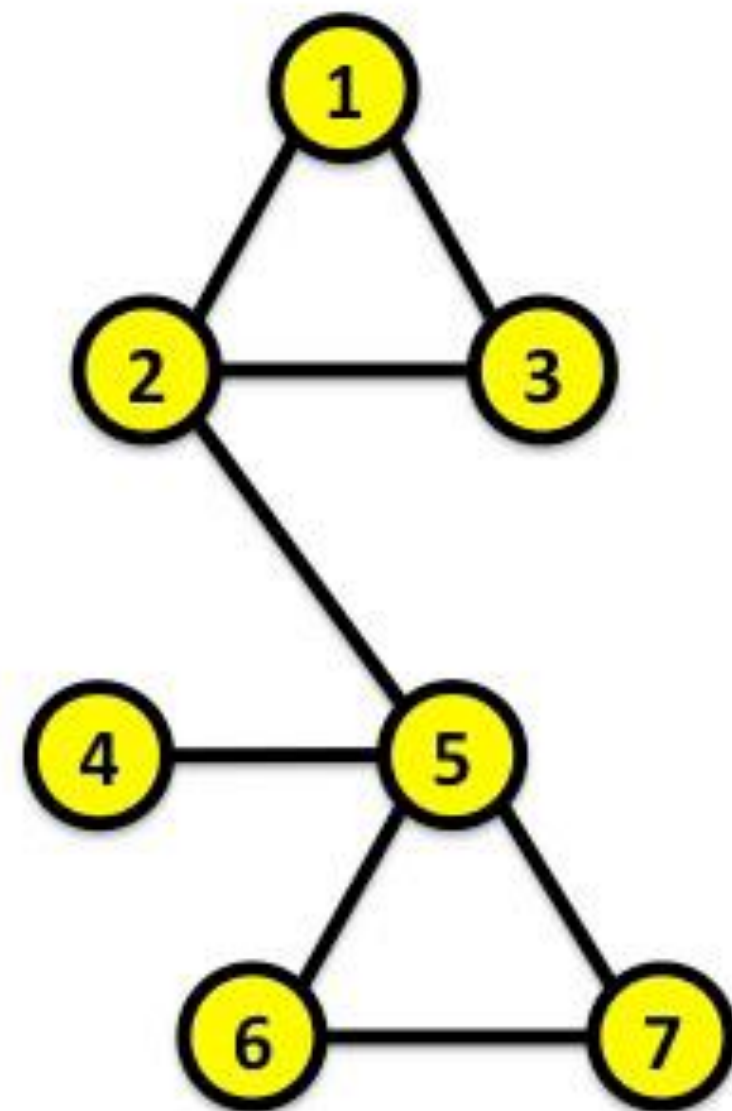


BASIC CONCEPTS

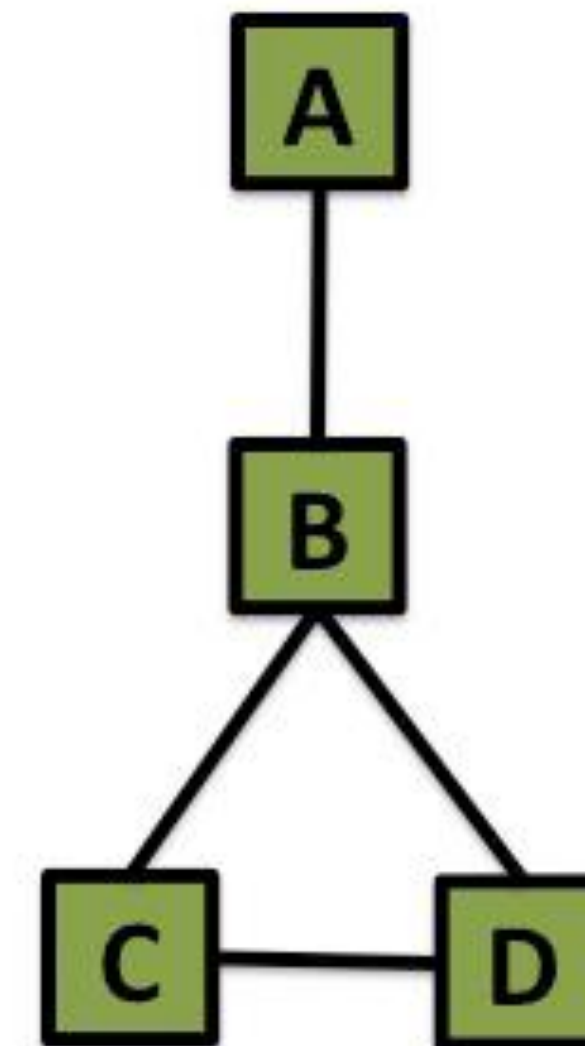
BIPARTITE GRAPH

Lecture 1

Projection U



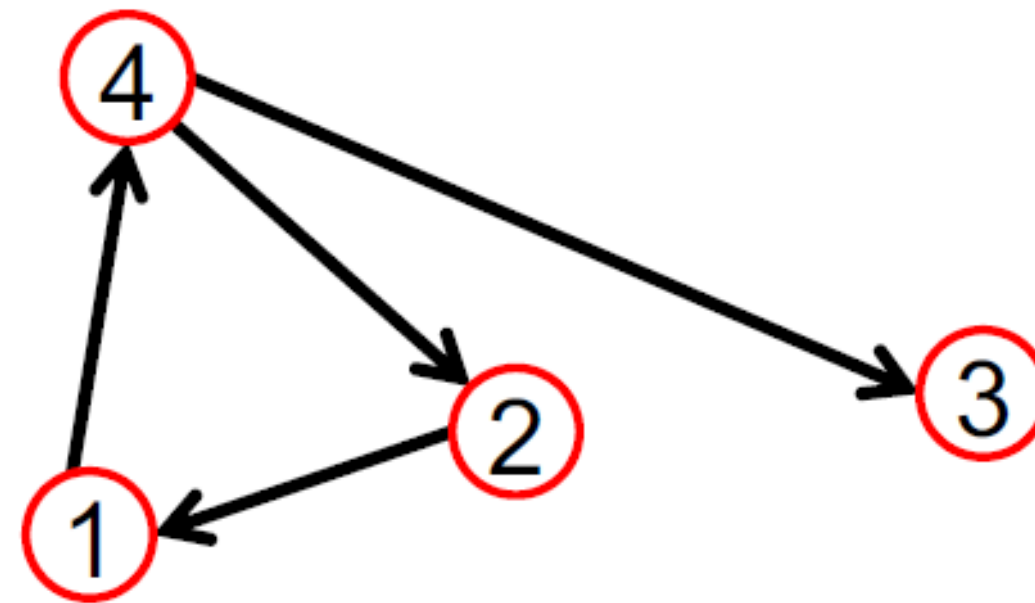
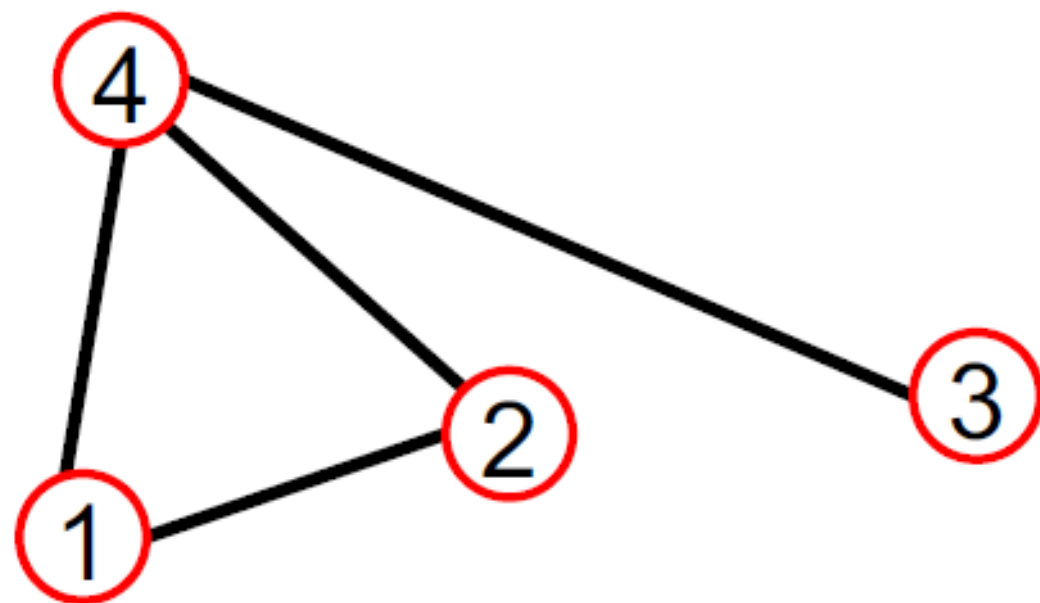
Projection V



BASIC CONCEPTS

Lecture 1

GRAPH REPRESENTATION: ADJACENCY MATRIX



$A_{ij} = 1$ if there is a link from node i to node j

$A_{ij} = 0$ otherwise

$$A = \begin{pmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

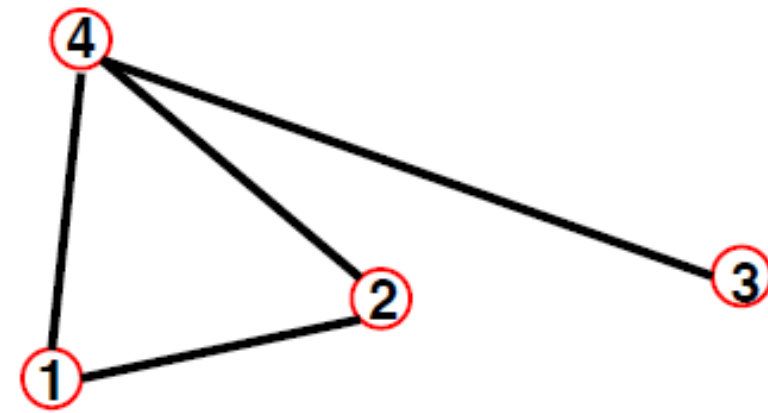
$$A = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \end{pmatrix}$$

Adjacency matrix $A^{n \times n}$ is a matrix with nonzero element a_{ij} when there is an edge e_{ij}

BASIC CONCEPTS

GRAPH REPRESENTATION: ADJACENCY MATRIX

Undirected



$$A_{ij} = \begin{pmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

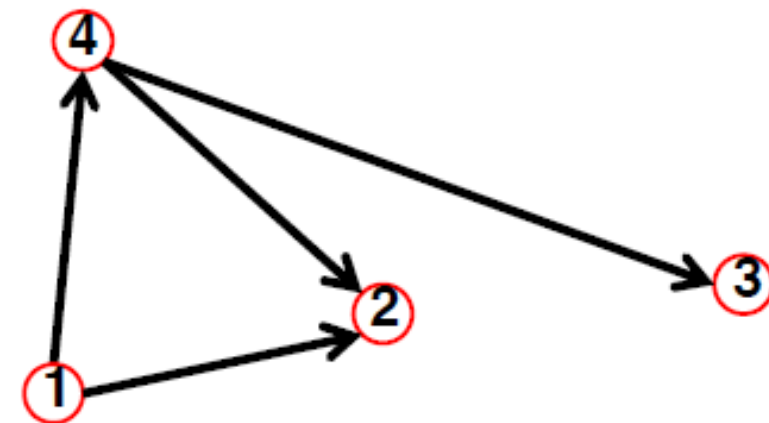
$$A_{ij} = A_{ji}$$
$$A_{ii} = 0$$

$$k_i = \sum_{j=1}^N A_{ij}$$

$$k_j = \sum_{i=1}^N A_{ij}$$

$$L = \frac{1}{2} \sum_{i=1}^N k_i = \frac{1}{2} \sum_{ij} A_{ij}$$

Directed



$$A = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \end{pmatrix}$$

$$A_{ij} \neq A_{ji}$$
$$A_{ii} = 0$$

$$k_i^{out} = \sum_{j=1}^N A_{ij}$$

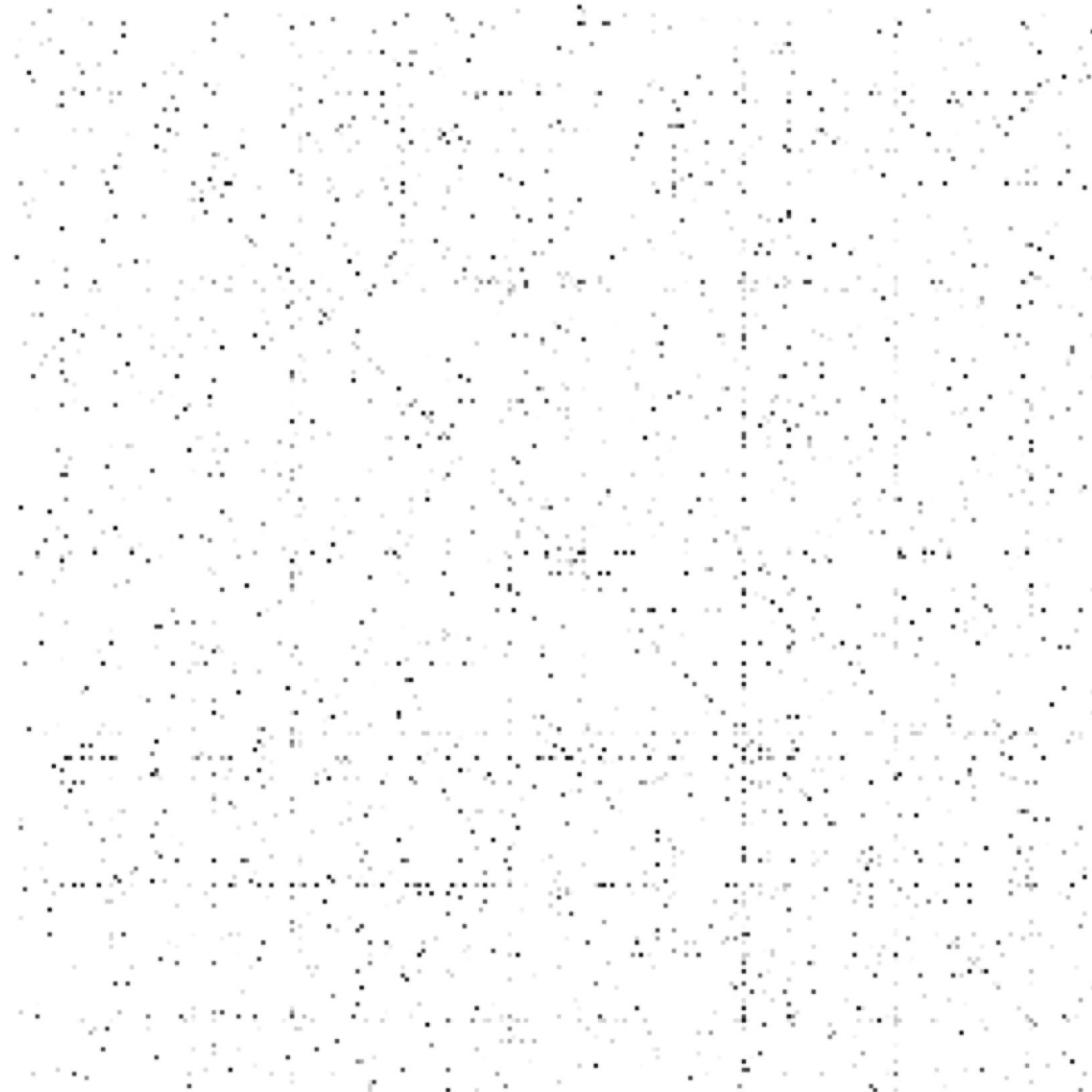
$$k_j^{in} = \sum_{i=1}^N A_{ij}$$

$$L = \sum_{i=1}^N k_i^{in} = \sum_{j=1}^N k_j^{out} = \sum_{i,j} A_{ij}$$

BASIC CONCEPTS

Lecture 1

GRAPH REPRESENTATION: ADJACENCY MATRIX



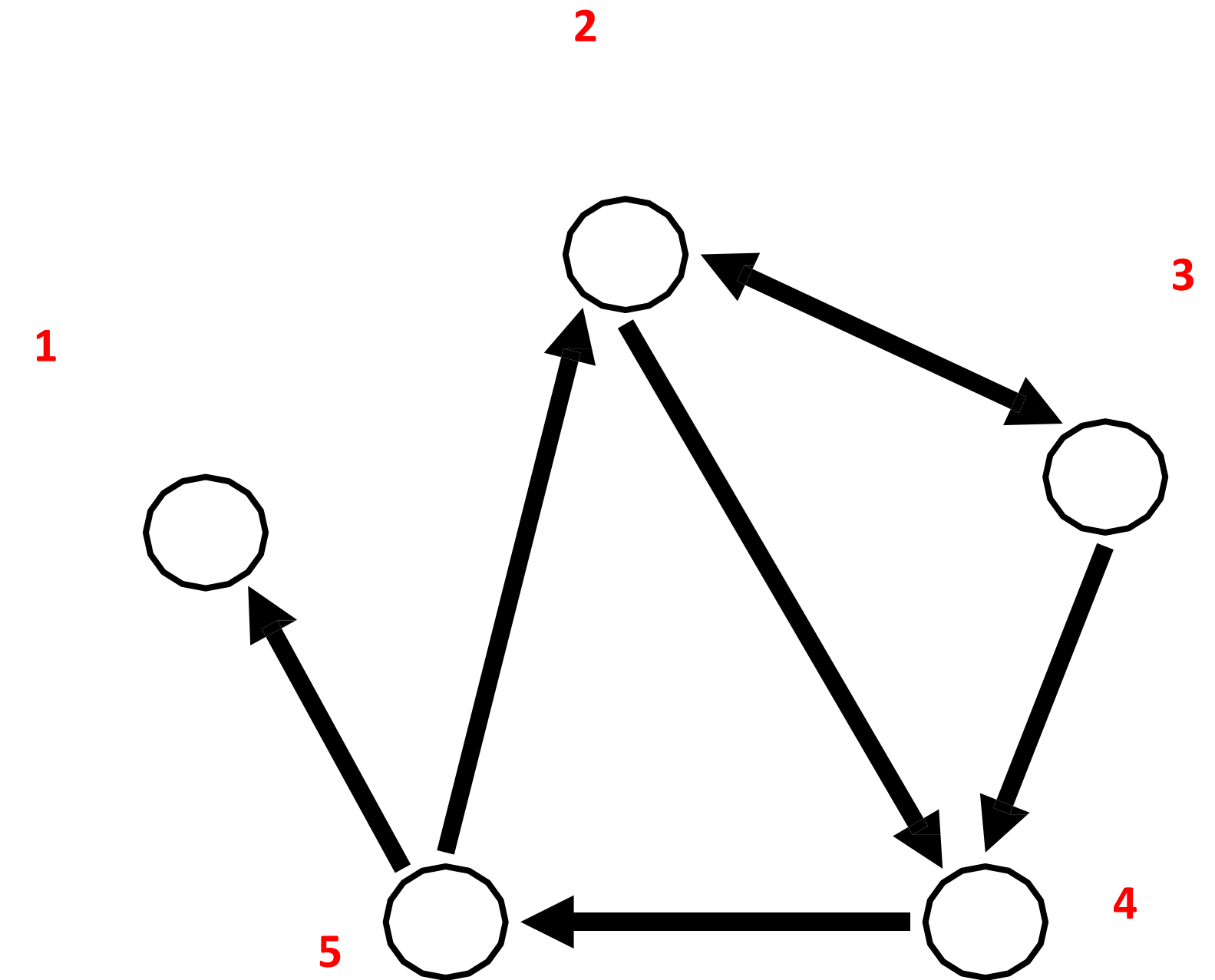
BASIC CONCEPTS

Lecture 1

GRAPH REPRESENTATION: EDGE LIST

Represent graph as a set of edges:

- (2, 3)
- (2, 4)
- (3, 2)
- (3, 4)
- (4, 5)
- (5, 2)
- (5, 1)



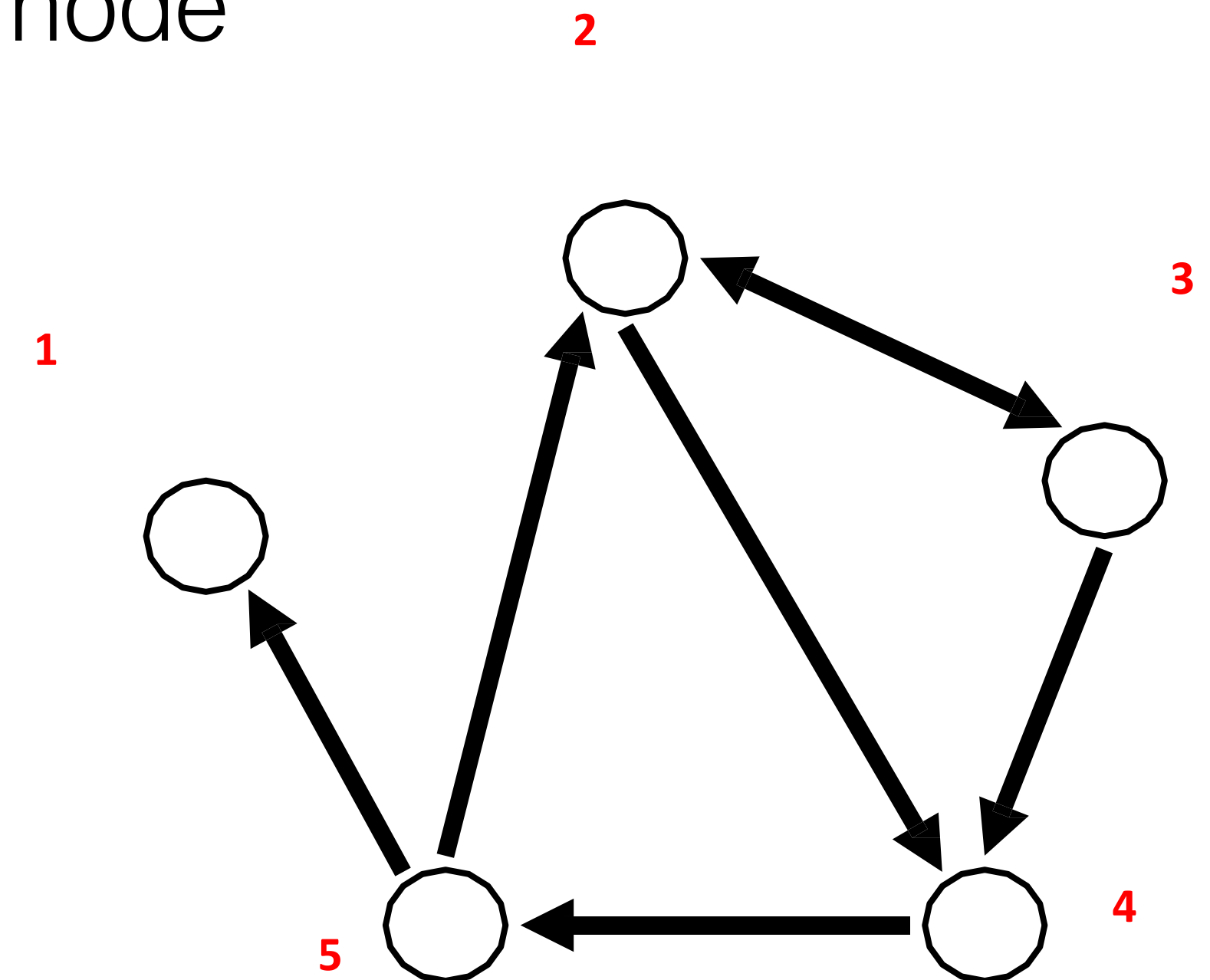
BASIC CONCEPTS

Lecture 1

GRAPH REPRESENTATION: EDGE LIST

Adjacency list:

- Easier to work with if network is
Large
Sparse
- Allows us to quickly retrieve all neighbors of a given node
1:
2: 3, 4
3: 2, 4
4: 5
5: 1, 2



BASIC CONCEPTS

EDGE ATTRIBUTES

Lecture 1

Possible options:

- Weight (e.g. frequency of communication)
- Ranking (best friend, second best friend...)
- Type (friend, relative, co-worker)
- Sign: Friend vs. Foe, Trust vs. Distrust
- Properties depending on the structure of the rest of the graph: number of common friends

BASIC CONCEPTS

EDGE ATTRIBUTES

Lecture 1

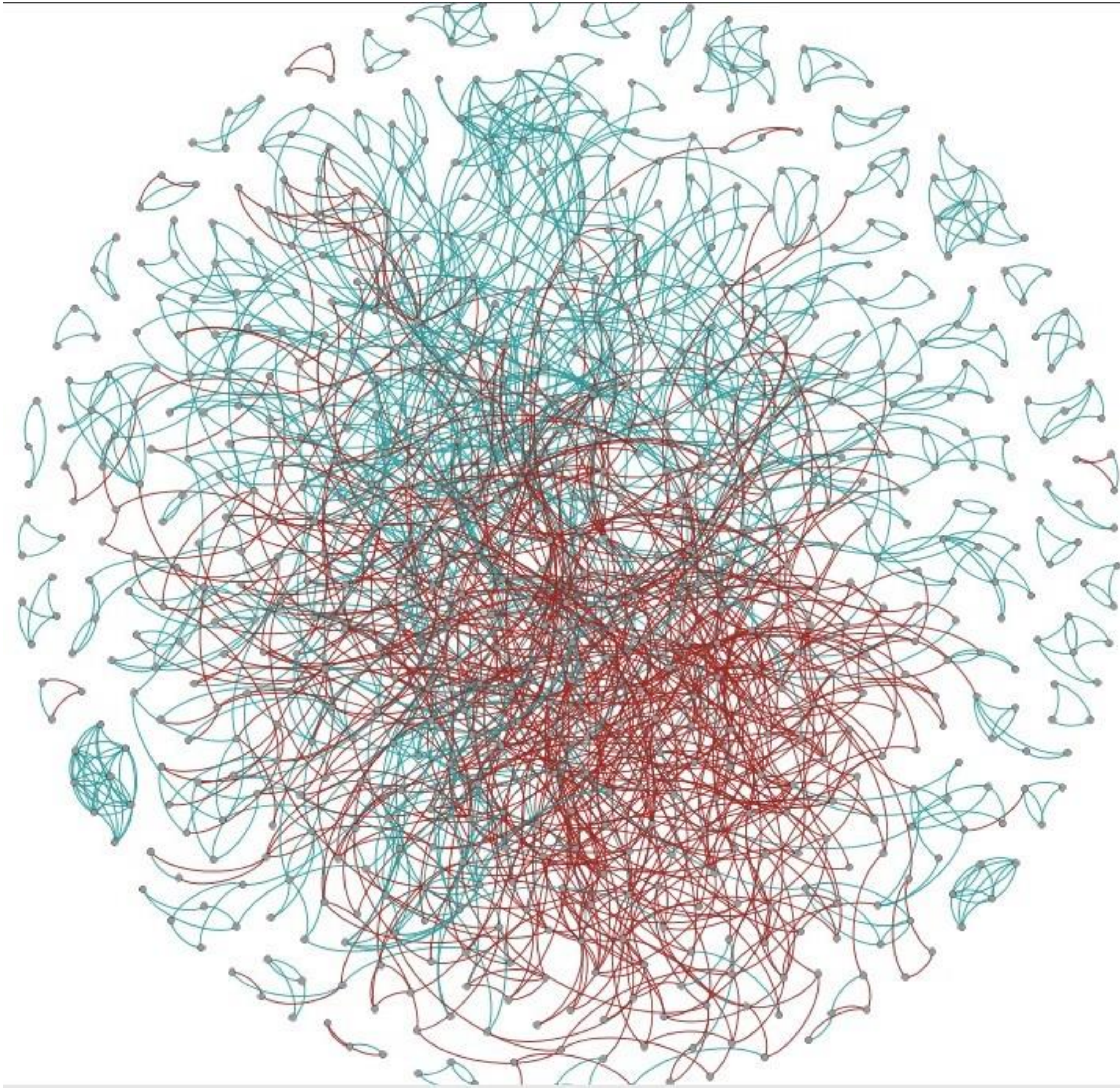
Possible options:

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- Ranking (best friend, second best friend...)
- Type (friend, relative, co-worker)
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BASIC CONCEPTS

POSITIVE AND NEGATIVE WEIGHTS

Lecture 1



BASIC CONCEPTS

POSITIVE AND NEGATIVE WEIGHTS

Lecture 1

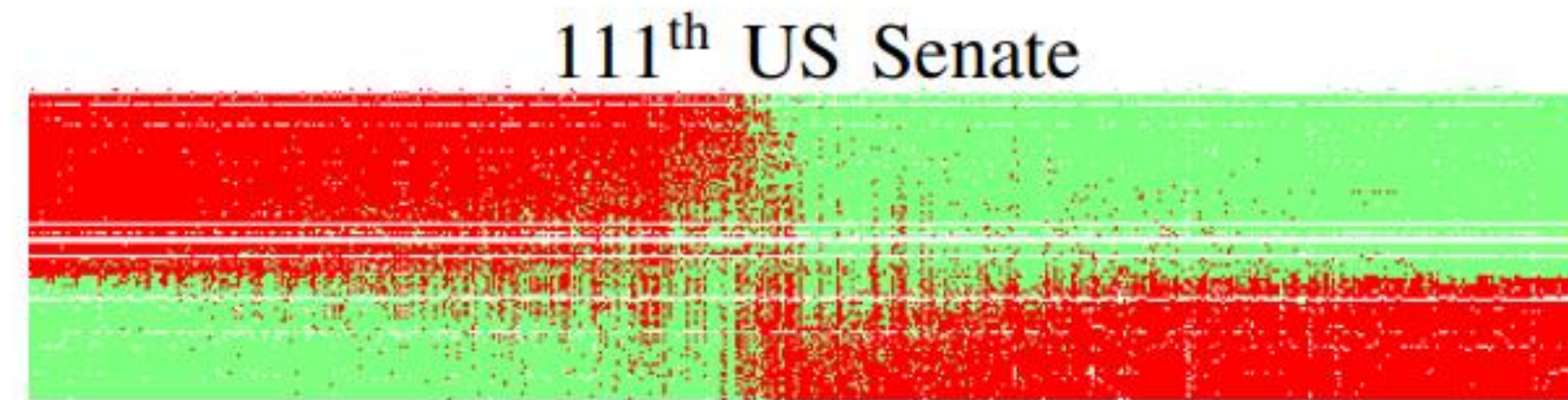


Fig. 4. Vote matrix of the 111th US Senate after scaling with ANCO-HITS

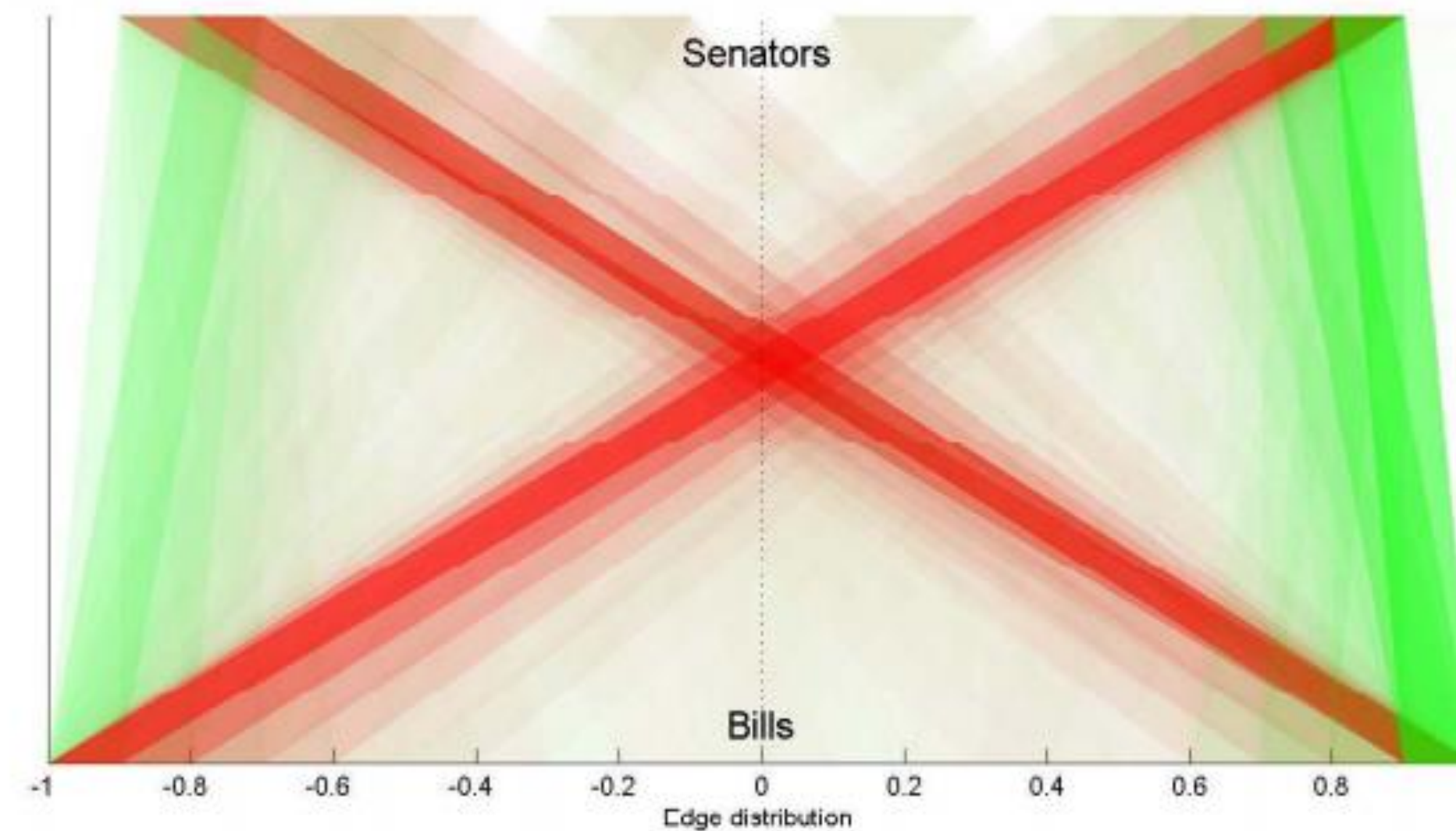


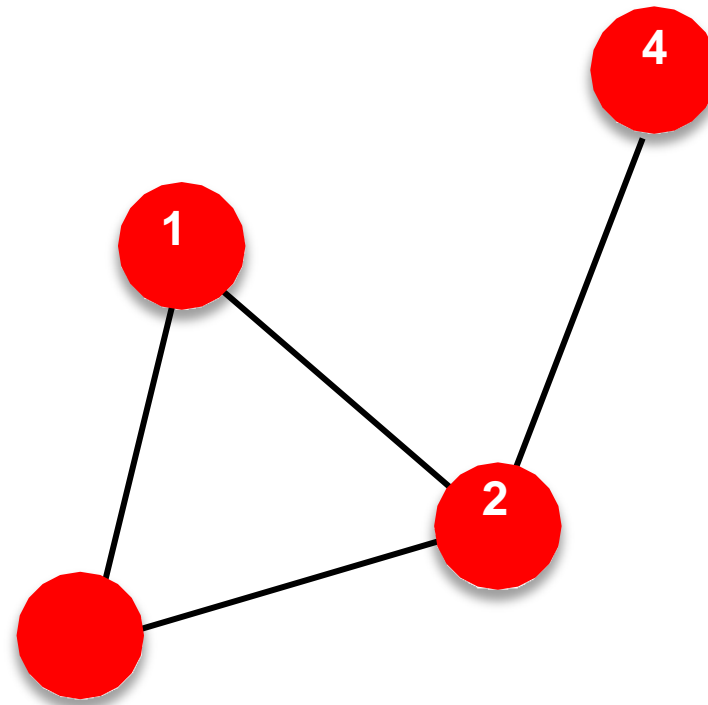
Fig. 5. Bipartite graph of the 111th US Senate after scaling with ANCO-HITS

BASIC CONCEPTS

POSITIVE AND NEGATIVE WEIGHTS

Lecture 1

Unweighted (undirected)



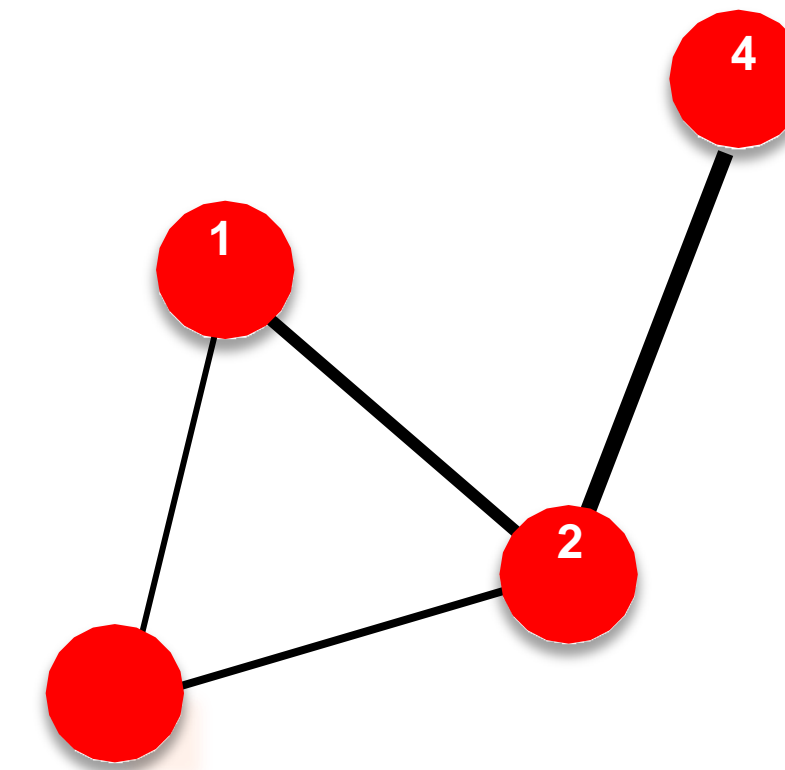
$$A_{ij} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

$$A_{ii} = 0$$

$$A_{ij} = A_{ji}$$

$$E = \frac{1}{2} \sum_{i,j=1}^N A_{ij} \quad \bar{k} = \frac{2E}{N}$$

Weighted (undirected)



$$A_{ij} = \begin{pmatrix} 0 & 2 & 0.5 & 0 \\ 2 & 0 & 1 & 4 \\ 0.5 & 1 & 0 & 0 \\ 0 & 4 & 0 & 0 \end{pmatrix}$$

$$A_{ii} = 0$$

$$A_{ij} = A_{ji}$$

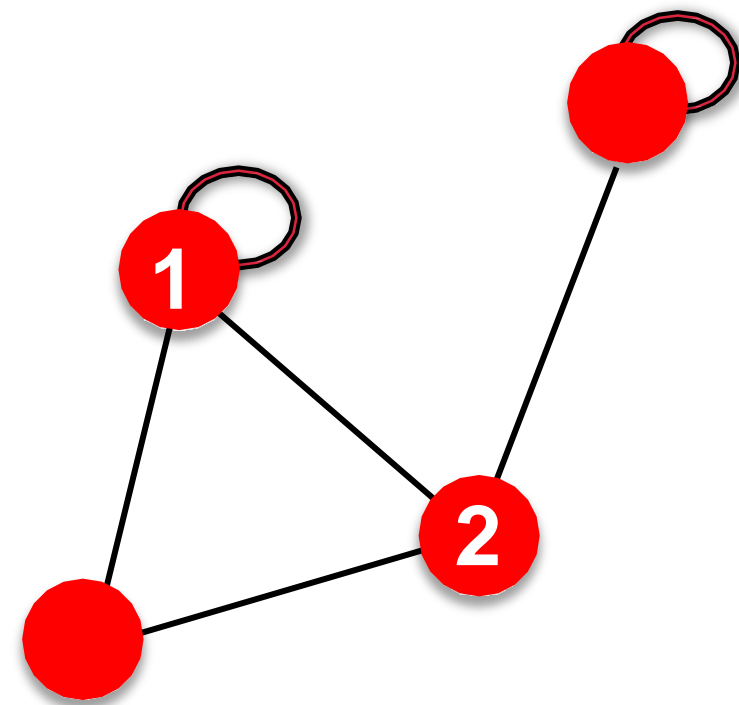
$$E = \frac{1}{2} \sum_{i,j=1}^N \text{nonzero}(A_{ij}) \quad \bar{k} = \frac{2E}{N}$$

BASIC CONCEPTS

POSITIVE AND NEGATIVE WEIGHTS

Lecture 1

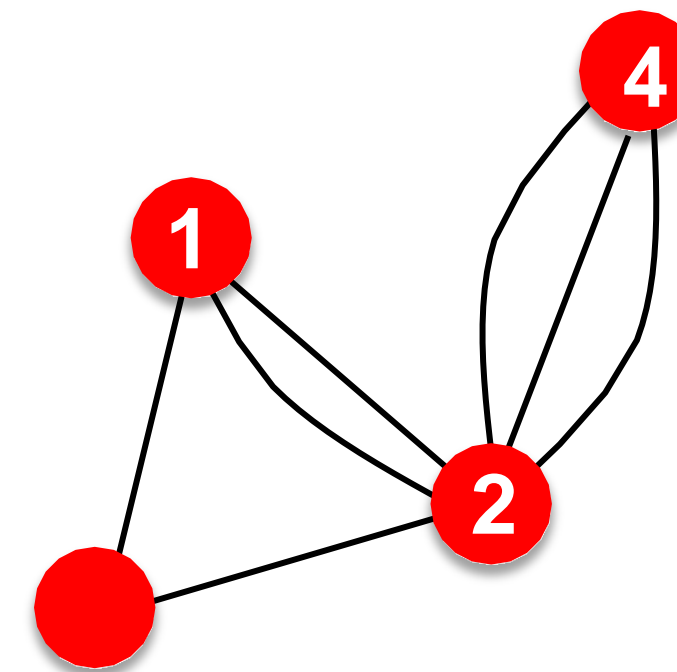
Self-edges (self-loops) (undirected)



$$A_{ij} = \begin{pmatrix} 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 \end{pmatrix}$$

$$E = \frac{1}{2} \sum_{i,j=1, i \neq j}^N A_{ij} + \sum_{i=1}^N A_{ii} \quad A_{ij} = A_{ji}$$

Multigraph (undirected)



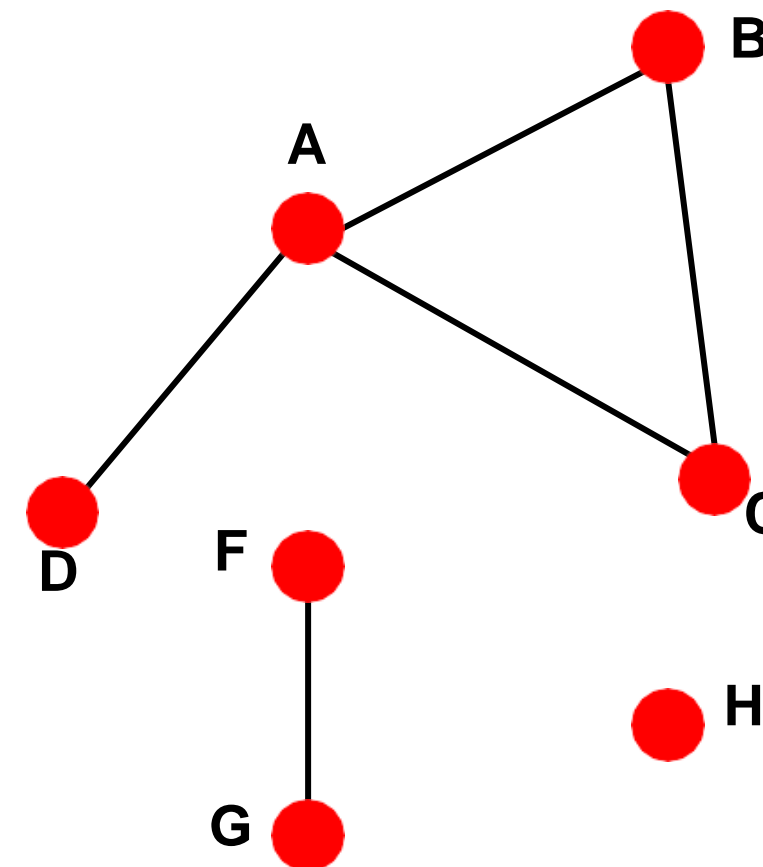
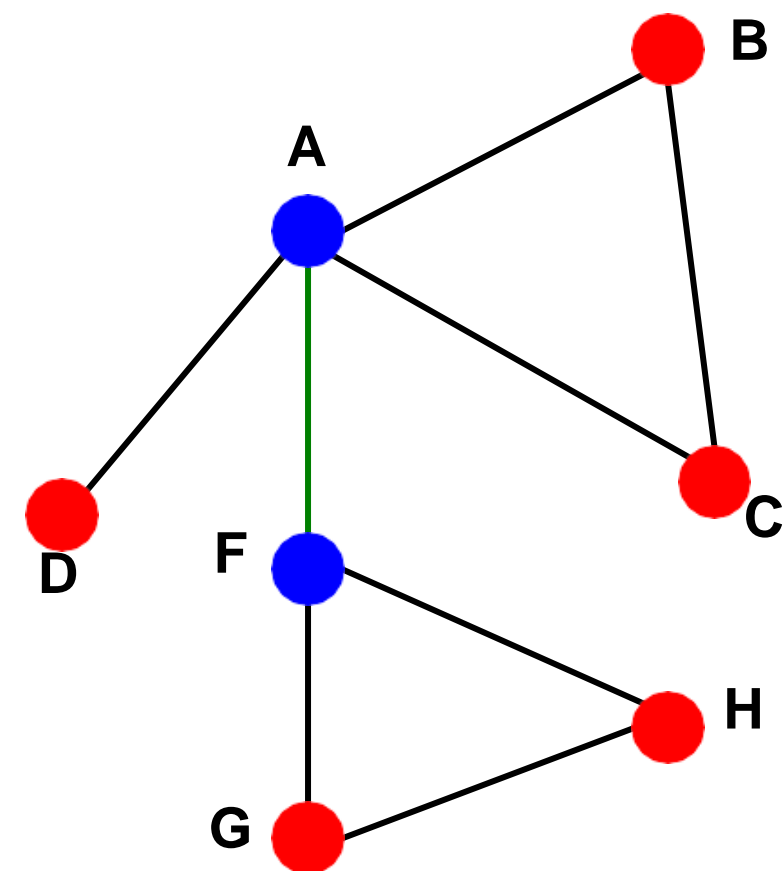
$$A_{ij} = \begin{pmatrix} 0 & 2 & 1 & 0 \\ 2 & 0 & 1 & 3 \\ 1 & 1 & 0 & 0 \\ 0 & 3 & 0 & 0 \end{pmatrix}$$

$$E = \frac{1}{2} \sum_{i,j=1}^N \text{nonzero}(A_{ij}) \quad A_{ii} = 0 \quad A_{ij} = A_{ji} \quad \bar{k} = \frac{2E}{N}$$

BASIC CONCEPTS

CONNECTIVITY OF UNDIRECTED GRAPHS

- A *path* from v_i to v_j is a sequence of edges that joins two vertices.
(It also ordered list of vertices such that that there is an edge to the next vertex on the list)
- A graph is *connected* if there a paths between any two vertices.
- *Connected component* is a maximal connected subgraph of G



Largest Component:
Giant Component

Isolated node (node H)

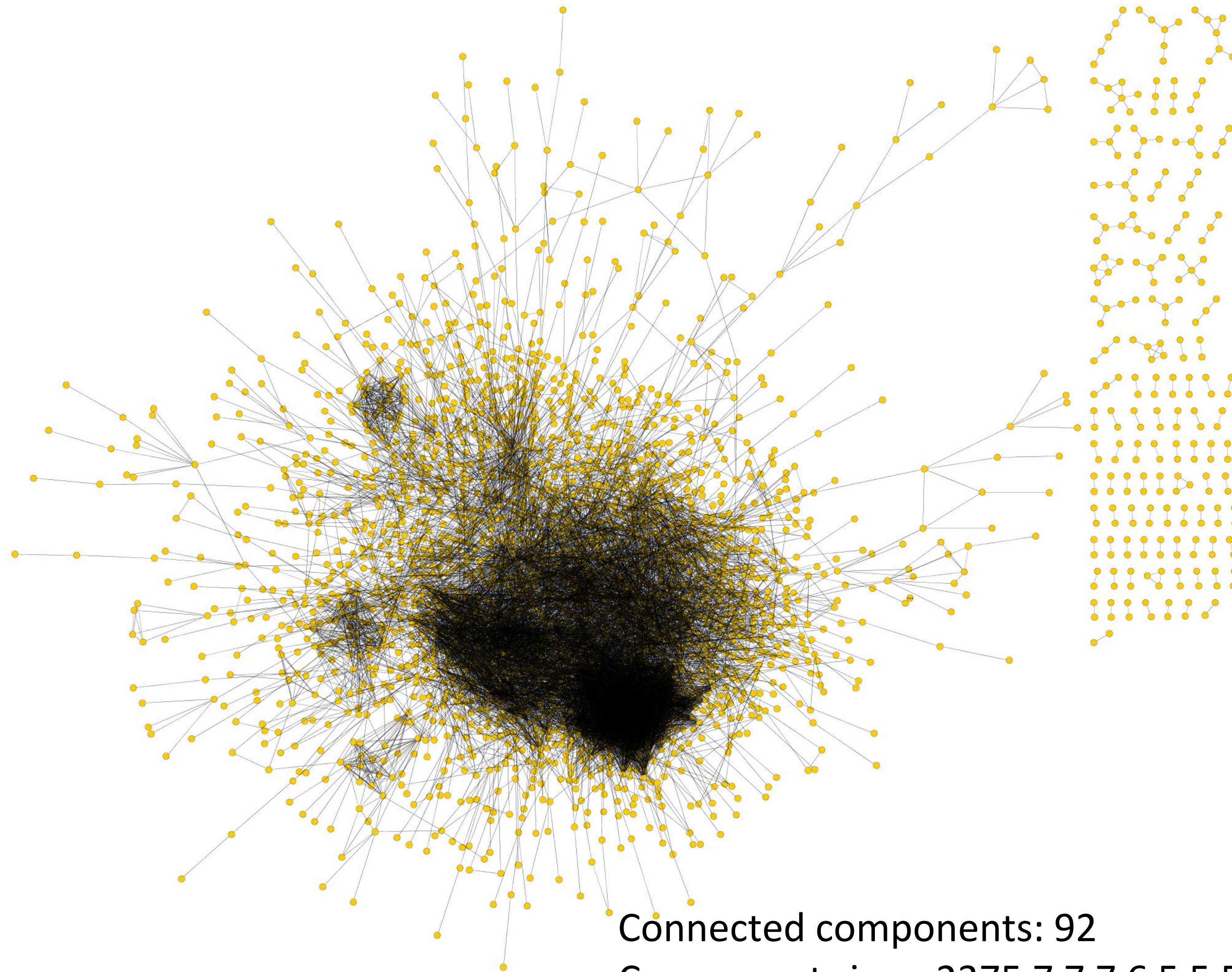
Bridge edge: If we erase the **edge**, the graph becomes disconnected

Articulation node: If we erase the **node**, the graph becomes disconnected

BASIC CONCEPTS

CONNECTIVITY OF UNDIRECTED GRAPHS

Lecture 1



Connected components: 92

Component sizes: 2375 7 7 7 6 5 5 5 5 5 5 4 4 4 4

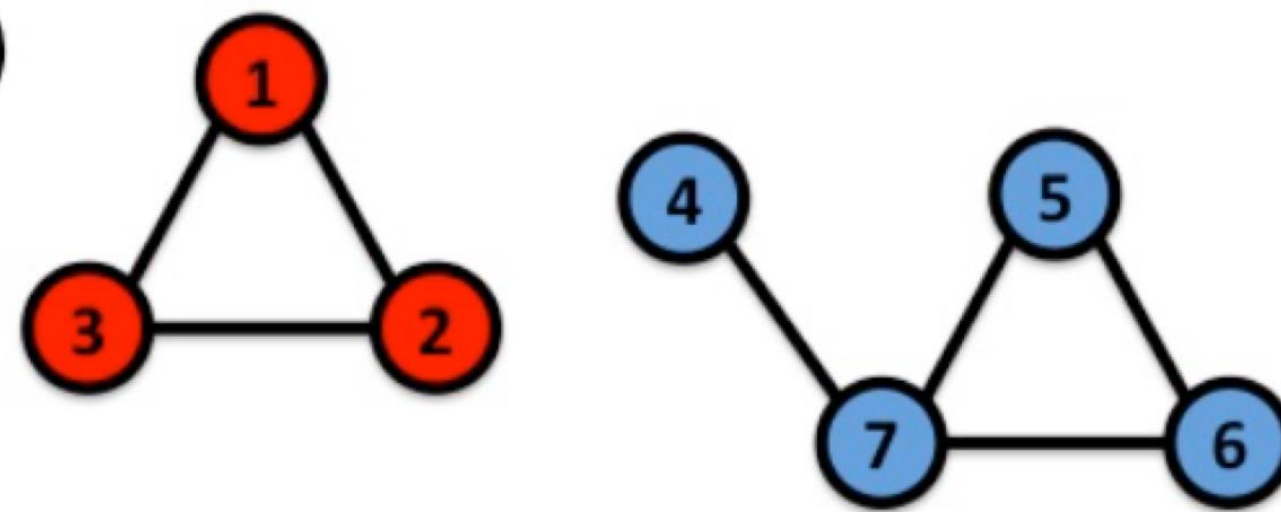
BASIC CONCEPTS

CONNECTIVITY EXAMPLE

The adjacency matrix of a network with several components can be written in a block-diagonal form, so that nonzero elements are confined to squares, with all other elements being zero:

Disconnected

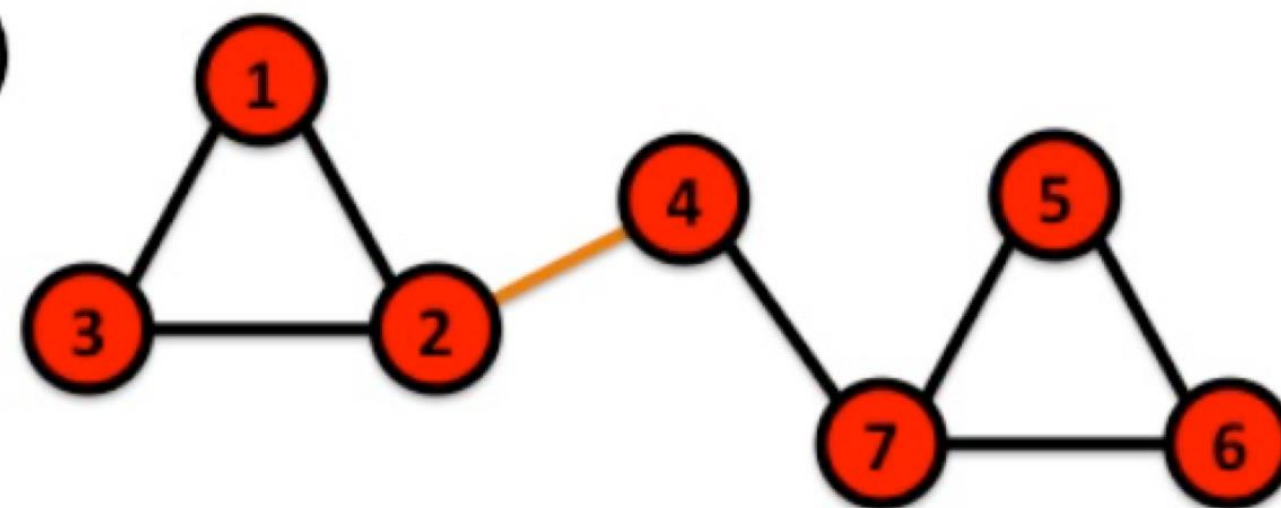
(a)



$$\begin{pmatrix} 0 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 \end{pmatrix}$$

Connected

(b)

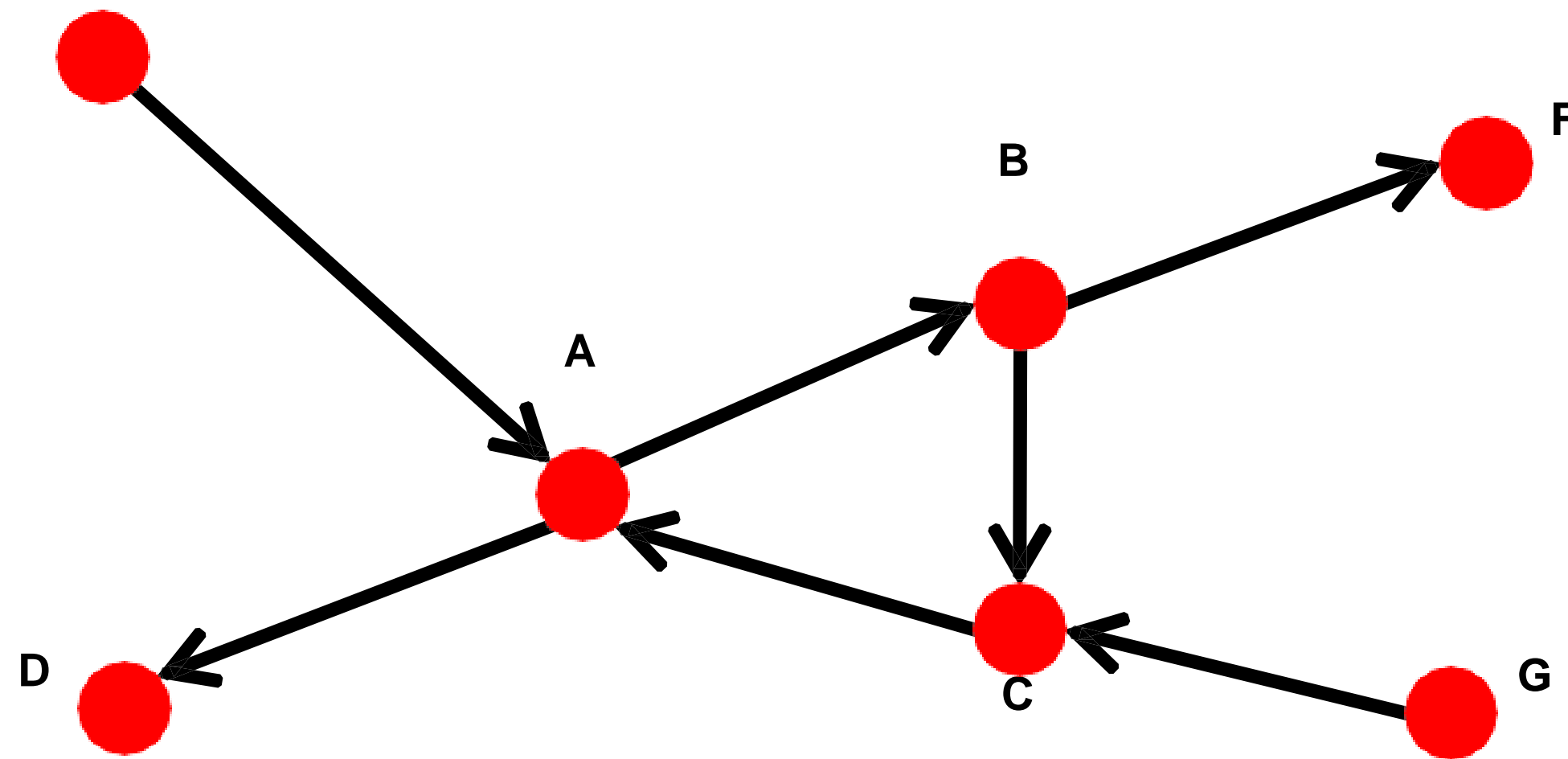


$$\begin{pmatrix} 0 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 \end{pmatrix}$$

BASIC CONCEPTS

CONNECTIVITY OF DIRECTED GRAPHS

- Strongly connected directed graph has a path from each node to every other node and vice versa e.g., A-B path and B-A path)
- Weakly connected directed graph is connected if we disregard the edge directions



Graph on the left is connected but not strongly connected (e.g., there is no way to get from F to G by following the edge directions).

NETWORK PROPERTIES

DEGREE DISTRIBUTION

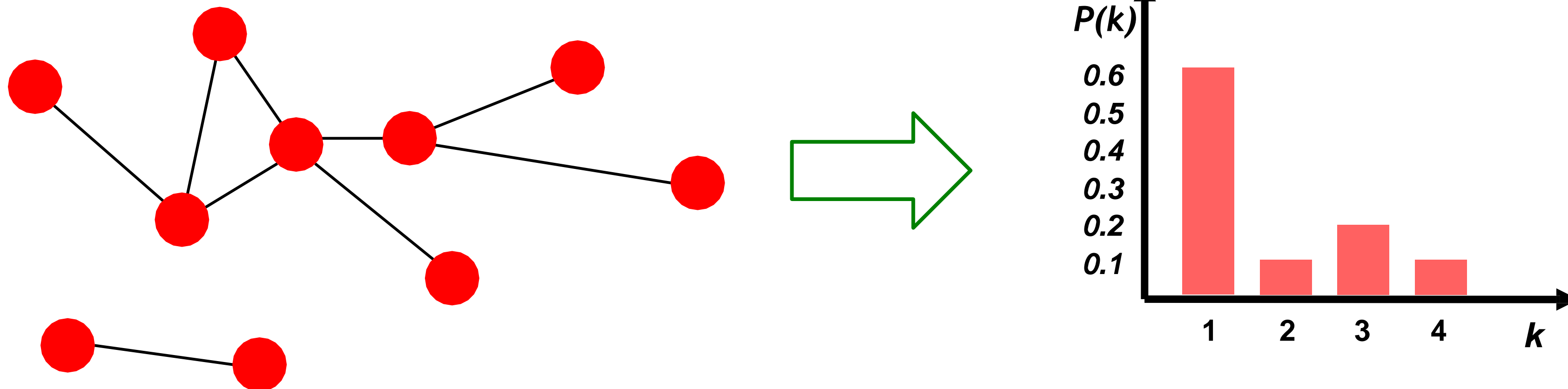
Lecture 1

Degree distribution $P(k)$: Probability that a randomly chosen node has degree k

$N_k = \#$ nodes with degree k

Normalized histogram:

$$P(k) = N_k / N \rightarrow \text{plot}$$

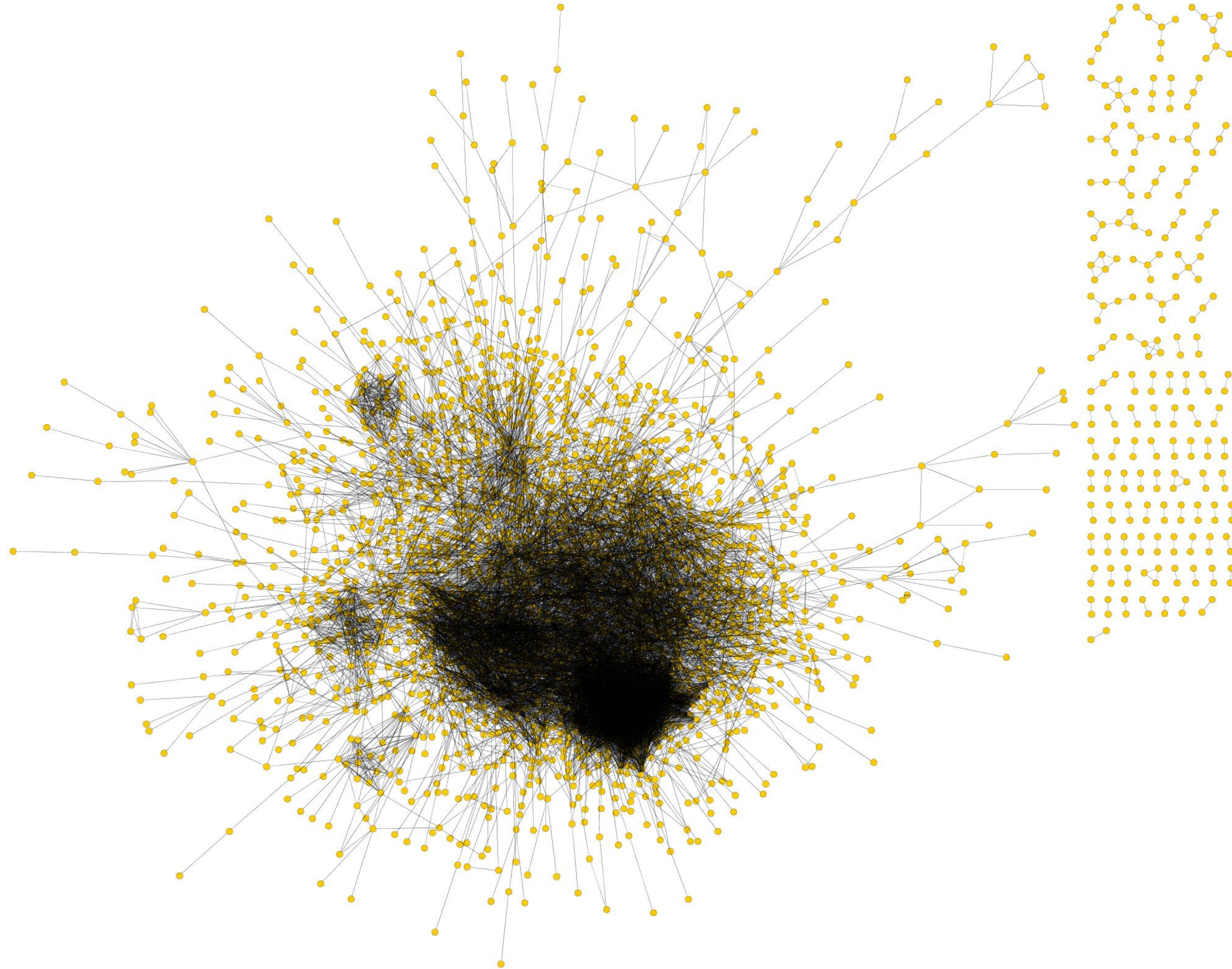


For directed graphs we have separate in- and out-degree distributions.

NETWORK PROPERTIES

DEGREE DISTRIBUTION

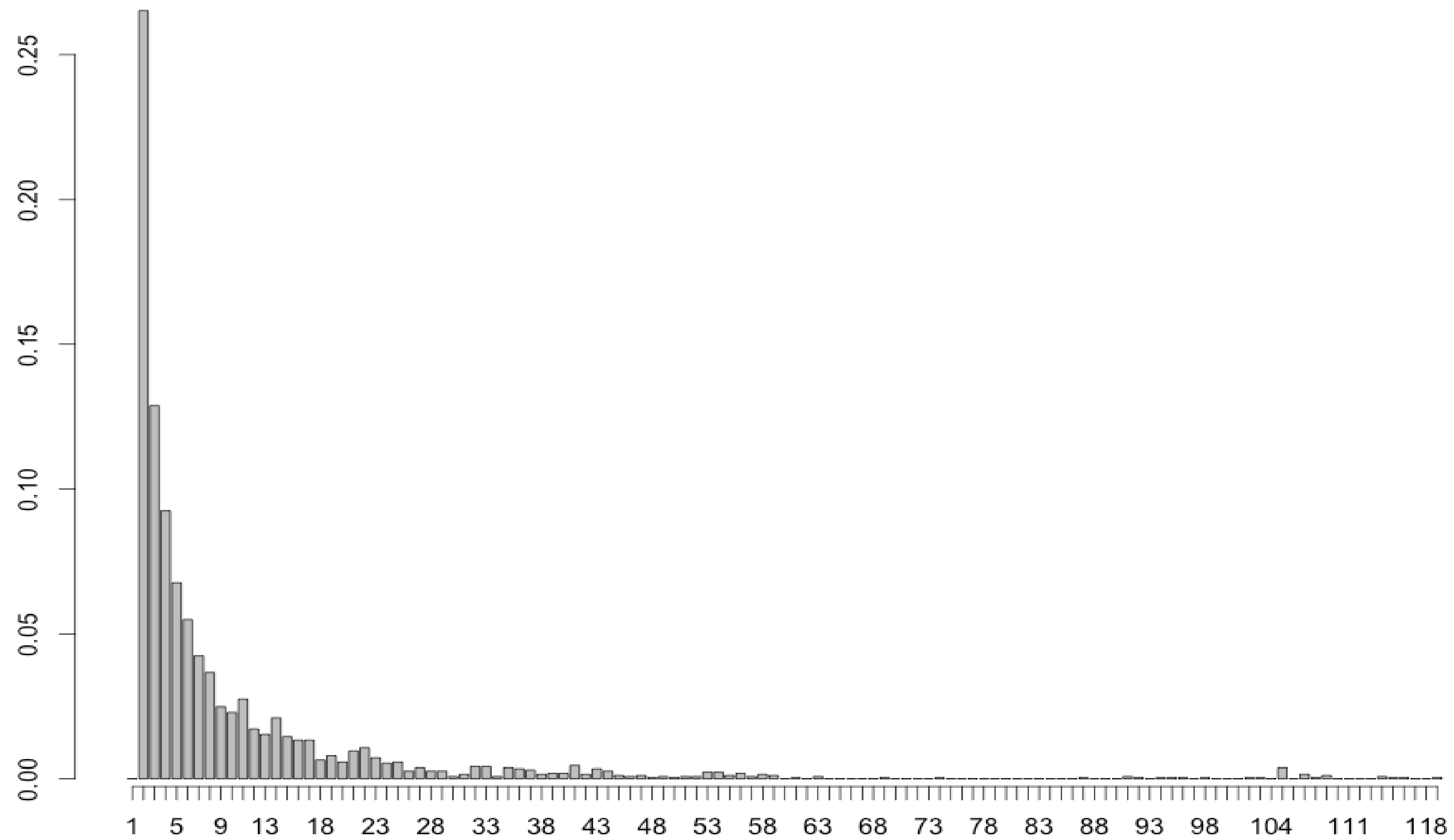
Lecture 1



NETWORK PROPERTIES

DEGREE DISTRIBUTION

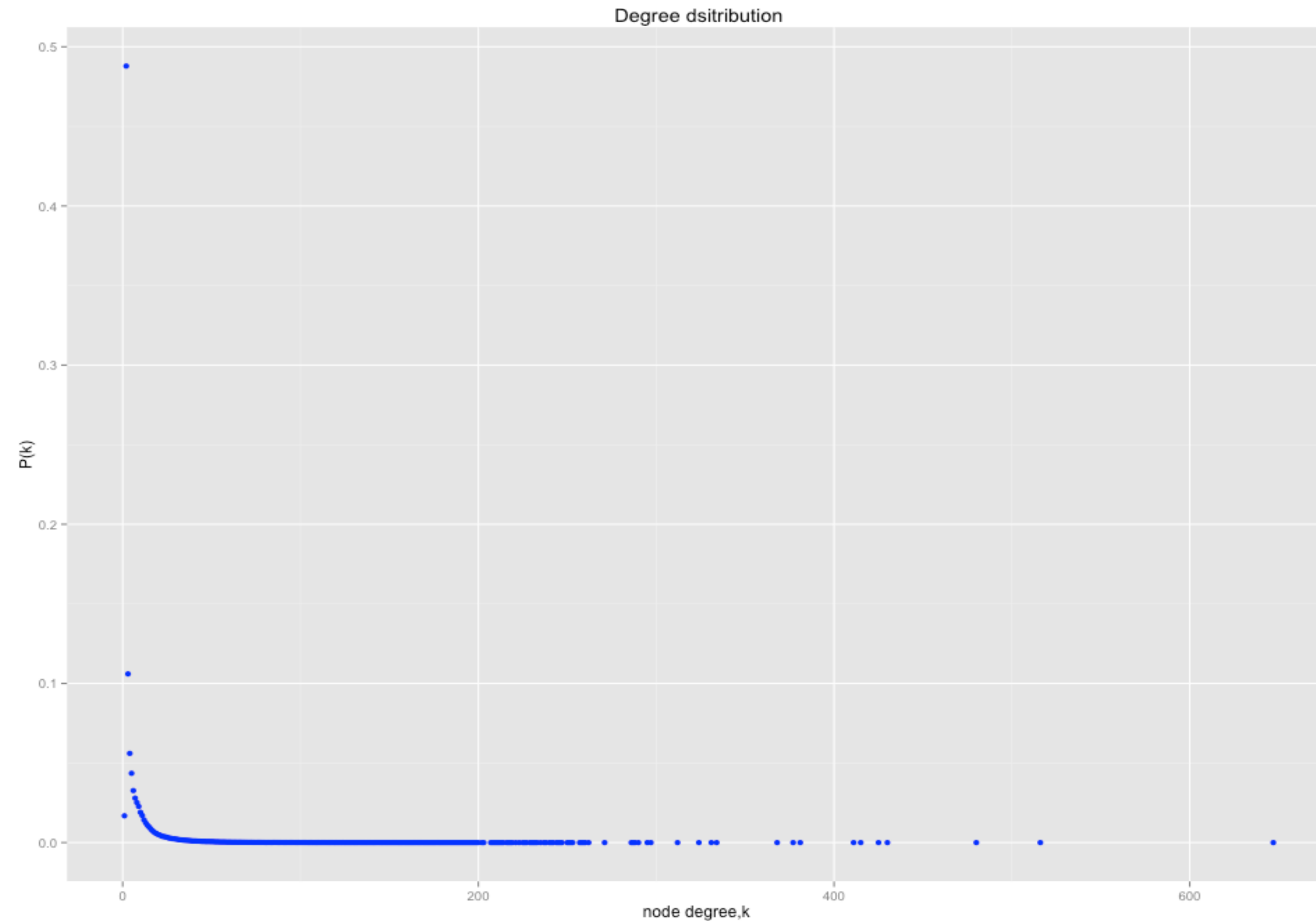
Lecture 1



NETWORK PROPERTIES

DEGREE DISTRIBUTION

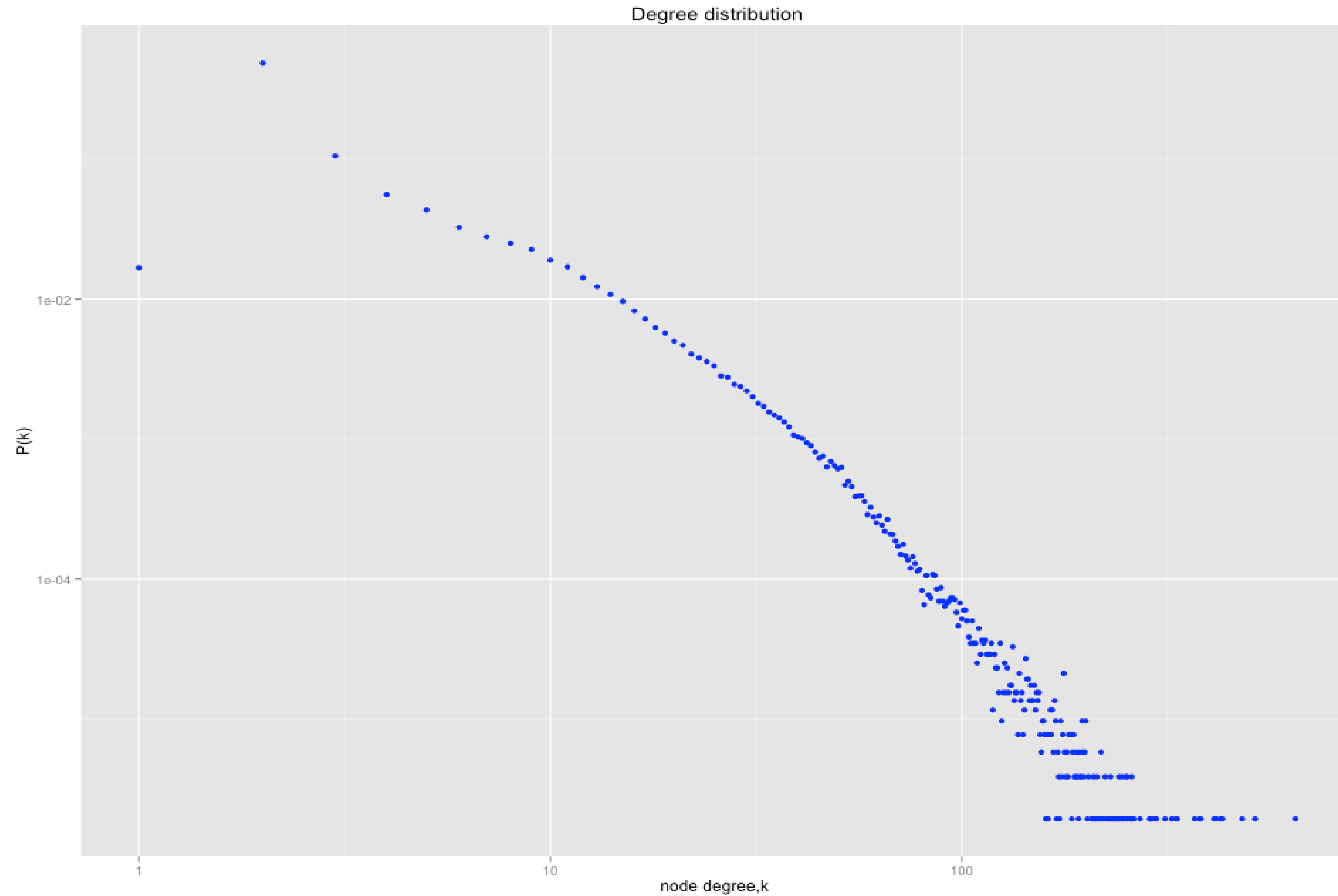
Lecture 1



NETWORK PROPERTIES

DEGREE DISTRIBUTION

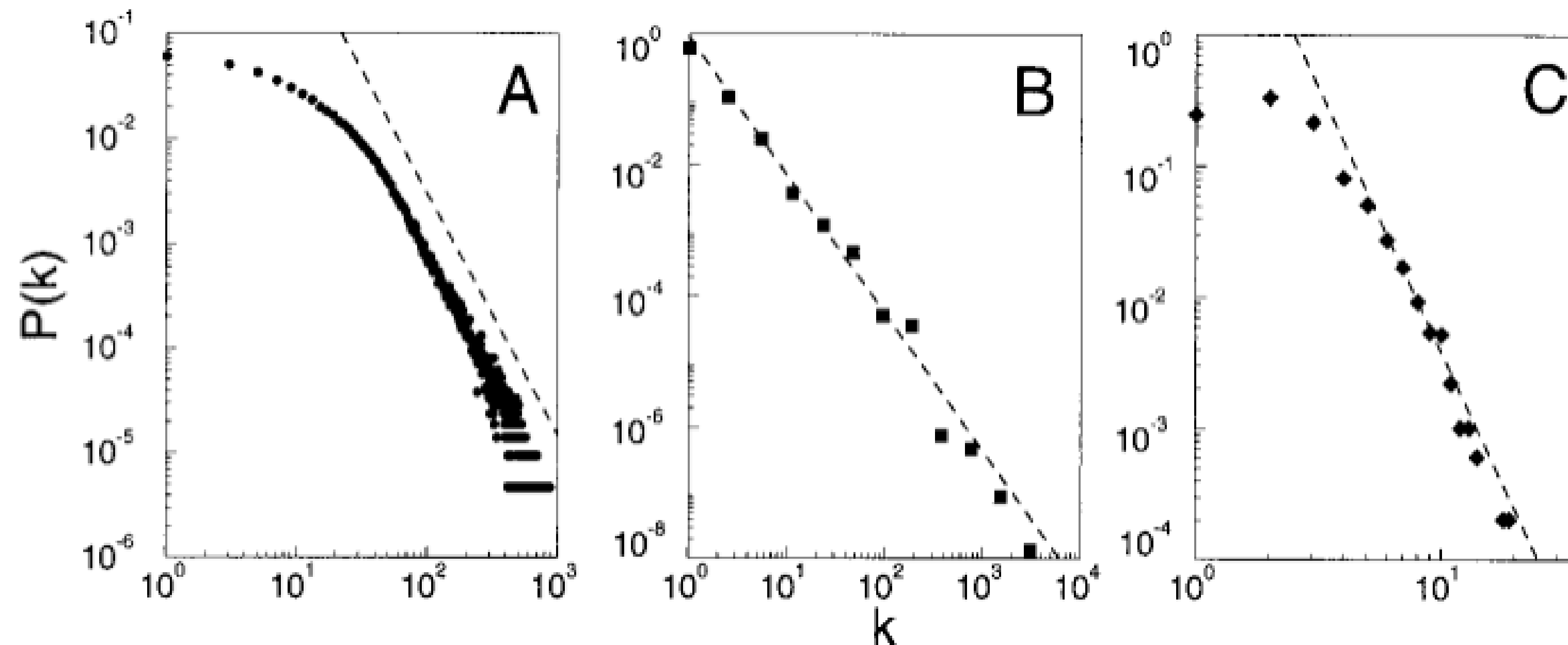
Lecture 1



NETWORK PROPERTIES

Lecture 1

POWER LAW NETWORKS



Actor collaboration graph, $N=212,250$ nodes, $\langle k \rangle = 28.8$, $\gamma = 2.3$

WWW, $N = 325,729$ nodes, $\langle k \rangle = 5.6$, $\gamma = 2.1$

Power grid data, $N = 4941$ nodes, $\langle k \rangle = 5.5$, $\gamma = 4$

NETWORK PROPERTIES

DEGREE DISTRIBUTION

Lecture 1

- Power law distribution

$$P(k) = Ck^{-\gamma} = \frac{1}{k^{\gamma}}C$$

- Log-log coordinates

$$\log P(k) = -\gamma \log k + \log C$$

$$y = -\gamma x + b$$

NETWORK PROPERTIES

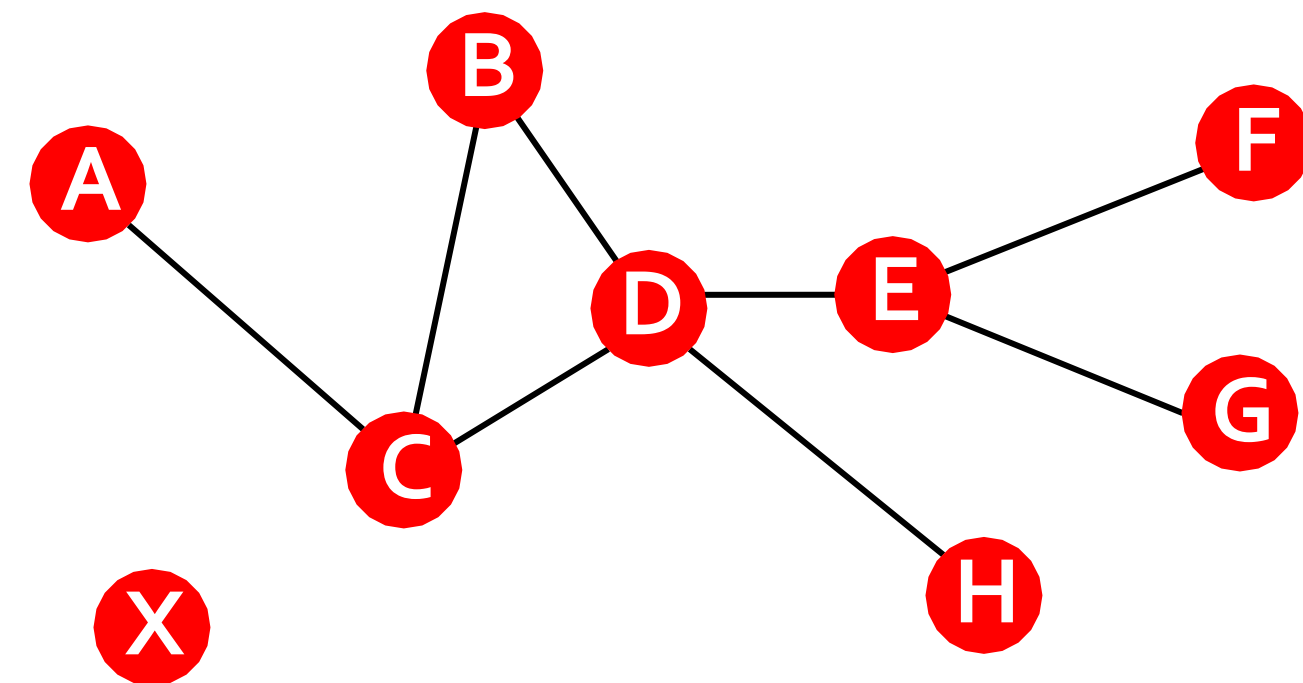
PATH IN GRAPHS

A ***path*** is a sequence of nodes in which each node is linked to the next one

$$P_n = \{i_0, i_1, i_2, \dots, i_n\} \quad P_n = \{(i_0, i_1), (i_1, i_2), (i_2, i_3), \dots, (i_{n-1}, i_n)\}$$

A path can intersect itself and pass through the same edge multiple times

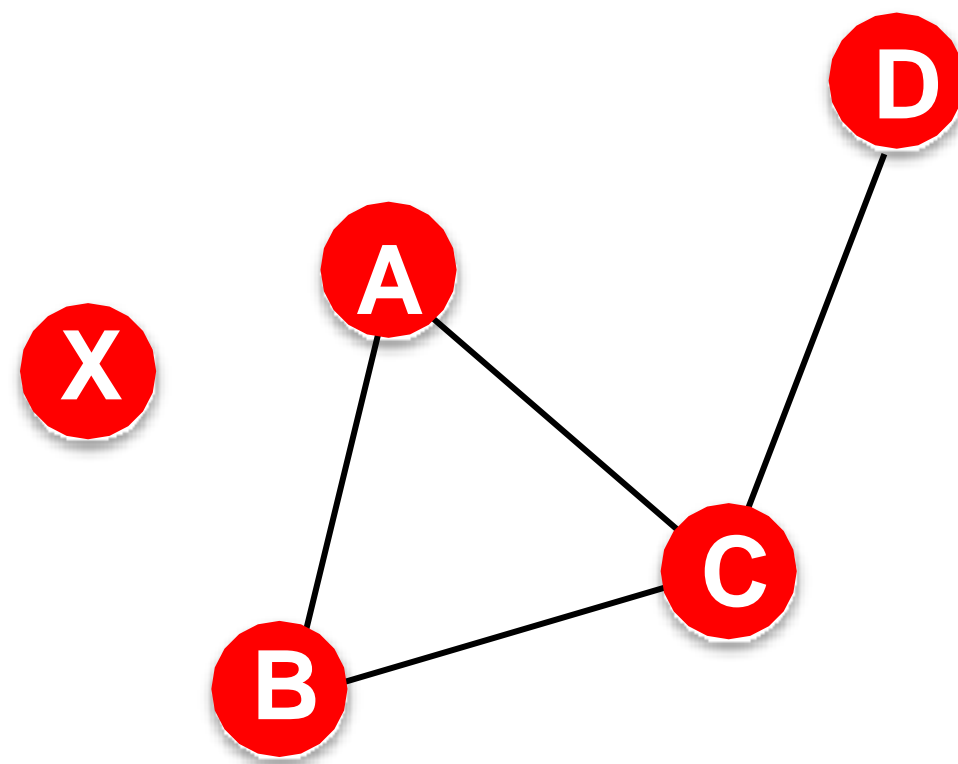
- E.g.: ACBDCDEG



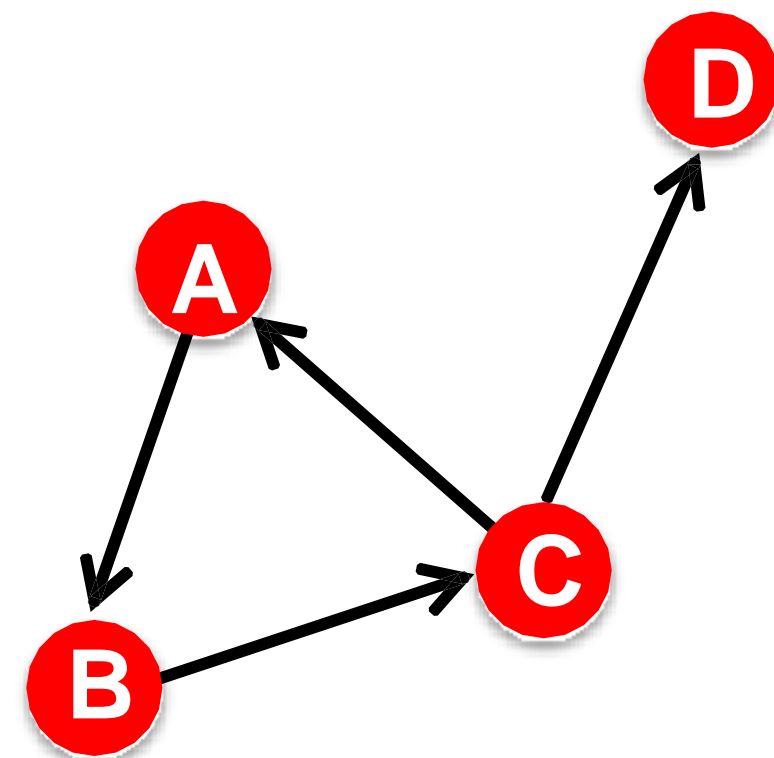
NETWORK PROPERTIES

DISTANCE IN GRAPHS

Lecture 1



$$h_{B,D} = 2$$
$$h_{A,X} = \infty$$



$$h_{B,C} = 1, h_{C,B} = 2$$

Distance (shortest path, geodesic)

between a pair of nodes is defined as the number of edges along the shortest path connecting the nodes

*If the two nodes are not connected, the distance is usually defined as infinite (or zero)

In **directed graphs**, paths need to follow the direction of the arrows

Consequence: Distance is **not symmetric**:

$$h_{B,C} \neq h_{C,B}$$

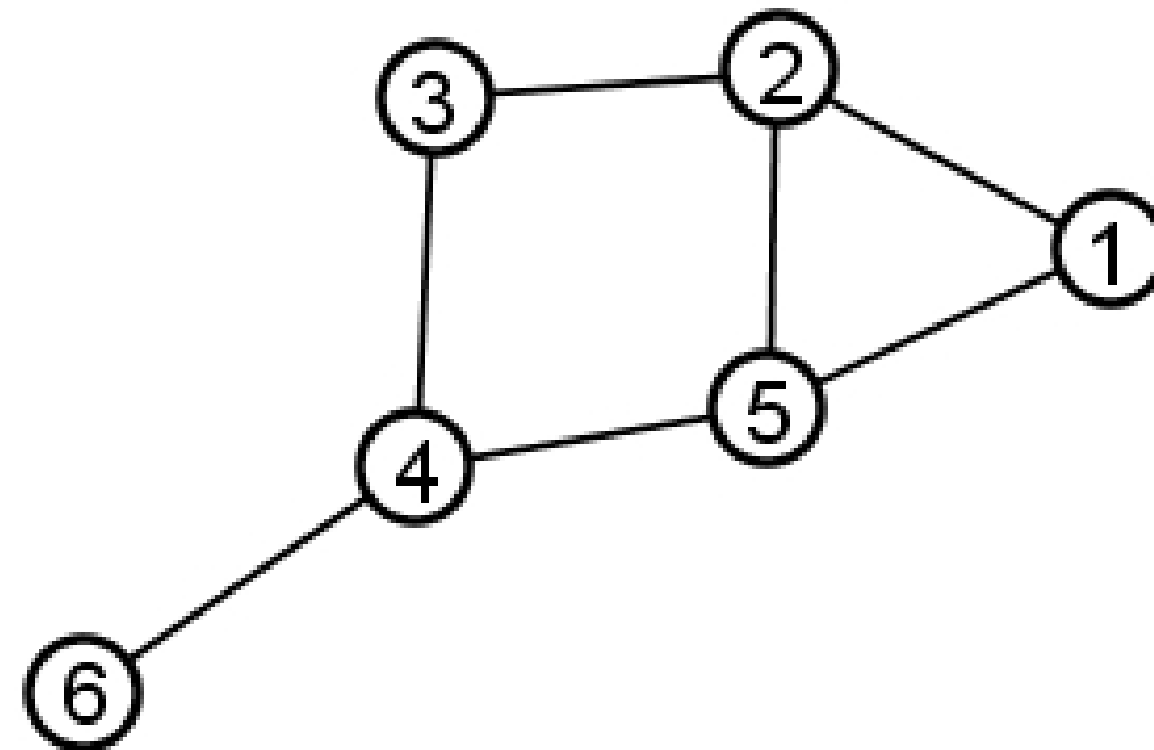
NETWORK PROPERTIES

Lecture 1

NETWORK DIAMETER

- The **distance** $d_{G(v_i; v_j)}$ between two vertices is the number of edges in the shortest path from v_i to v_j
- Graph **diameter** is the largest shortest path: $D = \max_{i,j} d_{G(v_i; v_j)}$
- **Average path length**:

$$\langle L \rangle = \frac{1}{n(n-1)} \sum_{i \neq j} d_G(v_i, v_j)$$



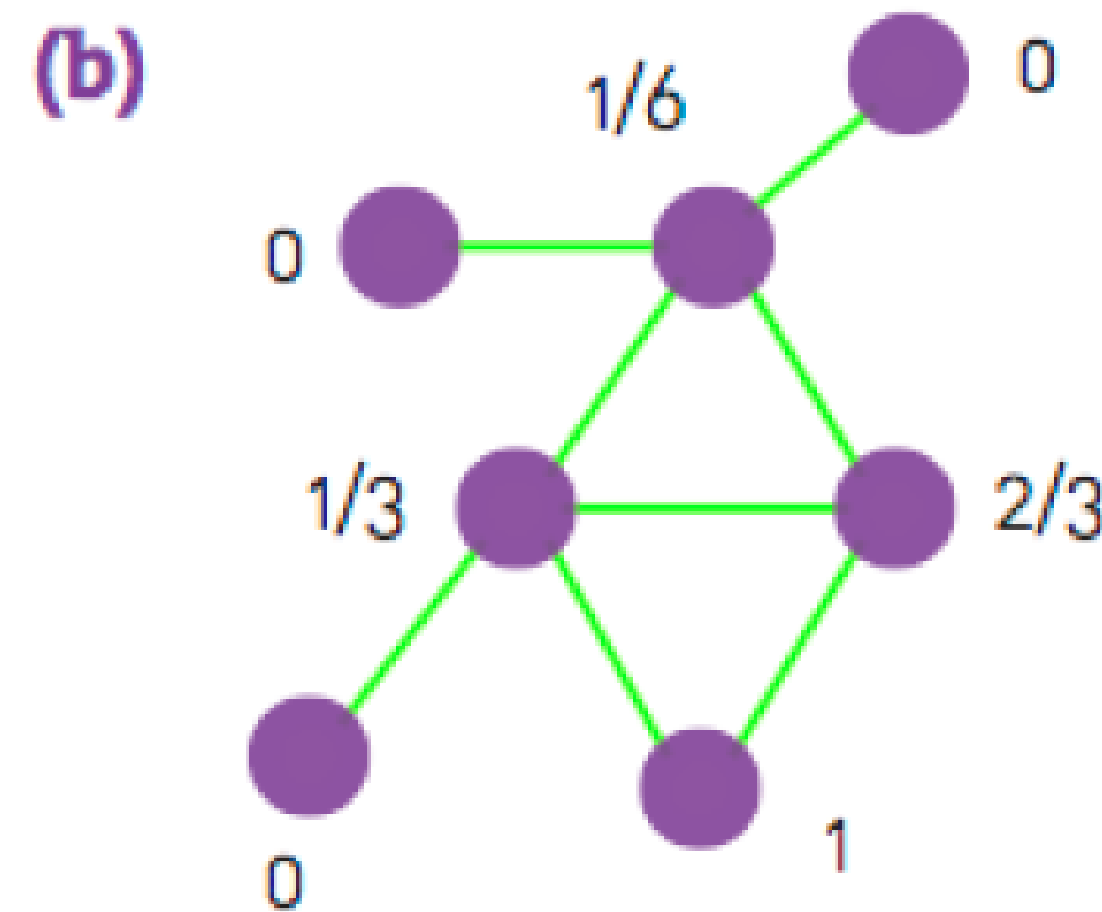
NETWORK PROPERTIES

CLUSTERING COEFFICIENT

Lecture 1

- Global clustering coefficient:

$$C = \frac{3 \times \text{number of triangles}}{\text{number of connected triplets of vertices}}$$



$$\langle C \rangle = \frac{13}{42} \approx 0.310$$

$$C_{\Delta} = \frac{3}{8} = 0.375$$

NETWORK PROPERTIES

CLUSTERING COEFFICIENT

Lecture 1

- Local clustering coefficient (per vertex):

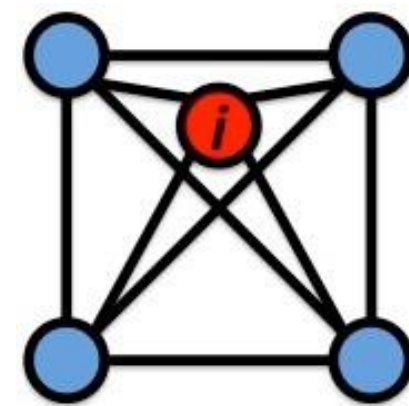
How connected are i 's neighbors to each other?

$C_i \in [0, 1]$ Node i with degree k_i

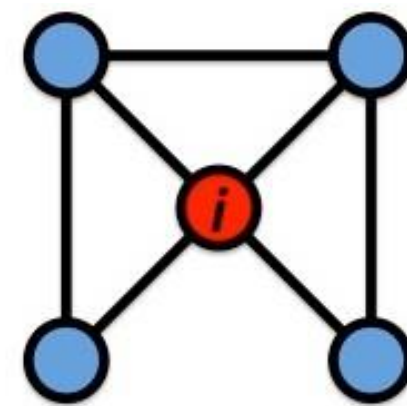
$$C_i = \frac{2e_i}{k_i(k_i - 1)}$$

where e_i is the number of edges between the neighbors of node i

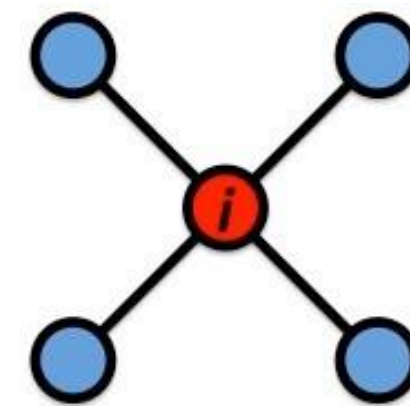
Note $k_i(k_i - 1)$ is
max number of
edges between the
 k_i neighbors



$$C_i = 1$$



$$C_i = 1/2$$



$$C_i = 0$$

Clustering coefficient
is undefined (or defined
to be 0) for nodes with
degree 0 or 1

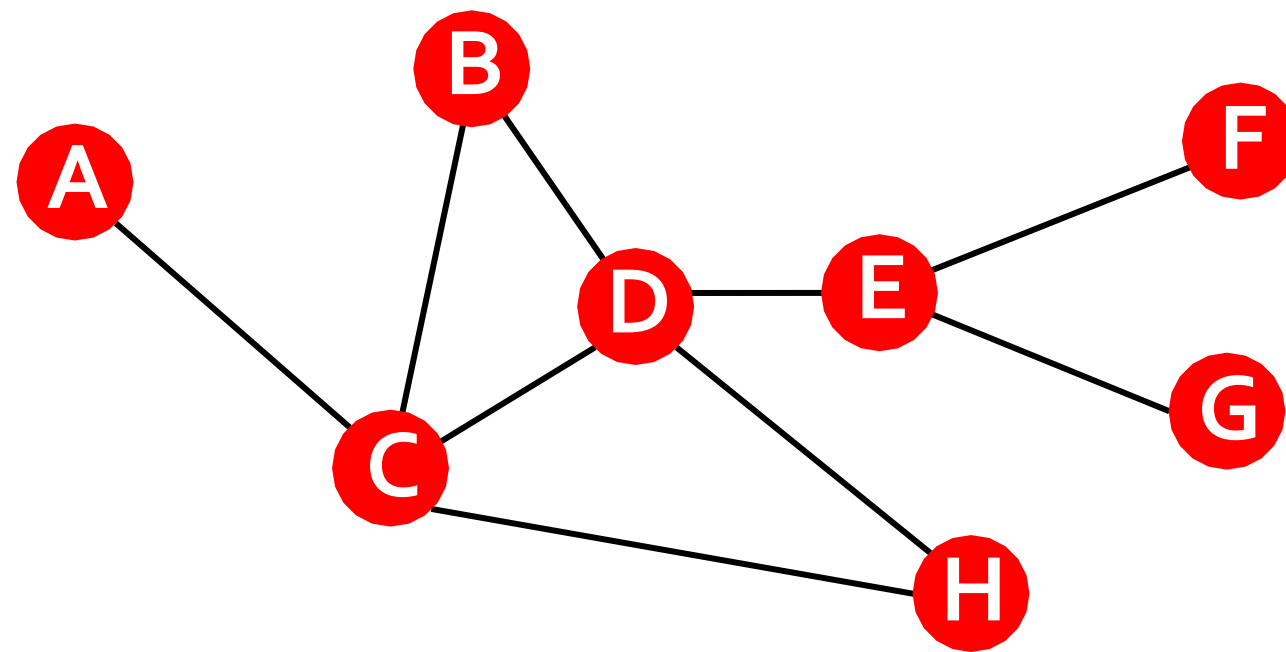
NETWORK PROPERTIES

CLUSTERING COEFFICIENT

Lecture 1

- Average clustering coefficient

$$\bar{C} = \frac{1}{n} \sum_{i=1}^n C_i$$



$$k_B=2, \quad e_B=1, \quad C_B=2/2 = 1$$

$$k_D=4, \quad e_D=2, \quad C_D=4/12 = 1/3$$

$$\text{Avg. clustering: } C=0.33$$

NETWORK PROPERTIES

NETWORK FRAMING

Lecture 1

