

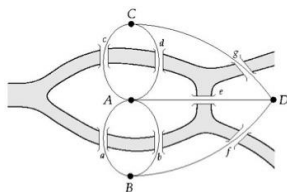
1

THE BRIDGES OF KONIGSBERG



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THE BRIDGES OF KONIGSBERG



Can one walk across the seven bridges and never cross the same bridge twice?

1735: Leonhard Euler's theorem:

- (a) If a graph has nodes of odd degree, there is no path.
- (b) If a graph is connected and has no odd degree nodes, it has at least one path.

Network Science: Graph Theory 101

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THE 7 BRIDGES OF KONIGSBERG

- Can one walk across all seven bridges and never cross the same one twice? Despite many attempts, no one could find such path. The problem remained unsolved until 1736, when Leonard Euler, a Swiss born mathematician, offered a rigorous mathematical proof that such path does not exist.
- Euler's insight was to represent each of the four land areas separated by the river as nodes, distinguishing them with letters A, B, C, and D. Next he connected with lines each piece of land that had a bridge between them. He thus built a {graph}, whose {nodes} were the pieces of land and {links} were the bridges.

Network Science: Graphs

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THE 7 BRIDGES OF KONIGSBERG

- Then Euler made a simple observation: if there is a path crossing all bridges, but never the same bridge twice, then nodes with odd number of links must be either the starting or the end point of this path. Indeed, with an odd number of links you can arrive to a node and have no unused link for you to leave it. Yet, a continuous path that goes through all bridges can have only one starting and one end point. Thus such a path cannot exist on a graph that has more than two nodes with an odd number of links. As the Königsberg graph had three such nodes, B, C, D, each with three links, no path could satisfy the problem.

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THE 7 BRIDGES OF KONIGSBERG

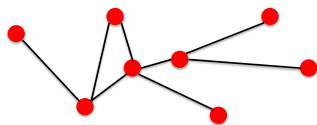
- Today we remember Euler's proof because it was the first time someone solved a mathematical problem by turning it into a graph. In hindsight the proof has two important messages: The first is that some problems become simpler and more treatable if they are represented as a graph. The second is that the existence of the path does not depend on our ingenuity to find it. Rather, it is a property of the graph. Indeed, given the layout of the Königsberg bridges, no matter how smart we are, we will never find the desired path

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COMPONENTS OF A COMPLEX SYSTEM



- **components:** nodes, vertices N
- **interactions:** links, edges L
- **system:** network, graph (N, L)

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NETWORKS OR GRAPHS?

network often refers to real systems

- WWW,
- social network
- metabolic network.

Language: (Network, node, link)

graph: mathematical representation of a network

- web graph,
- social graph (a Facebook term)

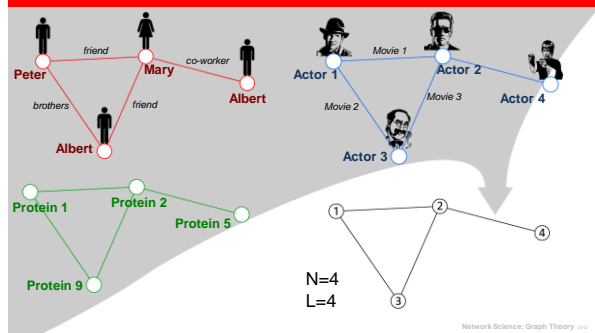
Language: (Graph, vertex, edge)

We will try to make this distinction whenever it is appropriate, but in most cases we will use the two terms interchangeably.

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A COMMON LANGUAGE



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CHOOSING A PROPER REPRESENTATION

The choice of the proper network representation determines our ability to use network theory successfully.

In some cases there is a unique, unambiguous representation. In other cases, the representation is by no means unique.

For example,, the way we assign the links between a group of individuals will determine the nature of the question we can study.

THIS IS MODELLING!

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CHOOSING A PROPER REPRESENTATION



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CHOOSING A PROPER REPRESENTATION

The structure of adolescent romantic and sexual networks

If you connect those that have a romantic and sexual relationship, you will be exploring the **sexual networks**.

Bearman PS, Moody J, Stovel K.
Institute for Social and Economic Research and Policy - Columbia University
<http://researchnews.osu.edu/archive/chainspix.htm>

Network Science: Graph Theory 104

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CHOOSING A PROPER REPRESENTATION

If you connect individuals based on their first name (*all Peters connected to each other*), you will be exploring what?

It is a network, nevertheless.

Network Science: Graph Theory 2019

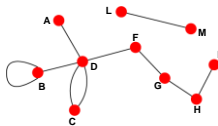
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UNDIRECTED VS. DIRECTED NETWORKS

Undirected

Links: undirected (symmetrical)

Graph:

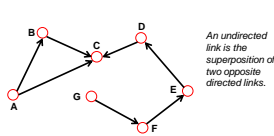


Undirected links :
coauthorship links
Actor network
protein interactions

Directed

Links: directed (*arcs*).

Digraph = directed graph:



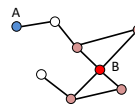
Directed links :
URLs on the www
phone calls
metabolic reactions

Network Science: Graph Theory 2020

NODE DEGREES

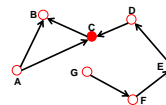
Node degree: the number of links connected to the node.

Undirected



$$k_A = 1 \quad k_B = 4$$

Directed



In *directed networks* we can define an **in-degree** and **out-degree**.

The (total) degree is the sum of in- and out-degree.

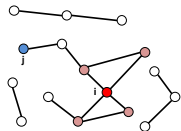
$$k_c^{in} = 2 \quad k_c^{out} = 1 \quad k_c = 3$$

Source: a node with $k^{\text{in}} = 0$; **Sink:** a node with $k^{\text{out}} = 0$.

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AVERAGE DEGREE

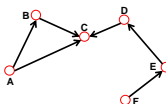
Undirected



$$\langle k \rangle \equiv \frac{1}{N} \sum_{i=1}^N k_i \quad \langle k \rangle \circ \frac{2L}{N}$$

N – the number of nodes in the graph

Directed



$$\langle k^{in} \rangle = \frac{1}{N} \sum_{i=1}^N k_i^{in}, \quad \langle k^{out} \rangle = \frac{1}{N} \sum_{i=1}^N k_i^{out}, \quad \langle k^{in} \rangle = \langle k^{out} \rangle$$

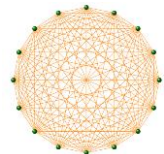
$$\langle k \rangle \propto \frac{L}{N}$$

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COMPLETE GRAPH

The maximum number of links a network of N nodes can have is: $L_{\max} = \frac{N(N-1)}{2}$

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



A graph with degree $L=L_{\max}$ is called a **complete graph**, and its average degree is $\langle k \rangle = N-1$

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REAL NETWORKS ARE SPARSE

Most networks observed in real systems are sparse:

$$L \ll L_{\max} \\ \text{or} \\ \langle k \rangle \ll N-1.$$

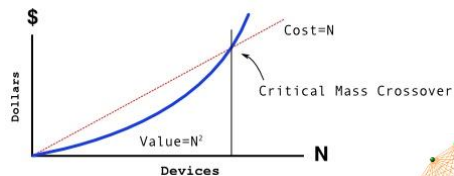
WWW (ND Sample):	N=325,729;	L=1.4 10^6	$L_{\max}=10^{12}$	$\langle k \rangle=4.51$
Protein (<i>S. Cerevisiae</i>):	N= 1,870;	L=4,470	$L_{\max}=10^7$	$\langle k \rangle=2.39$
Coauthorship (Math):	N= 70,975;	L=2 10^5	$L_{\max}=3 \cdot 10^{10}$	$\langle k \rangle=3.9$
Movie Actors:	N=212,250;	L=6 10^6	$L_{\max}=1.8 \cdot 10^{13}$	$\langle k \rangle=28.78$

(Source: Albert, Barabasi, RMP2002)

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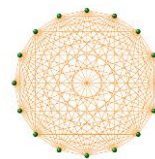
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METCALFE'S LAW



The maximum number of links a network of N nodes can have is:

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



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ADJACENCY MATRIX



$A_{ij}=1$ if there is a link between node i and j

$A_{ij}=0$ if nodes i and j are not connected to each other.

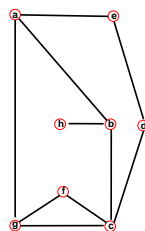
$$A_{ij} = \begin{pmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \end{pmatrix} \quad A_{ij} = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \end{pmatrix}$$

Note that for a directed graph (right) the matrix is not symmetric.

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ADJACENCY MATRIX



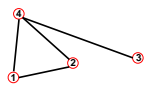
	a	b	c	d	e	f	g	h
a	0	1	0	0	1	0	1	0
b	1	0	1	0	0	0	0	1
c	0	1	0	1	0	1	1	0
d	0	0	1	0	1	0	0	0
e	1	0	0	1	0	0	0	0
f	0	0	1	0	0	0	1	0
g	1	0	1	0	0	1	0	0
h	0	1	0	0	0	0	0	0

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ADJACENCY MATRIX AND NODE DEGREES

Undirected



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

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

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

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

$$L = \frac{1}{2} \sum_{i=1}^N k_i = \frac{1}{2} \sum_{j=1}^N k_j$$

Directed



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

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

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

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

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

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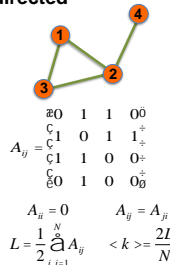


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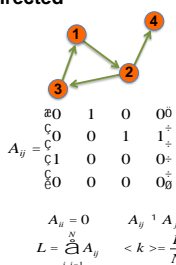
GRAPHOLOGY 1

Undirected



Actor network, protein-protein interactions

Directed

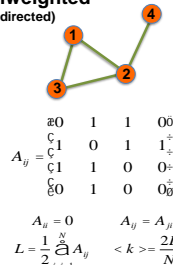


WWW, citation networks

Network Science: Graph Theory 100

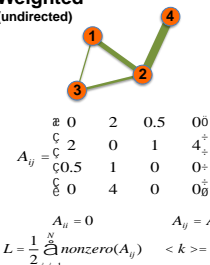
GRAPHOLOGY 2

Unweighted (undirected)



protein-protein interactions, www

Weighted (undirected)



Call Graph, metabolic networks

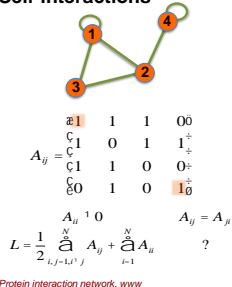
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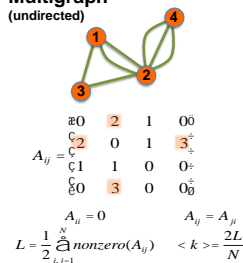
GRAPHOLOGY 3

Self-interactions



Protein interaction network, www

Multigraph (undirected)



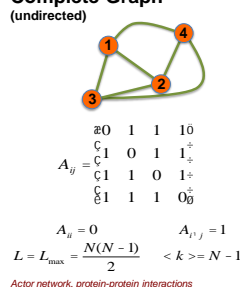
Social networks, collaboration networks

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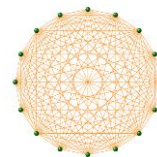
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GRAPHOLOGY 4

Complete Graph (undirected)



Actor network, protein-protein interactions



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GRAPHOLOGY: Real networks can have multiple characteristics

WWW > directed multigraph with self-interactions

Protein Interactions > undirected unweighted with self-interactions

Collaboration network > undirected multigraph or weighted.

Mobile phone calls > directed, weighted.

Facebook Friendship links > undirected, unweighted.

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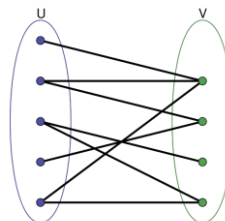
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BIPARTITE GRAPHS

bipartite graph (or **bigraph**) 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:

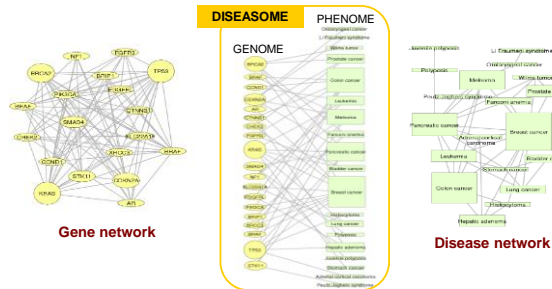
Hollywood actor network
Collaboration networks
Disease network (diseasome)



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GENE NETWORK – DISEASE NETWORK

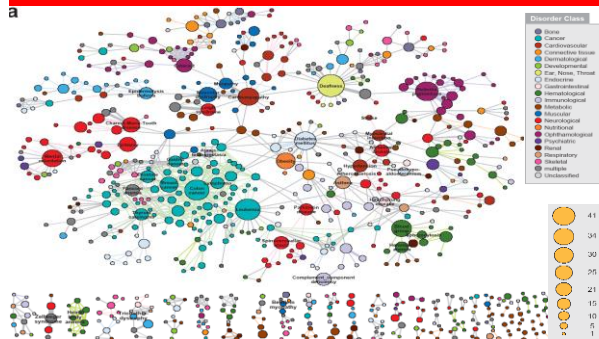


Goh, Cusick, Valle, Childs, Vidal & Barabási, PNAS (2007)

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HUMAN DISEASE NETWORK



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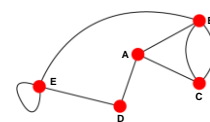
PATHS

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

P_{i_0, i_n} of length n between nodes i_0 and i_n is an ordered collection of $n+1$ nodes and n links

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

- A path can intersect itself and pass through the same link repeatedly. Each time a link is crossed, it is counted separately
- A legitimate path on the graph on the right:
ABCBCADEEBA
- In a directed network, the path can follow only the direction of an arrow.

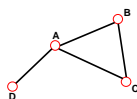


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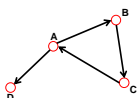
DISTANCE IN A GRAPH

Shortest Path, Geodesic Path



The *distance* (*shortest path*, *geodesic path*) between two nodes is defined as the number of edges along the shortest path connecting them.

*If the two nodes are disconnected, the distance is infinity.



In *directed graphs* each path needs to follow the direction of the arrows.

Thus in a digraph the distance from node A to B (on an AB path) is generally different from the distance from node B to A (on a BA path).

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NUMBER OF PATHS BETWEEN TWO NODES

Adjacency Matrix

N_{ij} , number of paths between any two nodes i and j :

Length $n=1$: If there is a link between i and j , then $A_{ij}=1$ and $A_{ij}=0$ otherwise.

Length $n=2$: If there is a path of length two between i and j , then $A_{ik}A_{kj}=1$, and $A_{ik}A_{kj}=0$ otherwise.

The number of paths of length 2:

$$N_{ij}^{(2)} = \sum_k A_{ik}A_{kj} = [A^2]_{ij}$$

Length n : In general, if there is a path of length n between i and j , then $A_{ik} \dots A_{kj}=1$ and $A_{ik} \dots A_{kj}=0$ otherwise.

The number of paths of length n between i and j is

$$N_{ij}^{(n)} = [A^n]_{ij}$$

*holds for both directed and undirected networks.

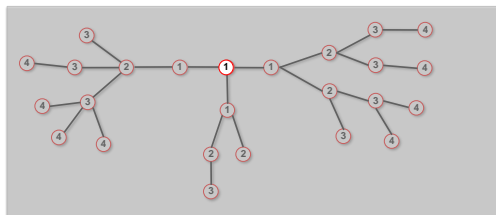
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FINDING DISTANCES: BREADTH FIRST SEARCH

Distance between node 1 and node 4:

1. Start at 1.



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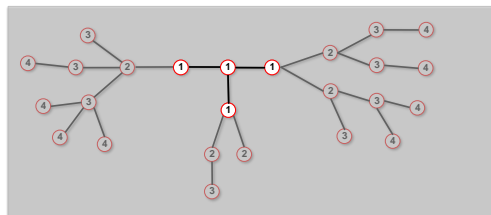
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FINDING DISTANCES: BREADTH FIRST SEARCH

Distance between node 1 and node 4:

1. Start at 1.

2. Find the nodes adjacent to 1. Mark them as at distance 1. Put them in a queue.



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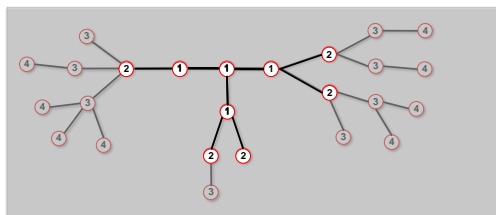
FINDING DISTANCES: BREADTH FIRST SEARCH

Distance between node 1 and node 4:

1. Start at 1.

2. Find the nodes adjacent to 1. Mark them as at distance 1. Put them in a queue.

3. Take the first node out of the queue. Find the unmarked nodes adjacent to it in the graph. Mark them with the label of 2. Put them in the queue.



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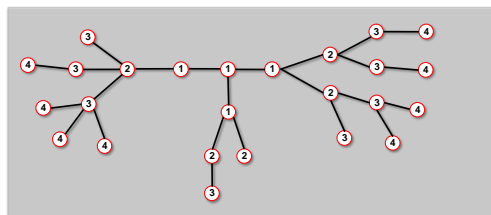
40

FINDING DISTANCES: BREADTH FIRST SEARCH

Distance between node 1 and node 4:

1. Repeat until you find node 4 or there are no more nodes in the queue.

2. The distance between 1 and 4 is the label of 4 or, if 4 does not have a label, infinity.



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NETWORK DIAMETER AND AVERAGE DISTANCE

Diameter: d_{\max} the maximum distance between any pair of nodes in the graph.Average path length/distance, $\langle d \rangle$, for a **connected graph**:where d_{ij} is the distance from node i to node j

$$\langle d \rangle \circ \frac{1}{2L_{\max}} \sum_{i,j} d_{ij}$$

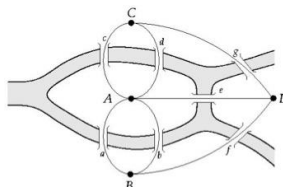
In an **undirected graph** $d_{ij} = d_{ji}$, so we only need to count them once:

$$\langle d \rangle \circ \frac{1}{L_{\max}} \sum_{i,j} d_{ij}$$

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THE BRIDGES OF KONIGSBERG

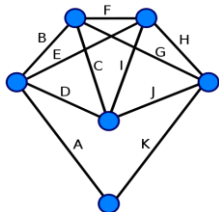


Can one walk across the seven bridges and never cross the same bridge twice?

Euler PATH or CIRCUIT: return to the starting point by traveling each link of the graph once and only once.

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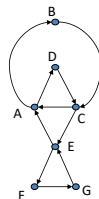
EULERIAN GRAPH: it has an Eulerian path

Every vertex of this graph has an even degree, therefore this is an Eulerian graph.
Following the edges in alphabetical order gives an Eulerian circuit/cycle.

http://en.wikipedia.org/wiki/Euler_circuit

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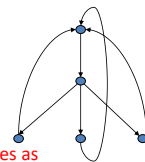
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EULER CIRCUITS IN DIRECTED GRAPHS

If a digraph is strongly connected and the in-degree of each node is equal to its out-degree, then there is an Euler circuit

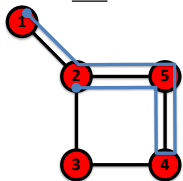
Otherwise there is no Euler circuit.

→ a circuit we need to enter each node as many times as we leave it.

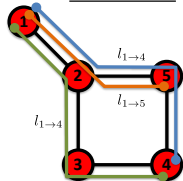


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PATHOLOGY: summaryPath

A sequence of nodes such that each node is connected to the next node along the path by a link.

Shortest Path

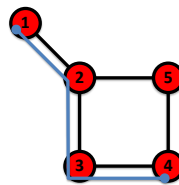
$$l_{1 \rightarrow 4} = 3$$

$$l_{1 \rightarrow 5} = 2$$

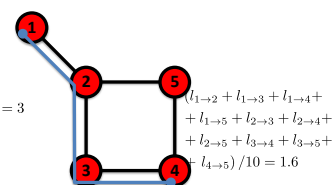
The path with the shortest length between two nodes (distance).

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PATHOLOGY: summaryDiameter

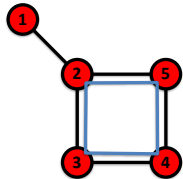
The longest shortest path in a graph

Average Path Length

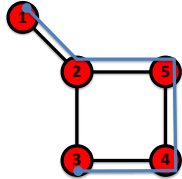
The average of the shortest paths for all pairs of nodes.

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PATHOLOGY: summaryCycle

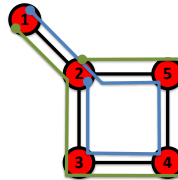
A path with the same start and end node.

Self-avoiding Path

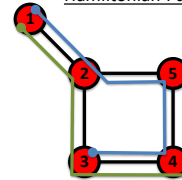
A path that does not intersect itself.

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PATHOLOGY: summaryEulerian Path

A path that traverses each link exactly once.

Hamiltonian Path

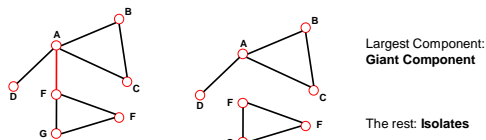
A path that visits each node exactly once.

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CONNECTIVITY OF UNDIRECTED GRAPHS

Connected (undirected) graph: any two vertices can be joined by a path.
A disconnected graph is made up of two or more connected components.



Bridge: if we erase it, the graph becomes disconnected.

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CONNECTIVITY OF UNDIRECTED GRAPHS

Adjacency Matrix

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:

$$A = \begin{pmatrix} \text{Red Square} & 0 & \cdots \\ 0 & \text{Red Square} & \cdots \\ \vdots & \vdots & \ddots \end{pmatrix}$$

Figure after Newman, 2010

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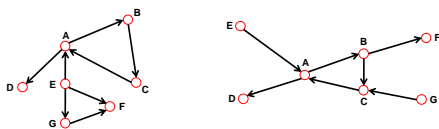
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CONNECTIVITY OF DIRECTED GRAPHS

Strongly connected graph: has a path from each node to every other node **and vice versa** (e.g. AB path and BA path).

Weakly connected directed graph: it is connected if we disregard the edge directions.

Strongly connected components can be identified, but not every node is part of a nontrivial strongly connected component.



In-component: nodes that can reach the scc.

Out-component: nodes that can be reached from the scc.

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THREE CENTRAL QUANTITIES IN NETWORK SCIENCE

Degree distribution

 p_k

Average path length

 $\langle d \rangle$

Clustering coefficient

 C

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DEGREE

DISTRIBUTION

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STATISTICS REMINDER

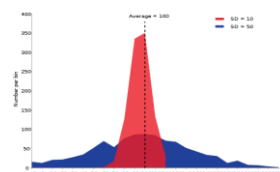
We have a sample of values x_1, \dots, x_N

Distribution of x (a.k.a. PDF): probability that a randomly chosen value is x

$$P(x) = (\# \text{ values } x) / N$$

$$\sum_i P(x_i) = 1 \text{ always!}$$

Histograms >>>



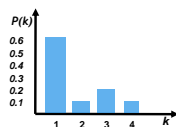
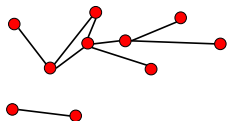
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DEGREE DISTRIBUTION

Degree distribution $P(k)$: probability that a randomly chosen vertex has degree k

N_k = # nodes with degree k
 $P(k) = N_k / N$ ● plot



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DEGREE DISTRIBUTION

discrete representation: p_k is the probability that a node has degree k .

continuum description: $p(k)$ is the pdf of the degrees, where

$$\int_{k_1}^{k_2} p(k) dk$$

represents the probability that a node's degree is between k_1 and k_2 .

Normalization condition:

$$\sum_{k=0}^{\infty} p_k = 1 \quad \int_{K_{\min}}^{\infty} p(k) dk = 1$$

where K_{\min} is the minimal degree in the network.

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CLUSTERING COEFFICIENT

* **Clustering coefficient:**

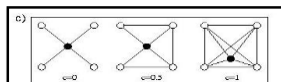
what portion of your neighbors are connected?

* Node i with degree k_i

* C_i in $[0,1]$

* e_i Number of edges between neighbors

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



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THREE CENTRAL QUANTITIES IN NETWORK SCIENCE

Degree distribution: $P(k)$

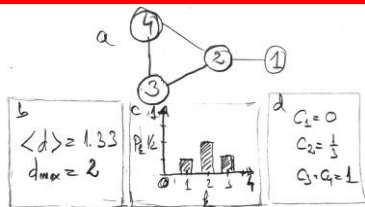
Path length: $\langle d \rangle$

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

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THREE CENTRAL QUANTITIES IN NETWORK SCIENCE



A. Degree distribution:

p_k

B. Path length:

$\langle d \rangle$

C. Clustering coefficient:

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

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