Graph theory basics

Introduction to Network Science Carlos Castillo Topic 02



Sources

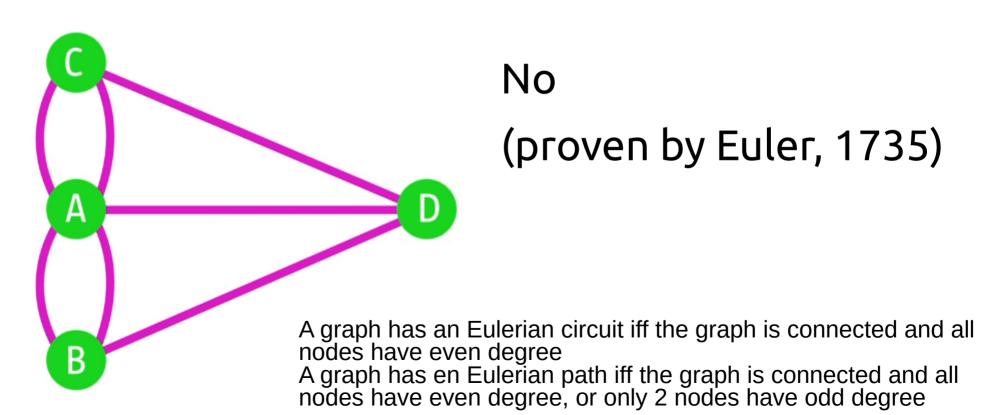
- Albert László Barabási: Network Science.
 Cambridge University Press, 2016.
 - Follows almost section-by-section chapter 02
- URLs cited in the footer of specific slides

The seven bridges of Königsberg



http://networksciencebook.com/images/ch-02/video-2-1.m4v

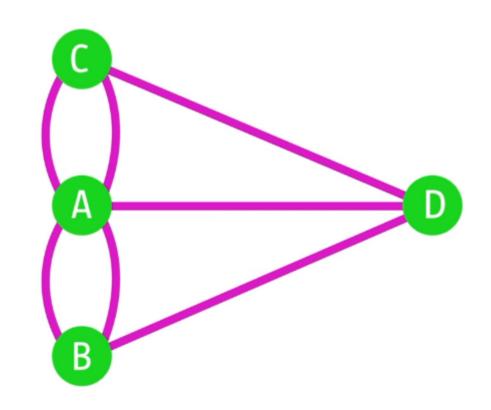
Can one walk across the 7 bridges w/o crossing the same path twice?



Basic concepts

Notation for a graph

- G = (V,E)
 - V: nodes or vertices
 - E: links or edges
- |V| = N size of graph
- |E| = L number of links



Typical notation variations

- You may find that G is denoted by (N, A), this is typical of directed graphs
- You may find that |V| is denoted by n, |E| is denoted by m

Directed vs undirected graphs

- In a directed graph, also known as "digraph", E is a symmetric relation
 - $(u,v) \in E \Rightarrow (v,u) \in E$
- In an undirected graph, E is not symmetric

Example graphs we will use

Network	[V]	E
Zachary's Karate Club (karate.gml)	34	78
Les Misérables (lesmiserables.gml)	77	254
E-mail exchanges (email-eu-core.csv)	868	25K
US companies ownership	1351	6721
Marvel comics (hero-network.csv)	6K	570K

Degree

- Node i has degree k_i
 - This is the number of links incident on this node
 - The total number of links L is given by $L=rac{1}{2}\sum_{i=1}^{N}k_{i}$
- Average degree $\langle k \rangle = \frac{1}{N} \sum_{i=1}^{N} k_i = \frac{2L}{N}$

In directed networks

- We distinguish in-degree from out-degree
 - Incoming and outgoing links, respectively
- Degree is the sum of both $k_i = k_i^{\rm in} + k_i^{\rm out}$
- Counting total number of links:

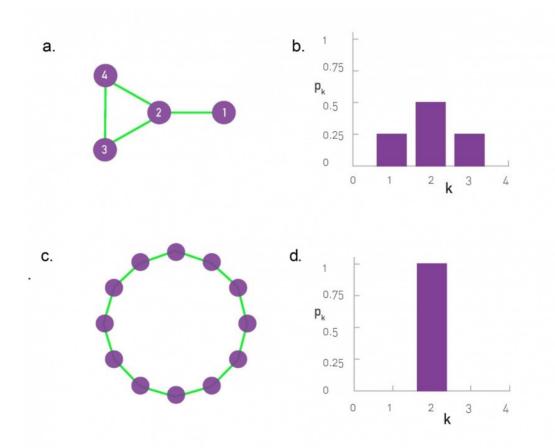
$$L = \sum_{i=1}^{N} k_i^{\text{in}} = \sum_{i=1}^{N} k_i^{\text{out}}$$

Degree distribution

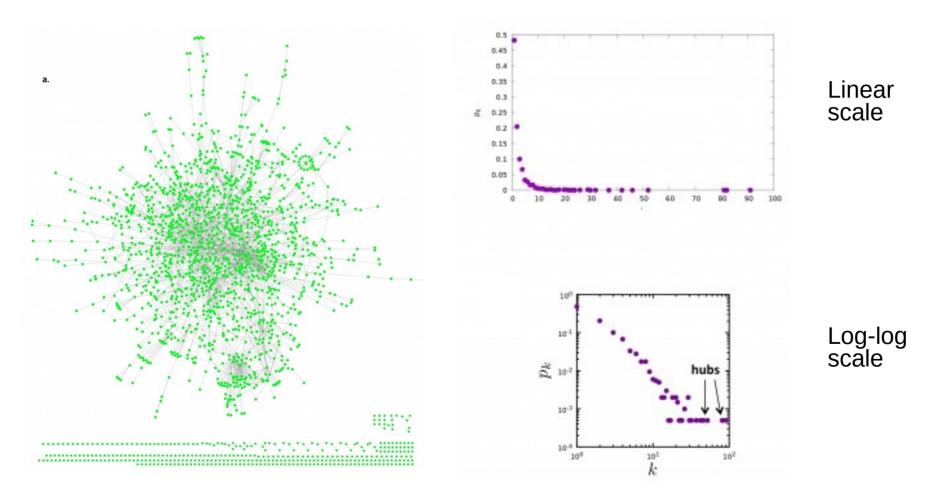
- If there are N_k nodes with degree k
- The degree distribution is given by $p_k = \frac{N_k}{N}$

• The average degree is then $\langle k \rangle = \sum_{k=0}^{\infty} k p_k$

Degree distribution; two toy graphs

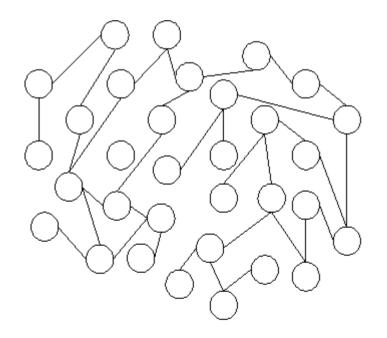


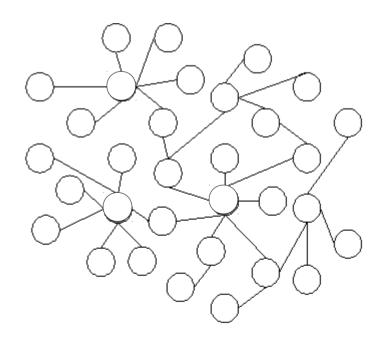
Degree distribution; real graph



Exercise

Draw the degree distribution of these graphs



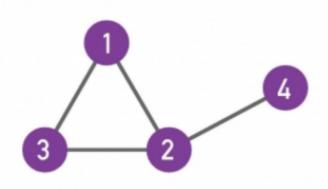


Adjacency matrix

What is an adjacency matrix

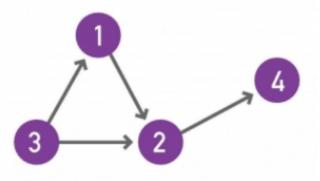
- A is the adjacency matrix of G = (V, E) iff:
 - A has |V| rows and |V| columns
 - $-A_{ii} = 1$ if $(i,j) \in E$
 - A_{ii} = 0 if (i,j) ∉ E

Examples



Undirected graph

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



Directed graph

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

Quick exercise

• In terms of A, what is the expression for:

$$k_i^{\text{in}} = k_i^{\text{out}} =$$

Some "graphology" ...

- G is undirected ⇔ A is symmetric
- G has a self-loop
 ⇔ A has a non-zero element in the diagonal
- G is complete

 A_{ii} ≠ 0 (except if i=j)

Real networks are sparse

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

• Most real networks are sparse, i.e., $L \ll L_{\rm max}$

How sparse are some networks?

Network	[V]	E	Max E
Zachary's Karate Club	34	78	561
Les Misérables	77	254	2962
E-mail exchanges	868	25K	376K
US companies ownership	1351	6721	911K
Marvel comics	6K	570K	17M

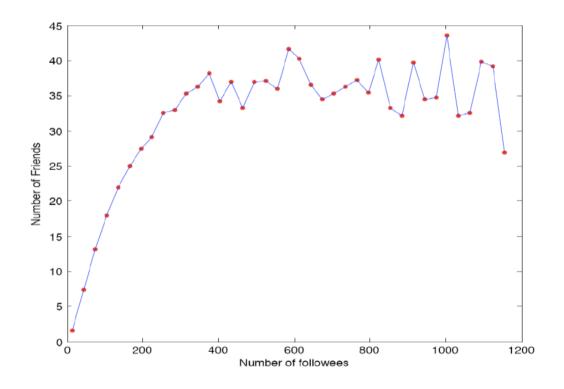
Example: protein interaction network

(N=2K, L=3K)

Why are networks sparse?

- Different mechanisms, think about it from the node perspective:
 - How many items could the node be connected to
 - Would it be **realistic** to connect to a large fraction of them?
- In social networks, Dunbar's number (\approx 150)

Example: actual friends in Twitter vs people you follow in Twitter



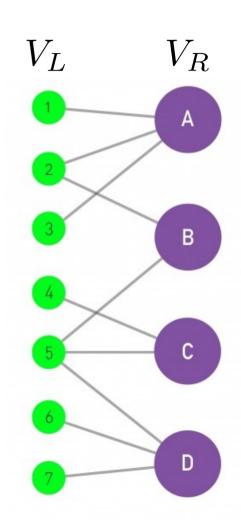
Weighted networks

- In weighted networks, instead of $A_{ij} \in [0,1]$
- We have that $A_{ij} \in \mathbb{R}$
- Weights may represent different tie strengths

Bipartite networks

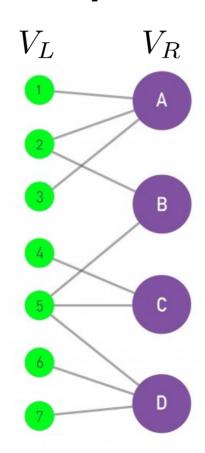
A bipartite graph is a graph
 G = (V,E) such that

$$V = V_L \cup V_R, V_L \cap V_R = \emptyset, E \subseteq V_L \times V_R$$



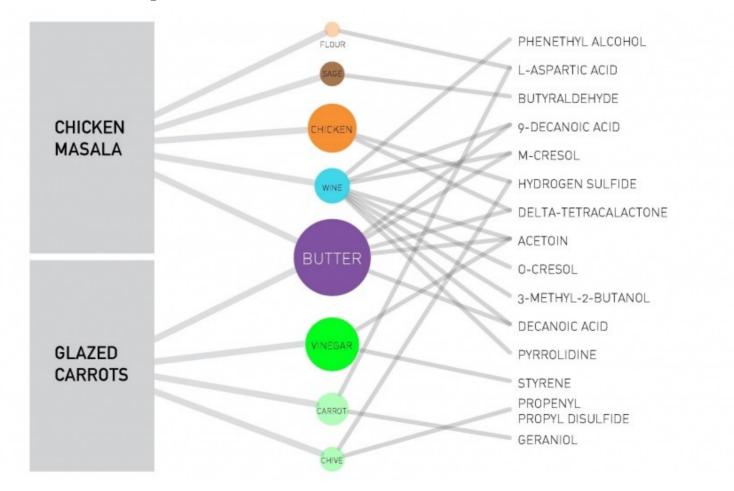
Projecting a bipartite network

Left projection:
graph where nodes
are 1, 2, ..., 7 and
nodes are connected
if they share a
neighbor



Right projection:
graph where nodes
are A, B, ..., D and
nodes are connected
if they share a
neighbor

Tripartite network



Clique and Bi-partite clique

- A clique is a complete graph: $E = (V \times V)$
- An **n-clique** is a complete graph of n nodes
- A **bi-partite clique** is such that

$$V = V_1 \cup V_2, V_1 \cap V_2 = \emptyset, E = (V_1 \times V_2)$$

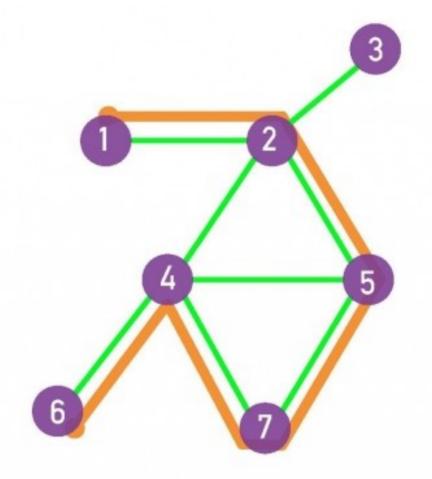
• A (n₁, n₂)-clique is a bipartite clique such that

$$|V_1| = n_1, |V_2| = n_2$$

Paths and distances

Paths

- A path is a sequence of edges from E
- The destination of each edge is the origin of the next edge
- The length of the path is the number of edges on it
- Example: a path marked in orange, having length 5

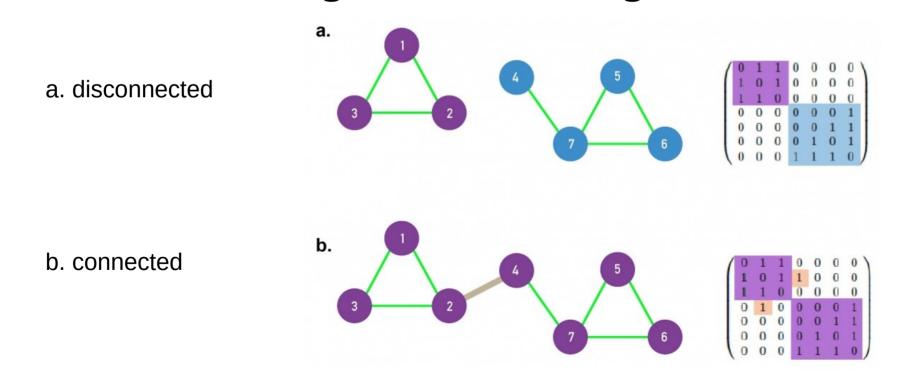


Connectedness

- If a path exists between two nodes i, j:
 - those nodes are part of the same connected component
- A graph that has only one connected component is called a connected graph

Connected graphs

A disconnected graph has an adjacency matrix that can be arranged in block diagonal form



Distance

- If two nodes i, j are in the same connected component:
 - the distance between i and j, denoted by d_{ij} is the length of the shortest path between them

Diameter

- The **diameter** of a network is the maximum distance between two nodes on it, d_{max}
- The effective diameter (or effective-90% diameter) is a number d such that 90% of the pairs of nodes (i,j) are at a distance smaller than d
- The average distance is <d>, and is measured only for nodes that are in the same connected component

Local clustering coefficient

- The local clustering coefficient C_i is a property of a node i
- Let L_i represent the number of links among neighbors of node i

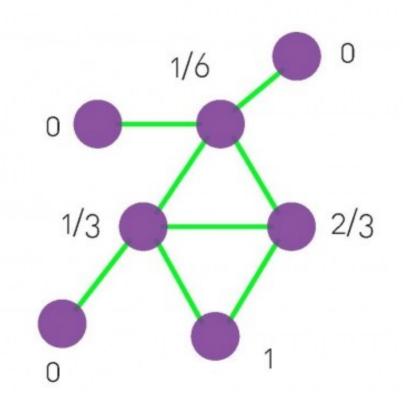
$$C_i = \frac{2L_i}{k_i(k_i - 1)}$$

Average clustering coefficient

 The average clustering coefficient is a property of the entire graph

$$\langle C \rangle = \frac{1}{N} \sum_{i=1}^{N} C_i$$

Example C_i (check to ensure you understood)



More to practice ...

 You can practice with exercises in section 2.11 of Barabási (2016)