

# Network science / Graph mining

## Metrics for analyzing a connected world

Erwan Le Merrer<sup>1</sup>

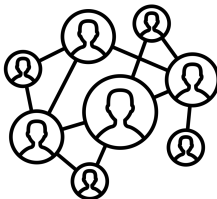
<sup>1</sup>Inria, Rennes

ESIR 2022

- 1 Graphs and representations
- 2 Classical metrics
- 3 Three important graph models & a generative method
- 4 Exploring graphs
- 5 Importance metrics
- 6 Community metrics
- 7 Comparing graphs
- 8 TVGs: time varying graphs

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# Graphs ?



**Figure:** A graph: entities (nodes/vertices) and connections (edges)

An abstraction/representation for reasoning about characteristics of

- physical networks (computers, roads, circuits).
- relational data.

Focus on the structure rather than on the details of modeled objects

# The omnipresence of graphs in applications

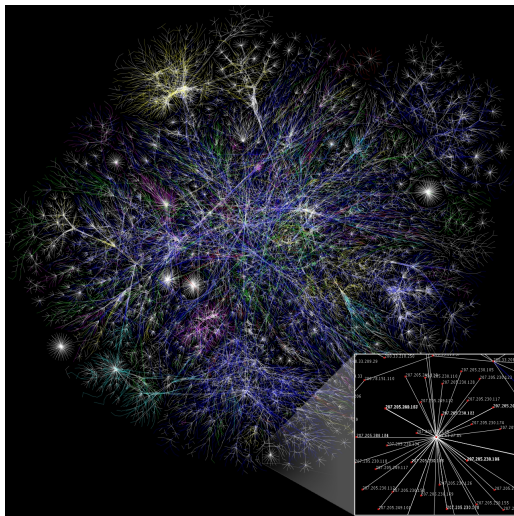


Figure: The Internet AS graph

# The omnipresence of graphs in applications

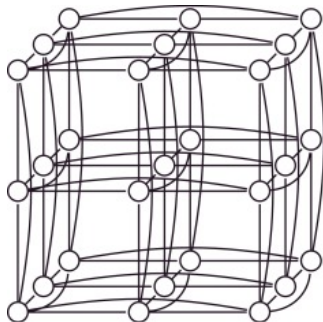


Figure: Interconnecting system-on-chips in a datacenter rack

# The omnipresence of graphs in applications

- exemple use in social nets, epidemics...

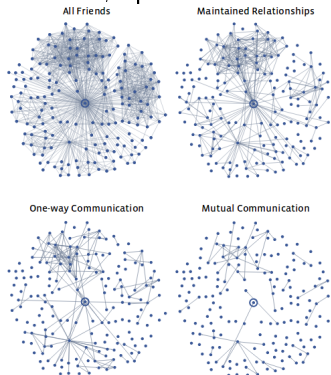


Figure 3.8: Four different views of a Facebook user's network neighborhood, showing the structure of links corresponding respectively to all declared friendships, maintained relationships, one-way communication, and reciprocal (i.e. mutual) communication. (Image from [286].)

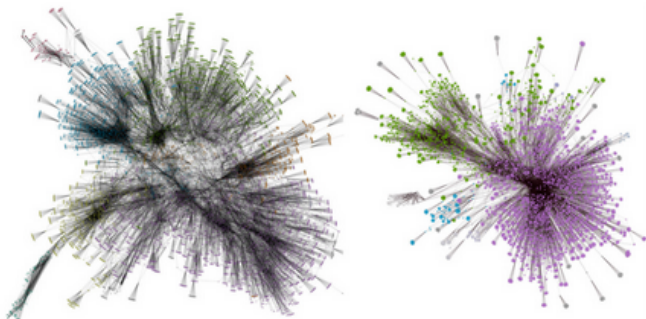
**Figure:** From Networks, Crowds, and Markets: Reasoning about a Highly Connected World . By David Easley and Jon Kleinberg. Cambridge University Press, 2010.

# e.g.: recommendations on YouTube

The image shows a YouTube video player interface. The main video is titled "Lady Gaga's FULL Pepsi Zero Sugar Super Bowl LI Halftime Show | NFL". The video player shows a performance by Lady Gaga on a stage with a red background. To the right of the video player is a sidebar of recommended videos. A red dotted circle highlights this sidebar, which contains several video thumbnails and titles, including "Beyonce - Super Bowl 2015 Halftime Show HD 1080p", "Mix - Lady Gaga's FULL Pepsi Zero Sugar Super Bowl LI Halftime Show | NFL", "Michael Jackson Super Bowl Complete Version HQ", "Lady Gaga | REAL VOICE (WITHOUT AUTO-TUNE)", "Grammy - Metallica & Lady Gaga (with James Mc ON)", "Techno Oldschool & New Beat - German (1988 a 1993)", "Dead Radio - Alarm (2017) (New Full Album)", and "Katy Perry - Super Bowl 2015 - HD".

Figure: Recommendations: contextual, personalized?





4-hops graphs from a YouTube video, new user (left) and returning user (right)

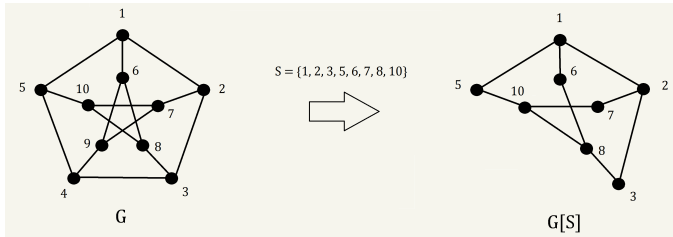
**Figure:** Blank profile vs. my recommendations

# Core notions (1)

- *Directed and undirected graphs:*
  - in directed graphs, edges have orientation (arrow end)

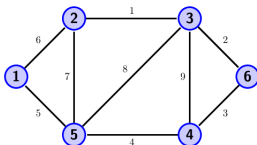


- A *subgraph* of  $G$ : formed by a subset of vertices and edges from  $G$ .

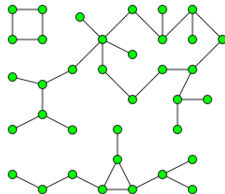


## Core notions (2)

- Edge *weight*: value assigned as a label to an edge.
  - e.g., distance in km of a road from city 1 to 2.

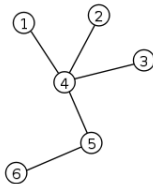


- Graph *connectivity*:
  - A graph is *connected* if there is a *path* btw any pair of vertices.
  - Otherwise, *connected components* are the subgraphs in which paths exist.

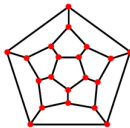


# Core notions (3)

- A *cycle*: a path in which a vertex is reachable from itself.
  - Example of an *acyclic* connected graph: a *tree*



- A *planar* graph: can be drawn without any edges crossing each other.



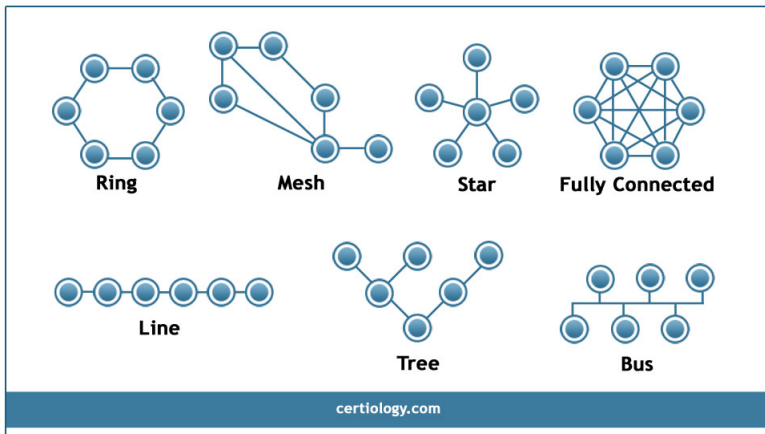
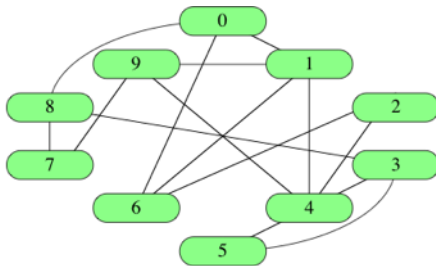


Figure: Graphs to remember, often used as illustrations

# Adjacency list or edge list representations



Edge list:

[ [0,1], [0,6], [0,8], [1,4], [1,6], [1,9], [2,4], [2,6], [3,4], [3,5], [3,8], [4,5], [4,9], [7,8], [7,9] ]

$O(|V|)$  access time to find an edge, but  $O(|E|)$  space in memory.

Adjacency list:

[ [1, 6, 8], [0, 4, 6, 9], [4, 6], [4, 5, 8], [1, 2, 3, 5, 9], [3, 4], [0, 1, 2], [8, 9], [0, 3, 7], [1, 4, 7] ]

$O(1)$  access time to vertex, but  $O(|V|)$  to access a given edge.<sup>1</sup>

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<sup>1</sup><https://www.khanacademy.org/computing/computer-science/algorithms/graph-representation/a/representing-graphs>

# Matrix representation

|   | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
|---|---|---|---|---|---|---|---|---|---|---|
| 0 | 0 | 1 | 0 | 0 | 0 | 0 | 1 | 0 | 1 | 0 |
| 1 | 1 | 0 | 0 | 0 | 1 | 0 | 1 | 0 | 0 | 1 |
| 2 | 0 | 0 | 0 | 0 | 1 | 0 | 1 | 0 | 0 | 0 |
| 3 | 0 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 1 | 0 |
| 4 | 0 | 1 | 1 | 1 | 0 | 1 | 0 | 0 | 0 | 1 |
| 5 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 | 0 | 0 |
| 6 | 1 | 1 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 1 |
| 8 | 1 | 0 | 0 | 1 | 0 | 0 | 0 | 1 | 0 | 0 |
| 9 | 0 | 1 | 0 | 0 | 1 | 0 | 0 | 1 | 0 | 0 |

Figure: Matrix representation of previous graph

Find edge presence in  $O(1)$  time, but  $\Theta(V^2)$  space in memory.  
1's to be replaced by edge weights for weighted graphs.

# Example tool families for manipulating graphs

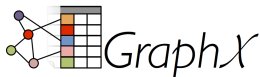


Figure: For massive graphs (cannot fit into on server's memory)

*X – Stream*

Figure: Big graph processing on a single machine



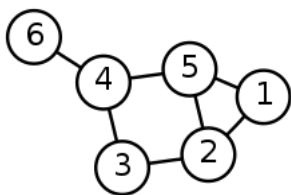
Figure: For a database-like handling of graphs

*NetworkX*

Figure: Prototyping in Python, lots of contributions



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- $G(V, E)$ : graph  $G$  with node set  $V$ , connected by edge set  $E$ .
  - $V = \{1, 2, 3, 4, 5, 6\}$ ;  
 $E = [[1, 5], [1, 2], [2, 3], [2, 5], [3, 4], [4, 5], [4, 6]]$
- Number of nodes is  $n = |V|$ , edges is  $m = |E|$ .
- Neighbors of node  $i$  are set  $\Gamma(i)$ .
  - $\Gamma(1) = \{2, 5\}$

# Degree of a node

- The degree  $d_v$  of node  $v$  is equal to  $|\Gamma(v)|$  (its number of neighbors).
- Degree span:  $0 \leq d_v \leq n - 1$  (if no self loops).
- *Degree distribution*  $P(d)$  is the probability distribution of each degree in the current graph:

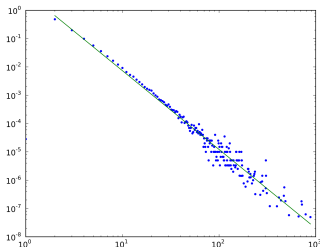


Figure: Degree distribution: x-axis is degree, y-axis is probability

- In(out)-degree of  $v$ : counts incoming(outgoing) edges only.

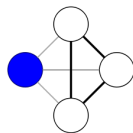
# Clustering coefficient

- Every two nodes in a *clique* are neighbors.
- *Local clustering coefficient* of a node  $i$  measures “how close are  $\Gamma(i)$  from being a clique”:

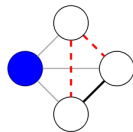
$$C_i = \frac{2|e_{jk} : v_j, v_k \in \Gamma(v_i), e_{jk} \in E|}{d_i(d_i - 1)}$$

- Average clustering coefficient:

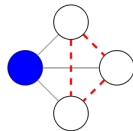
$$\bar{C} = \frac{1}{n} \sum_{i=1}^n C_i$$



$$c = 1$$



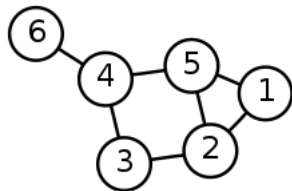
$$c = 1/3$$

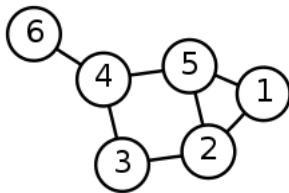


$$c = 0$$

# Path lengths

- *Path*: sequence of adjacent nodes connecting two nodes (if exists).
  - e.g., two paths btw 6 and 1: (4,5,1) and (4,3,2,5,1).
  - One *hop*: one transition from a node to another.
- *Shortest path*: path of minimal cardinality.
  - Distance  $\text{dist}(6,1) = |(4,5,1)| = 3$
- *Single-source shortest path (SSSP)*: shortest paths from node  $i$  to all other nodes ( $V \setminus i$ ).
- *All-pairs shortest paths (APSP)*: SSSP from  $\forall i \in V$ .





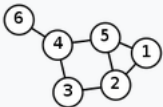
- *Average path length*: average of all-pair shortest distances in the graph.
- *Diameter*: longest path of the APSP, i.e., greatest distance between any pair of vertices.
  - $diam(G) = |(4,5,1)| = 3$ , starting at node 6.

# Spectral analysis

The Laplacian matrix  $L_G = D - A$ :

- $D$  is the degree matrix a diagonal-matrix with  $D(i,i)$  is the degree of the  $i$ th node in  $G$
- $A$  is the adjacency matrix, with  $A(i,j) = 1$  if and only if  $(i,j) \in E$

$$L_G(i,j) = \begin{cases} \deg(i) & \text{if } i = j \\ -1 & \text{if } (i,j) \in E \equiv 1 \\ 0 & \text{otherwise} \end{cases}$$

| Labelled graph  | Degree matrix  | Adjacency matrix   | Laplacian matrix   |
|---|--|--|--|
|  | $\begin{pmatrix} 2 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$ | $\begin{pmatrix} 0 & 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{pmatrix}$ | $\begin{pmatrix} 2 & -1 & 0 & 0 & -1 & 0 \\ -1 & 3 & -1 & 0 & -1 & 0 \\ 0 & -1 & 2 & -1 & 0 & 0 \\ 0 & 0 & -1 & 3 & -1 & -1 \\ -1 & -1 & 0 & -1 & 3 & 0 \\ 0 & 0 & 0 & -1 & 0 & 1 \end{pmatrix}$ |

For an (undirected) graph  $G$  and its **Laplacian matrix**  $L = D - A$  with eigenvalues  $\lambda_0 \leq \lambda_1 \leq \dots \leq \lambda_{n-1}$ :

- $\lambda_0 = 0$ , as  $v_0 = (1, 1, \dots, 1)$  satisfies  $Lv_0 = 0$  (row sum and column sum of  $L$  are 0)
- # of connected components in  $G$  is the algebraic multiplicity of the 0 eigenvalue ( $\implies \lambda_2 = 0$  iff  $G$  is disconnected)
- the smallest non-zero eigenvalue of  $L$  is called the **spectral gap**
- the second smallest eigenvalue of  $L$  (could be zero) is the **algebraic connectivity** of  $G$
- ...



*An example of a result: the diameter of a non complete graph  $G$  satisfies:*

$$\text{diam}(G) \leq \left\lceil \frac{\log(\text{vol}(G)/\delta)}{\log \frac{\lambda_{n-1} + \lambda_1}{\lambda_{n-1} - \lambda_1}} \right\rceil,$$

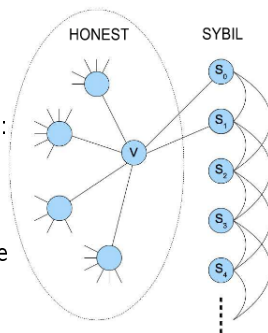
*with  $\delta$  the minimum degree of  $G$ .*

*...and multiple results from graph theory, in general or for specific graphs*

- The conductance  $\Phi(C)$  of a set  $C$  of vertices in a given graph  $G$  is the ratio between the number of edges going out from  $C$  and the number of edges inside  $C$ :

$$\Phi(C) = \frac{|cut(C)|}{vol(C)},$$

where  $vol(C)$  is the sum of the degrees of the vertices in  $C$ .

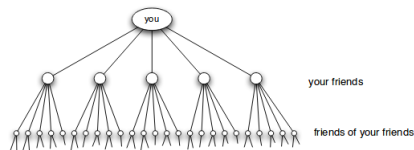


# Expansion

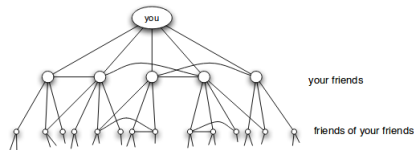
- *Expansion* of  $G$ : mean number of nodes that are reached in  $h$  hops from all nodes:

$$e_G(h) = \frac{1}{n^2} \sum_{v \in V} |C_v(h)|,$$

with  $C_v(h)$  the set of reachable nodes from  $v$  in  $h$  hops.



(a) Pure exponential growth produces a small world



(b) Triadic closure reduces the growth rate

- Measures the robustness of a graph:

$$r_G(h) = \frac{1}{|E|} \sum_{v \in V} l(v, |C_v(h)|),$$

with  $l(v, |C_v(h)|)$  the number of edges that need to be removed to split  $C_v(h)$  into 2 sets (of roughly the same size).  $h$ : distance (hops).