

# Graphs and graph algorithms

## An introduction for Python

Victor Morel

<https://victor-morel.net/>

Chalmers University of Technology

*morelv@chalmers.se*



11<sup>th</sup> November 2022

# Outline

## 1 Introducing graphs

- Mathematical tool
- Graphs to represent networks

## 2 Graphs representation

- Diagram
- Adjacency matrix
- Adjacency list
- Lists

## 3 Graphs in Python

- Basic representation
- NetworkX

## 4 Operations on graph

## 5 Graphs algorithms

- Exploring graphs
- Shortest path

## 6 Conclusion

# You use graphs everyday

## Graphs are everywhere

- When you use a GPS 
- When you run a request on a search engine 
- When you connect on social media 
- When anything you use is based on a neural network

## Sorry I lied

I was not accurate:

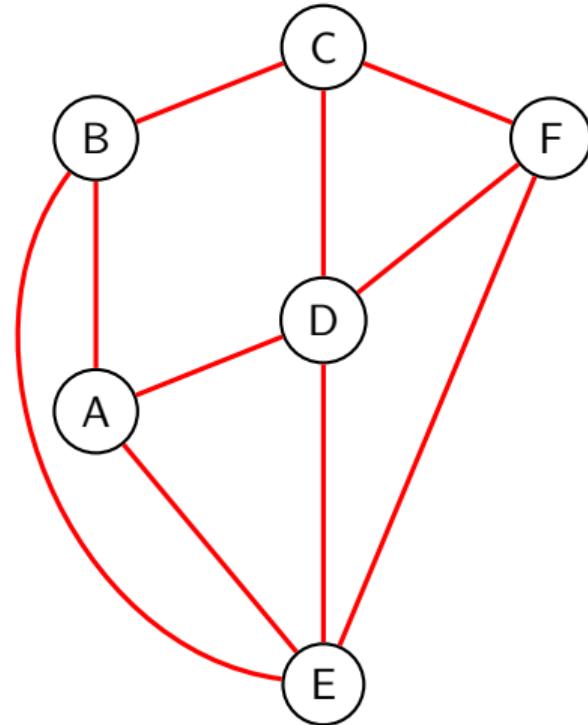
- What I described were networks
- But networks are represented with graphs

So what is a graph? 

# Basic definition

In its simplest form

- A graph  $G$  is composed of
  - ▶ A set of vertices  $V$
  - ▶ Also called nodes
  - ▶ Connected through edges  $E$
  - ▶ Also called links
- $G = (V, E)$



A graph can be represented like this

The end

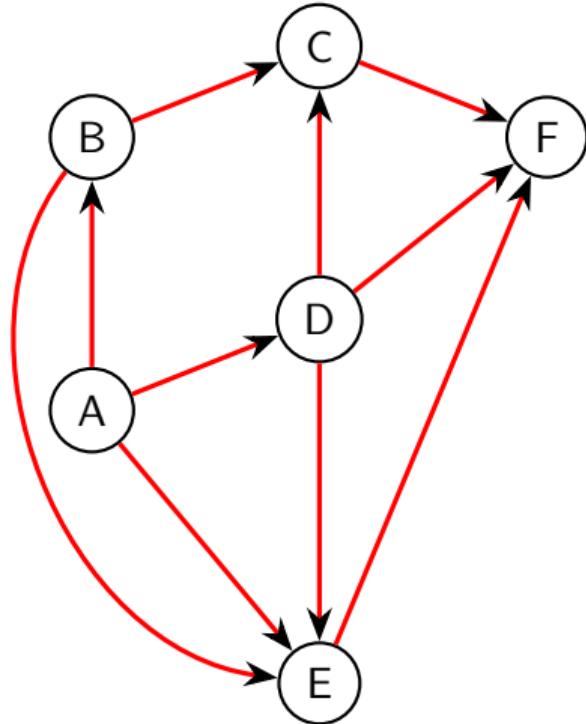


For the less perspicacious: that's a joke, there's plenty of other stuff to see.

# Directed graphs

Some graphs are directed

- Constraint on the edges
- Edges have a direction

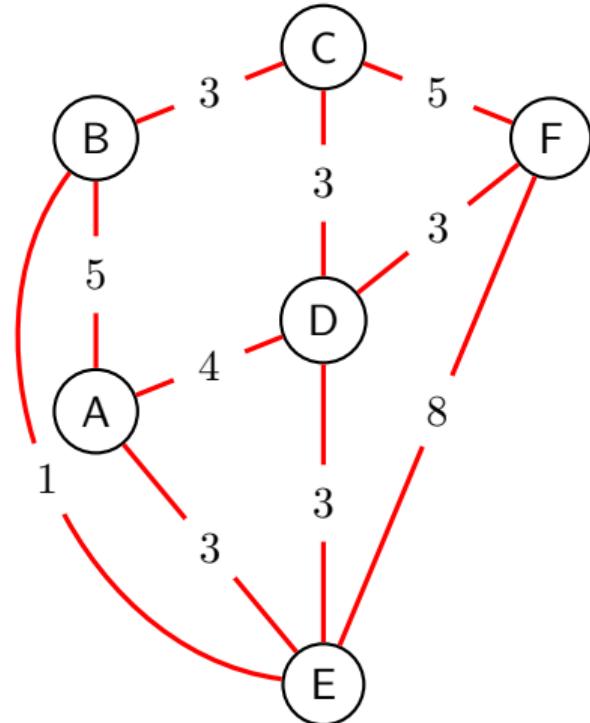


A directed graph can be represented like this

# Weighted graphs

Some graphs are weighted

- Constraint on the edges here again
- Edges have a weight
- Also called a cost
- Typical of a certain kind of graph we will study later



A weighted graph can be represented like this

# Vocabulary

Degree Number of vertices

Adjacent nodes Nodes connected through an edge

Arc In a directed graph, an edge (yup that's all)

Loop In a directed graph, an edge that connects a node to itself

Successor In a directed graph, node j is a successor of i if  $\exists$  arc  $(i,j)$

Predecessor In a directed graph, node j is a predecessor of i if  $\exists$  arc  $(j,i)$

Connected For a regular graph, it means all nodes are connected. For a directed graph, it means that for every nodes  $(i,j) \exists$  a path between i and j (semiconnected to be accurate)

Planar graph A graph is planar if it can be embedded in a plane (edges don't cross)

You want a planar graph?

Don't cross →



# Outline

## 1 Introducing graphs

- Mathematical tool
- Graphs to represent networks

## 2 Graphs representation

- Diagram
- Adjacency matrix
- Adjacency list
- Lists

## 3 Graphs in Python

- Basic representation
- NetworkX

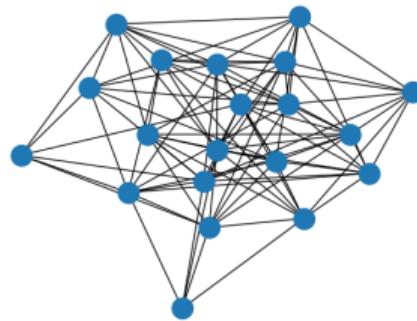
## 4 Operations on graph

## 5 Graphs algorithms

- Exploring graphs
- Shortest path

## 6 Conclusion

# Random networks are random



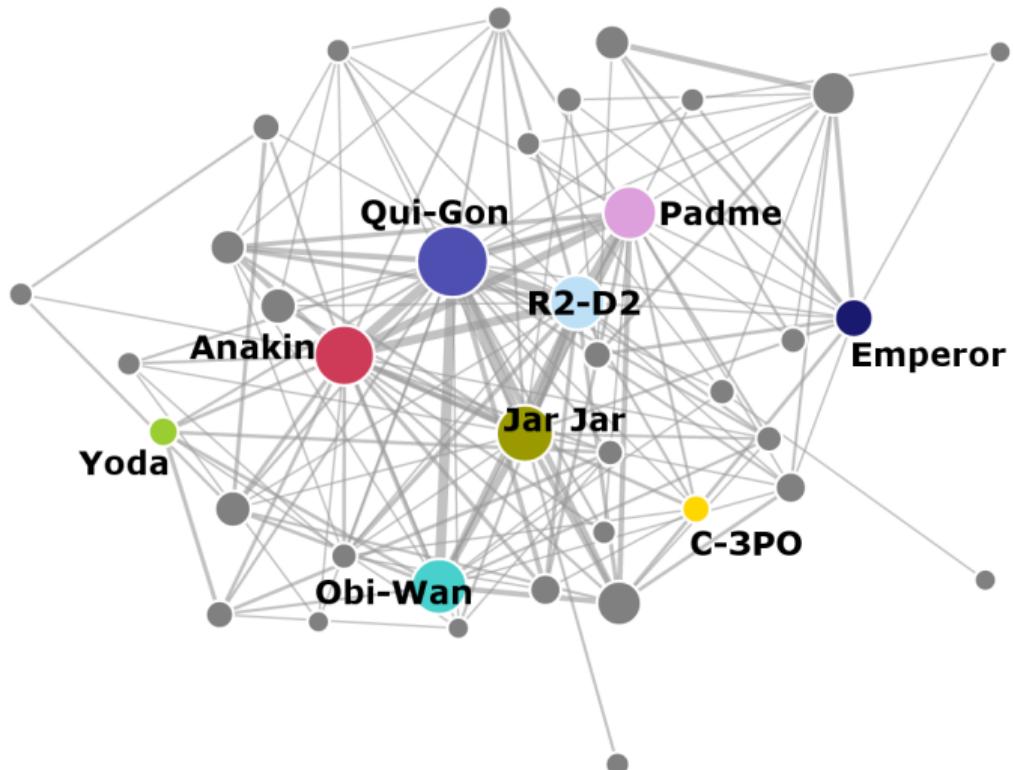
Beautiful isn't it?

## Too random

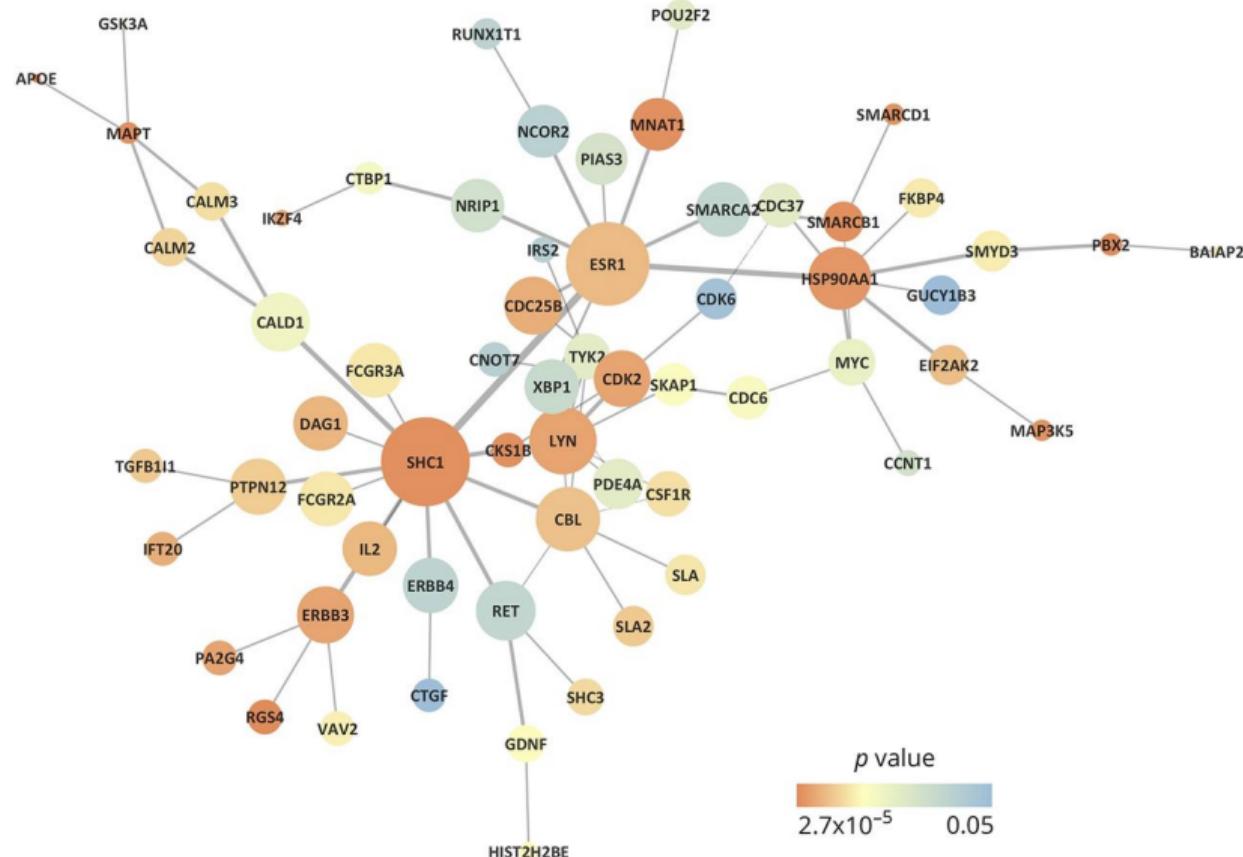
Random networks miss some characteristics:

- Scale-free (few super connected nodes, lots of little connected)
- Small world (can easily reach any node from any other)
- High clustering coefficient (high density of ties)

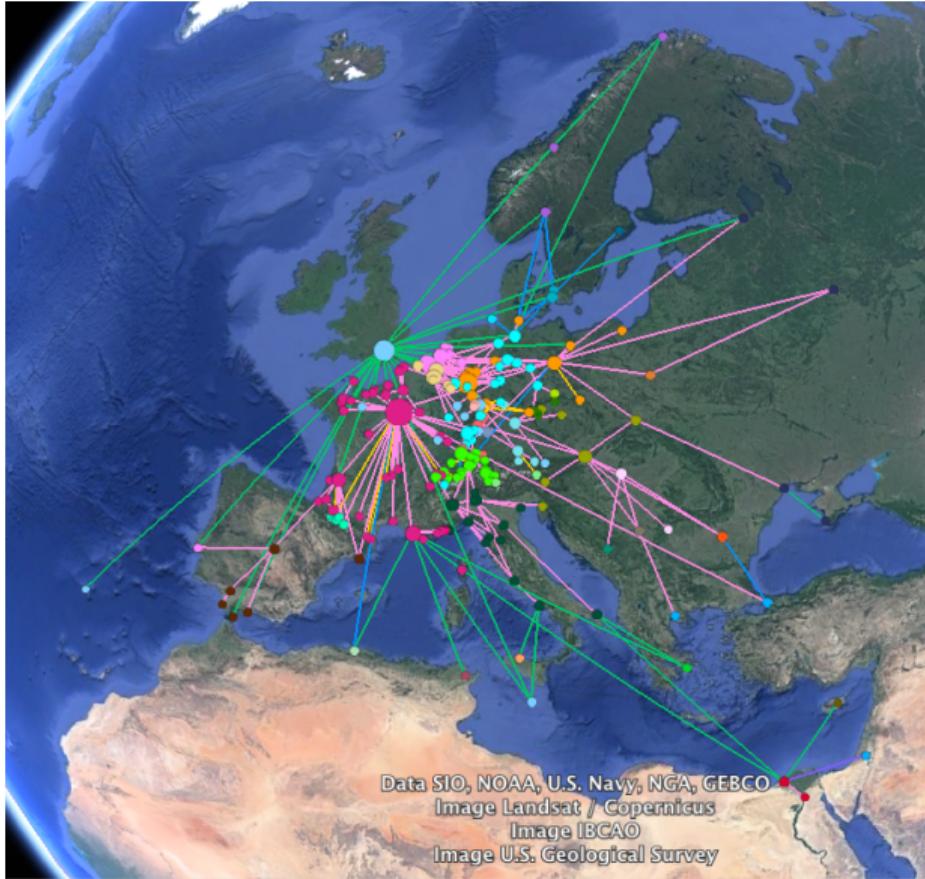
# Social networks



# Biological networks



# Transport networks



# Use-case: tram network in Göteborg



Network graphs are usually weighted and undirected  
They can be directed (but not in our case)

# Outline

## 1 Introducing graphs

- Mathematical tool
- Graphs to represent networks

## 2 Graphs representation

- Diagram
- Adjacency matrix
- Adjacency list
- Lists

## 3 Graphs in Python

- Basic representation
- NetworkX

## 4 Operations on graph

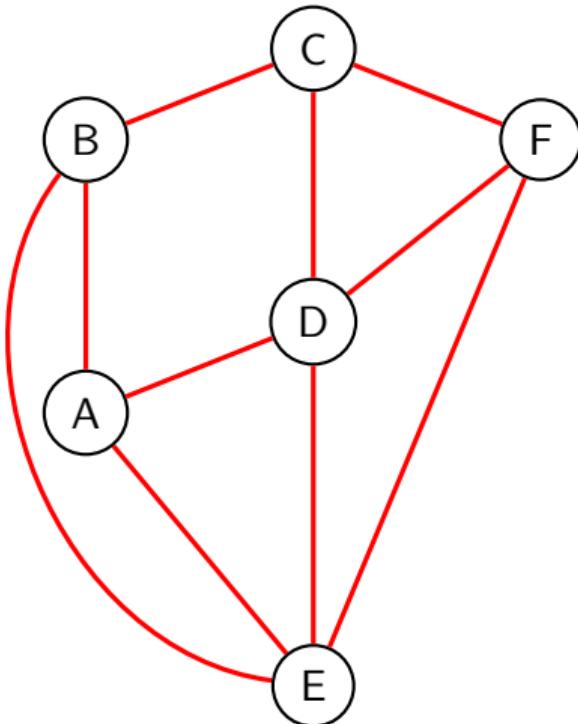
## 5 Graphs algorithms

- Exploring graphs
- Shortest path

## 6 Conclusion

## Graphs for humans: diagrams

We humans like this:

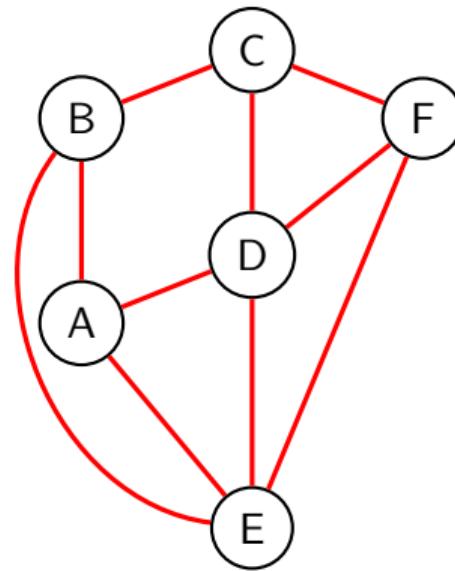


But not exactly machine-readable

## Graphs for machines: adjacency matrix

Mathematicians like this:

$$A_{6 \times 6} = \begin{bmatrix} 0 & 1 & 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 1 & 1 & 0 \end{bmatrix}$$



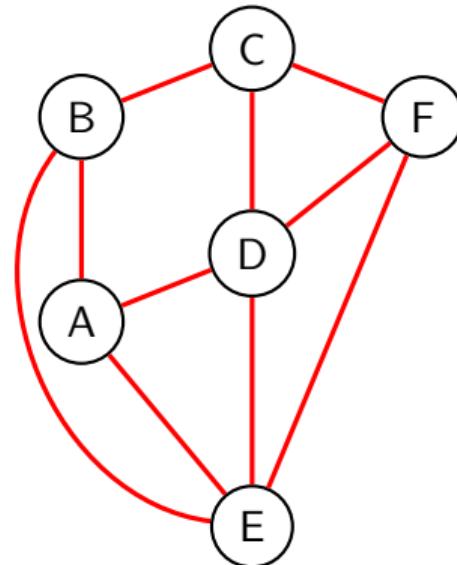
Previous graph

Good machine-representation! 

## Graphs for machines: adjacency list

Computers also like this:

```
adjlist = {  
    A: [B, D, E],  
    B: [A, C, E],  
    C: [B, D, F],  
    D: [A, C, E, F],  
    E: [A, B, D, F],  
    F: [C, D, E]  
}
```



Previous graph

## Graphs for snakes: two lists

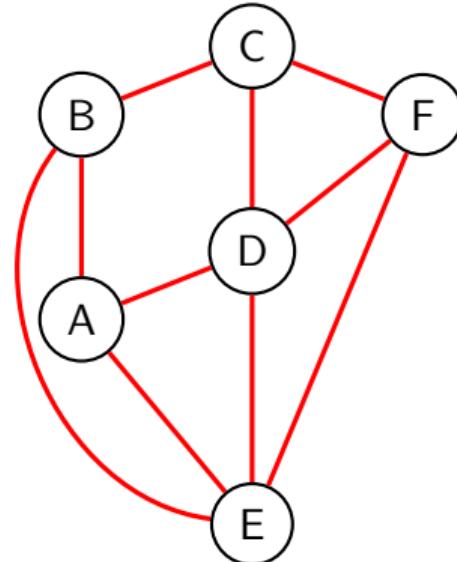
Reptiles 🐍 also like this representation:

vertices = [A, B, C, D, E, F]

edges =

[(A,B),(A,D),(A,E),(B,C),(B,E),(C,D),  
(C,F),(D,E),(D,E),(E,F)]

↑ This is a typical  notation



Still the same graph

# Outline

## 1 Introducing graphs

- Mathematical tool
- Graphs to represent networks

## 2 Graphs representation

- Diagram
- Adjacency matrix
- Adjacency list
- Lists

## 3 Graphs in Python

- Basic representation
- NetworkX

## 4 Operations on graph

## 5 Graphs algorithms

- Exploring graphs
- Shortest path

## 6 Conclusion

# Graph in python - manual implementation

```
graph_skeleton.py
1 class Graph:
2     def __init__(self, start=None, values = None,
3      |   |   |directed=False):
4         self._adjlist = {}
5         if values is None:
6             |   values = {}
7         self._valuelist = values
8         self._isdirected = directed
9         # plus some code for building a graph from a 'start' object
10        # such as a list of edges
11        # here are some of the public methods to implement
12    def vertices(self):
13    def edges(self):
14    def neighbours(self,v):
15    def add_edge(self,a,b):
16    def add_vertex(self,a):
17    def is_directed(self):
18    def get_vertex_value(self, v):
19    def set_vertex_value(self, v, x):
20
21
22 class WeightedGraph(Graph):
23     #missing something?
24
25     def set_weight(self, a, b, w):
26     def get_weight(self, a, b):
```

## Graph in python - NetworkX



**NetworkX**  
Network Analysis in Python

# Outline

## 1 Introducing graphs

- Mathematical tool
- Graphs to represent networks

## 2 Graphs representation

- Diagram
- Adjacency matrix
- Adjacency list
- Lists

## 3 Graphs in Python

- Basic representation
- NetworkX

## 4 Operations on graph

## 5 Graphs algorithms

- Exploring graphs
- Shortest path

## 6 Conclusion

# Diameter

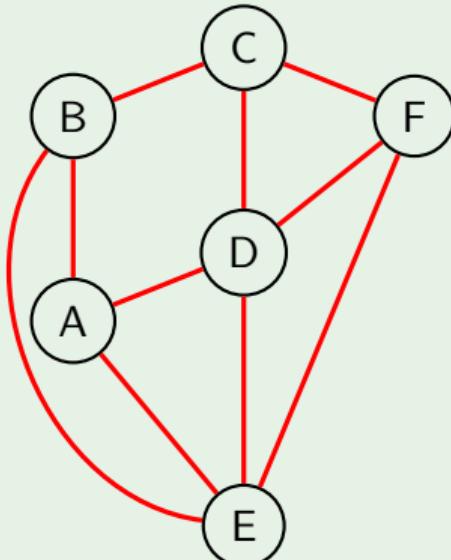
## Definition

The diameter  $d$  of a graph is the greatest distance between any pair of vertices.

## In NetworkX

```
nx.diameter(G)
```

## Example



Diameter is 2

# Degree centrality

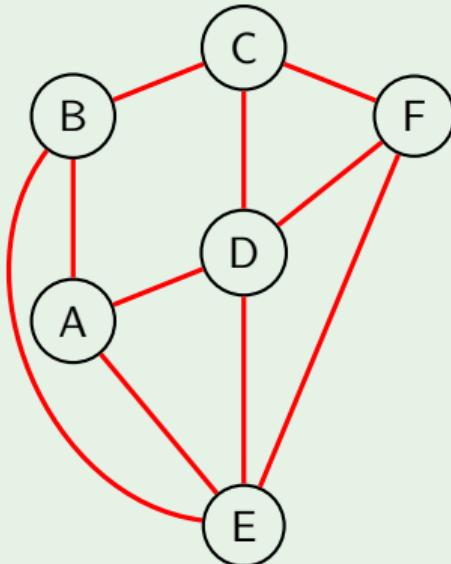
## Definition

The degree centrality for a node  $v$  is the fraction of nodes it is connected to (minus itself).

## In NetworkX

```
nx.degree(G)
```

## Example



Degree centrality for each node is:

A: 0.6, B: 0.6, C: 0.6, D: 0.8, E: 0.8, F: 0.6

# Clustering coefficient

## Definition

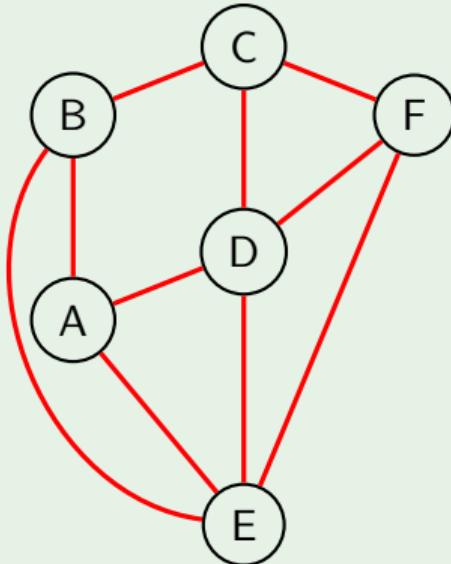
The clustering coefficient  $C_i$  for a vertex  $v_i$  is given by a proportion of the number of links between the vertices within its neighbourhood divided by the number of links that could possibly exist between them (in simple terms).

$$C_i = \frac{2|\{e_{jk} : v_j, v_k \in N_i, e_{jk} \in E\}|}{k_i(k_i-1)} \text{ otherwise}$$

## In NetworkX

```
nx.clustering(G), nx.average_clustering(G)
```

## Example



Clustering coefficient for each node is:  
A: 0.66, B: 0.33, C: 0.33, D: 0.5, E: 0.5, F: 0.66. Average is: 0.5

# Outline

## 1 Introducing graphs

- Mathematical tool
- Graphs to represent networks

## 2 Graphs representation

- Diagram
- Adjacency matrix
- Adjacency list
- Lists

## 3 Graphs in Python

- Basic representation
- NetworkX

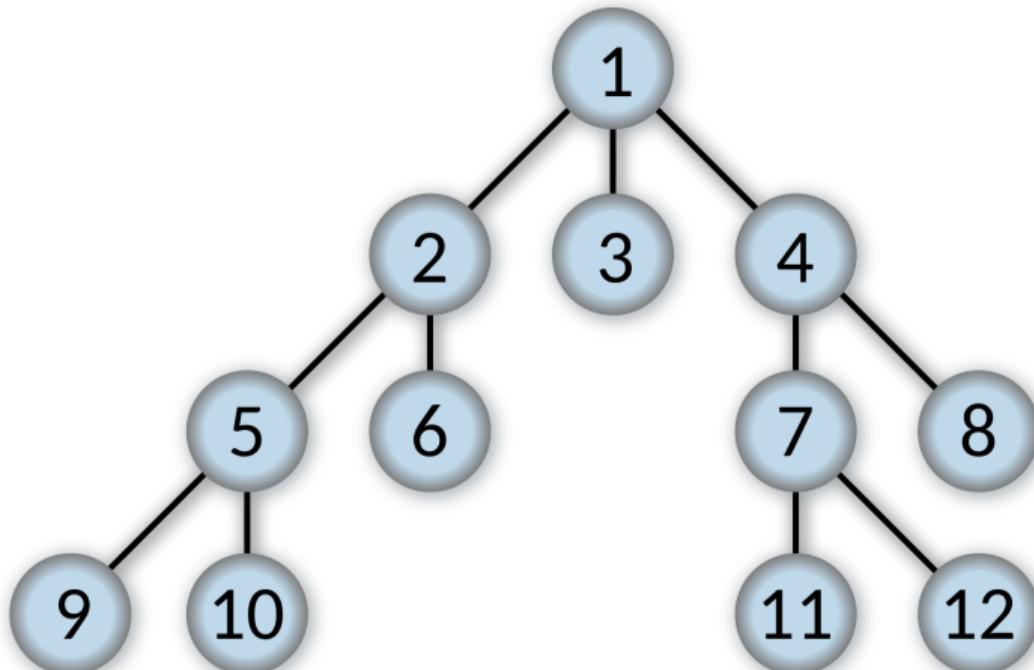
## 4 Operations on graph

## 5 Graphs algorithms

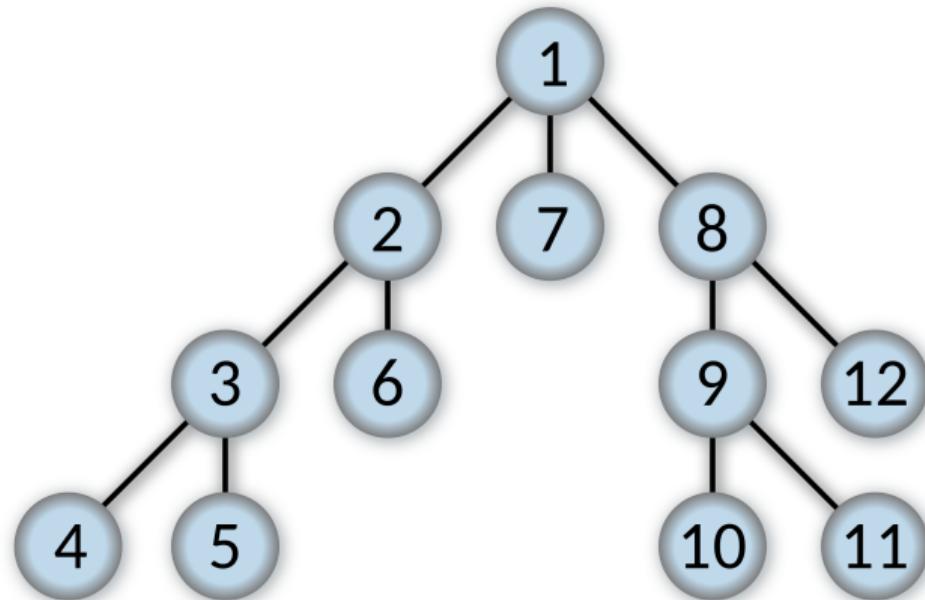
- Exploring graphs
- Shortest path

## 6 Conclusion

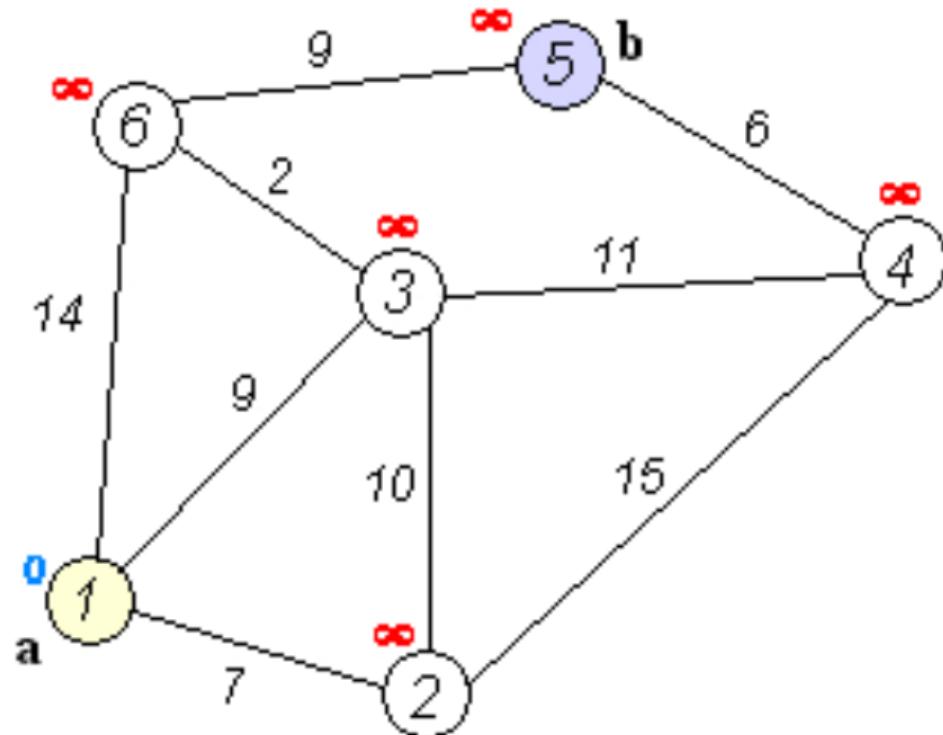
## Breadth-first search



## Depth-first search



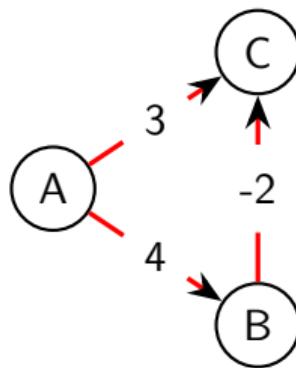
## Dijkstra's algorithm



# Going further

Shortest path with negative values?

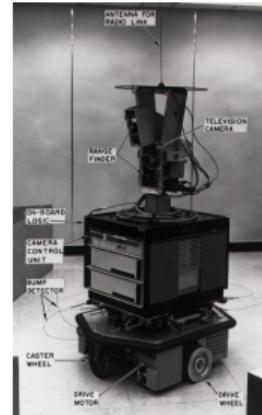
Go for Bellman-Ford



Not so clever, Dijkstra hey? (Start with A)

Want something bit more efficient?

Go for A\* (uses heuristics)



Exterminaaate

# Outline

## 1 Introducing graphs

- Mathematical tool
- Graphs to represent networks

## 2 Graphs representation

- Diagram
- Adjacency matrix
- Adjacency list
- Lists

## 3 Graphs in Python

- Basic representation
- NetworkX

## 4 Operations on graph

## 5 Graphs algorithms

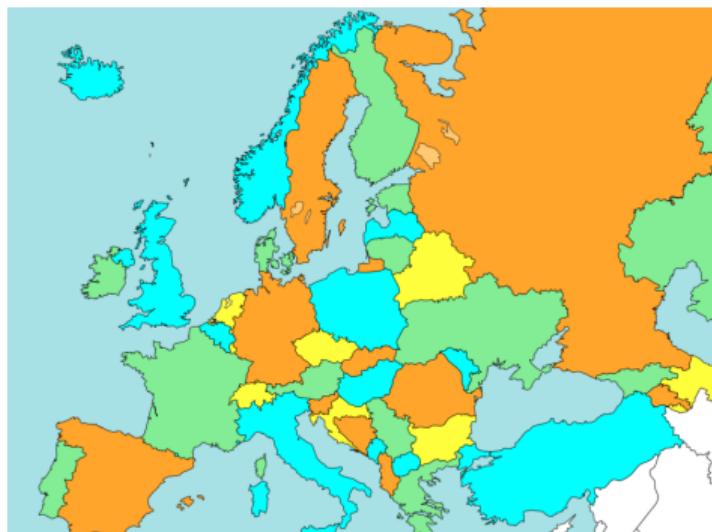
- Exploring graphs
- Shortest path

## 6 Conclusion

# Graphs: other applications

## Coloring

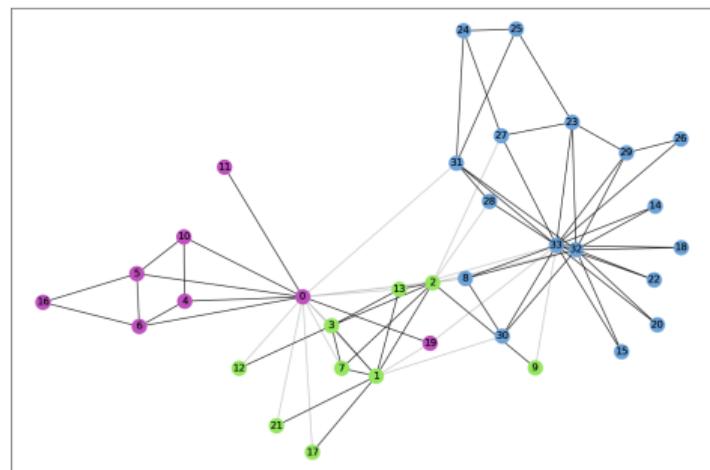
Adjacent vertices always have different colors



Only four colors needed

## Social networks

- Clustering
- Community detection



A famous example: the Karate Club