

# Online Social Networks and Media

Introduction

## Instructors:

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## Goal

Understand the importance of networks in life, technology and applications

Study the theory underlying social networks

Learn about algorithms that make use of network structure

Learn about the tools to analyze them

## Today:

Some logistics

A taste of the topics to be covered

Introduction to graph theory and notebooks

# Logistics

## Textbooks:

Easley and Kleinberg free text-book on [Networks, Crowds and Markets](#)

M. E. J. Newman, [The structure and function of complex networks](#), SIAM Reviews, 45(2): 167-256, 2003

Reza Zafarani, Mohammad Ali Abbasi, Huan Liu, free text-book on [Social Media Mining](#)

Michele Coscia, [The Atlas for the Aspiring Network Scientist](#)

## Web page:

[www.cs.uoi.gr/~tsap/teaching/cs-d6](http://www.cs.uoi.gr/~tsap/teaching/cs-d6)

20% Presentations and class participation

30% Assignments

50% Term Project (in 2 Phases)

*No Final Exam*

**CONSIDER THE FOLLOWING  
COMPLEX SYSTEMS**

# The Economy



# The Human Cell

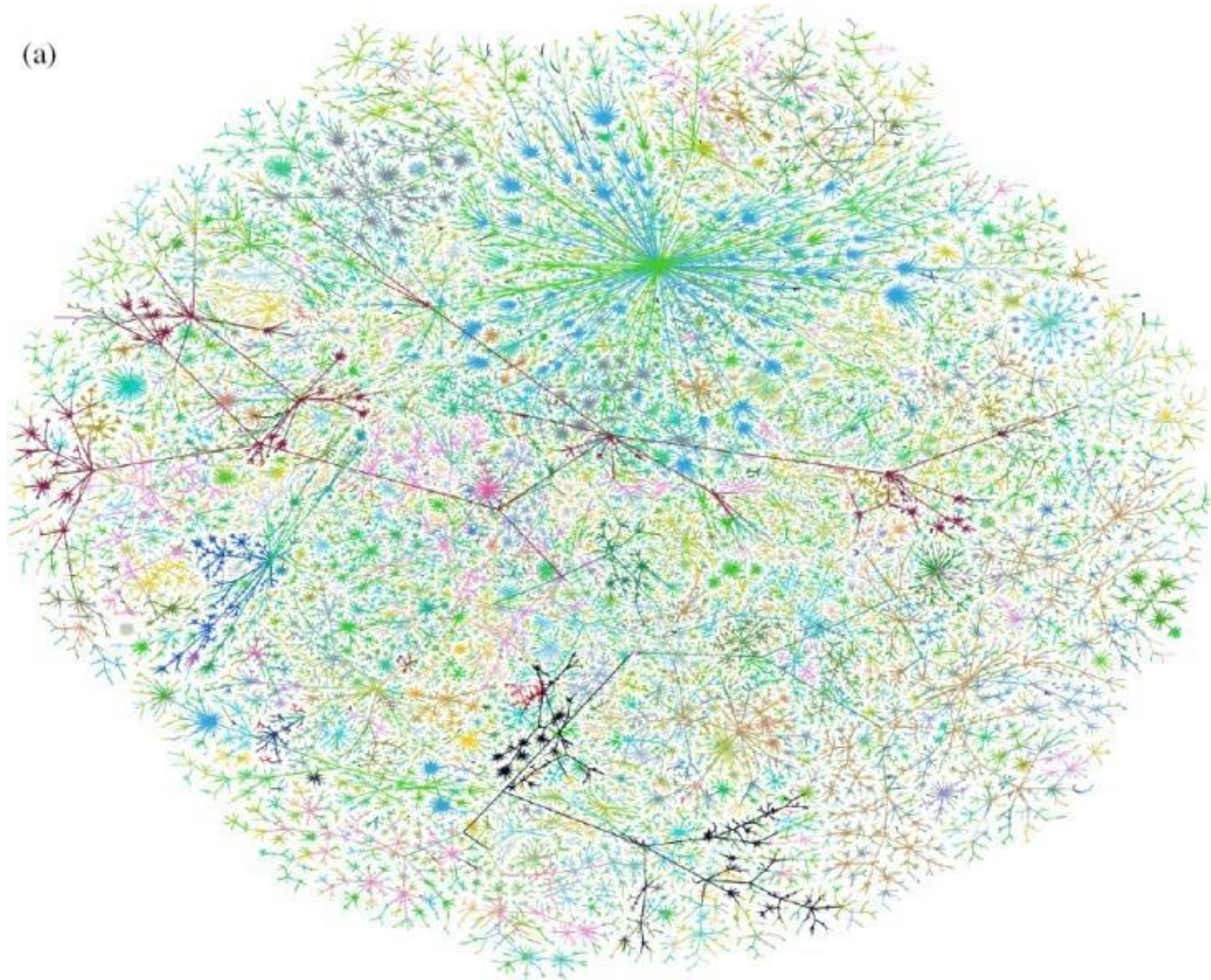


# Traffic and roads



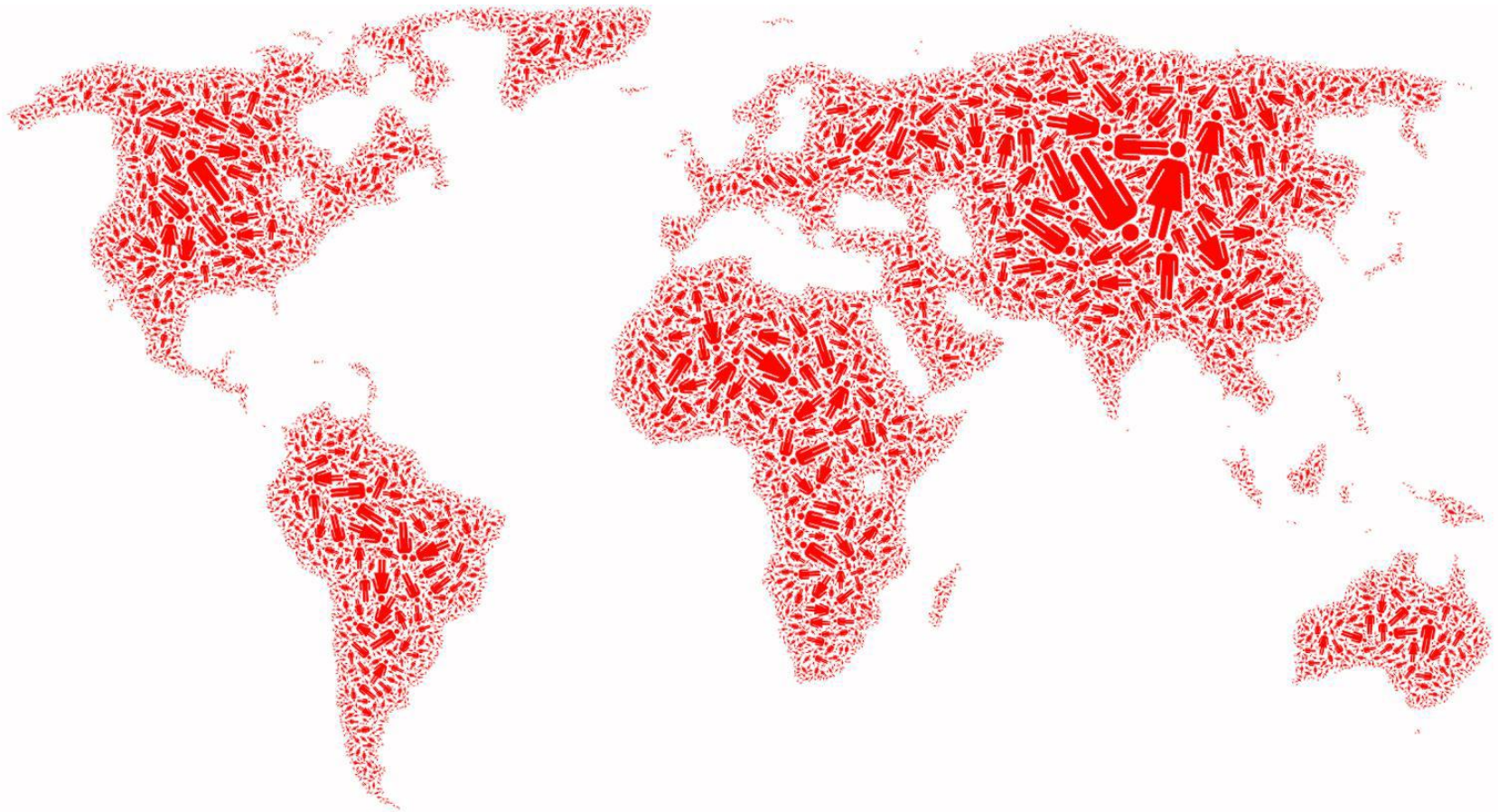


# Internet

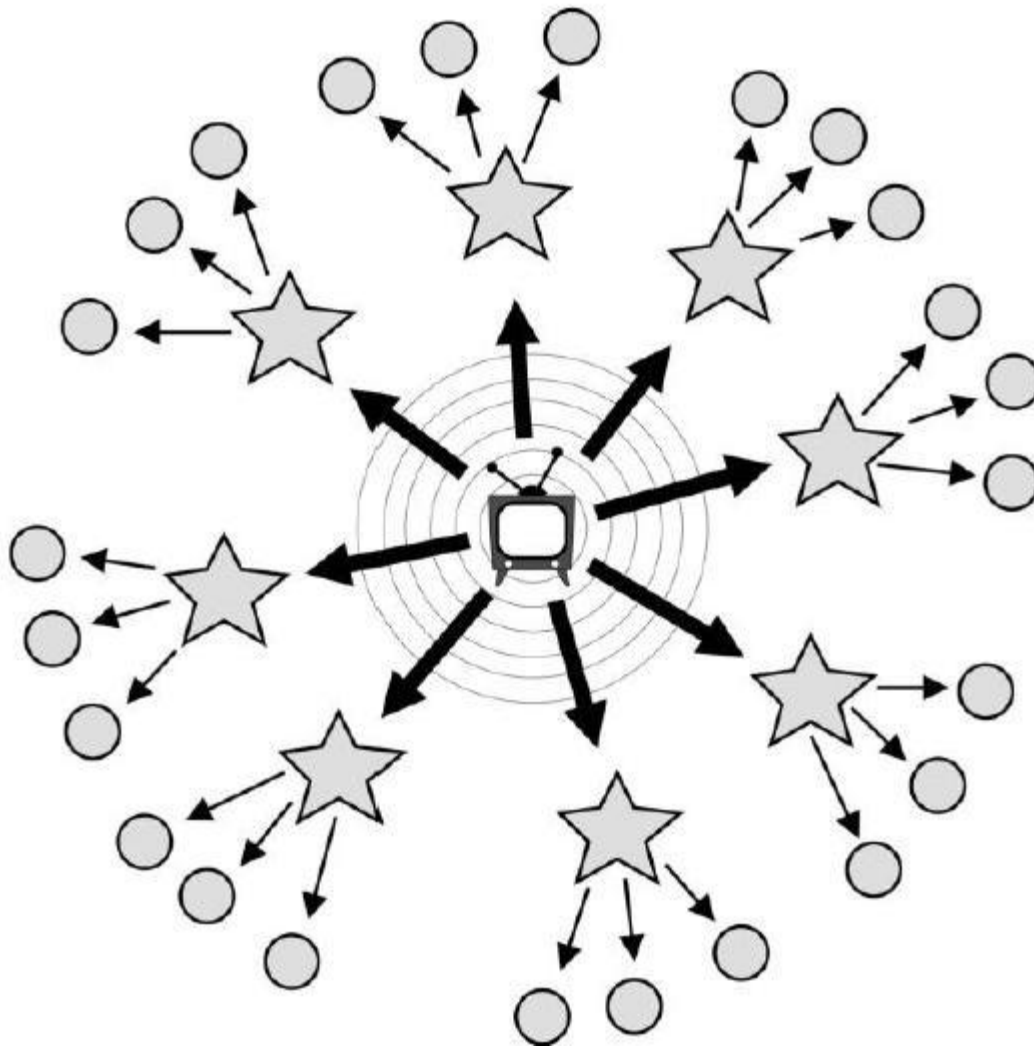




# Society



# Media and Information



WHAT DO THEY HAVE IN COMMON?

# THE NETWORK!

All of these systems can be modeled as  
**networks**

# What is a network?

- Network: a collection of **entities** that are interconnected with **links**.



# Social networks



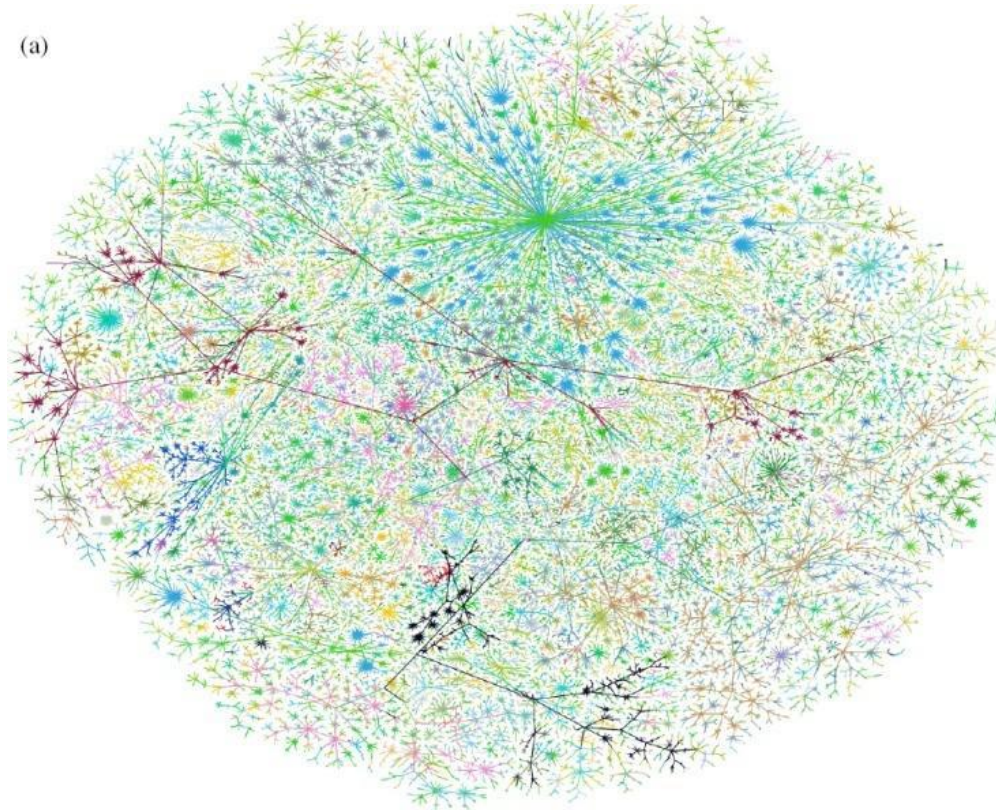
- **Entities:** People
- **Links:** Friendships

# Communication networks



- **Entities:** People
- **Links:** email exchange

# Communication networks

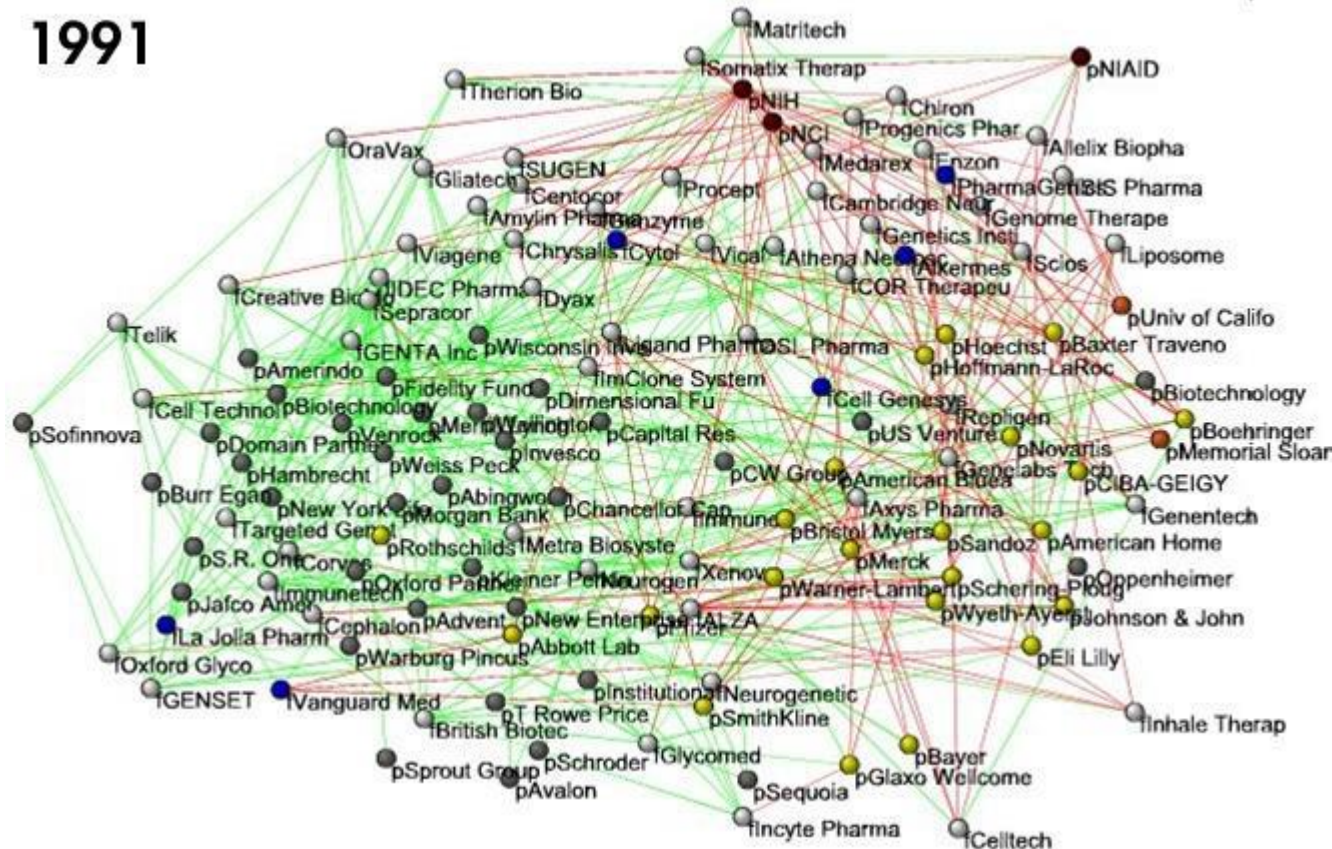


- **Entities:** Internet nodes
- **Links:** communication between nodes



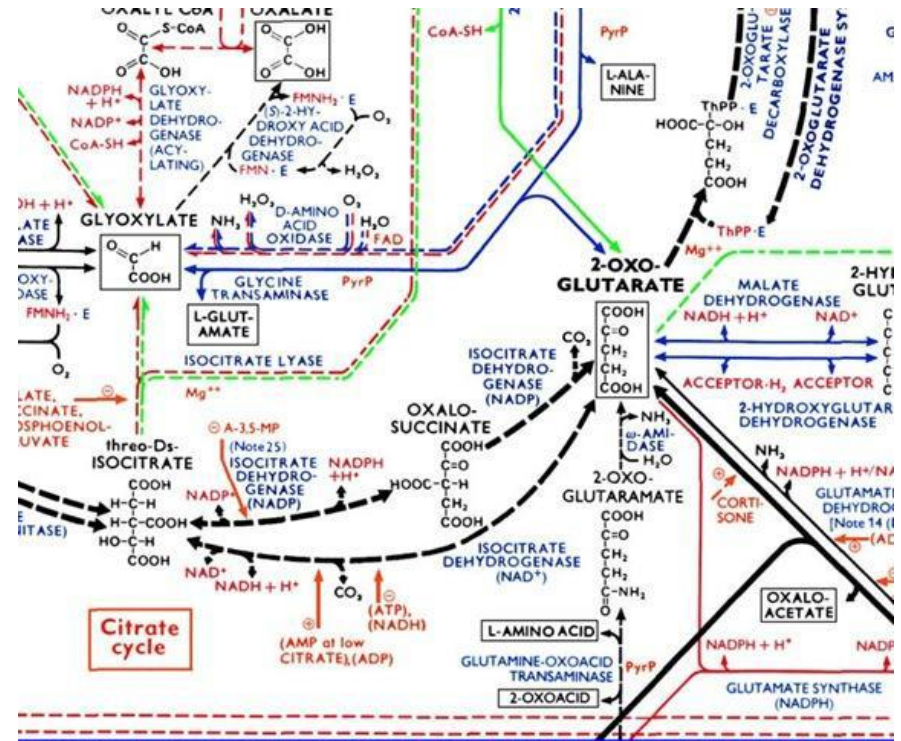
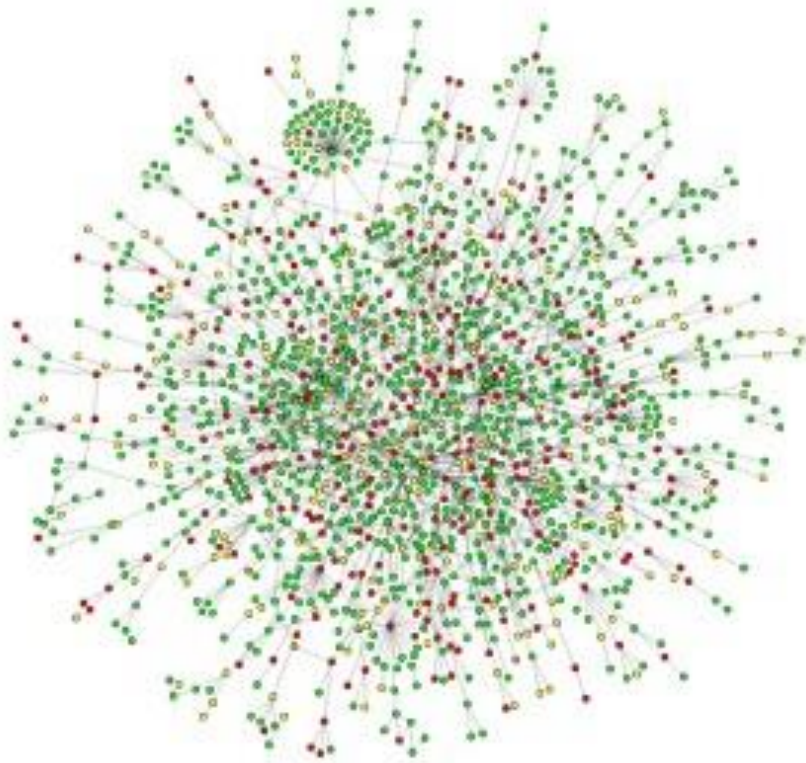
# Financial Networks

1991



- **Entities:** Companies
- **Links:** relationships (financial, collaboration)

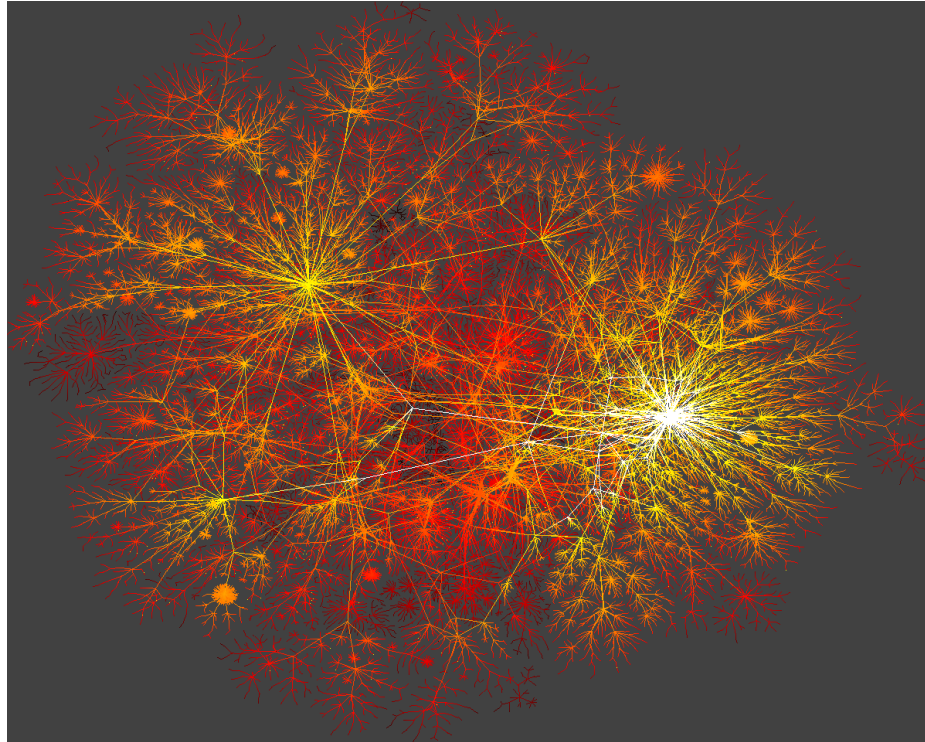
# Biological networks



- **Entities:** Proteins
- **Links:** interactions
- **Entities:** metabolites, enzymes
- **Links:** chemical reactions

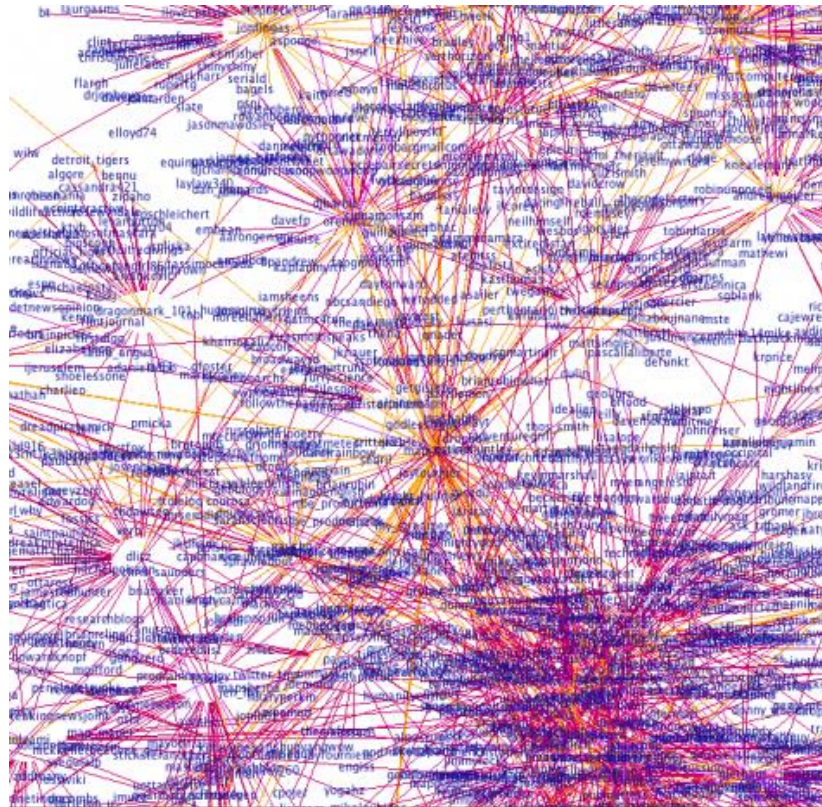


# Information networks



- Entities: Web Pages
- Links: Links

# Information/Media networks



- Entities: Twitter users
- Links: Follows/conversations

# Many more

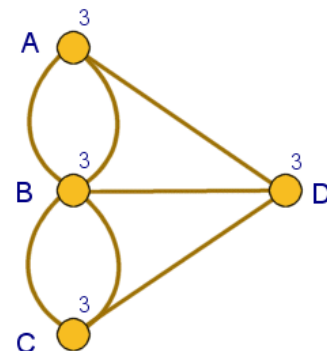
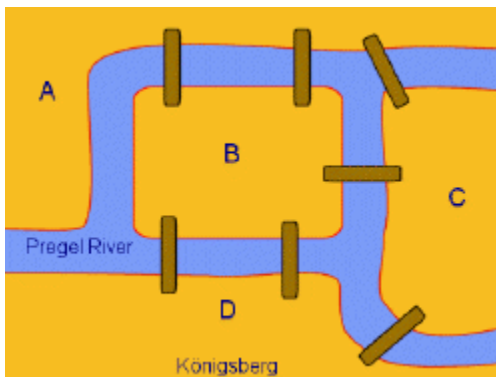
- Wikipedia
- Brain
- Highways
- Software
- Etc...

# Why networks are important?

- We cannot truly understand a **complex system** unless we understand the **underlying network**.
  - Everything is **connected**, studying individual entities gives only a partial view of a system
- Two main themes:
  - What are the **structural properties** of the network?
  - How do **processes** happen in the network?
  - How can we use networks to **extract knowledge**

# Graphs

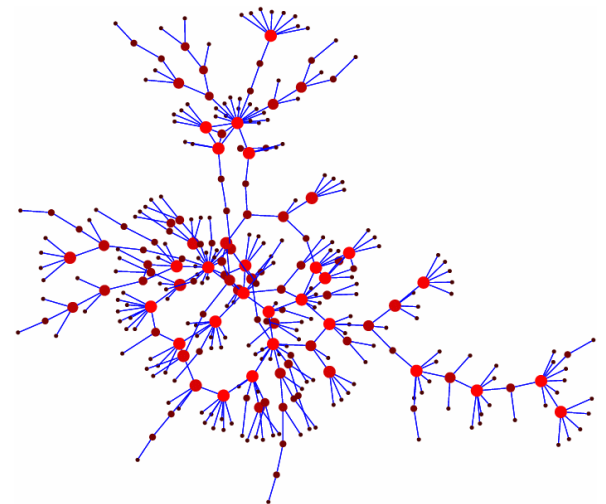
- In mathematics, networks are called **graphs**, the entities are **nodes**, and the links are **edges**
- Graph theory starts in the 18th century, with Leonhard Euler
  - The problem of Königsberg bridges
  - Since then graphs have been studied extensively.





# Networks in the past

- Graphs have been used in the past to model existing networks (e.g., networks of highways, social networks)
  - usually these networks were **small**
  - network can be studied **visual inspection** can reveal a lot of information



# Networks now

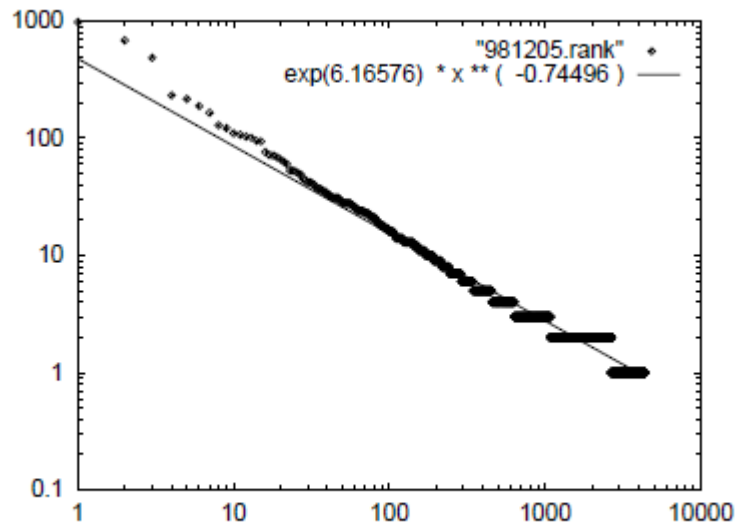
- More and larger networks appear
  - Products of technology
    - e.g., Internet, Web, Facebook, Twitter
  - Result of our ability to collect more, better, and more complex data
    - e.g., gene regulatory networks
  - Result of the willingness of users to contribute data
    - e.g., users making their relationships public online
- Networks of thousands, millions, or billions of nodes
  - Impossible to process visually
  - Problems become harder
  - Processes are more complex
- Networks as a source of data rather than combinatorial objects

# Topics

- Measuring Real Networks
- Modeling the evolution and creation of networks
- Identifying important nodes in the network
- Understanding information cascades and virus contagions
- Finding communities in graphs
- Link Prediction
- Storing and processing huge networks
- Network embeddings
- Machine learning in networks
- Network Fairness
- Other special topics

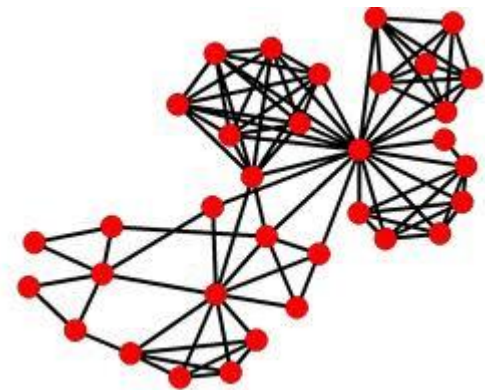
# Understanding large graphs

- What does a network **look like**?
  - Measure different properties to understand the structure



(a) Int-12-98

degree of nodes



Triangles in the graph

# Real network properties

- Most nodes have only a small number of neighbors (degree), but there are some nodes with very high degree (**power-law degree distribution**)
  - **scale-free** networks
- If a node **x** is connected to **y** and **z**, then **y** and **z** are likely to be connected
  - high **clustering coefficient**
- Most nodes are just a few edges away on average.
  - **small world** networks
- Networks from very diverse areas (from internet to biological networks) have similar properties
  - Is it possible that there is a unifying underlying generative process?



# Generating random graphs

- Classic graph theory model (Erdős-Renyi)
  - each edge is generated independently with probability  $p$
- Very well studied model but not realistic:
  - most vertices have about the same degree
  - the probability of two nodes being linked is independent of whether they share a neighbor
  - the average paths are short

# Modeling real networks

- Real life networks are **not “random”**
- Can we define a **model** that **generates** graphs with statistical properties similar to those in real life?
- The **rich-get-richer** model

We need to accurately model the mechanisms that govern the evolution of networks (for prediction, simulations, understanding)

# Ranking of nodes on the Web

- Is my home page as important as the facebook page?
- We need algorithms to compute the **importance of nodes** in a graph
- The **PageRank** Algorithm
  - A success story of network use



It is impossible to create a web search engine without understanding the web graph

# Information/Virus Cascade

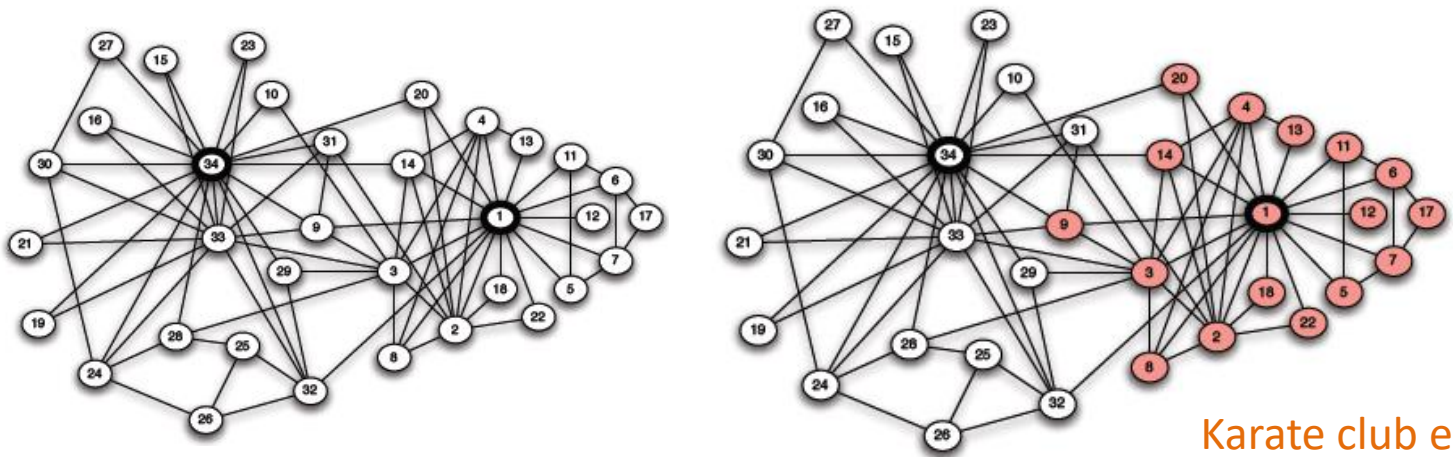
- How do **viruses spread** between individuals? How can we stop them?
- How does **information propagate** in social and information networks? What items become **viral**? Who are the **influencers** and trend-setters?
- We need **models and algorithms** to answer these questions

Online advertising relies heavily on online social networks and word-of-mouth marketing.

COVID-19 demonstrated the importance of understanding virus spread.

# Clustering and Finding Communities

- What is **community**?
  - “Cohesive subgroups are subsets of actors among whom there are relatively strong, direct, intense, frequent, or positive ties.” [Wasserman & Faust '97]



Karate club example  
[W. Zachary, 1970]

How do we define and discover communities in large graphs?

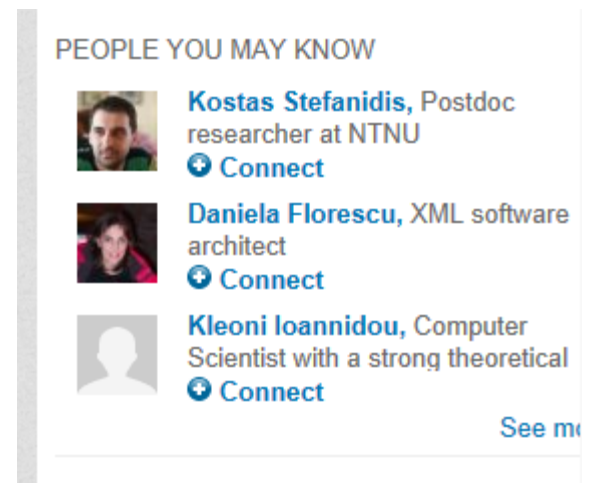
# Homophily

- **Homophily:** “Birds of a feather flock together”. People with similar interests tend to connect to each other
  - Caused by two related social forces [Friedkin98, Lazarsfeld54]
    - *Social influence:* People become similar to those they interact with
    - *Selection:* People seek out similar people to interact with
  - Homophily is widely observed in social networks and has important implications on the network
- Applications in online marketing
  - *viral marketing* relies upon social influence affecting behavior
  - *recommender systems* predict behavior based on similarity

How does homophily affect the properties and the processes in the network?

# Link Prediction

- Given a snapshot of a social network at time  $t$ , we seek to accurately **predict** the edges that will be added to the network during the interval from time  $t$  to a given future time  $t'$ .
- **Applications:**
  - Accelerate the growth of a social network (e.g., Facebook, LinkedIn, Twitter) that would otherwise take longer to form.
  - Identify suspect relationships



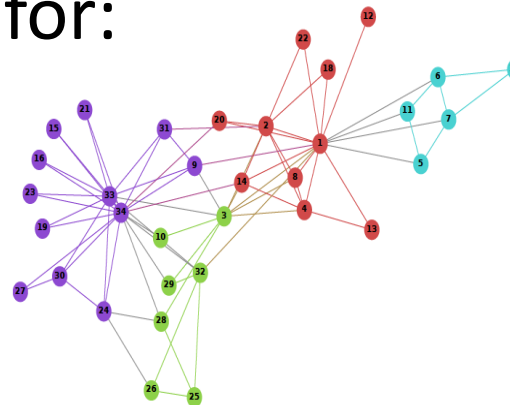
How do we predict future links?

How do we make friendship recommendations?

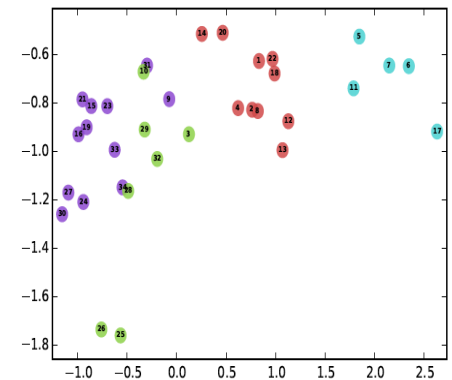


# Network Embeddings

- Project the nodes of the network into a **multidimensional real space**, such that the nodes that are close to each other in the network (or, semantically similar) are mapped to near-by points in this space
- An application of machine learning on graphs.
- Multiple applications for:
  - Link prediction
  - Node classification
  - More



Zachary's Karate Club Network:



2-dimensional node embedding

# Machine Learning on Graphs

- The network information can be used for **Machine Learning** tasks such as classification:
  - “You will be known by the company you keep”: Use the network information to classify the nodes of the graph.
  - Simple **label propagation** techniques have been used very successfully in the past
- **Graph Machine Learning (GML)**: Application of sophisticated ML techniques on graphs. The AI revolution has also reached networks
  - **GNN: Graph Neural Networks**. Incorporate the network structure into the Neural Network computation.

# Network content

- Users on online social networks generate **content**.
- Mining the content **in conjunction** with the **network** can be useful
  - Do friends post similar content on Facebook?
  - Can we understand a user's interests by looking at those of their friends?
    - The importance of homophily
  - Social recommendations: Can we predict a movie rating using the social network?

# Social Media

- Today **Social Media** (Twitter, Facebook, Instagram) have supplanted the traditional media sources
  - Information is generated and disseminated mostly online by users
  - Twitter has become a global “**sensor**” detecting and reporting everything
  - Social media have become the main platform for disseminating ideas
- Interesting problems:
  - Automatically detect events using Twitter
  - Sentiment mining
  - Misinformation and fake news detection

# The dark side of Social Networks

- Do algorithms make **fair** and correct decisions?
- Do algorithms create filter bubbles and echo chambers, increase polarization, and promote misinformation?
- Are online social networks and media a threat to democracy and our mental health?



# Tools

**NetworkX**: a Python language software package for the creation, manipulation, and study of the structure, dynamics, and functions of complex networks.

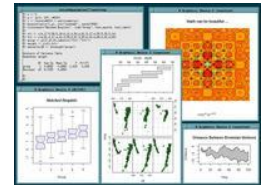
<http://networkx.lanl.gov/>



**Stanford Network Analysis Platform (SNAP)**: general purpose, high performance system for analysis and manipulation of large networks written in C++ <http://snap.stanford.edu/snap/index.html>



**R**: free software environment for statistical computing and graphics. <http://www.r-project.org/>



**Gephi**: interactive visualization and exploration platform for all kinds of networks and complex systems, dynamic and hierarchical graphs  
<http://gephi.org/>





# Frameworks for Processing Large Graphs

Large scale (in some cases billions of vertices, trillions of edges)

How to process graphs in parallel?

- Write *your own code*
- Use *MapReduce (general parallel processing)* \*
- *Pregel* (bulk synchronous parallel model) introduced by Google in 2010\*
- *Giraph* <http://incubator.apache.org/giraph/> (part of Hadoop software)

Storage?

\*J. Dean, S. Ghemawat. *MapReduce: Simplified Data Processing on Large Clusters*. OSDI 2004: 137-150

\*\* G. Malewicz, M. H. Austern, A. J. C. Bik, J. C. Dehnert, I. Horn, N. Leiser: *Pregel: a system for large-scale graph processing*. SIGMOD Conference 2010: 135-146

# Data

Collected using available APIs (Twitter, Facebook, etc)

Using existing collections, e.g., from SNAP (more in the webpage), permission may be required

## **Stanford Large Network Dataset Collection**

[60 large social and information network datasets](#)

## **Coauthorship and Citation Networks**

[DBLP](#): Collaboration network of computer scientists

[KDD Cup Dataset](#)

## **Internet Topology**

[AS Graphs](#): AS-level connectivities inferred from Oregon route-views, Looking glass data and Routing registry data

## **Yelp Data**

[Yelp Review Data](#): reviews of the 250 closest businesses for 30 universities for students and academics to explore and research

## **Youtube dataset**

[Youtube data](#): YouTube videos as nodes. Edge  $a \rightarrow b$  means video  $b$  is in the related video list (first 20 only) of a video  $a$ .

## **Amazon product copurchasing networks and metadata**

[Amazon Data](#): The data was collected by crawling Amazon website and contains product metadata and review information about 548,552 different products (Books, music CDs, DVDs and VHS video tapes).

## **Wikipedia**

[Wikipedia page to page link data](#): A list of all page-to-page links in Wikipedia

[DBpedia](#): The DBpedia data set uses a large multi-domain ontology which has been derived from Wikipedia.

[Edits and talks](#): Complete edit history (all revisions, all pages) of Wikipedia since its inception till January 2008.

## **Movie Ratings**

[IMDB database](#): Movie ratings from IMDB

[User rating data](#): Movie ratings from MovieLens

# Acknowledgements

- Thanks to Jure Leskovec for some of the material from his course notes.
- M. E. J. Newman, **The structure and function of complex networks**, SIAM Reviews, 45(2): 167-256, 2003

# Graph Theory Reminder

For a more extensive reminder

([pptx](#), [pdf](#))

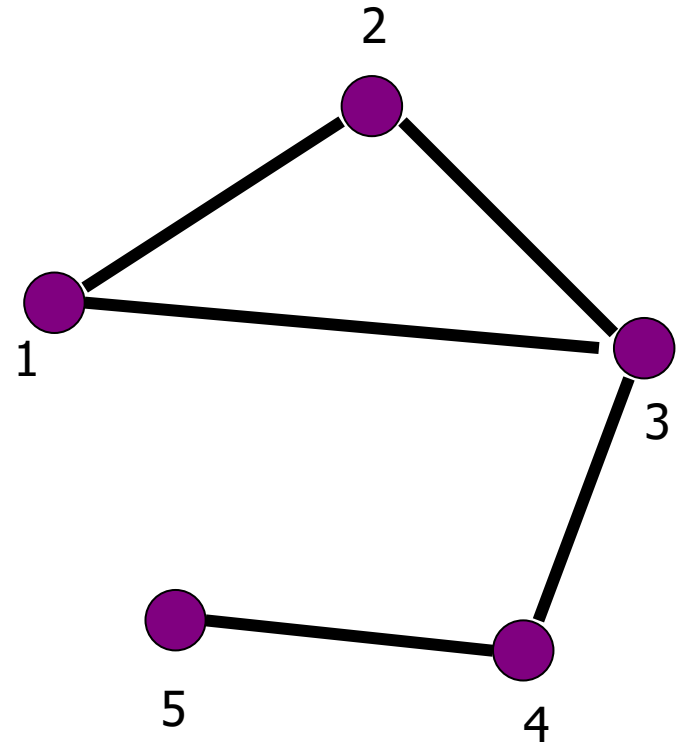
# Undirected Graph

- Graph  $G=(V,E)$ 
  - $V$  = set of vertices (nodes)
  - $E$  = set of edges

undirected graph

$V = \{1, 2, 3, 4, 5\}$

$E = \{(1,2), (1,3), (2,3), (3,4), (4,5)\}$



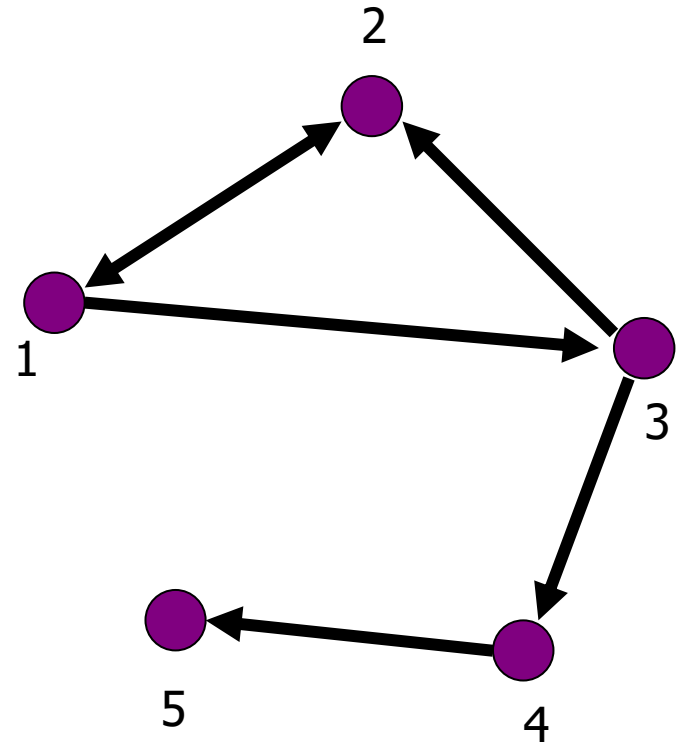
# Directed Graph

- Graph  $G=(V,E)$ 
  - $V$  = set of vertices (nodes)
  - $E$  = set of edges

directed graph

$V = \{1, 2, 3, 4, 5\}$

$E = \{\langle 1,2 \rangle, \langle 2,1 \rangle, \langle 1,3 \rangle, \langle 3,2 \rangle, \langle 3,4 \rangle, \langle 4,5 \rangle\}$

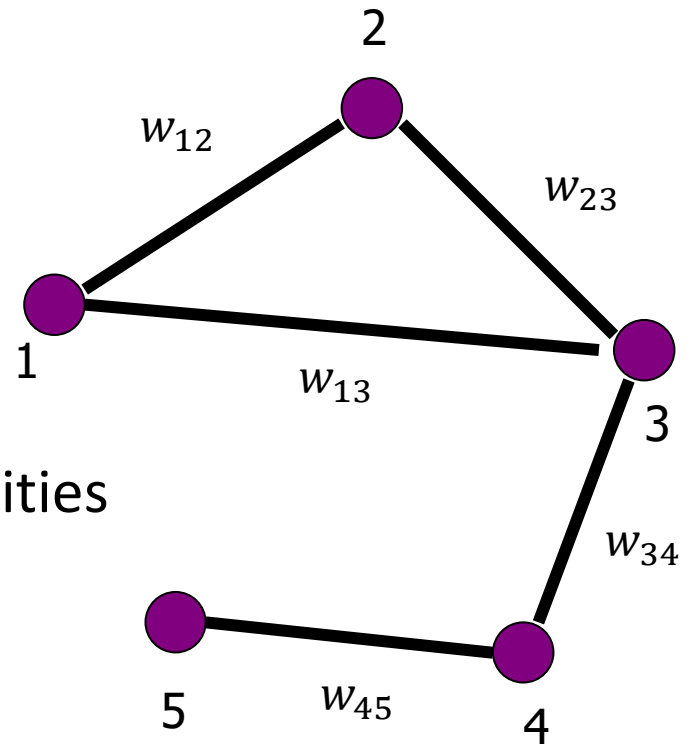




# Weighted Graph

- Graph  $G=(V,E)$ 
  - $V$  = set of vertices (nodes)
  - $E$  = set of edges and their weights

Weights can be either distances or similarities



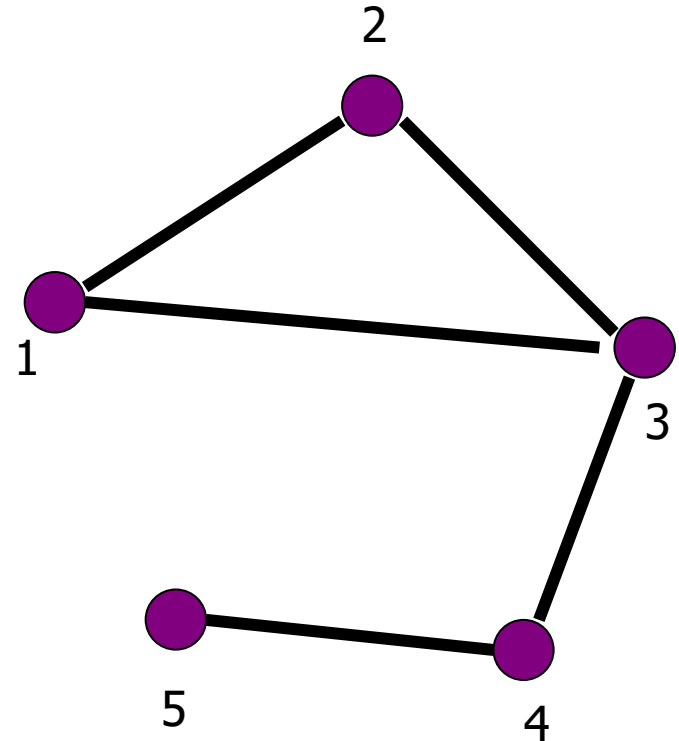
Weighted graph

$V = \{1, 2, 3, 4, 5\}$

$E = \{(1,2,w_{12}), (1,3,w_{13}), (2,3,w_{23}), (3,4,w_{34}), (4,5,w_{45})\}$

# Undirected graph

- Neighborhood  $N(i)$  of node  $i$ 
  - Set of nodes adjacent to  $i$
- degree  $d(i)$  of node  $i$ 
  - Size of  $N(i)$
  - number of edges incident on  $i$



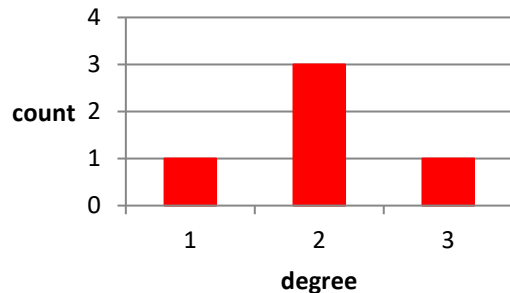
# Undirected graph

- degree sequence

- $[d(1), d(2), d(3), d(4), d(5)]$
- $[2, 2, 3, 2, 1]$

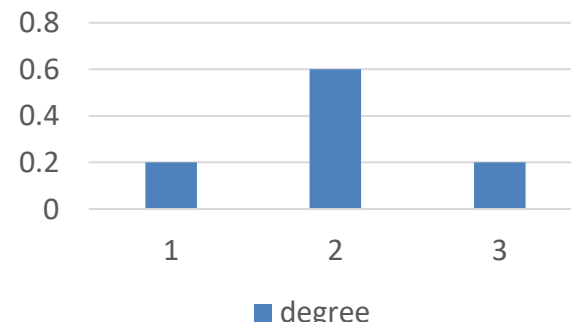
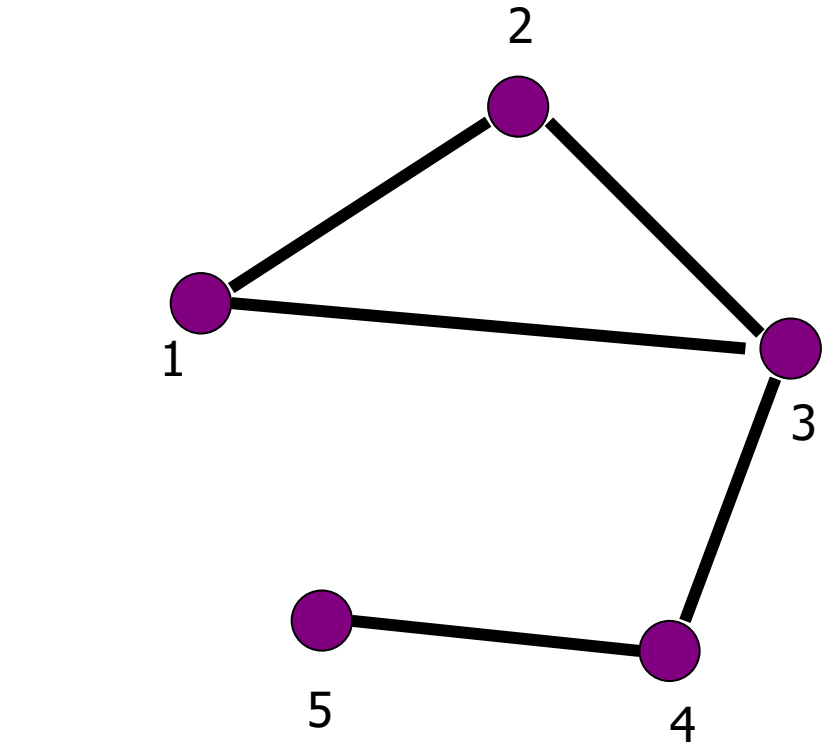
- degree histogram

- $[(1:1), (2:3), (3,1)]$



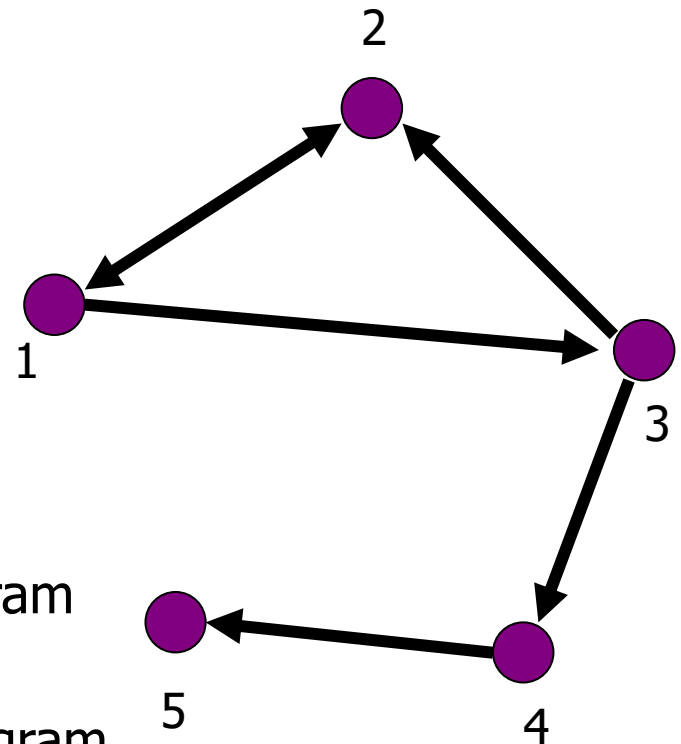
- degree distribution

- $[(1:0.2), (2:0.6), (3,0.2)]$



# Directed Graph

- **in-degree**  $d_{in}(i)$  of node  $i$ 
  - number of edges incoming to node  $i$
- **out-degree**  $d_{out}(i)$  of node  $i$ 
  - number of edges leaving node  $i$
- in-degree sequence
  - [1,2,1,1,1]
- out-degree sequence
  - [2,1,2,1,0]
- in-degree histogram
  - [(1:4),(2:1)]
- out-degree histogram
  - [(0:1),(1:2),(2:2)]



# Graph Traversals

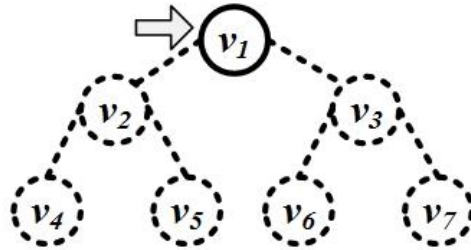
- A **traversal** is a procedure for visiting (going through) all the nodes in a graph

# Depth First Search Traversal

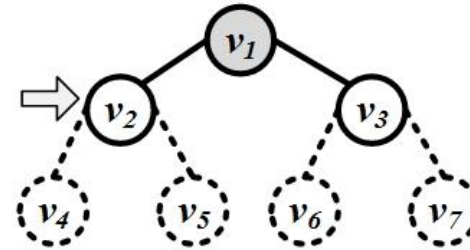
- Depth-First Search (**DFS**) starts from a node  $i$ , selects one of its neighbors  $j$  from  $N(i)$  and performs Depth-First Search on  $j$  before visiting other neighbors in  $N(i)$ .
  - The algorithm can be implemented using a *stack structure*



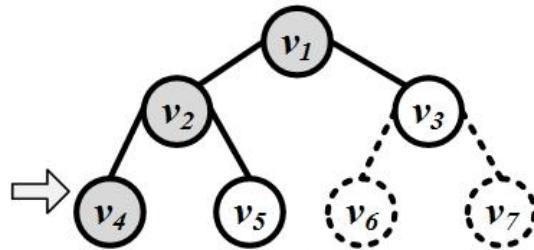
# Example for a tree graph



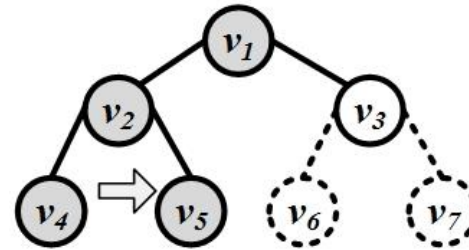
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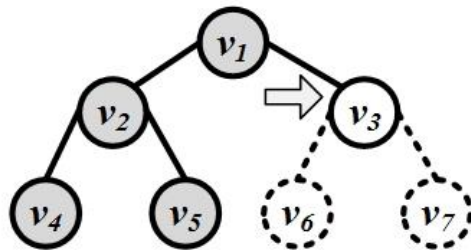
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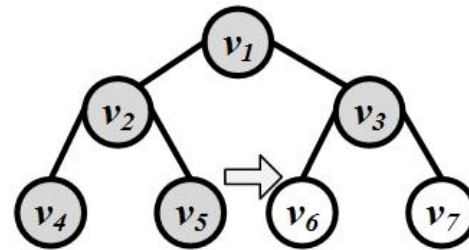
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(4)



(5)

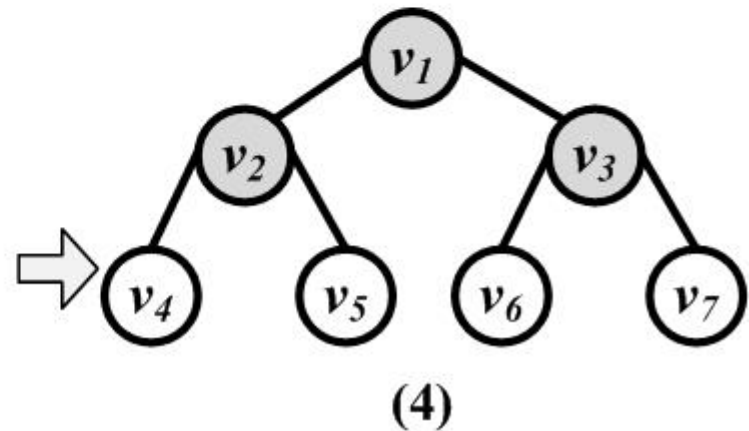
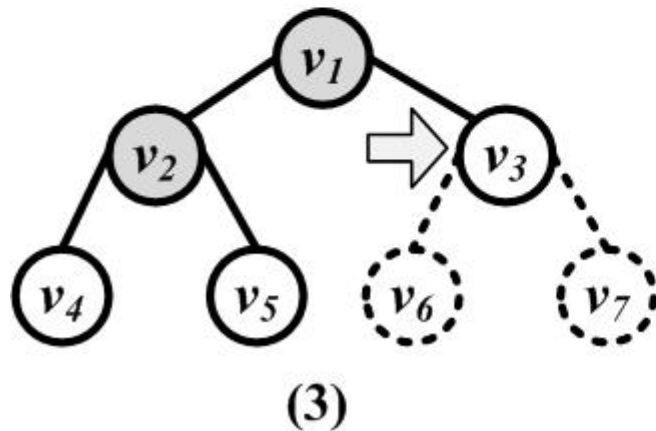
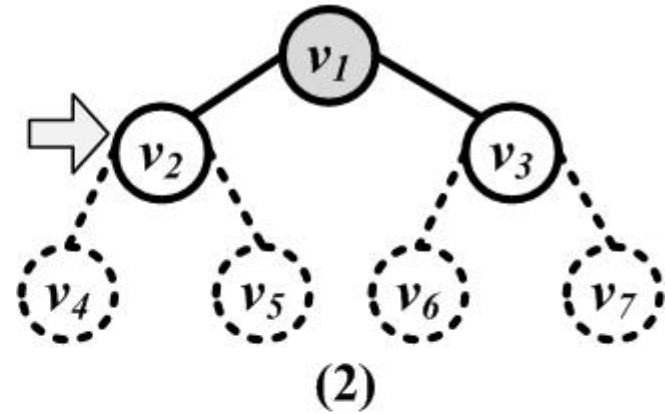
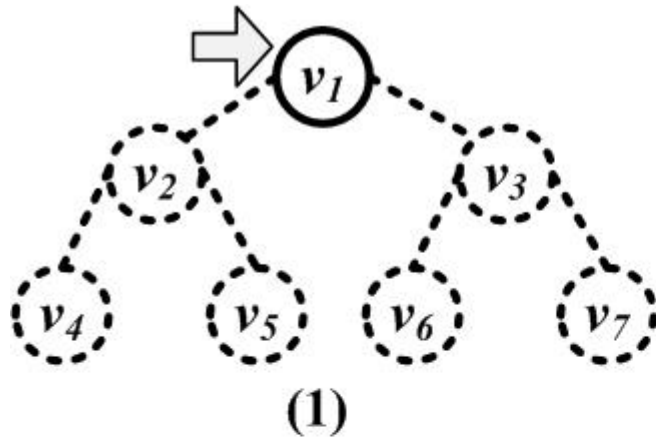


(6)

# Breadth First Search Traversal

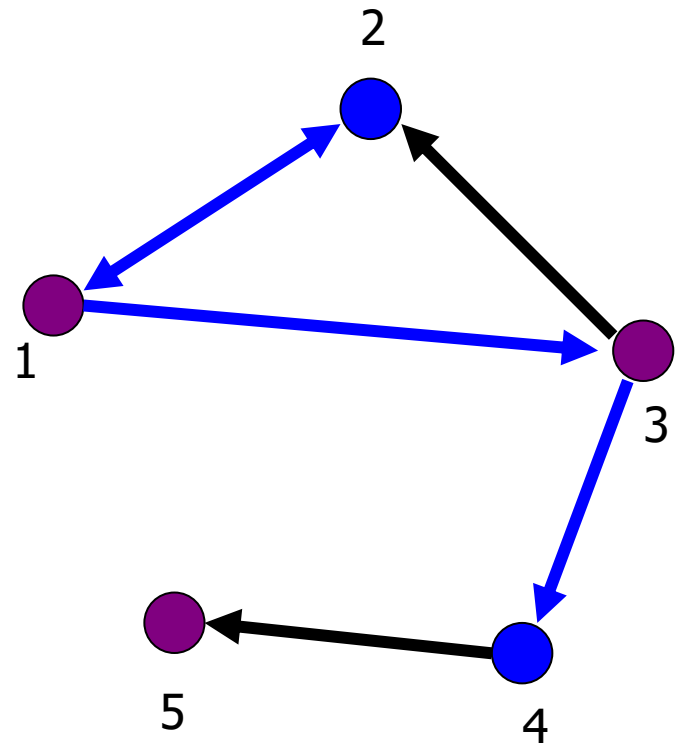
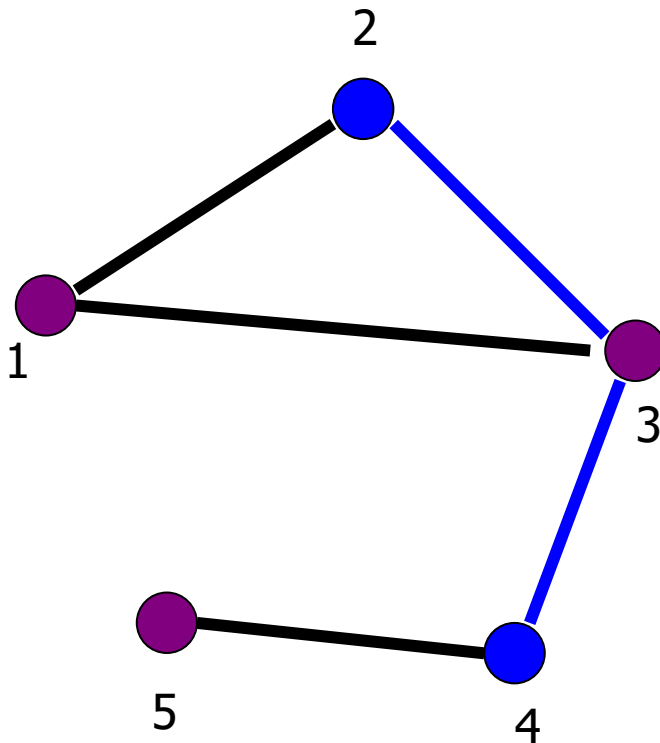
- Breadth-First-Search (**BFS**) starts from a node, visits all its immediate neighbors first, and then moves to the second level by traversing their neighbors.
  - The algorithm can be implemented using a *queue structure*

# Example of BFS on a tree



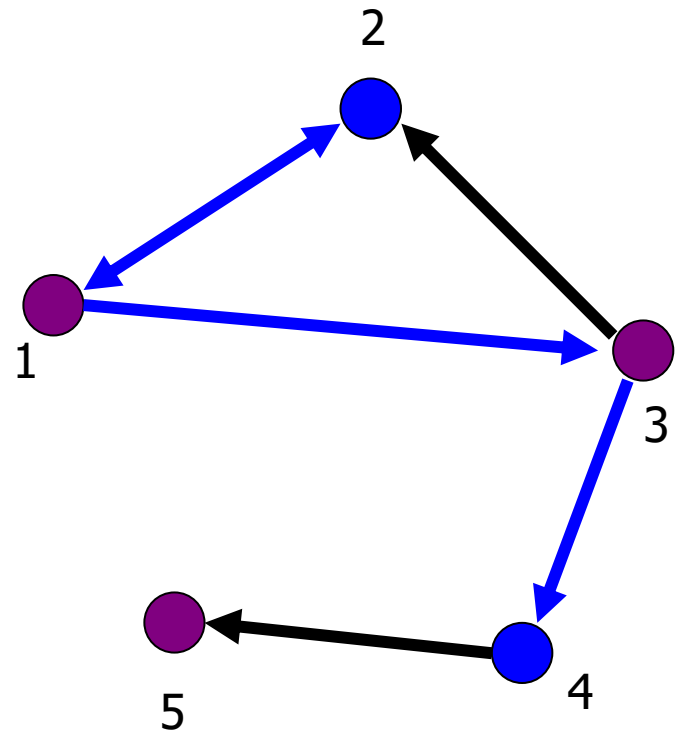
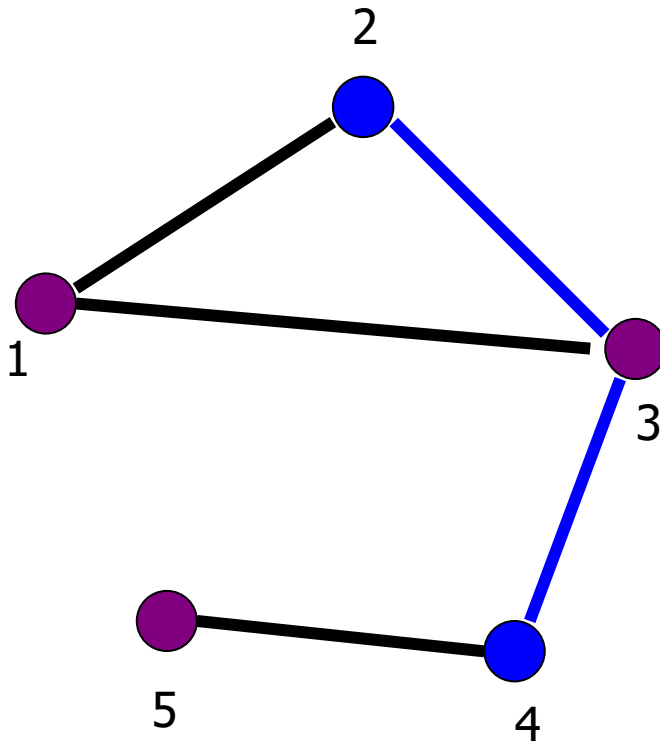
# Paths

- Path from node  $i$  to node  $j$ : a sequence of edges (directed or undirected) from node  $i$  to node  $j$ 
  - path **length**: number of edges on the path
  - nodes  $i$  and  $j$  are **connected**
  - **cycle**: a path that starts and ends at the same node



# Shortest Paths

- Shortest Path from node  $i$  to node  $j$ 
  - also known as **BFS path**, or **geodesic path**
  - We can find all shortest paths from a node using BFS

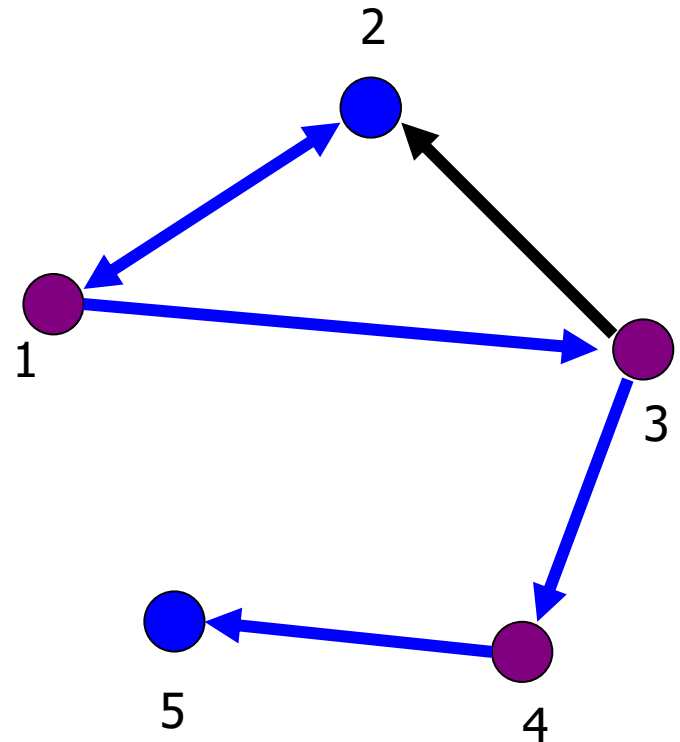
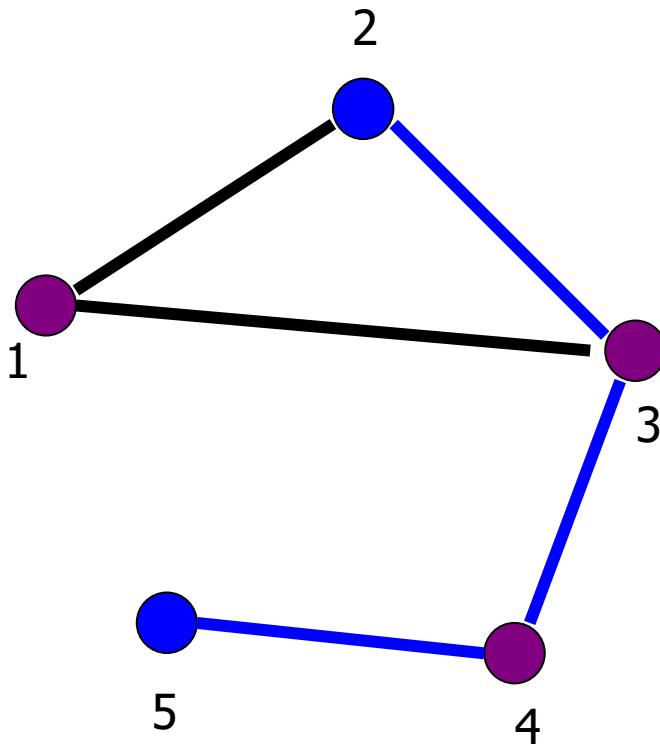


# Shortest paths on weighted graphs

- Shortest paths on **weighted** graphs are harder to construct
  - There are several well known algorithms for finding **single-source**, or **all-pairs** shortest paths
  - For example: **Dijkstra's Algorithm**

# Diameter

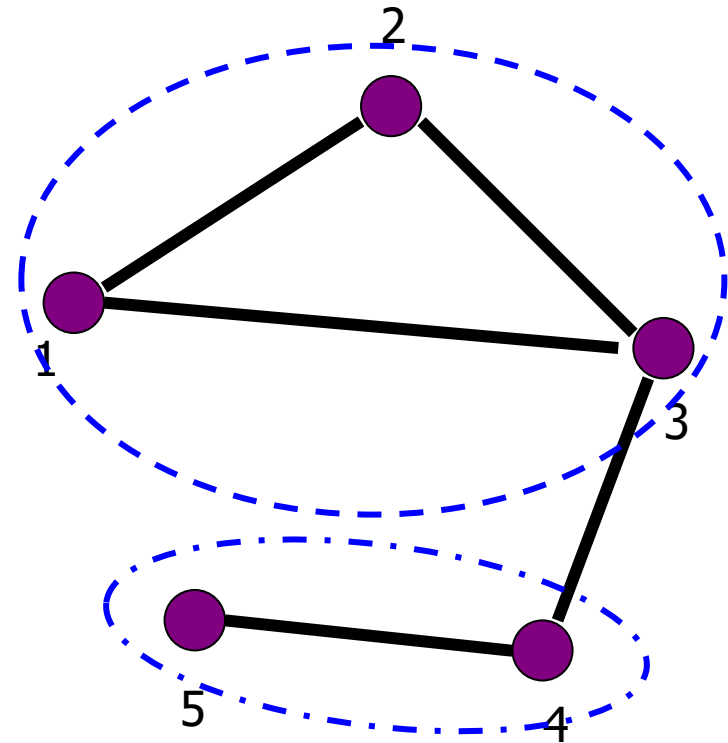
- The **longest shortest path** in the graph





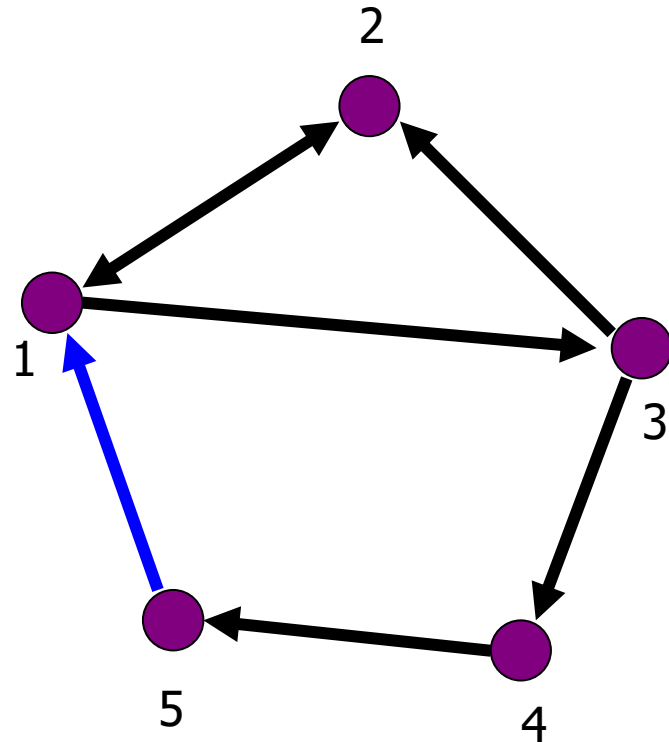
# Undirected graph

- **Connected** graph: a graph where every pair of nodes is connected
- **Disconnected** graph: a graph that is not connected
- **Connected Components:** subsets of vertices that are connected



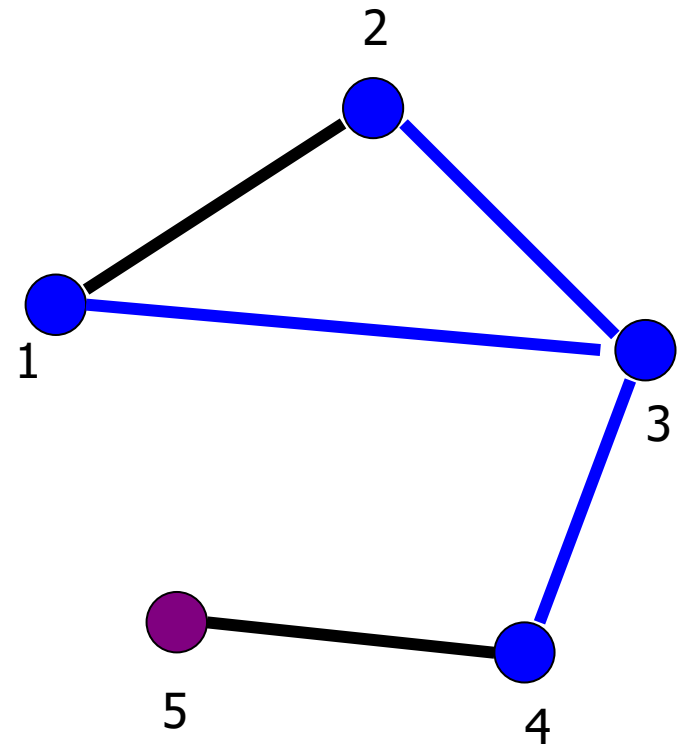
# Directed Graph

- **Strongly connected graph:** there exists a path from every  $i$  to every  $j$
- **Weakly connected graph:** If edges are made to be undirected the graph is connected



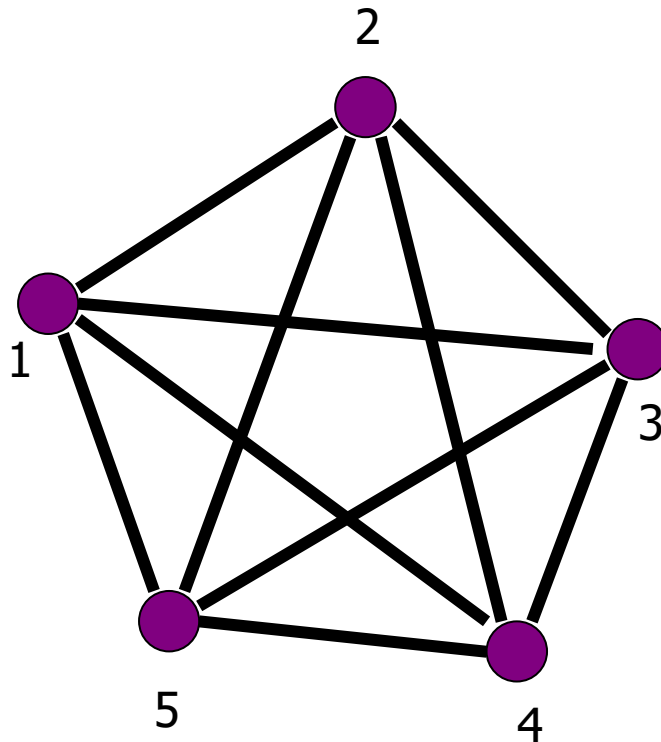
# Subgraphs

- **Subgraph**: Given  $V' \subseteq V$ , and  $E' \subseteq E$ , the graph  $G'=(V',E')$  is a subgraph of  $G$ .
- **Induced subgraph**: Given  $V' \subseteq V$ , let  $E' \subseteq E$  is the set of all edges between the nodes in  $V'$ . The graph  $G'=(V',E')$ , is an induced subgraph of  $G$



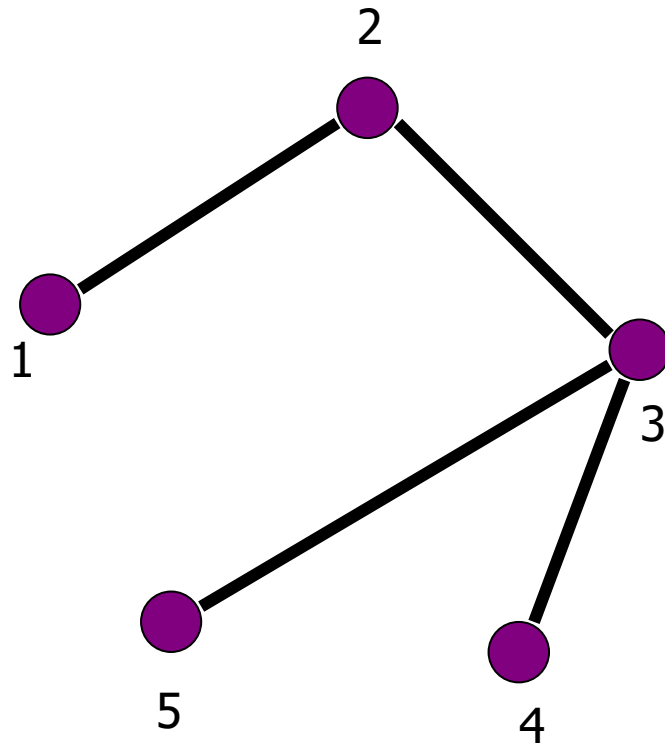
# Fully Connected Graph

- **Clique**  $K_n$
- A graph that has all possible  $n(n-1)/2$  edges



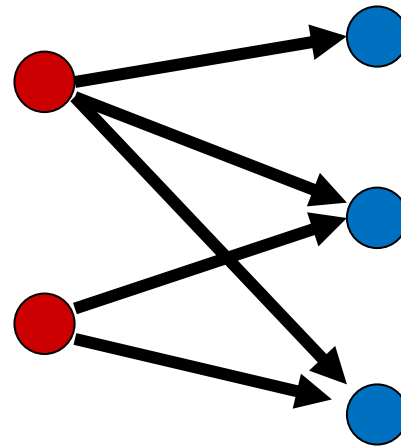
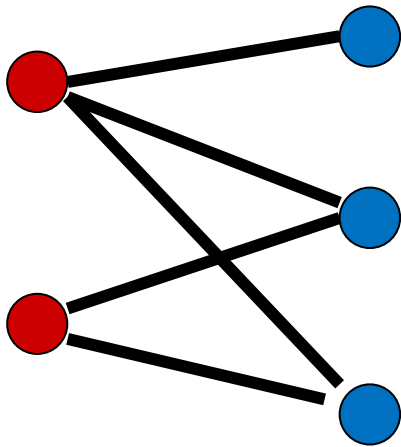
# Trees

- Connected Undirected graphs without cycles



# Bipartite graphs

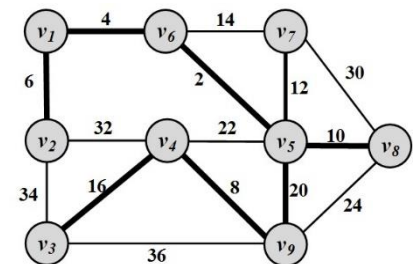
- Graphs where the set of nodes  $V$  can be partitioned into two sets  $L$  and  $R$ , such that there are edges only between nodes in  $L$  and  $R$ , and there is no edge within  $L$  or  $R$



# Spanning Tree

- For any connected graph, the **spanning tree** is a subgraph and a tree that includes all the nodes of the graph
- There may exist multiple spanning trees for a graph.
- For a weighted graph and one of its spanning tree, the weight of that spanning tree is the summation of the edge weights in the tree.
- Among the many spanning trees found for a weighted graph, the one with the minimum weight is called the

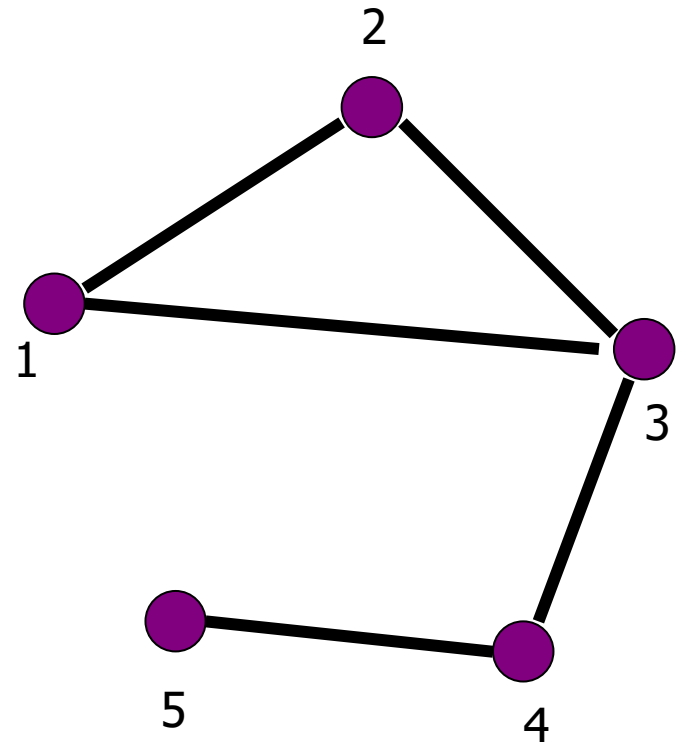
**minimum spanning tree (MST)**



# Graph Representation

- Adjacency Matrix
  - **symmetric** matrix for undirected graphs

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

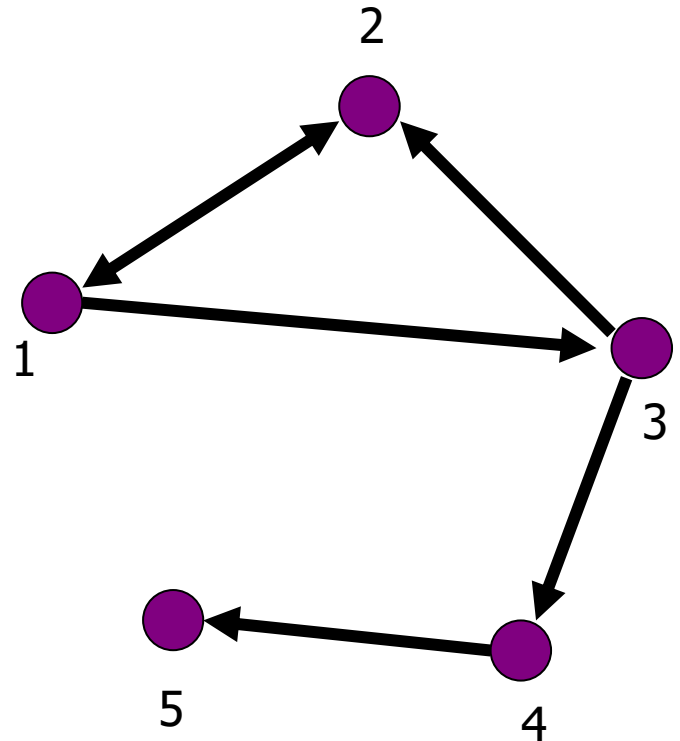




# Graph Representation

- Adjacency Matrix
  - **unsymmetric** matrix for undirected graphs

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



# Graph Representation

- Adjacency List
  - For each node keep a list with neighboring nodes

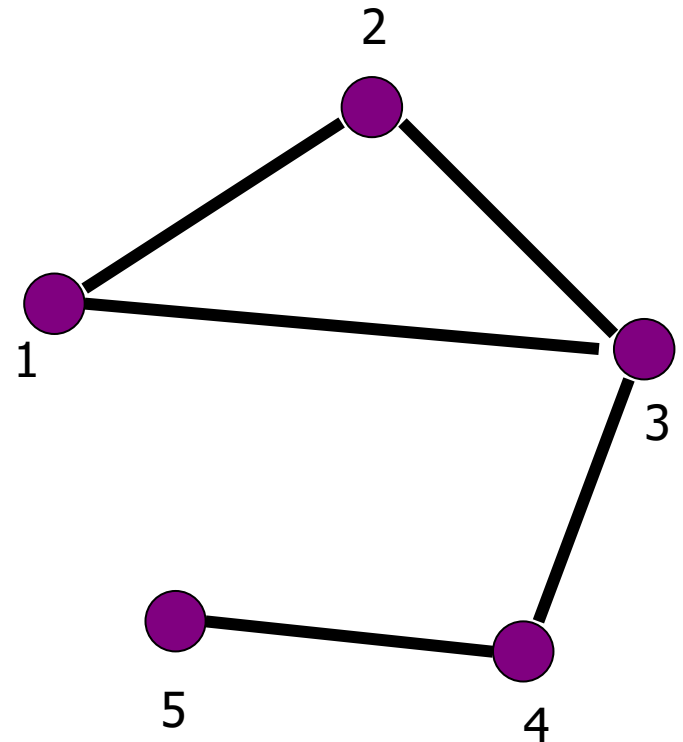
1: [2, 3]

2: [1, 3]

3: [1, 2, 4]

4: [3, 5]

5: [4]



# Graph Representation

- Adjacency List
  - For each node keep a list of the nodes it points to

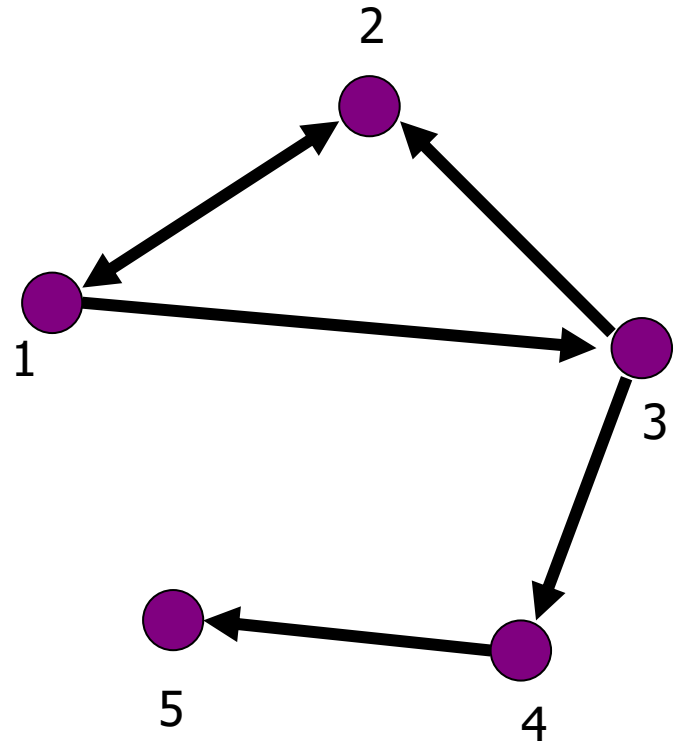
1: [2, 3]

2: [1]

3: [2, 4]

4: [5]

5: [null]



# Graph Representation

- List of edges
  - Keep a list of all the edges in the graph

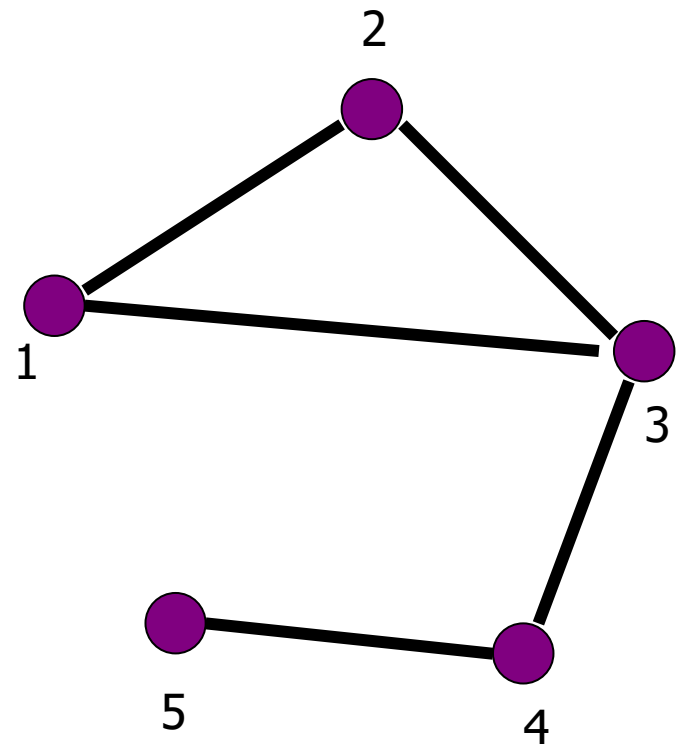
(1,2)

(2,3)

(1,3)

(3,4)

(4,5)



# Graph Representation

- List of Edges
  - Keep a list of all the directed edges in the graph

(1,2)

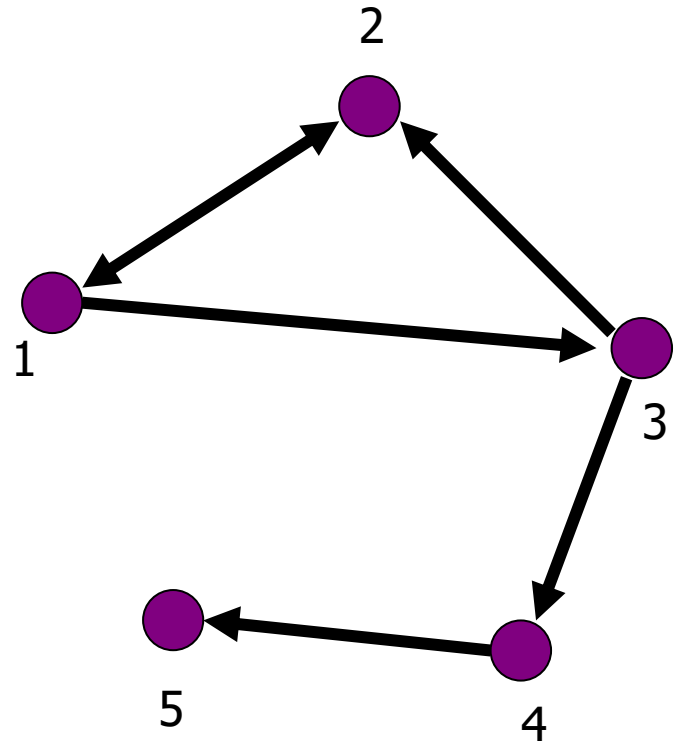
(2,1)

(1,3)

(3,2)

(3,4)

(4,5)



# P and NP

- **P**: the class of problems that can be **solved** in polynomial time
- **NP**: the class of problems that can be **verified** in polynomial time, but there is **no known solution** in polynomial time
- **NP-hard**: problems that are at least as hard as any problem in **NP**

# Introduction to Notebooks

# Notebooks

- Notebooks offer an interactive Python programming environment for coding, that combines code, text, and output.
- It is often used in Data Science for presenting the analysis output
- We will use Notebooks in class for testing in practice the different algorithms and models we present.
- For the course assignment you will maintain one or more notebooks that you will update with assignments on the latest material we cover in class.



# Python

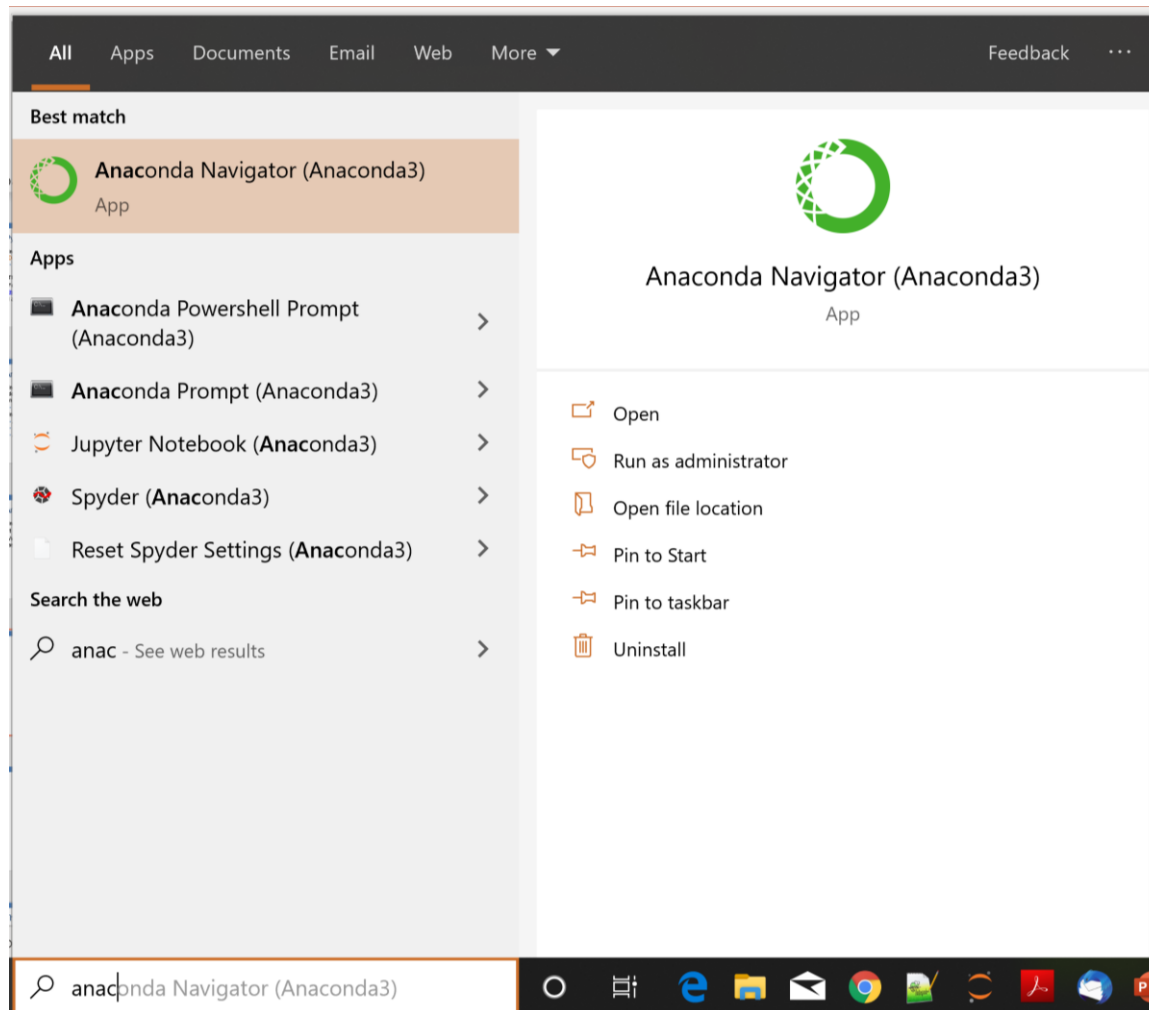
- In the last few years there is an increasing community that creates **Data Mining tools in Python**
  - Python is overwhelmingly used today for data science tasks
  - It is also heavily used in industry
  - We will use Python for this class.
- There are tons of resources online for Python.
- For an introduction you can also look at the slides of the [Introduction to Programming](#) course by prof. N. Mamoulis
- I will assume you have installed Python to your laptop, and that you have a good knowledge of programming in Python.

# Anaconda

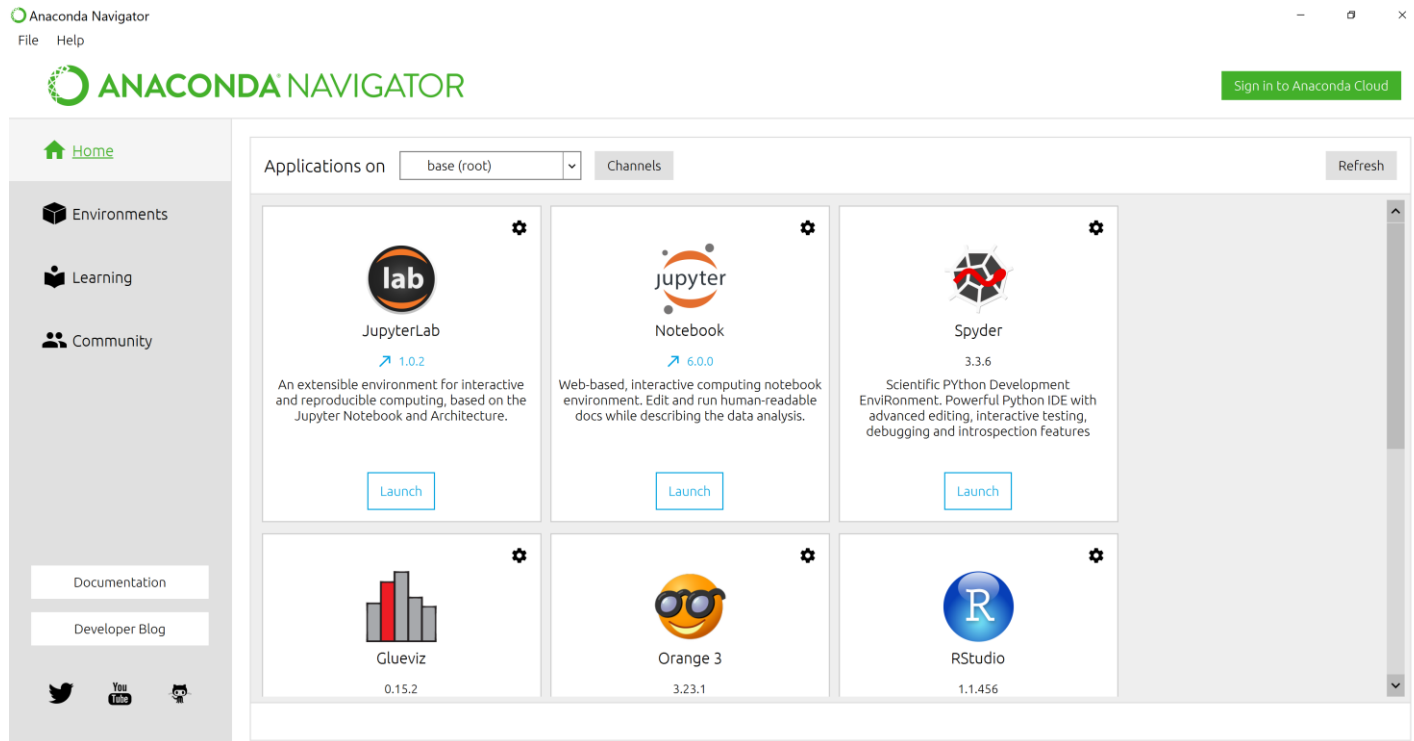
- Installing libraries in Python can be complicated, so you should download the **Anaconda Scientific Python** distribution which will install most of the libraries that we will use.
  - Use Python 3.0
- Installing Anaconda installs a lot of libraries and also:
  - Anaconda Navigator
  - **Jupyter Notebook**: An interactive web-based interface for running python.
  - Anaconda Powershell: terminal for running commands

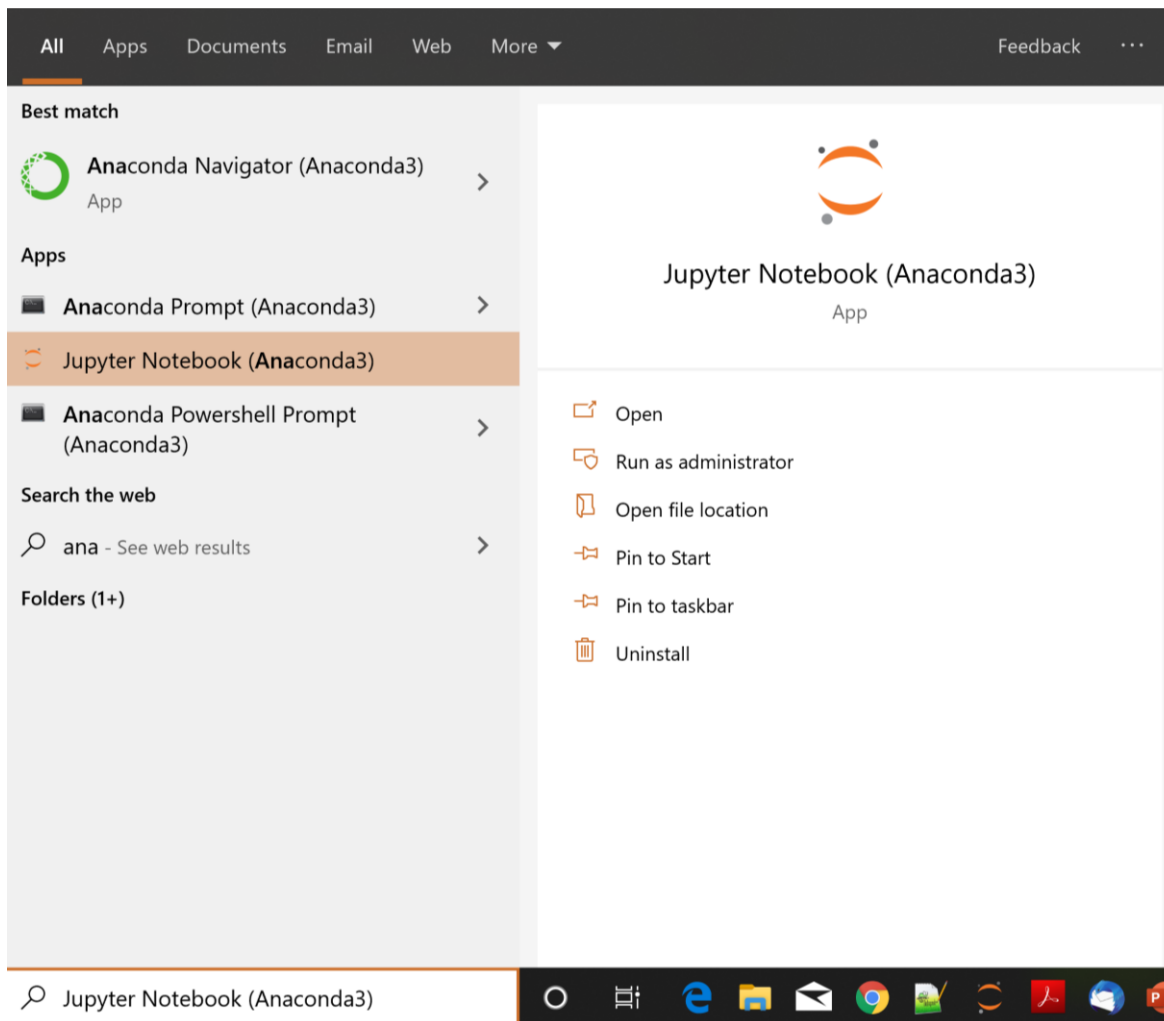
# Jupyter Notebook

- Installing Anaconda will also install Jupyter Notebook.
- If you wish to install it in a different way, together with the relevant libraries you are free to do so.
- We will use Notebook for our examples and it is **required** for the assignments.
  - In almost all assignments you are required to submit a Notebook.



# The Anaconda Navigator







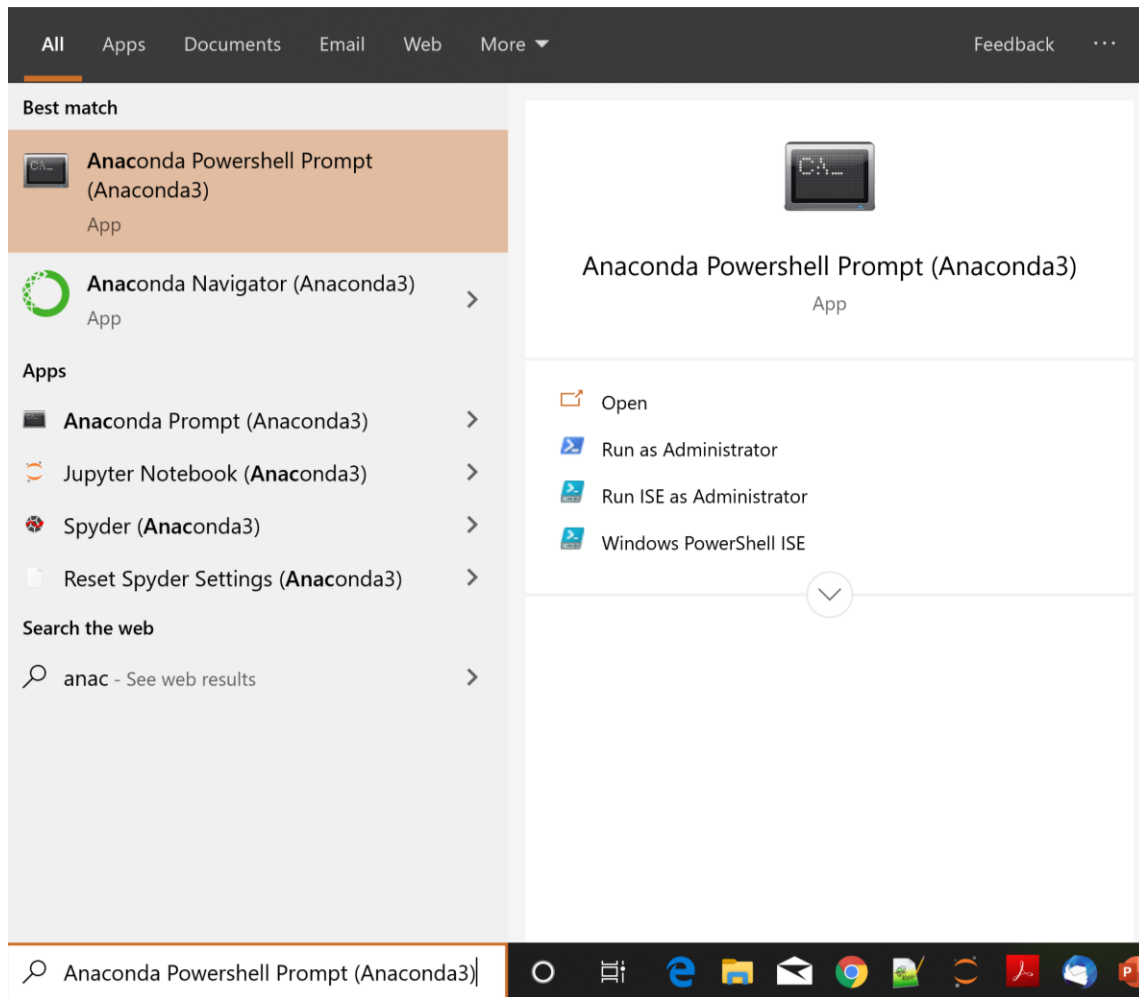
Quit Logout

Files Running Clusters

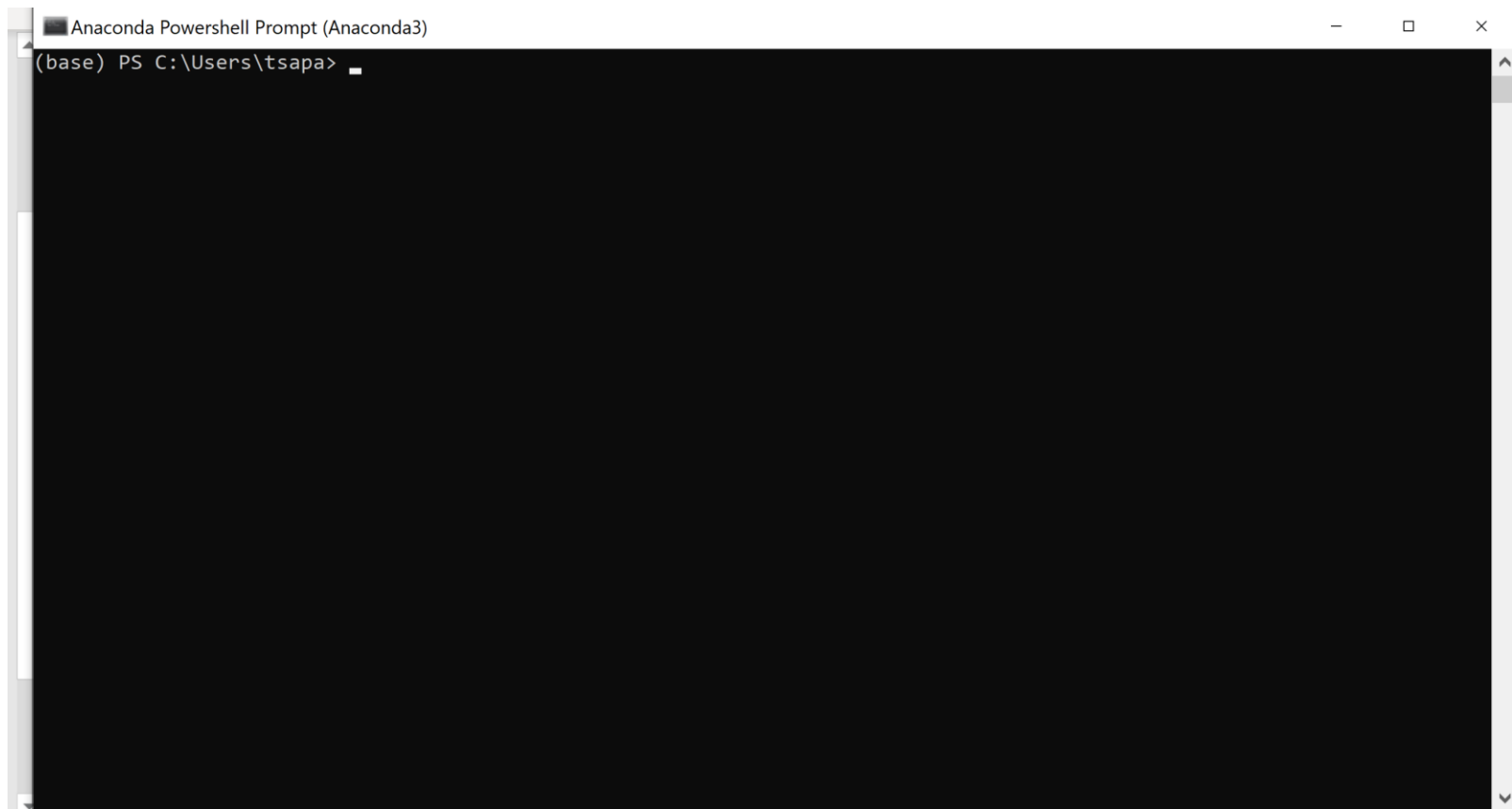
Select items to perform actions on them.

Upload New ↕

<input type="checkbox"/> 0 ▾	/	Name ▾	Last Modified	File size
<input type="checkbox"/>	3D Objects		21 days ago	
<input type="checkbox"/>	Contacts		21 days ago	
<input type="checkbox"/>	Documents		3 months ago	
<input type="checkbox"/>	Downloads		4 hours ago	
<input type="checkbox"/>	Dropbox (Personal)		8 days ago	
<input type="checkbox"/>	Dropbox (Uol)		2 months ago	
<input type="checkbox"/>	Favorites		10 days ago	
<input type="checkbox"/>	Google Drive		3 days ago	
<input type="checkbox"/>	Links		21 days ago	
<input type="checkbox"/>	Music		21 days ago	
<input type="checkbox"/>	OneDrive		2 days ago	
<input type="checkbox"/>	Roaming		5 months ago	
<input type="checkbox"/>	Saved Games		21 days ago	
<input type="checkbox"/>	Searches		21 days ago	





A screenshot of an Anaconda Powershell Prompt window. The window has a title bar with the text "Anaconda Powershell Prompt (Anaconda3)" and standard Windows window controls (minimize, maximize, close). The main area is black, and the prompt text "(base) PS C:\Users\tsapa>" is displayed in white at the top left. A small white cursor is visible at the end of the prompt line. The window is slightly offset from the top-left corner of the screen.

```
(base) PS C:\Users\tsapa> 
```

# Installing Packages

- You can install packages from the Anaconda terminal using the **conda install** command:
  - `conda install <name of package>`
- For example, [Seaborn](#) is a package for Statistical Data Visualization.
  - `conda install seaborn`
- [panda-datareader](#) is a package for loading online datasets.
  - `conda install pandas-datareader`
- You can also use the **pip install** command.
  - `pip install snap-stanford`

# Notebooks

- Jupyter Notebook offers an interactive web-based interface for running code.
- The Notebook runs inside a browser.
- It allows you to interact with the code, running different parts of the code
- The results also appear in the browser, so you can have together the code and the results
- You can also add text, commenting on the results.
- We will now see some details on how to create notebooks

# Changing the notebook default directory

- This used to be important before but now Jupyter Notebook takes you to the home directory
- From the Anaconda terminal type the command:  
➤ `jupyter notebook --generate-config`
- This will generate `.jupyter/jupyter_notebook_config.py` file under your home directory.
- Find, un-comment and modify the line  
`# c.NotebookApp.notebook_dir = ''` in the config file to point to the desired directory



Quit

Logout

Files

Running

Clusters

Nbextensions

Select items to perform actions on them.

<input type="checkbox"/> 0	/		
<input type="checkbox"/>	3D Objects		
<input type="checkbox"/>	Contacts		
<input type="checkbox"/>	Documents		
<input type="checkbox"/>	Downloads		
<input type="checkbox"/>	Dropbox (Personal)		10 months ago
<input type="checkbox"/>	Dropbox (Uol)		2 months ago
<input type="checkbox"/>	Favorites		8 months ago
<input type="checkbox"/>	gensim-data		a year ago
<input type="checkbox"/>	Google Drive		4 days ago
<input type="checkbox"/>	IdeaProjects		18 days ago
<input type="checkbox"/>	Links		8 months ago
<input type="checkbox"/>	Music		8 months ago
<input type="checkbox"/>	OneDrive		2 days ago
<input type="checkbox"/>	OneDrive - ΠΑΝΕΠΙΣΤΗΜΙΟ ΙΩΑΝΝΙΝΩΝ		2 days ago
<input type="checkbox"/>	Roaming		2 years ago

Upload New

Notebook:

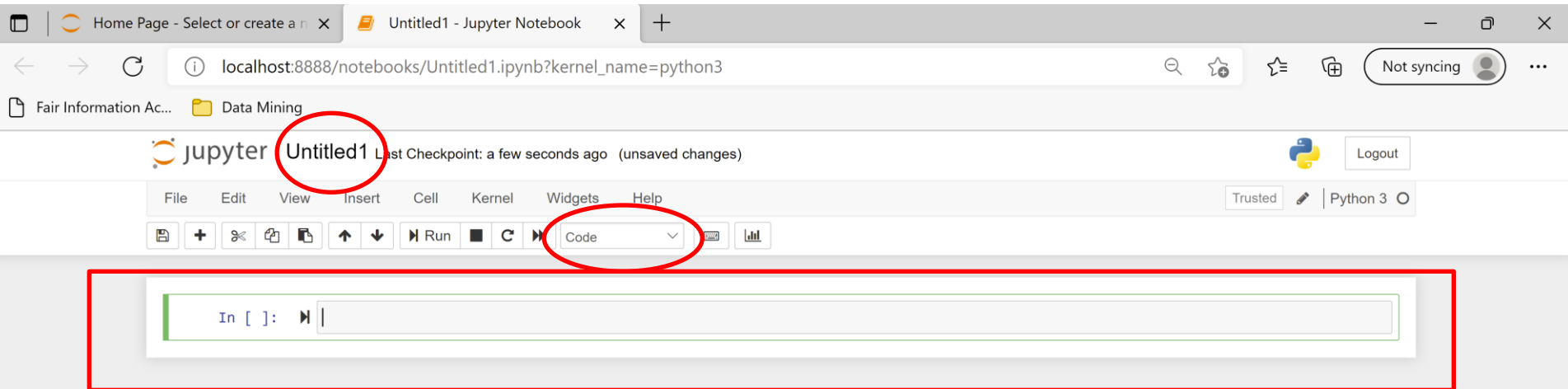
Python 3

Other:

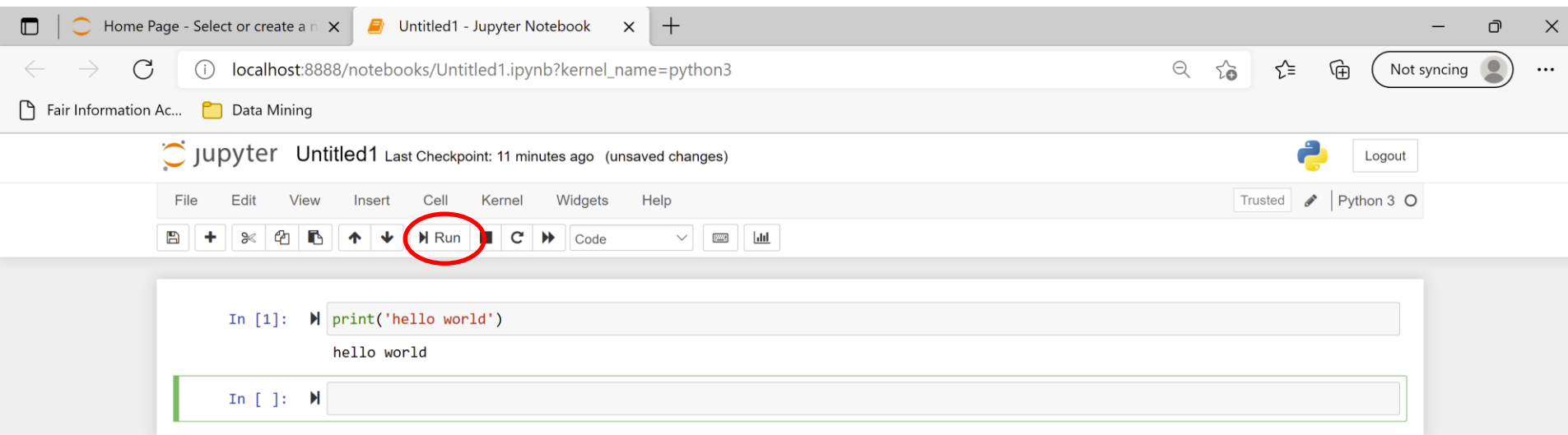
Text File

Folder

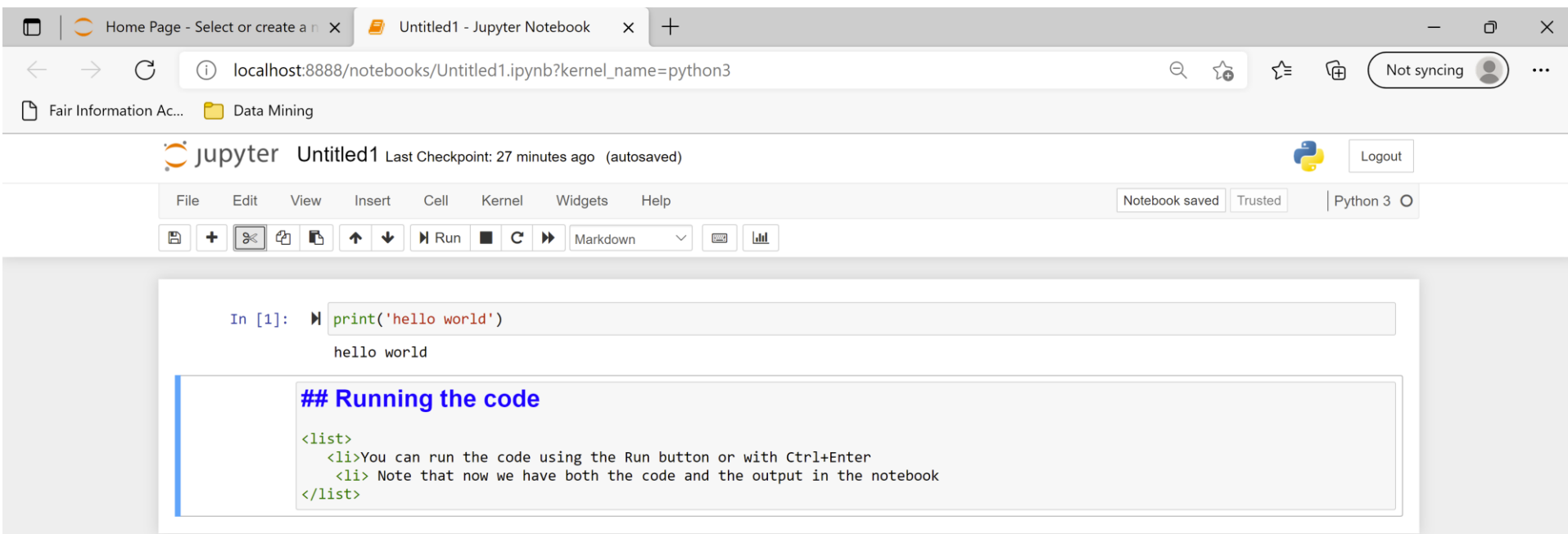
Terminal



- The notebook is organized in cells
- In each cell you can write either code or text
- The default behavior is code



- You can run the code using the Run button or with Ctrl+Enter
- Note that now we have both the code and the output in the notebook



- You can also write text in Markdown language
  - You can combine HTML, and Latex, and there are some other commands
- You can learn more about Markdown by searching online, e.g.:
  - [Learn How to Write Markdown & LaTeX in The Jupyter Notebook | by Khelifi Ahmed Aziz | Towards Data Science](#)
- You need to Run the Markdown cell as well



Run button circled in red

```
In [1]: print('hello world')
hello world
```

### Running the code

- You can run the code using the Run button or with Ctrl+Enter
- Note that now we have both the code and the output in the notebook

```
In [ ]:
```

The screenshot shows a web browser window with the URL `localhost:8888/notebooks/Untitled1.ipynb?kernel_name=python3`. The Jupyter Notebook interface is visible, with the 'File' menu open. The 'Download as' option is selected, and a sub-menu is displayed with the following options: AsciiDoc (.asciidoc), HTML (.html), LaTeX (.tex), Markdown (.md), Notebook (.ipynb), PDF via LaTeX (.pdf), reST (.rst), Python (.py), and Reveal.js slides (.slides.html). The 'HTML (.html)' and 'PDF via LaTeX (.pdf)' options are circled in red. The background of the notebook shows a code cell with `print('hello world')` and its output, 'hello world', followed by a text cell titled 'Running the code'.

- You can export the notebook into HTML or PDF

# Attention!

- A notebook is run interactively, each time running a specific cell
- The state of the program remains in memory while the notebook is running
- Each cell has access to the current state of the memory
- You can jump between cells in a non-linear way
- You should be aware of the state of the memory of the notebook when you run a specific cell.

# A simple example

The screenshot shows a Jupyter Notebook with three code cells. The first cell, labeled 'In [1]:', contains the code `x = 2`. The second cell, labeled 'In [3]:', contains the code `y = 2*x+z` and `print(y)`, with the output `10` displayed below. The third cell, labeled 'In [2]:', contains the code `x = 4` and `z = 2`. A red circle highlights the cell numbers [1], [3], and [2] to illustrate the execution order.

- The order in which the cells are executed is shown in the increasing numbers (not always useful)
- The second in order cell is executed third
  - So it has access to `z`, and uses the value 4 for `x`

# Restarting

The screenshot displays the Jupyter Notebook web interface in a browser. The browser's address bar shows the URL `localhost:8888`. The Jupyter interface includes a top bar with the notebook title "Untitled1" and a status message "Last Checkpoint: Last Monday at 12:34 AM (autosaved)". Below this is a menu bar with options: File, Edit, View, Insert, Cell, Kernel, Widgets, and Help. A toolbar contains icons for file operations and running code. The "Kernel" menu is open, showing a list of actions: Interrupt, Restart, Restart & Clear Output, Restart & Run All, Reconnect, Shutdown, and Change kernel. The "Restart" option is circled in red. The notebook area contains three code cells:

```
In [1]: x = 2
```

```
In [3]: y = 2*x+z
print(y)
10
```

```
In [2]: x = 4
z = 2
```

The Windows taskbar at the bottom shows the search bar, task view button, and several open applications including Edge, File Explorer, and various productivity tools. The system clock indicates the time is 3:59 AM on 11/11/2021.

# Useful Links

- You can find some more information about notebooks in the [tutorials](#) for the Data Mining class
- Useful libraries for network processing:
  - [NetworkX](#)
  - [Snap](#)