

# **Graph Neural Networks**

## **Lecture 1**

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# Overview

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1. About the Course.
2. What are Graphs?
3. Examples of Graph Data
4. Motivation
5. Approaches to Graph Data Analysis Without Neural Networks
6. Choosing a Proper Representation
7. Types of Graph Tasks
8. Node-Level Tasks
9. Edge-level Tasks
10. Graph-level Tasks
11. Features of Graph Algorithms

## About the Course. Organizational Information

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- 12 lectures and seminars
- 5 homework assignments (3(2) theoretical, 2(3) practical)
- Lectures and seminars on Friday at 9:00

## About the Course. Grading

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- Final Grade =  $0.7 \cdot O_n + 0.3 \cdot O_e$ , rounded arithmetically
- $O_n = (\text{Total points scored})/7$
- 3(2) practical homework assignments (some points), 2(3) theoretical homework assignments (70 – some points)
- Exam can include:
  - Presentation with a discussion of an article
  - Or a conversational oral exam
- Automatic passing grade from 6.5 or higher based on cumulative score (provided all homework assignments are submitted)
- No block grading

## About the Course. Deadlines

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- Deadlines are flexible; a penalty of -10% is applied for each day of delay. After three days, the deadline becomes strict
- Twice per course, assignments can be submitted after the strict deadline (an additional 4 days are granted)
- Submissions are not accepted more than 7 days after the deadline
- The original score (without penalties) is considered for grading purposes

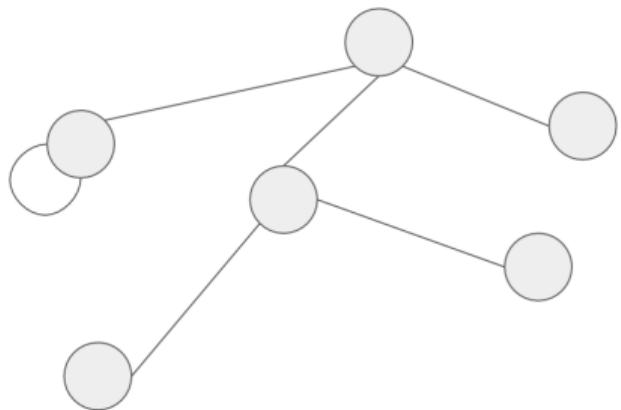
## About the Course. Homework Policy

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- Homework assignments must be completed individually
- No GPT, please

# What are Graphs?

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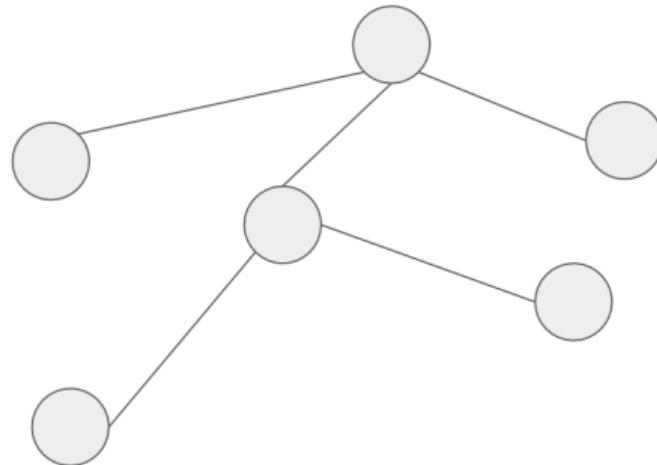


# What are Graphs?

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$G = (V, E)$ , where:

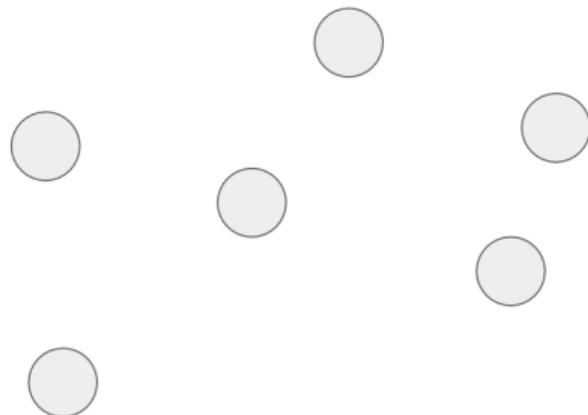
- $V$  is the set of vertices (or nodes), representing the objects.
- $E$  is the set of edges, representing the relationships or connections between the vertices.



# What are Graphs?

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Graph?



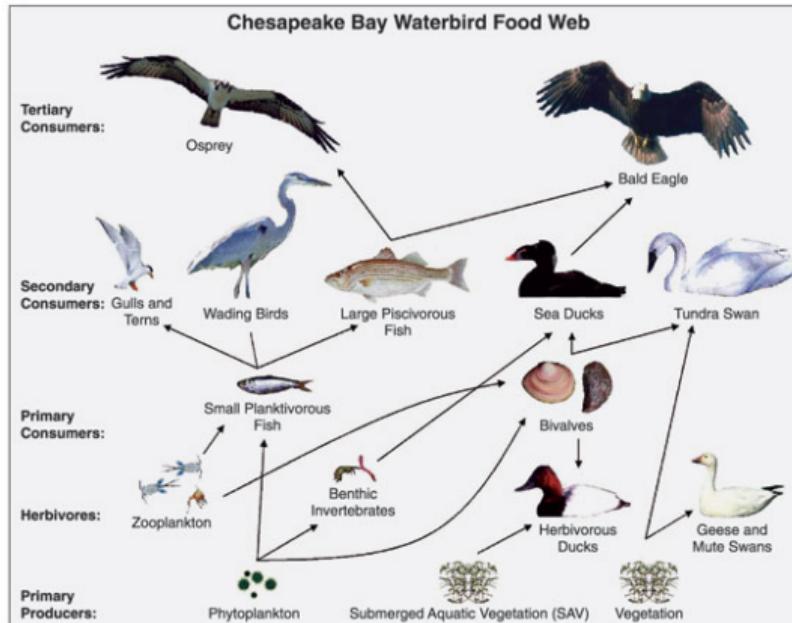
# Why Graphs are Needed in the Context of Data?

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- Graphs are a generalized way of representing any information related to interactions between objects.
- There are many different methods for graph analysis, making it convenient to transform data from various domains into a unified format.

# Examples of Graph Data

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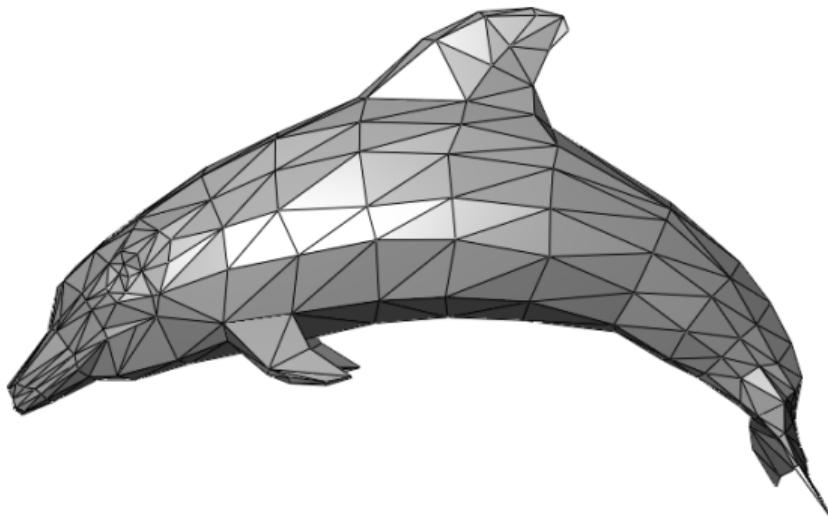
# Examples of Graph Data

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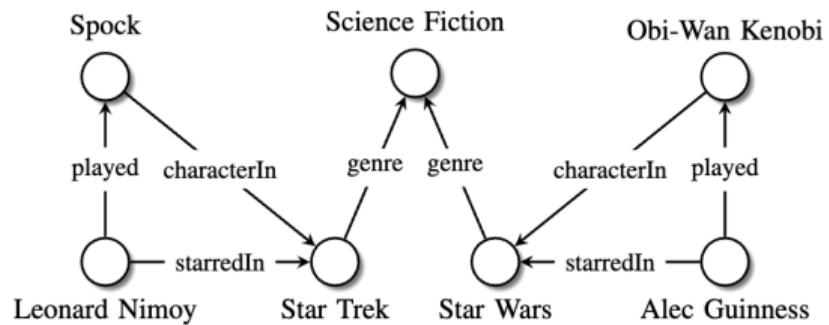
## Examples of Graph Data

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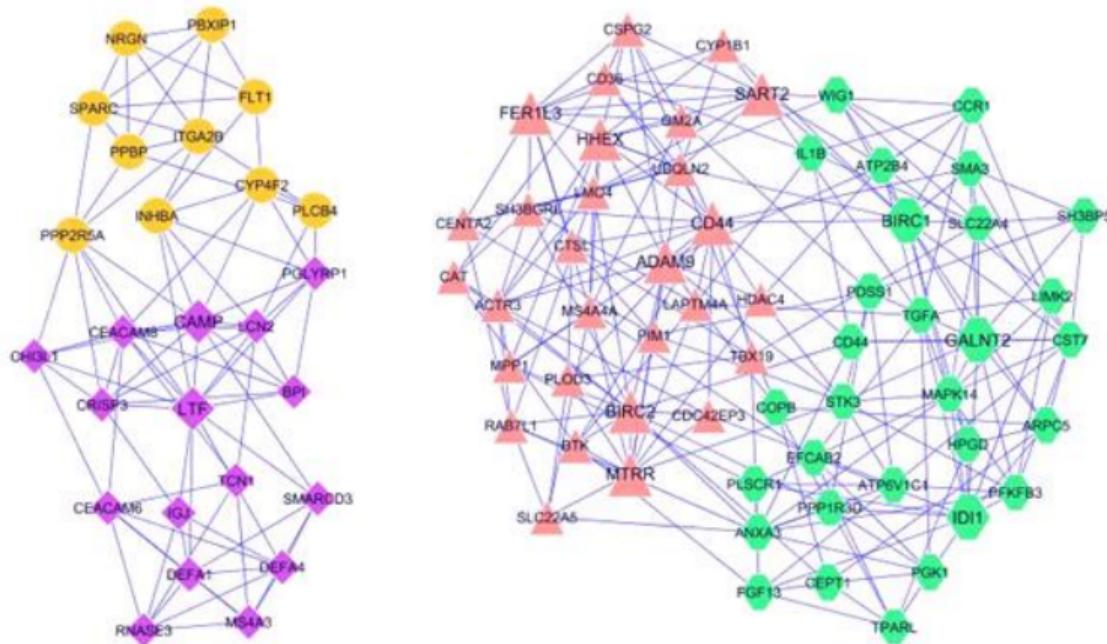
# Examples of Graph Data

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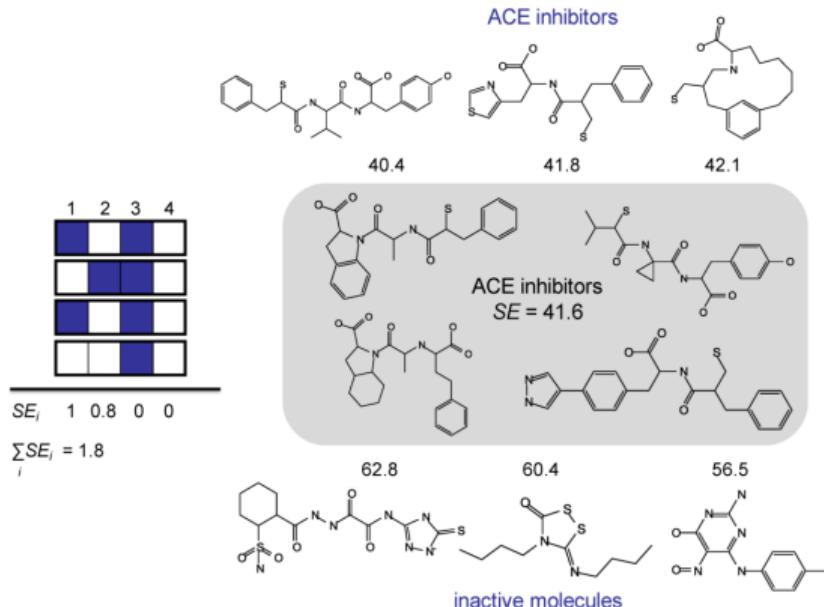


# Examples of Graph Data

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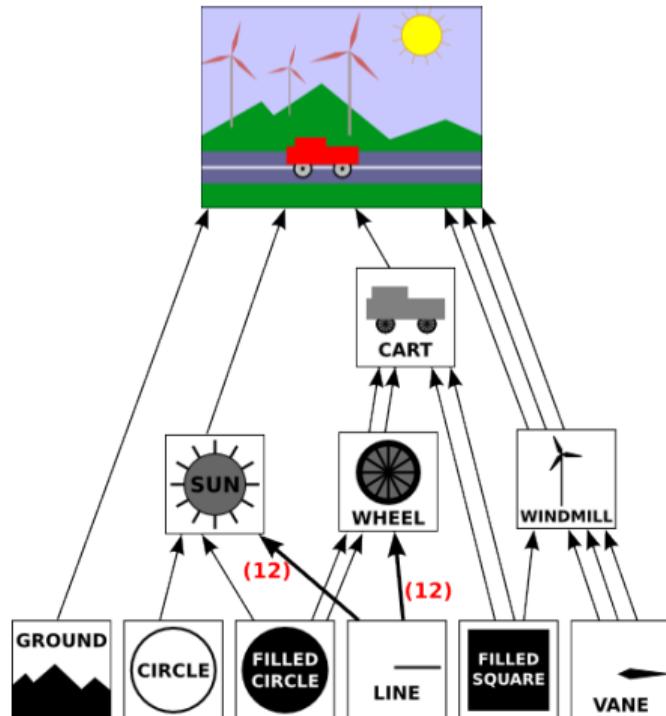


# Examples of Graph Data



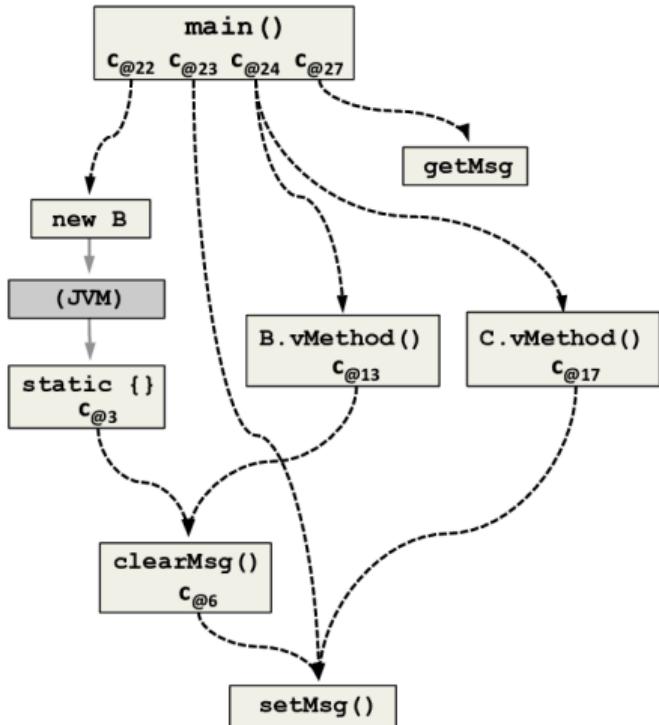
# Examples of Graph Data

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# Examples of Graph Data

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# Motivation

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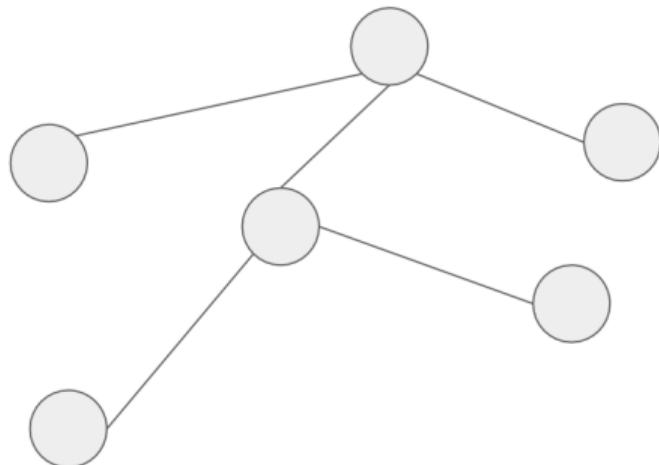
The primary applications of deep learning today involve sequential data:

- **NLP:** "askfnkksnf" - sequences of text
- **CV:** Structured data from image pixels
- **Audio:** Sound waves

# Motivation

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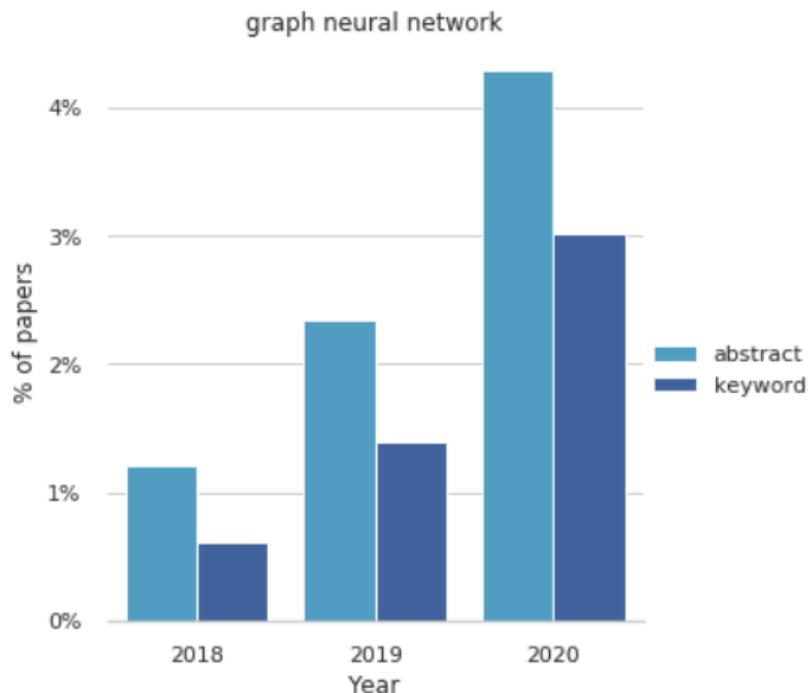
The distinctive feature of graphs is the relationships between objects, allowing information to be processed more broadly and differently.



**Where is the start and end?**

# Motivation

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# Approaches to Graph Data Analysis Without Neural Networks

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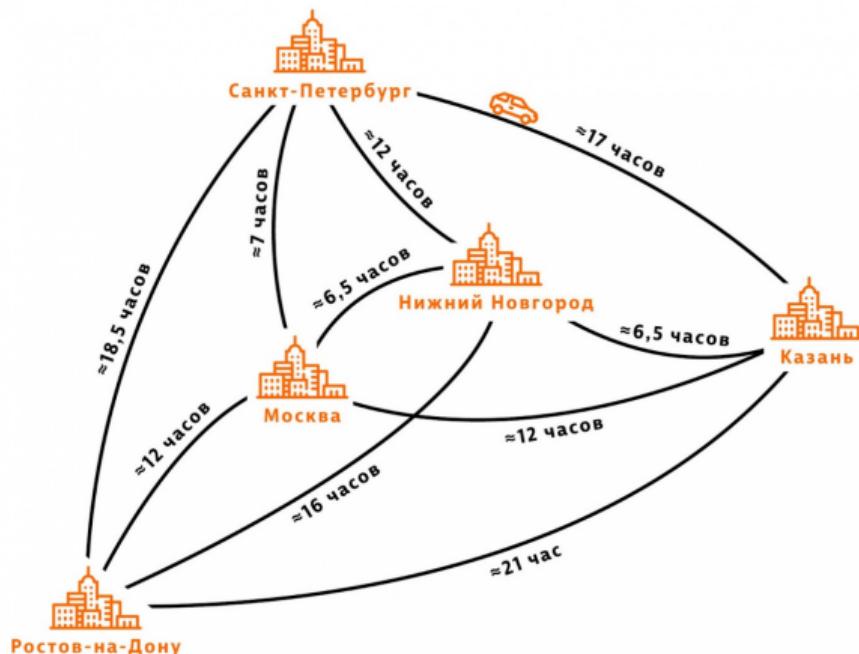
Many tasks that are currently solved using deep learning have existed for a long time in the context of graph data. Correspondingly, various methods for solving these tasks using conventional graph algorithms have also existed.

# Approaches to Graph Data Analysis Without Neural Networks



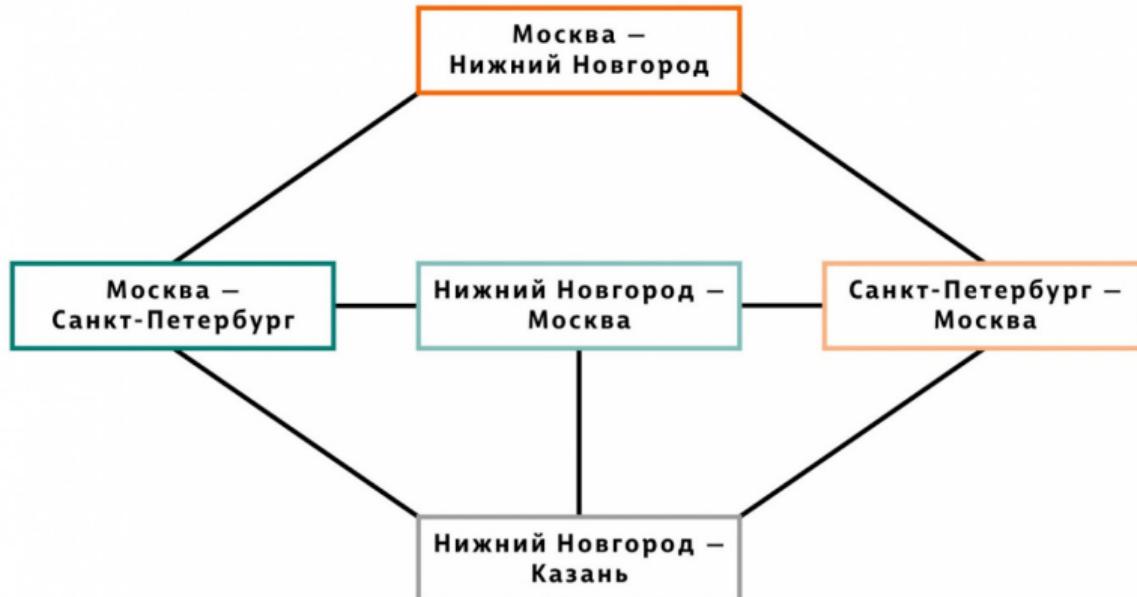
- Групповые вкусы пользователя - **начальные вершины обхода графа**
- Делаем 3 случайных шага по графу
- В каждой вершине есть вероятность телепортироваться в начало обхода
- Конечные вершины графа - **кандидаты для рекомендации**

# Approaches to Graph Data Analysis Without Neural Networks

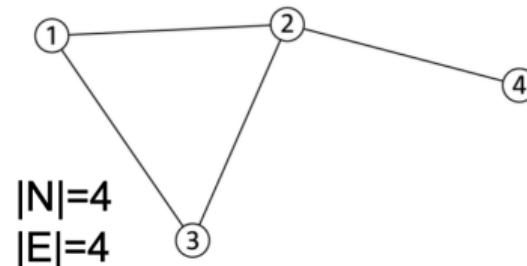
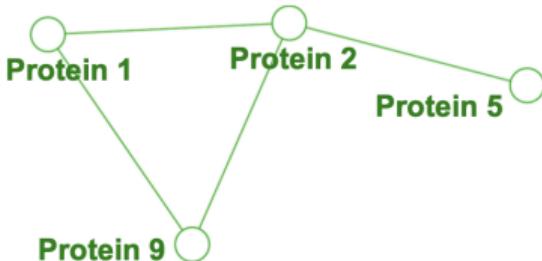
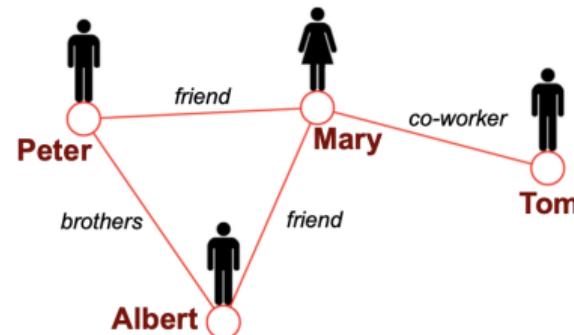
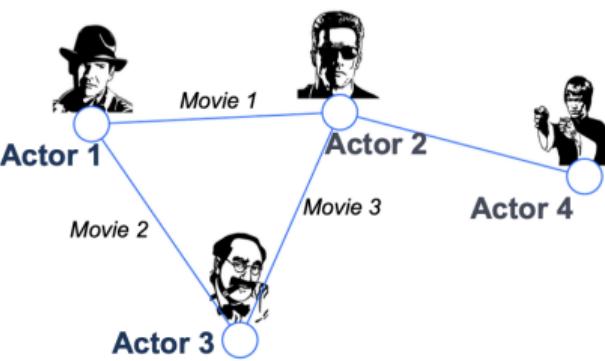


# Approaches to Graph Data Analysis Without Neural Networks

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# Finally, Graphs!



# Finally, Heterographs!

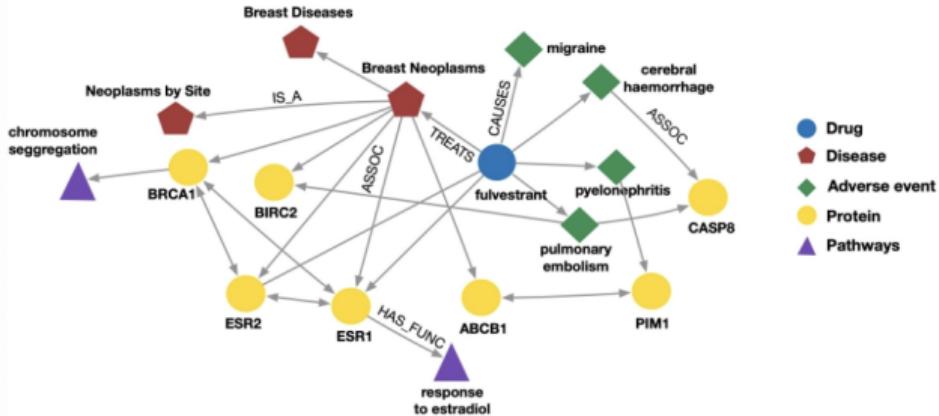
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A heterogeneous graph is defined as:

$$G = (\textcolor{blue}{V}, \textcolor{green}{E}, \textcolor{red}{R}, \textcolor{orange}{T})$$

- **Nodes with node types:**  $v_i \in \textcolor{blue}{V}$
- **Edges with relation types:**  $(v_i, r, v_j) \in \textcolor{green}{E}$
- **Node type:**  $\textcolor{orange}{T}(v_i)$
- **Relation type:**  $r \in \textcolor{red}{R}$
- **Nodes and edges have attributes/features.**

# Finally, Heterographs!



## Biomedical Knowledge Graphs

Example node: Migraine

Example edge: (fulvestrant, Treats, Breast Neoplasms)

Example node type: Protein

Example edge type (relation): Causes



## Academic Graphs

Example node: ICML

Example edge: (GraphSAGE, NeurIPS)

Example node type: Author

Example edge type (relation): pubYear

# Choosing a Proper Representation

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## How to build a graph?

- What are nodes?
- What are edges?

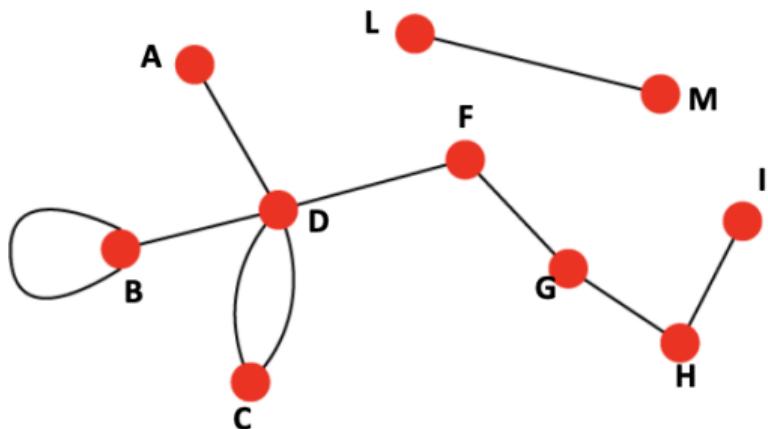
Choice of the proper network representation of a given domain/problem determines our ability to use networks successfully:

- In some cases, there is a unique, unambiguous representation.
- In other cases, the representation is by no means unique.
- The way you assign links will determine the nature of the question you can study.

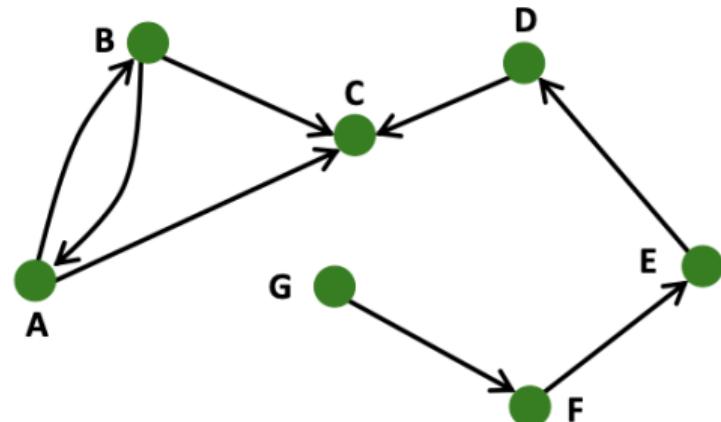
# Undirected vs Directed Graphs

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- **Links:** undirected (symmetrical, reciprocal)



- **Links:** directed



- Weights
- Properties

- Types
- Attributes

# Bipartite graph

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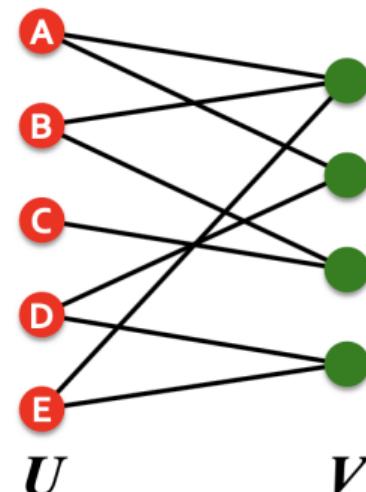
**Bipartite graph** is a graph whose nodes can be divided into two disjoint sets  $U$  and  $V$  such that every link connects a node in  $U$  to one in  $V$ ; that is,  $U$  and  $V$  are **independent sets**.

- **Examples:**

- Authors-to-Papers (they authored)
- Actors-to-Movies (they appeared in)
- Users-to-Movies (they rated)
- Recipes-to-Ingredients (they contain)

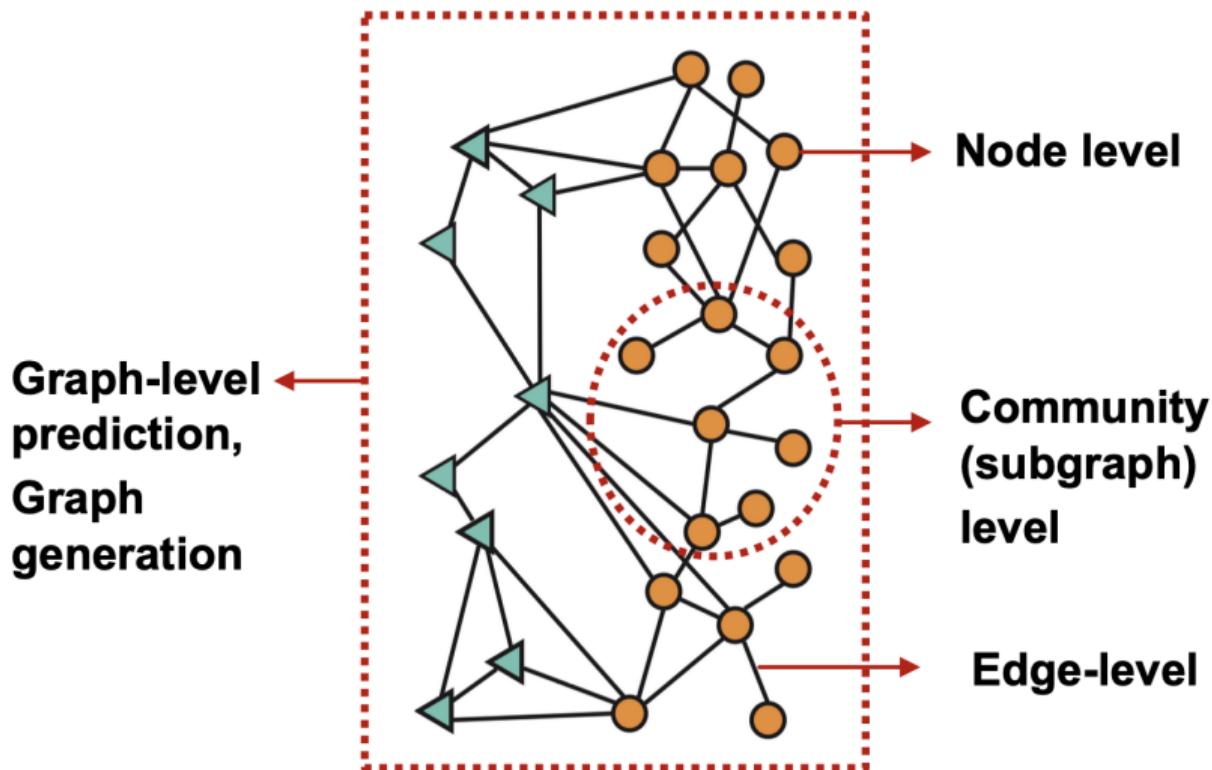
- **"Folded" networks:**

- Author collaboration networks
- Movie co-rating networks



# Types of Graph Tasks

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# Types of Graph Tasks

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1. **Edge-level** – Tasks related to edges:

- Link classification
- Regression
- Graph completion

2. **Node-level** – Tasks related to nodes:

- Node classification
- Regression
- Clustering

3. **Graph-level** – Tasks specific to the entire graph:

- Graph classification
- Regression
- Generation
- Evolution

# Node-Level Tasks

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**Determining a person's interests based on their surroundings**

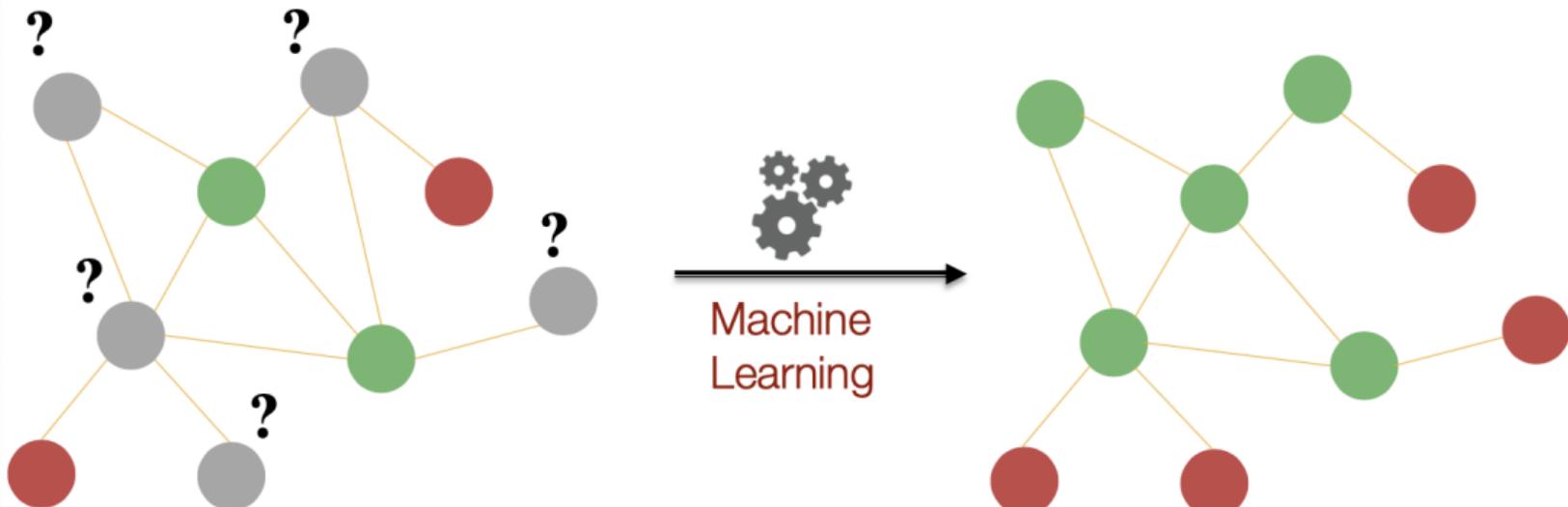
We aim to learn the function:

$$f : (V, E) \rightarrow \mathbb{R}^{|V|}$$

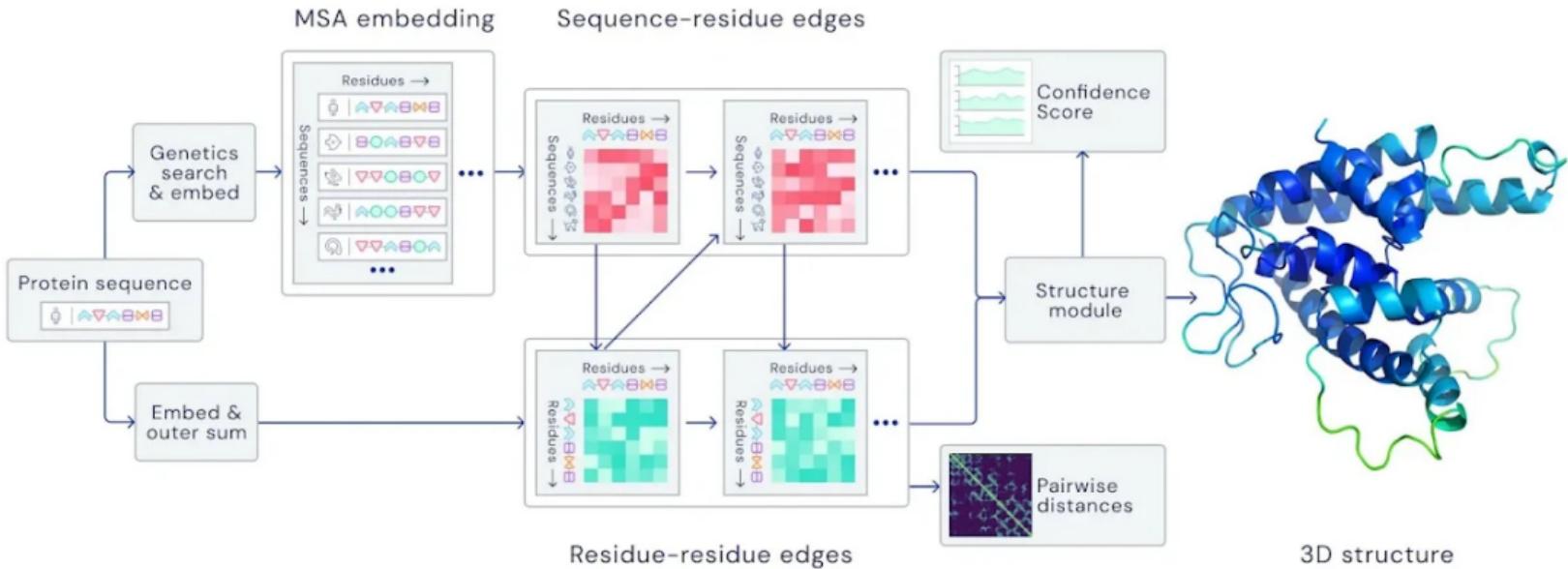
where  $V$  is the set of nodes,  $E$  is the set of edges, and the result is a numerical representation for each node in the graph.

# Node-Level Tasks

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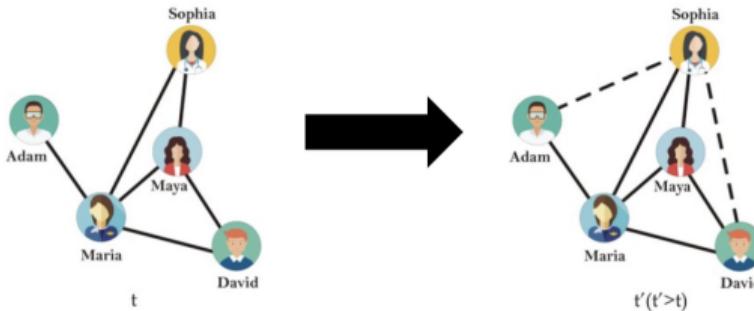


# Node-Level Tasks



# Edge-level Tasks

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**Make friend recommendations in a social network**

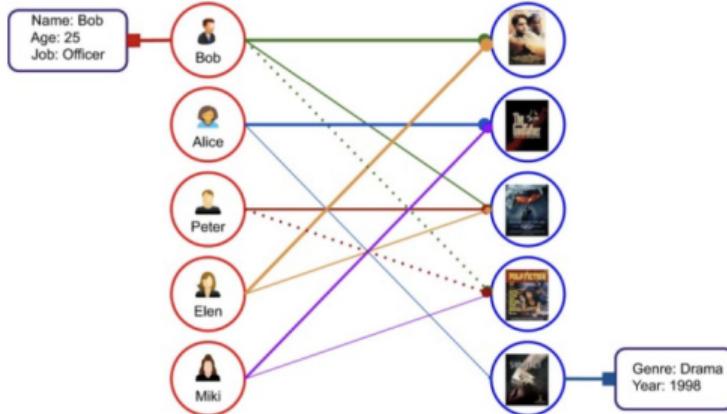
We aim to make predictions for all pairs  $(v_i, v_j)$ :

$$f(v_i, v_j)$$

where  $v_i$  and  $v_j$  are nodes in the graph.

# Edge-level Tasks

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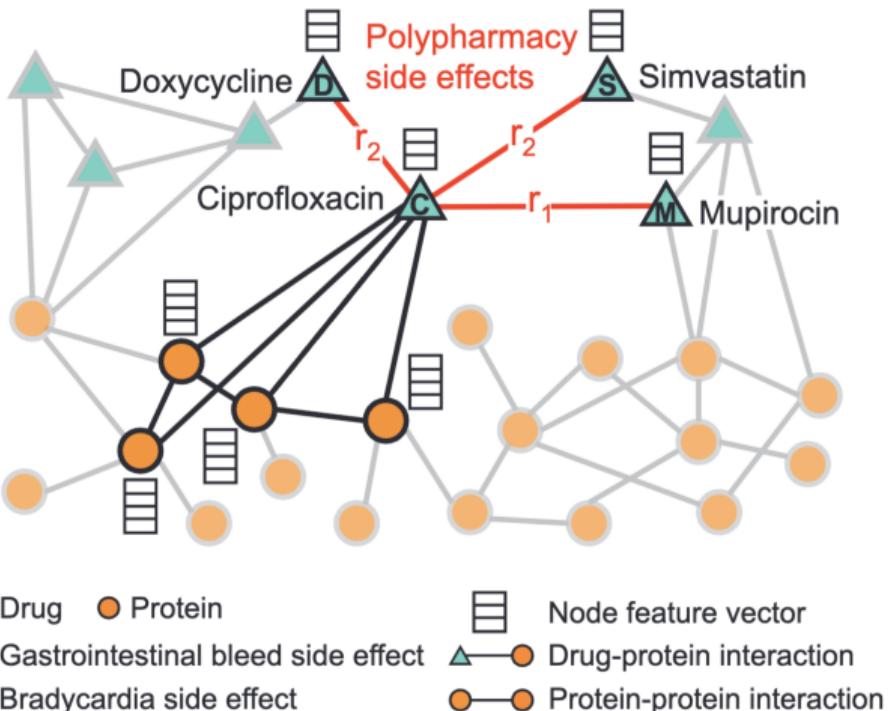
**Make recommendations for movies or music**

We aim to make predictions for all pairs  $(v_i, v_j)$ :

$$f(v_i, v_j)$$

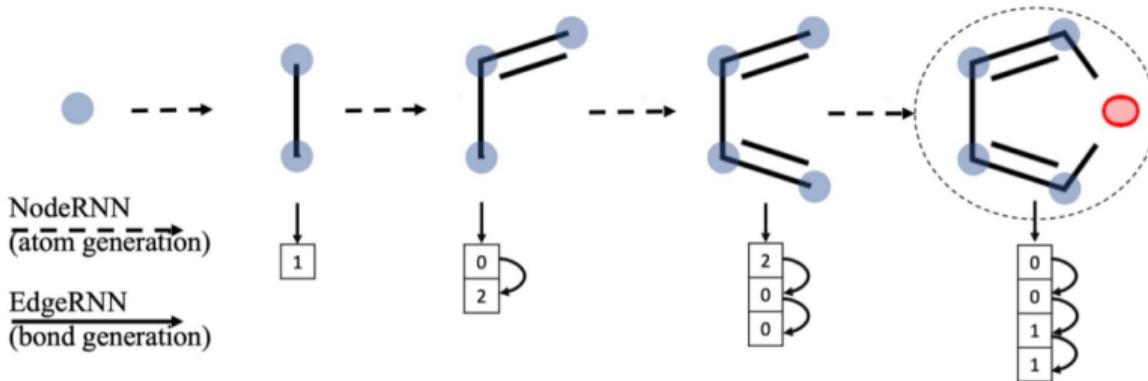
where  $v_i$  represents a user and  $v_j$  represents an item (e.g., movie or music).

# Edge-level Tasks



# Graph-level Tasks

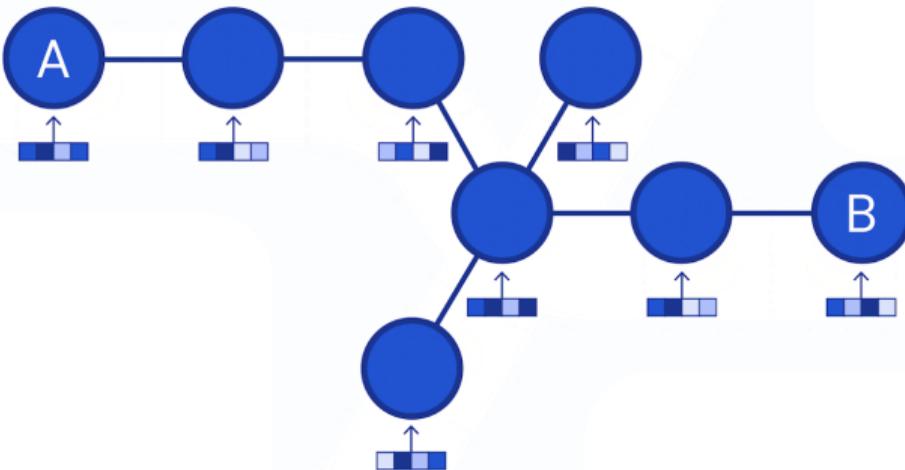
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We aim to learn how to generate graphs with specific characteristics.

## Subgraph-level Tasks

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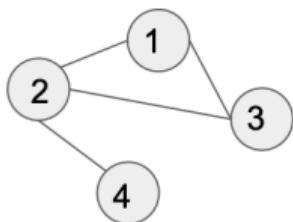
**Predicting vehicle arrival time:**

- The road is divided into segments (nodes).
- Accessibility between segments (edges) determines connectivity.

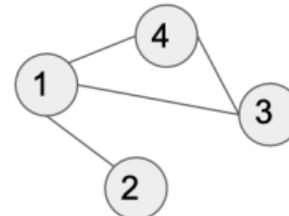
# Features of Graph Algorithms

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Graph algorithms must operate independently of the permutation of vertex indices.



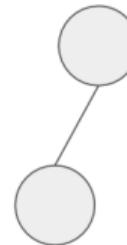
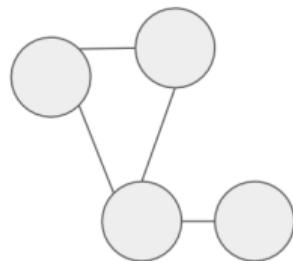
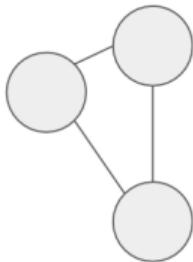
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# Features of Graph Algorithms

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Graph algorithms must be able to handle graphs of different sizes as input.



# The End?