



Mining & Learning on Graphs

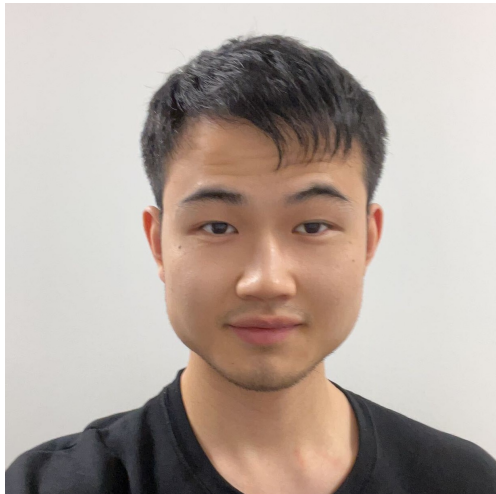
Course Overview and Logistics

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Assistant Professor
Computer and Information Science
University of Oregon
CS 410/510 - Fall 2024

Welcome

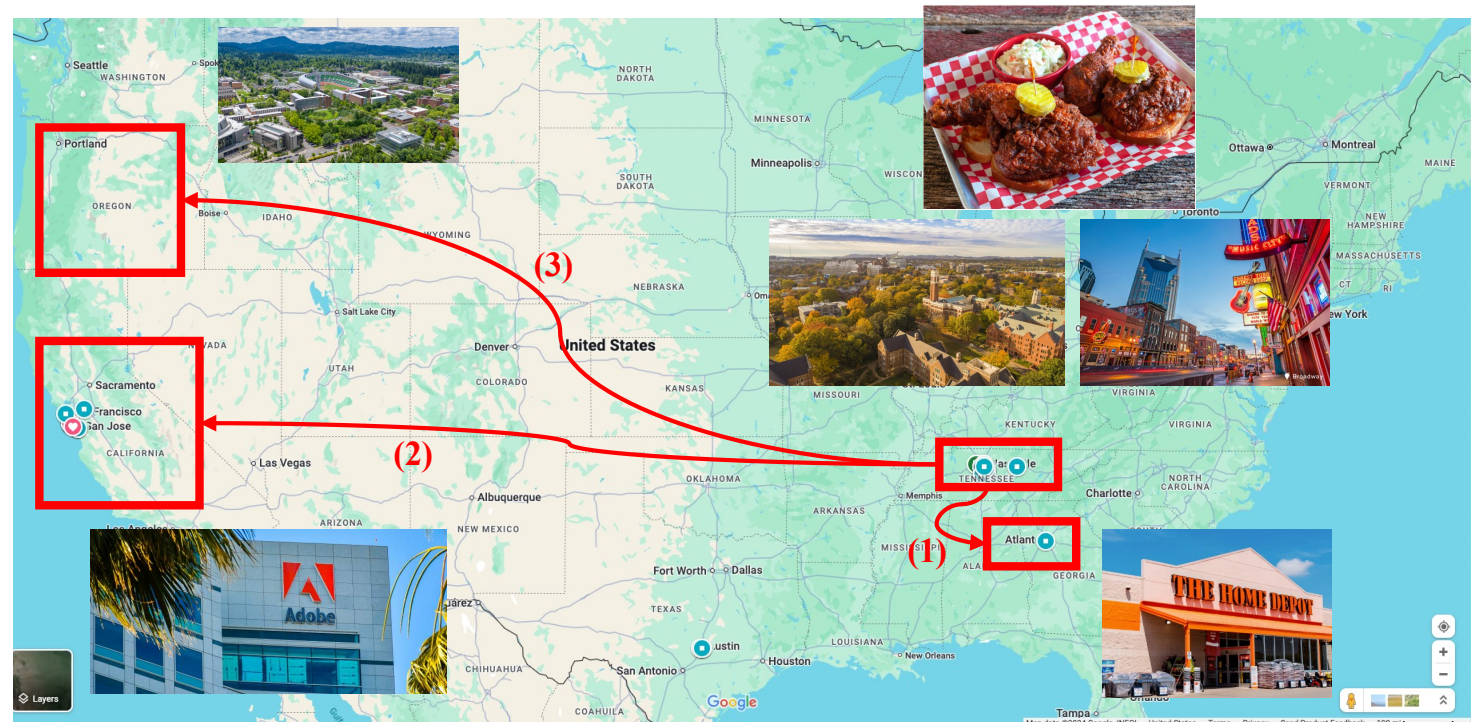


Welcome to the CS410/510 - Mining & Learning on Graphs!



Yu (Jack) Wang
You

Contact:
yuwang@uoregon.edu



Vanderbilt University, Nashville
The Home Depot Intern, Atlanta
Adobe Intern, San Jose
University of Oregon, Eugene



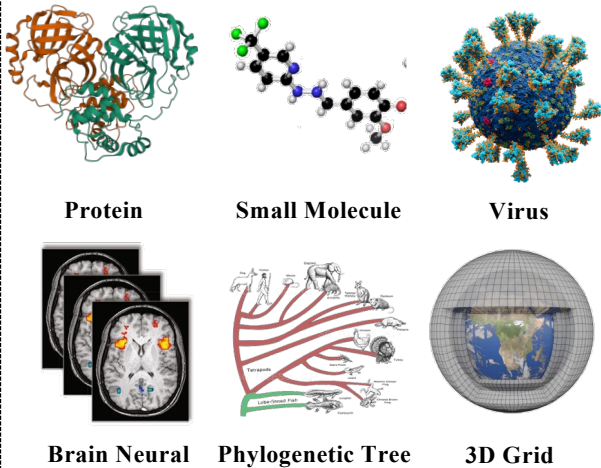
Our lab is currently recruiting!

- Data Mining and Machine Learning
- Graph and Geometric Machine Learning
- Data-centric + GenAI
- Trustworthy + Textual-attributed Graph
- AI/ML Application: Information Retrieval/Science/Cyber-security

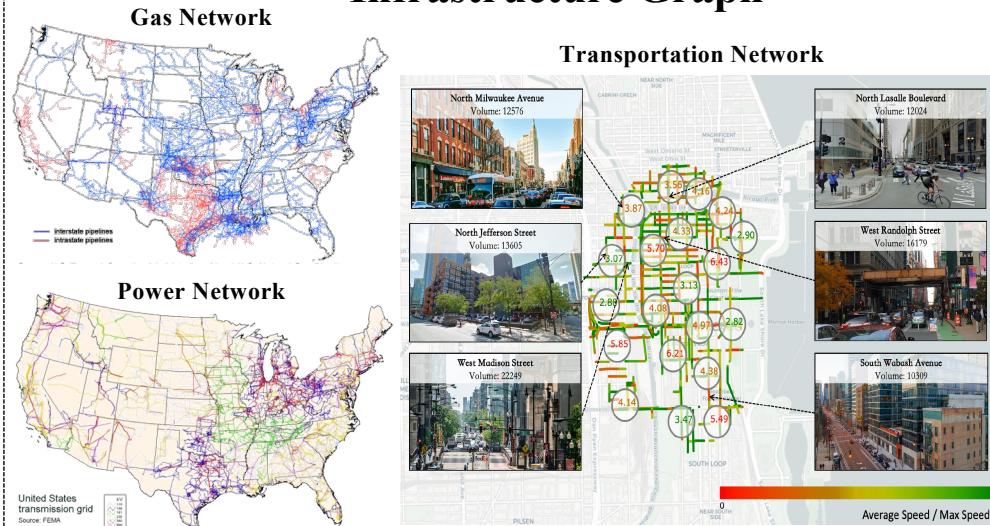
Graph-structured Data is Everywhere!



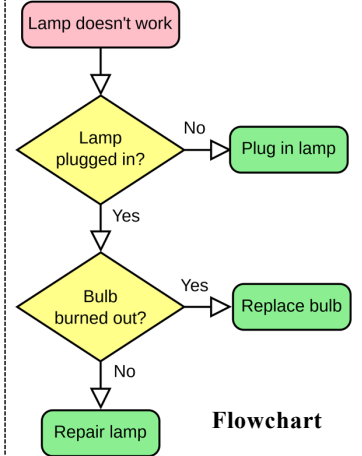
Scientific Graph



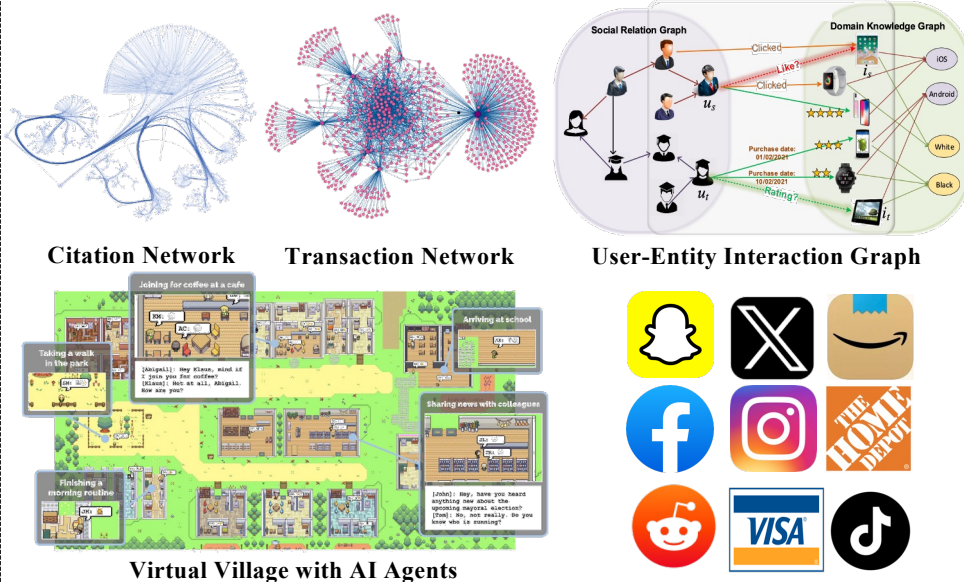
Infrastructure Graph



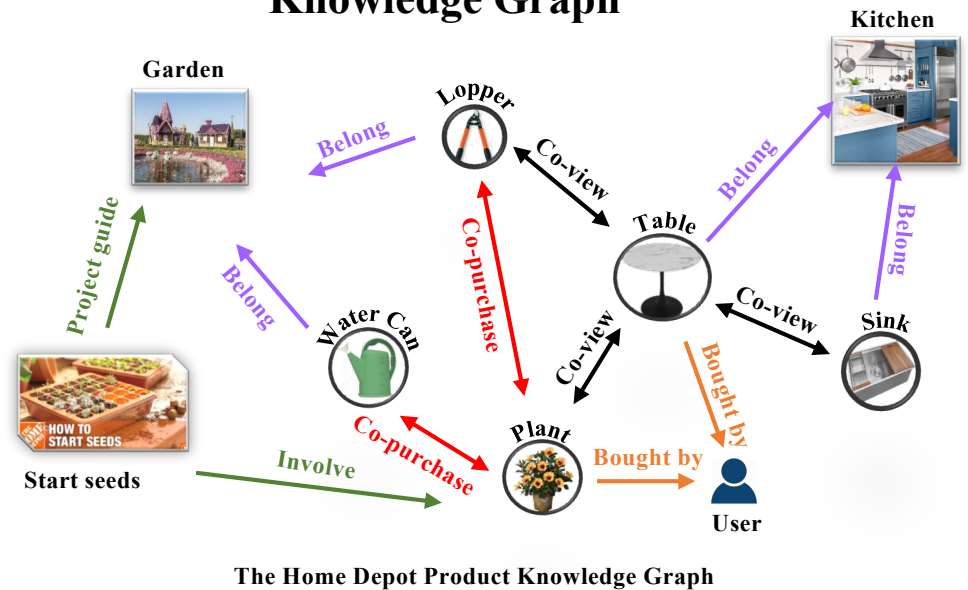
Decision Graph



Social Interaction Graph



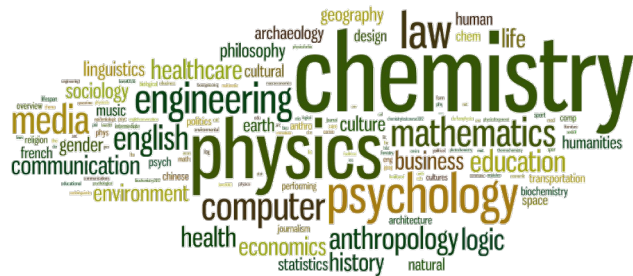
Knowledge Graph



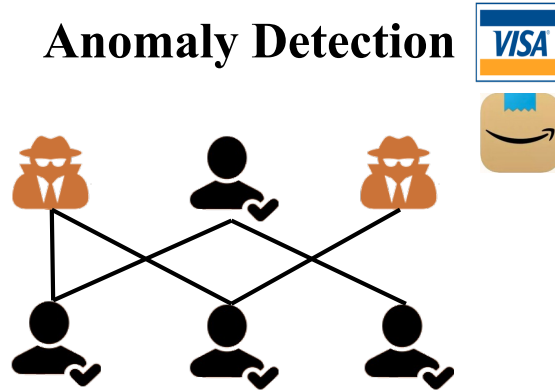
Graph-based Task is Everywhere!



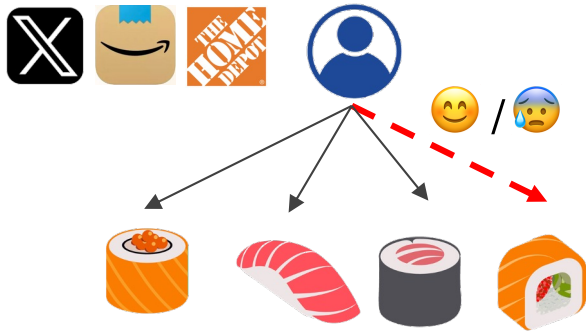
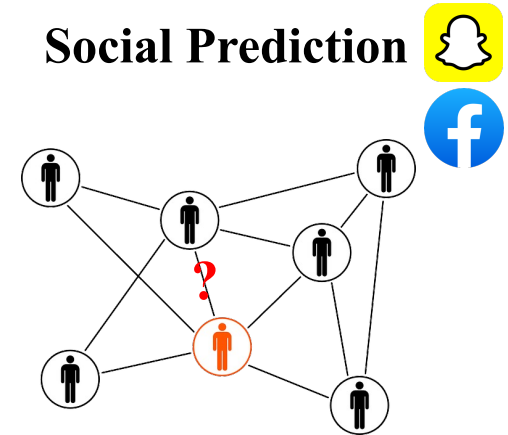
Paper Management



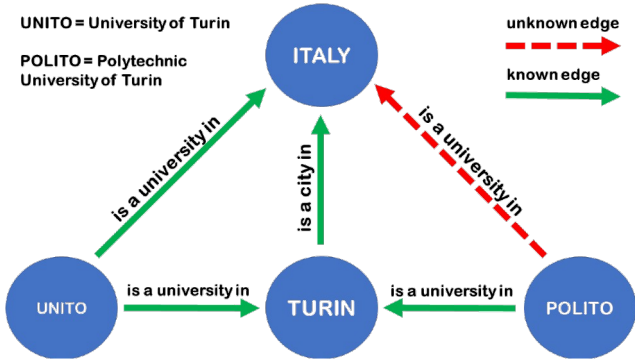
Anomaly Detection



Social Prediction



Recommender system



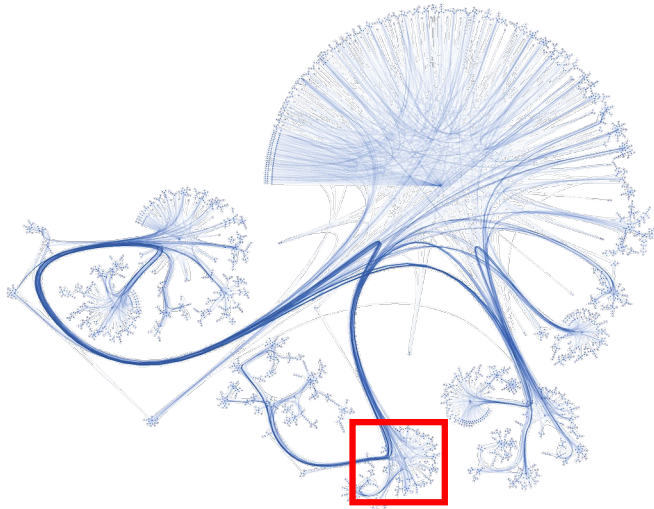
Question Answering



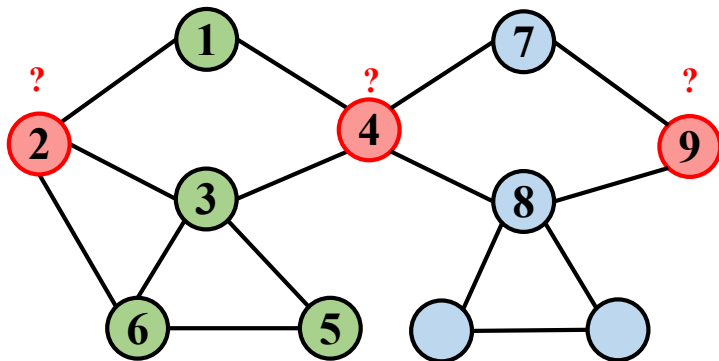


Example 1: Academic Paper Management

Cora - Paper Citation Networks



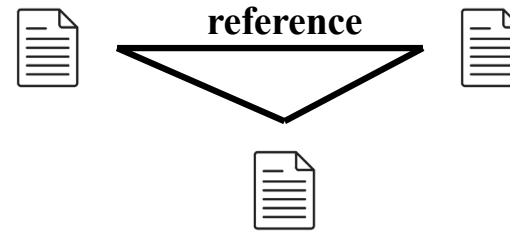
Zoom in



1 INTRODUCTION

We consider the problem of classifying nodes (such as documents) in a graph (such as a citation network), where labels are only available for a small subset of nodes. This problem can be framed as graph-based semi-supervised learning, where label information is smoothed over the graph via some form of explicit graph-based regularization (Zhu et al., 2003; Zhou et al., 2004; Belkin et al., 2006; Weston et al., 2012), e.g. by using a graph Laplacian regularization term in the loss function:

$$\mathcal{L} = \mathcal{L}_0 + \lambda \mathcal{L}_{\text{reg}}, \quad \text{with} \quad \mathcal{L}_{\text{reg}} = \sum_{i,j} A_{ij} \|f(X_i) - f(X_j)\|^2 = f(X)^\top \Delta f(X). \quad (1)$$



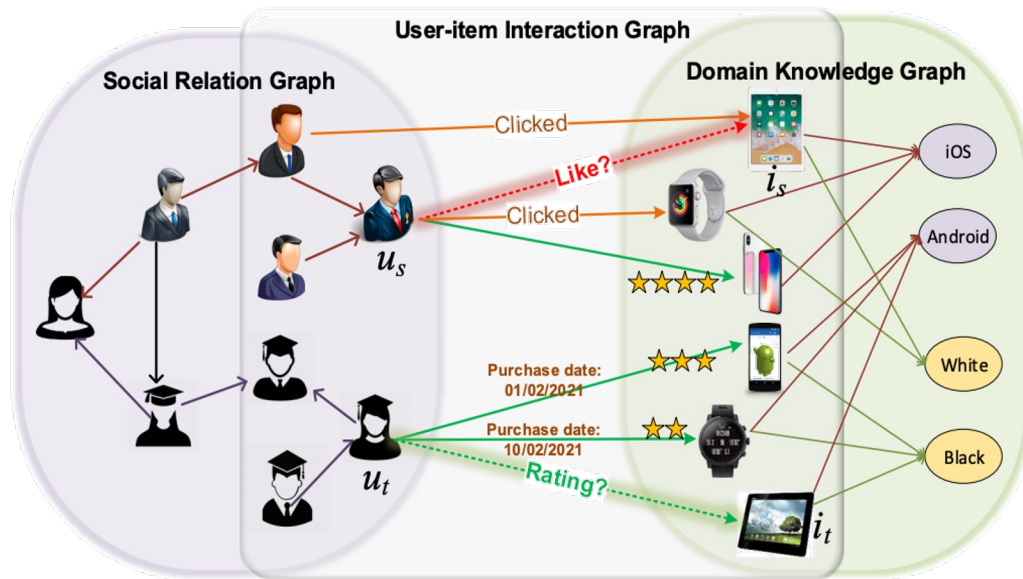
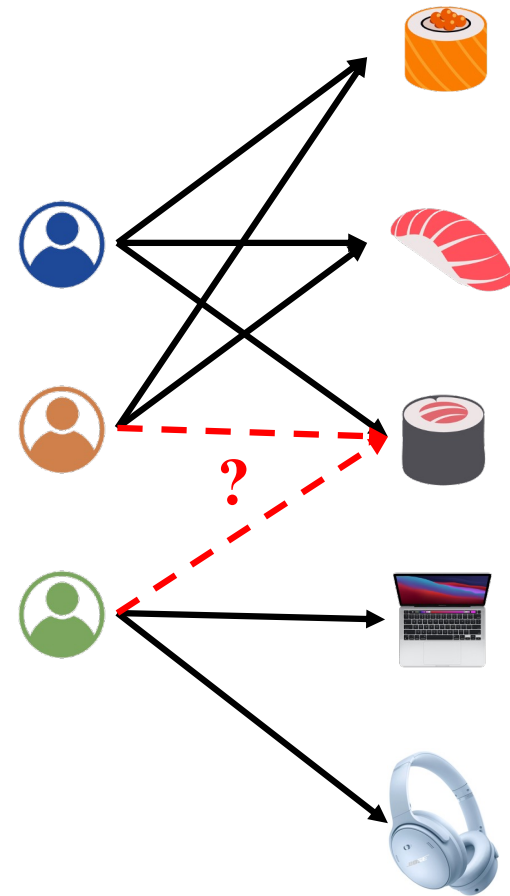
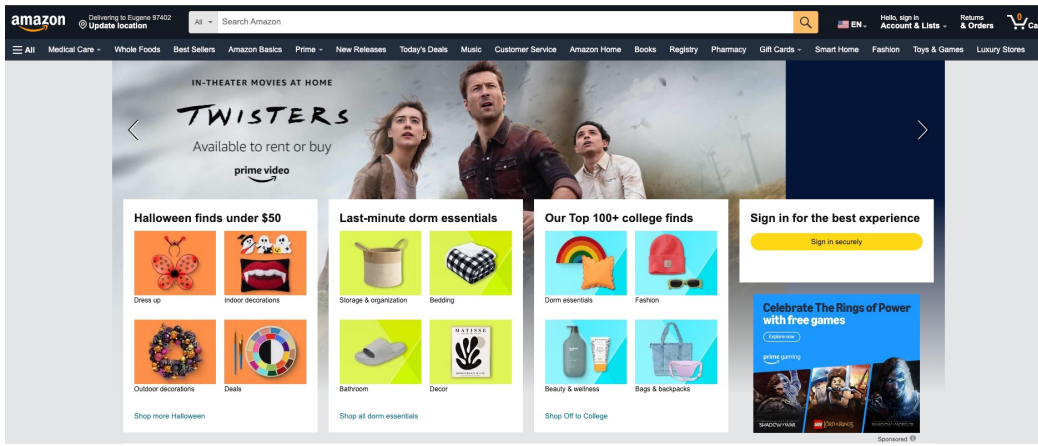
- Machine Learning
- Computer System
- What do you think?



Example 2: Personalization



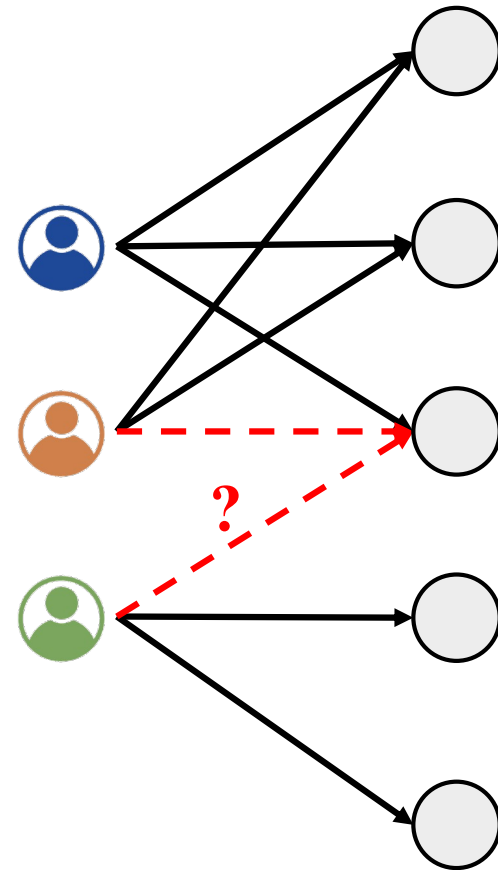
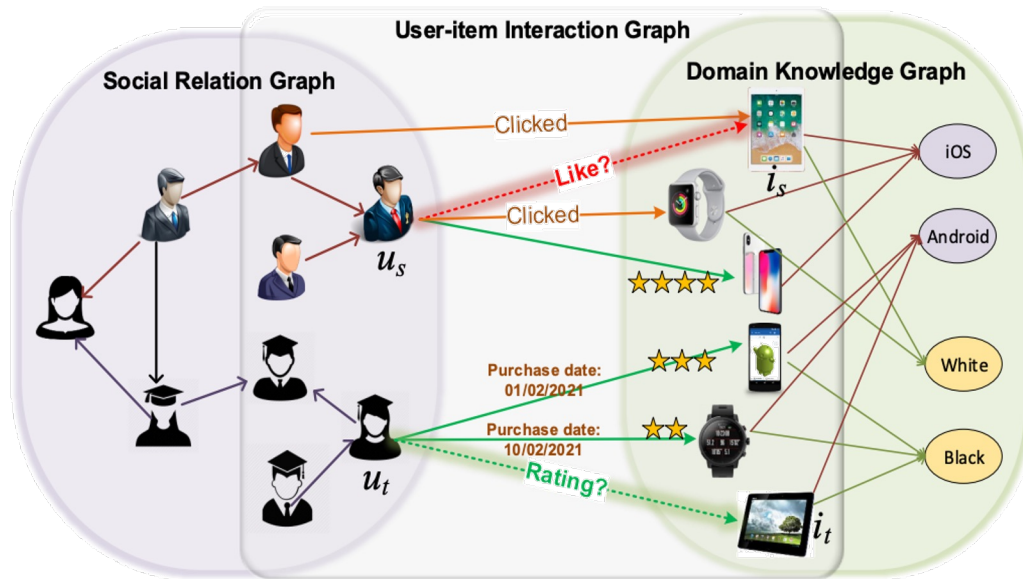
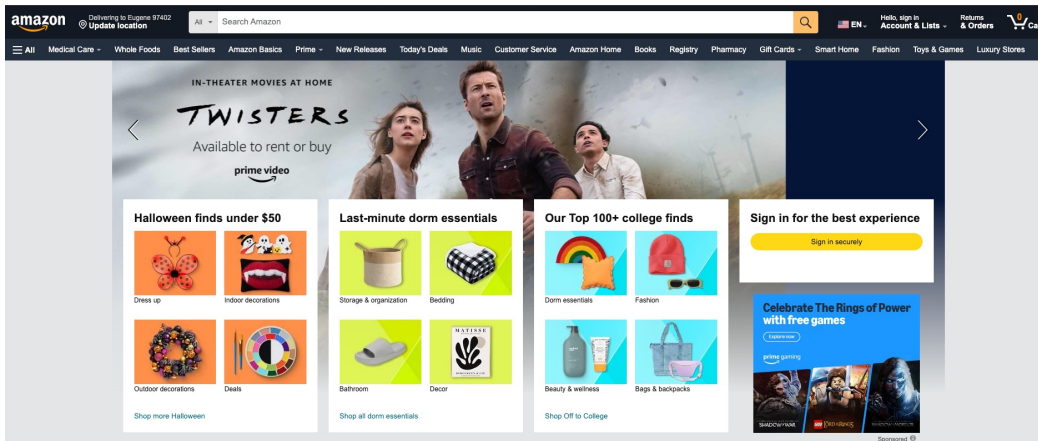
Amazon – Customer-Product Network



Example 2: Personalization



Amazon – Customer-Product Network

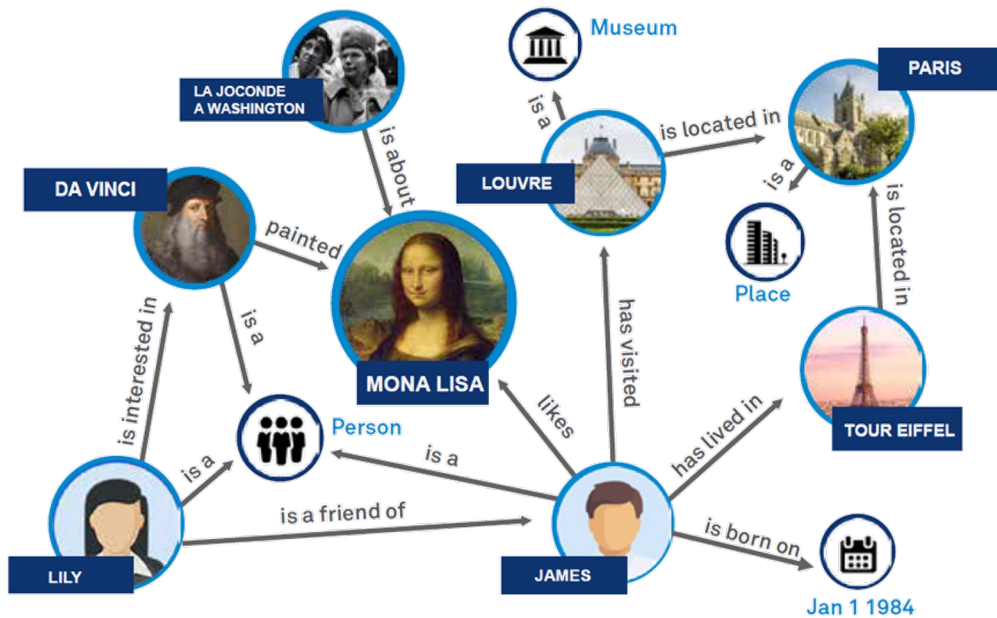


Recommendation based on customer-product interaction

Example 3: Question-answering



Knowledge Graph



Who painted Mona Lisa?

- (1) Locate Mona Lisa
- (2) Find her 1-hop neighbor
- (3) Da Vinci Painted Mona Lisa

How about LLMs?

Who painted Mona Lisa? Answer in 5 words

 Leonardo da Vinci painted Mona Lisa.

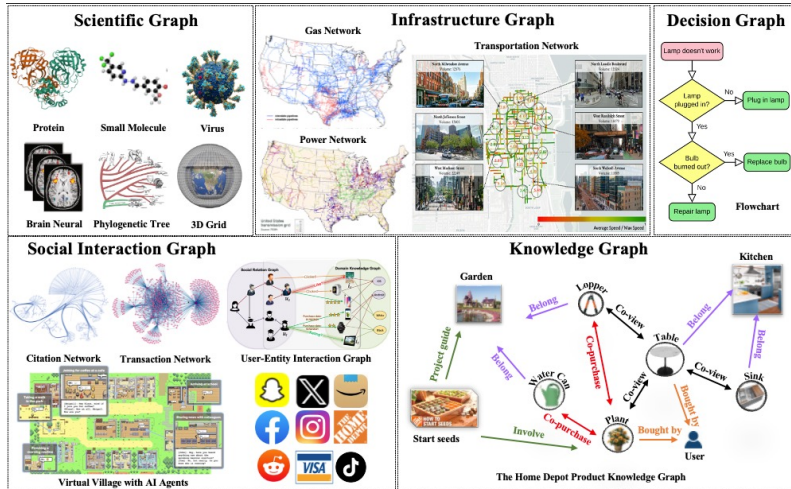


Store a large amount of factual knowledge in a symbolic format

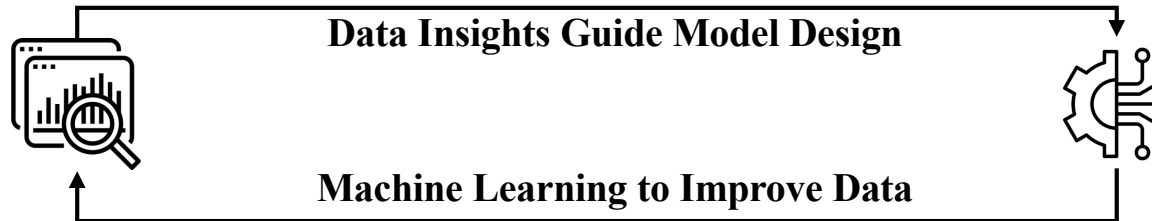
LLMs store knowledge in their parameters



Data Mining and Machine Learning on Graphs



Random Forests



Data mining

Analyze data

Derive patterns and relationships

Solve real-world problems

Machine Learning

Design Model

Allow Computer to Learn and Improve

Without being explicit programmed

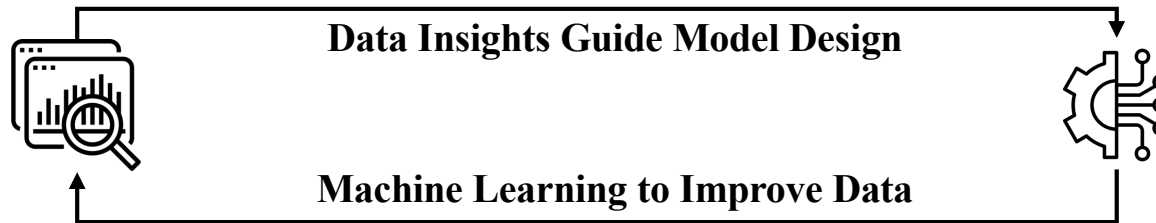


Data Mining and Machine Learning on Graphs



Data mining

Machine Learning



Real-world Applications

Data Structure

- **Image**
- **Language**
- **Time-series Data**
- **Graph**
 - Document Graph
 - Infrastructure Graph
 - Scientific Graph
 - Social Graph
 -

Purpose

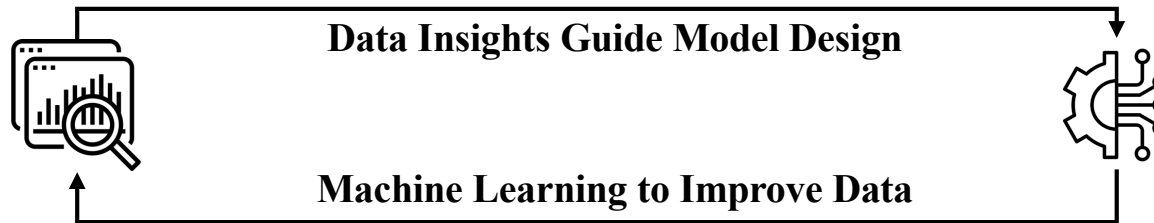
- **Utility - Performance**
- **Fairness**
- **Privacy**
- **Diversity**
- **Robustness**
- **Adversarial attack**
- **Eco-friendly**
- **Efficiency**

Data Mining and Machine Learning on Graphs



Data mining

Machine Learning



Real-world Applications

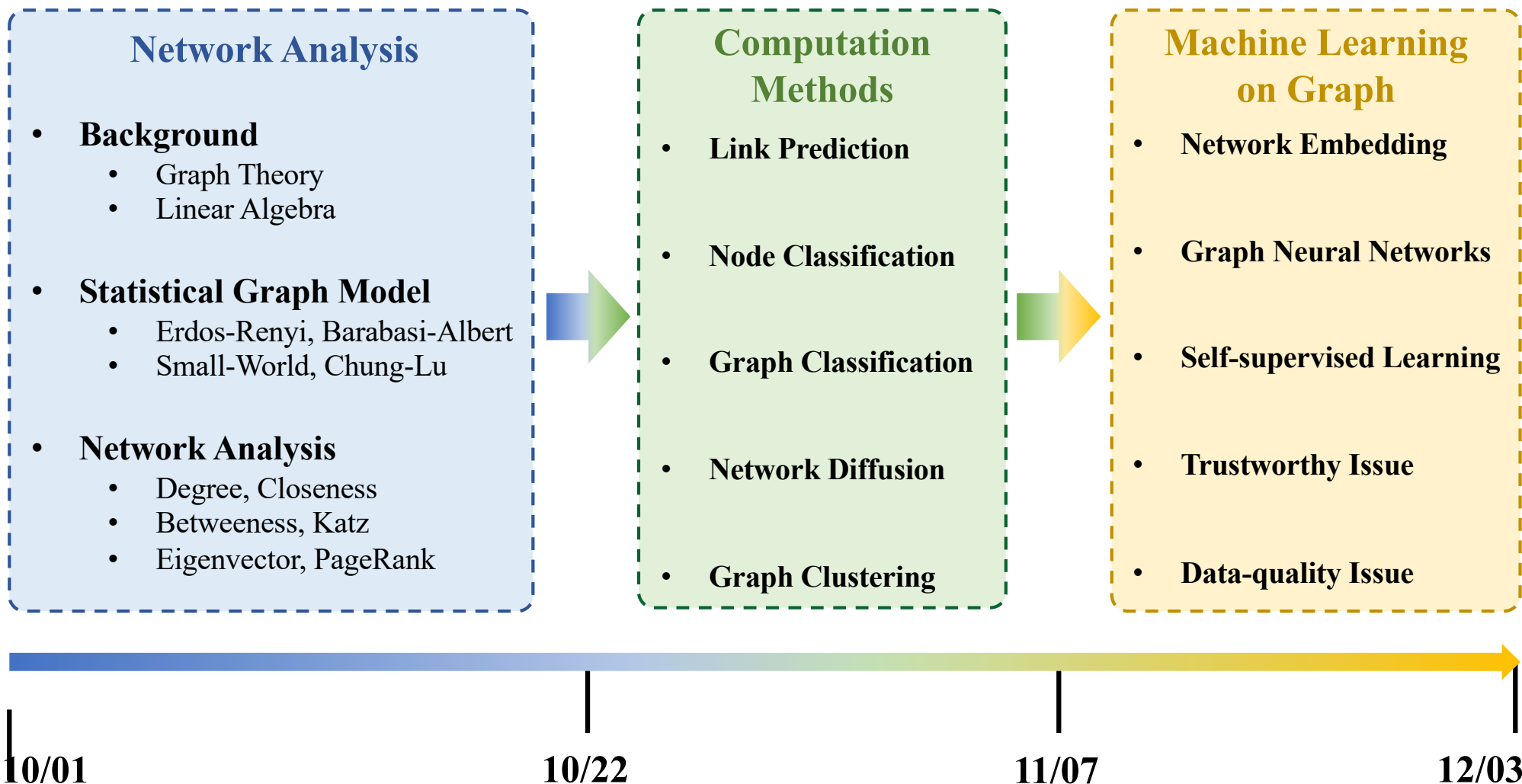
Data Structure

- Image
- Language
- Time-series Data
- **Graph**
 - Document Graph
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 - Scientific Graph
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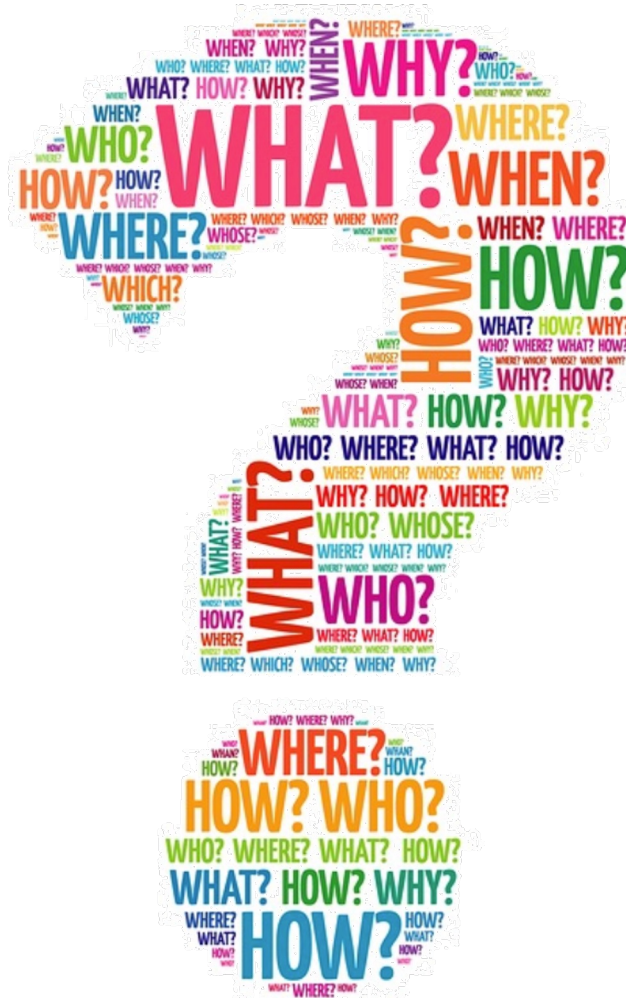
Purpose

- **Utility - Performance**
- Fairness
- Privacy
- Diversity
- Robustness
- Adversarial attack
- Eco-friendly
- Efficiency

Data Mining & Machine Learning on Graphs



Any Question?





Goals:

- Broad overview of basic knowledge and algorithm foundations of ML/DM on Graphs
- Hands-on experience with solving GML/DM problems
- Master real-world GML/DM applications

Requirements:

- Little to no background in ML
- Basic linear algebra, probability and statistics, and calculus
- Programming – Python
- Jupyter Notebooks for homework assignments

Course Logistics – Basic Contents



Times:

- **Classes:** Tuesday/Thursday 4:00-5:20 pm, 132 GSH
- **Office hours:** Friday 3:30-5:00 pm PST, other time by appointment
- **Zoom:** <https://uoregon.zoom.us/j/4052006678>

Components:

- 3 homework assignments (35%)
- **Midterm (30% - CS410, 25% - CS510)**
- Final Project (30%)
- Participation (5%)
- **Paper Presentation (5% - CS510)**
- Homework submitted with Overleaf (5% Bonus)

Course Logistics – Homework Assignments



Assignments

- Writing Assignment 1
- Programming Assignment 2
- Programming Assignment 3

No collaborations are allowed for assignments unless otherwise specified. Late Assignments will receive:

- 20% reduction, if submitting within (0, 24) hours late
- 40% reduction, if submitting within [24, 48) hours late
- 100% reduction, if submitting within [48, ∞) hours late, unless having documental special circumstances

You can download the assignments here. Also check out each assignment page for any additional info.

[Assignment 1](#)

[Assignment 2](#)

[Assignment 3](#)



Will Release Soon!



<https://ml-graph.github.io/fall-2024/paper/>

1. **Introduction and Background** – What is the general impact and background of the topic?
 2. **Motivation and Problem** – What is the core research problem and why do we study it?
 3. **Related Work and Challenges** – How did previous works on this problem and what are some challenges?
 4. **Proposed Solutions/Methods and Rationale** – What are the proposed methods/techniques and why propose them? What specific reasons that solving this problem would require these proposed methods/techniques
 5. **Experimental Setting, Results and Analysis** – What experiments are designed to verify the proposed method? How are results being discussed and analyzed? Are there any interesting findings?
5. **Conclusion and Future Work**



Paper

Graph Retrieval Augmented Generation

[ArXiv 2024] **Graph Retrieval-Augmented Generation: A Survey** [Paper]

[ArXiv 2024] **From Local to Global: A Graph RAG Approach to Query-Focused Summarization** [Paper]

Social Network Analysis

[ArXiv 2024] **Exploring Collaboration Mechanisms for LLM Agents: A Social Psychology View** [Paper]

[ArXiv 2024] **Scaling Large-Language-Model-based Multi-Agent Collaboration** [Paper]

[ArXiv 2024] **Network Formation and Dynamics Among Multi-LLMs** [Paper]

[ArXiv 2024] **Large Language Models Empowered Agent-based Modeling and Simulation: A Survey and Perspectives** [Paper]

[ArXiv 2024] **LLMs generate structurally realistic social networks but overestimate political homophily** [Paper]

GraphAI for Science

[NeurIPS 2024] **Learning to Group Auxiliary Datasets for Molecule** [Paper]

[ICLR 2023] **DiffDock: Diffusion Steps, Twists, and Turns for Molecular Docking** [Paper]

[ICML 2022] **Equivariant Diffusion for Molecule Generation in 3D** [Paper]

[ICLR 2023] **DiGress: Discrete Denoising diffusion for graph generation** [Paper]

GraphAI for Cybersecurity and System

[JSAC 2020] **RouteNet: Leveraging Graph Neural Networks for Network Modeling and Optimization in SDN** [Paper]

[ASCE 2022] **Graph Neural Networks for State Estimation in Water Distribution Systems: Application of Supervised and Semisupervised Learning** [Paper]

[ASCE 2022] **Optimal Power Flow using Graph Neural Networks** [Paper]

[ArXiv 2024] **PowerGraph: A power grid benchmark dataset for graph neural networks** [Paper]

[IEEE SmartGridComm] **On Graph Theory vs. Time-Domain Discrete-Event Simulation for Topology-Informed Assessment of Power Grid Cyber Risk** [Paper]

Course Logistics – Grades



Department of Computer Science, University of Oregon

Data Mining & Machine Learning on Graphs

Fall-2024



SYLLABUS



SCHEDULE



PAPER



ASSIGNMENTS



PROJECT



MATERIALS



GRADE

Course Assessment and Grading Scale

Category	CS-410 (%)	CS-510 (%)
Assignment	35%	35%
Midterm Exam	30%	25%
Final Project	30%	35%
Participation	5%	5%
Paper Presentation	0%	5%

Grade	Range
A	A+: 98-100, A: 93-97, A-: 90-92
B	B+: 87-89, B: 83-86, B-: 80-82
C	C+: 77-79, C: 73-76, C-: 60-72
F	F: <60

scds.uoregon.edu/cs

[username](#)

Any Question?

