## Course program and reading list



Semester 1 Year 2021

**School:** Efi Arazi School of Computer Science M.Sc.

### Recommendation Systems

#### Lecturer:

Dr. Asnat Messica Asnat.Messica@post.idc.ac.il

**Teaching Assistant:** 

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Course No.: Course Type: Weekly Hours: Credit:

3639 Elective 3 3

**Course Requirements : Group Code : Language:**Final Paper 211363901 English

**Prerequisites** 

#### Prerequisite:

52 - Calculus I

53 - Calculus II

54 - Linear Algebra I

55 - Linear Algebra II

56 - Discrete Mathematics

59 - Data Structures

69 - Logic And Set Theory

417 - Introduction To Computer Science

3141 - Machine Learning from Data

# Course Description

#### Introduction

Recommender systems are widely used by online companies worldwide to provide the most relevant content to each user. They changed the way users find relevant items in various areas such as e-commerce, content websites, social media, online advertisement and more.

The goal of this course is to provide the students with both theoretical and practical aspects of recommender systems. We will study machine learning algorithms for personalized recommendations, including recent algorithms, advanced evaluation, research topics, and implementation tools.

#### Meetings plan:

Meeting	Topic	Assignment
1	Introduction to Recommender Systems	
2 - 3	Collaborative filtering and Recommender Systems evaluation.	
4	Matrix Factorization	Exercise #1
5	Content based Recommender Systems and hybrid approach	

Meeting	Tagtic ization Machines. Context aware Recommender Systems.	Assignment
7	Loss types, multiple objectives optimization. Advanced Recommender Systems evaluation.	Exercise #2
8	Cold start, explore – exploit optimization	
9 - 10	Deep learning for Recommender Systems	Exercise #3
11	Explainability and application domains	
12	Practical aspects	
13	Summary and future trends	

Course Goals

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#### **Course Grading:**

Exam-40%

Exercises - 60%

Exercise #1: Matrix Factorization implementation and evaluation;

Exercise #2: Kaggle competition submission.

Exercise #3: Deep Factorization Machines and custom loss: implementation, evaluation and comparison to traditional algorithms.



### **Learning Outcomes**

#### **Learning Outcomes:**

Upon successful completion of this course students should be able to:

- Describe different approaches and algorithms for recommender systems
- Develop a recommender system to address specific use case needs and constraints
- Follow and take part in recent research in the area of recommender systems

**Lecturer Office Hours** 

Friday 14-15

Reading List

#### **Bibliography**

- Ricci, F., Rokach, L., Shapira, B., Kantor, P.B., 2011. Recommender Systems Handbook.
- Aggarwal, C.C., 2016. Recommender systems (Vol. 1). Cham: Springer International Publishing.
- Koren, Y., Bell, R. and Volinsky, C., 2009. Matrix factorization techniques for recommender systems. Computer, (8), pp.30-37.
- Hu, Y., Koren, Y. and Volinsky, C., 2008, December. Collaborative filtering for implicit feedback datasets.
  In Data Mining, 2008. ICDM'08. Eighth IEEE International Conference on (pp. 263-272). IEEE

• Rendle, S., 2010, December. Factorization machines. In 2010 IEEE International Conference on Data Mining (pp. 995-1000). IEEE

Additional papers will be provided before every meeting