

## **MULTIMEDIA INFORMATICS**

# MMI 712- Machine Learning Systems Design and Deployment



#### SYLLABUS

Year, Semester: 2022-2023 Fall

Course Conduct: Face-to-face in class II-06 Wednesday@13:40

Lecture videos on YouTube & Lecture notes on ODTUClass

Students are expected to watch the lecture videos and study the course material

before attending the weekly face-to-face sessions.

There will be regular quizzes from the content and attendance is expected in these

sessions.

**Lecturer:** Prof. Dr. Alptekin Temizel, <u>atemizel@metu.edu.tr</u>

**Teaching Assistant:** Ayberk Aydın, aayberk@metu.edu.tr

#### **Course Objective**

The course covers several aspects of designing reliable and scalable machine learning systems for real-world deployment. It deals with development of production quality models and introduces the machine learning pipeline, concepts on machine learning system design and data engineering. It provides know-how on model development, and how to scale up the training for large models as well as evaluation, calibration and debugging of these models. Generation of reproducible models via experiment tracking tools and model versioning is also covered. Hardware platforms and frameworks for deployment are introduced, followed by basic deployment concepts, containerized deployment and testing.

#### **Reference Material:**

CS 329S: Machine Learning Systems Design - https://stanford-cs329s.github.io/

#### **Reading Material:**

Rules of Machine Learning: Best Practices for ML Engineering <a href="https://developers.google.com/machine-learning/guides/rules-of-ml">https://developers.google.com/machine-learning/guides/rules-of-ml</a>

Challenges in Deploying Machine Learning: a Survey of Case Studies https://arxiv.org/pdf/2011.09926.pdf

#### **Grade Distribution:**

Assignments	45%
Final project	30%
5x Quizzes	25%

#### **Deliverables**

Documents and necessary files of the assignments must be uploaded to ODTUClass by students before the specified due dates.

### **University Policies**

All students are **expected to obey** the university code of integrity and avoid academic dishonesty or plagiarism.

No	Date	
1	5 Oct	Introduction to the Course and Machine Learning Life-Cycle
		Designing a Machine Learning System – I
2	12 Oct	Main Requirements of Machine Learning Systems
		Reliability, scalability, maintainability, adaptability
		ML in research vs. in production
		Traditional software vs. ML software, ML Production Myths
		Designing a Machine Learning System - II
3	19 Oct	Batch vs. online, Edge vs. cloud computing, Offline vs. Online Learning Iterative Development
	19 000	Phases of ML Adoption
		Versioning and Experiment Tracking
		Experiment tracking tools
4	26 Oct	Data versioning
		ML pipeline versioning
		Continuous integration/continuous delivery for ML
5	2 Nov	Hands-on Lab
		Experiment tracking
		Data versioning
		Data Engineering
		Data centric approach
6	9 Nov	Data basics and data formats
		Creating training datasets, labelling Semi-supervised and self-supervised learning
		Data Engineering-II
7	16 Nov	Sampling, Class imbalance problems
	10 1101	Data Augmentation
		Data Leakage
		Data Analysis with FiftyOne
8		Model Development and Training
		Model Selection
	23 Nov	AutoML, Neural Architecture Search
		Optimizers Model Calibration
		Model Optimization for Deployment  Model compression
9	30 Nov	Quantization
		Pruning
		Knowledge distillation
10		Case Study – Autonomous Driving
10	7 Dec	Invited speaker from industry
11		Training and Deployment Platforms and Packaging
		GPUs, TPUs
	14 Dec	IoT devices and TinyML
		Packaging  Deployment Franciscular
		Deployment Frameworks TensorRT
12	21 Dec	Triton inference server
		Google Cloud Platform (GCP)
		(***)
13	28 Dec	Evaluation
		Debugging, System evaluation and testing
		Data testing, profiling and visualization
		Benchmarking
		Perturbation evaluation, ablation study
4.4	4 15.5	Case Study – Deep Learning Solutions for Retail Stores
14	4 Jan	Invited speaker from industry