ECGR 6126/8126: Optimization for Machine Learning (X-list: ITCS 6116/8116)

Instructor: Prof. Ahmed Arafa, Spring 2025 (MW: 2:30–3:45 pm, EPIC 2230)

Suppose we are training a machine learning (ML) algorithm to solve an optimization problem, e.g., training a classifier on a certain dataset.

- Can we provide a guarantee that such algorithm gives a good solution? If so, how good is "good"?
- Can we quantify how fast the algorithm converges to a desired performance?

If you are interested in finding out answers for these (and related) questions, \Rightarrow then this course is for you.



What does the course provide?

This course provides an overview of optimization methods used in modern ML systems. By the end of the course, students will be able to (a) assess the performance of popular ML algorithms on various function (model) classes, (b) accurately implement *optimized* ML algorithms for different applications, and (c) characterize time-accuracy tradeoffs for ML training.

What are the course contents?

Topics include: convex optimization, (stochastic) gradient methods, second-order methods, nonconvex optimization, distributed optimization, and federated learning.

Grading:

- homework (30%): consisting of theoretical + practical (implementation) components
- exam (20%): after about 2/3 of the semester
- project (40%): in groups, to explore how the studied algorithms and methods perform in real-world example settings
- class participation (10%): attending and engaging in discussions

Are there any prereqs?

Good command of linear algebra, calculus, and programming (e.g., Python). Suggested prereqs: ECGR 4115/5115: Convex Optimization and AI Applications, ECGR 4105/5105: Introduction to Machine Learning.