



# Special Topics in Machine Learning Spring 2024

Course code: BME i9400 - 3 credits

**Course Description** This course provides a broad overview of modern-day machine learning with an emphasis of techniques that are commonly used in practice to classify biomedical data sets. The course begins with a review of probability theory and linear algebra, which are the foundation for statistical learning. We will then survey a variety of supervised and unsupervised architectures, beginning with logistic regression and including state-of-the-art techniques such as the transformers that power large language models. Throughout the course, students acquire hands-on experience with the presented concepts via application to real-world data sets from a variety of domains. For the Spring 2024 iteration, we will work primarily in the iPython (Jupyter Notebook) environment.

*An effort will be made to make this year's course "hands-on", taking the form of Python code snippets that demonstrate the concepts wherever possible. All students should bring a laptop with a working Python environment (instructions will be provided) to class.*

## Course Objectives

1. To be able to formulate problems in the framework of regression/classification.
2. To become comfortable working in the iPython environment.
3. To be able to apply state-of-the-art machine learning techniques to your data with an understanding of how to optimize performance.

**Lectures** Monday 02:00 PM-4:30 PM  
Steinman Hall 560  
Online TBA

**Instructor** Jacek Dmochowski (jdmochowski@ccny.cuny.edu)  
Office: ST 460

**Office Hours** By email appointment

## Helpful Textbooks (all optional):

Deep Learning with Python (Francois Chollet) [Publisher](#)  
Deep Learning (Ian Goodfellow, Yoshua Bengio, Aaron Courville) [HTML version](#)  
Pattern Classification (Richard O. Duda, Peter E. Hart, David G. Stork) [Amazon](#)

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## Spring 2024

### Topics Covered

*Dates and topics subject to change*

Week	Date	Topic
1	09.09.24	Probability theory and Bayes' Rule
2	09.16.24	Linear algebra primer
3	09.23.24	Optimization, maximum likelihood, gradient descent
4	09.30.24	Independent Components Analysis
5	10.07.24	Logistic regression
6	10.15.24	Multilayer perceptrons (deep learning)
7	10.21.24	Convolutional networks
8	10.28.24	Sequence models
9	11.04.24	Transformers and language models
10	11.11.24	<i>Midterm exam</i>
11	11.18.24	Autoencoders
12	11.25.24	Self-supervised learning
13	12.02.24	Biomedical ML applications
14	12.09.24	<i>Final project presentations</i>

**Assessment Tools** Performance Criterion 1: Evaluation of homework assignments.  
Performance Criterion 2: Grading of midterm exam  
Performance Criterion 3: Evaluation of course project.

**Final Grade**

Homework Assignments	30%
Midterm Exam	30%
Course Project	40%

**Lecture Policy** Students are expected to attend all lectures.

**Academic integrity** *Students may discuss homework but the turned-in work must be that of the individual student only. Submission of plagiarized homework solutions will result in referral to the Department Chair.*