

Math 198 — Math for Machine Learning

Spring 2021 Syllabus

1 Course Description

This DeCal is intended for any student interested in the mathematics behind machine learning models, and particularly those who meet the math prerequisites for CS 189 but want to become more comfortable with the material and its applications to machine learning. The course assumes general knowledge in linear algebra, vector calculus, optimization problems, and statistics; however, students do not need to feel like experts in any of these fields to take this course. We will give general descriptions of things like vector spaces and random variables, but will not cover all of the important results from these fields. Rather, we will focus on the results which have the most direct applications to machine learning, and will go over some of these applications so they are more familiar when encountered in CS 189.

Importantly, this course is not meant to replace CS 189, Math 54/110, Math 53, or any of the many statistics and optimization courses available. We will be building on results from these classes which may be presented without proof, and familiarity with the basics of linear algebra, vector calculus, and statistics are essential. This course is meant to bridge the gap between these courses and CS 189, so students feel more comfortable taking 189 in a later semester.

2 Course Location and Instructor Contacts

Course Website: <https://MathForMLdecal.github.io/>

Course Location and Time: TBD

Facilitators:

Parth Nobel, EECS '21, parthnobel@berkeley.edu

Kesav Viswanadha, EECS MS '21, kesav@berkeley.edu

Faculty Sponsor: Prof. Per-Olof Persson, persson@berkeley.edu

3 Topics

Tuesday lecture will be the week's topic, and Thursday will introduce and explain the application.

Week 1: Syllabus/Introduction to Topics

Application: Perceptrons

Week 2: Review of Vector Spaces, Linear Maps, Matrices

No Application Lecture — Labor Day Holiday

Week 3: Properties of Matrices, Spectral Theorem

Application: Orthogonal Projections

Week 4: Special Classes of Matrices

Application: Ordinary Least Squares

Week 5: Singular Value Decomposition

Application: PCA and Four Fundamental Subspaces

Week 6: Linear Algebra Wrap-Up

Application: Total Least Squares

Week 7: Vector and Matrix Calculus

Application: Re-Proving OLS

Week 8: Extrema and Convexity

Application: Gradient Descent

Week 9: Optimization of Non-Convex Functions

Application: Newton's Method, Gauss-Newton Algorithm

Week 10: Random Variables/Vectors, EV, Variance

Application: Chebyshev's Inequality and LLN

Week 11: Joint Distributions, Covariance, Correlation

Application: Bayes Nets

Week 12: Gaussian Distribution and Estimations

Application: Maximum A Priori

4 Grading and Structure

This course will meet twice per week, and we intend to alternate between math and applications, following the schedule above. Because this material is difficult to digest without practice, homework will be assigned with each math-based lecture and due two lectures later. However, homework will be graded on completion, not correctness. Homework may be submitted by emailing a PDF to both facilitators before lecture. Homework should not be completed in a group, nor should homework answers be shared.

There is no textbook for this course — notes will be posted on the course website, and are intended to present results and proofs from lecture in a more formal format. Nonetheless, lecture attendance is required. Students may miss up to two lectures unexcused, and must request excuses in advance from the facilitators for any further missed lectures.

Grade Breakdown

- 50%: Attendance
- 50%: Homework

For the first class missed beyond the two excused ones, the student's attendance points will be cut in a third. For the second class missed beyond the two excused ones, the student's attendance points will be cut in a ninth. This pattern follows for all further missed classes. *i.e.* if you miss k classes you will receive $\frac{1}{3^{\max(0, k-2)}}$ of the attendance points.

Homework grades are computed by multiplying together the scores of all but the lowest two of your homeworks. Scores on each homework are assigned from $\{0, 1\}$ based off good faith effort on at least 75% of the assigned problems.

Accordingly, to earn a “P” in this course, students must attend lectures in accordance with the requirements and complete all but two of their homework assignments. There are no quizzes or tests. If at any point you would like to know your official grade in the course, feel free to email one or both facilitators.

Facilitators welcome feedback throughout the course, in particular compliments can be sent to Parth, and complaints to Kesav.