# Math 198 – Math for Machine Learning

#### Spring 2020 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: ocf.io/seanvernon/decal

Course Location and Time: MW 4-5 PM, Haviland 12

Facilitators: Sean Vernon, CS '20, seanvernon@berkeley.edu Nick Park, Math '20, npark@berkeley.edu

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

### 3 Topics

Week 1: Syllabus/Introduction to Topics	Week 7: Vector and Matrix Calculus
Application: Perceptrons	Application: Gradient Descent
Week 2: Review of Vector Spaces, Linear Maps, Matrices	Week 8: Convexity
Application: Orthogonal Projections	Application: Proofs using Convexity
Week 3: Properties of Matrices, Spectral Theorem	Week 9: Optimization of Non-Convex Functions
No Application Lecture – President's Day Holiday	Application: Newton's Method, Gauss-Newton, etc.
Week 4: Special Classes of Matrices	Week 10: Random Variables/Vectors, EV, Variance
Application: Ordinary Least Squares	Application: Chebyshev's Inequality and LLN
Week 5: Singular Value Decomposition	Week 11: Joint Distributions, Covariance, Correlation
Application: PCA and Four Fundamental Subspaces	Application: Bayes Nets
Week 6: Further Applications: Ridge Regression, Total	Week 12: Gaussian Distribution and Estimations
Least Squares, Feature Augmentation, Kernel Trick	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 in-person or electronically (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.

To earn a "P" in this course, students must attend lectures in accordance with the requirements and complete at least 75% of the assigned homework problems. The lowest two homework scores will be dropped. 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. Compliments can be sent to Sean, and complaints to Nick.