

| | | Date | Lecture Topics | Deliverables | Notes | | |
|---------|--------------|------------------|--|--|--|--|--|
| Week 1 | Lecture 1 | 3/28/2022 | Introduction | | Slides | | |
| | Lecture 2 | 3/30/2022 | Supervised learning setup. LMS. | | Draft , Template , Notes ; Section 1 of Main Notes | | |
| | TA Lecture 1 | 4/1/2022 | Linear Algebra Review | | Notes ; Slides ; Annotated Slides | | |
| Week 2 | Lecture 3 | 4/4/2022 | Weighted Least Squares. Logistic regression. Newton's Method | | Draft , Template , Notes ; Section 2 of Main Notes | | |
| | Lecture 4 | 4/6/2022 | Exponential family. Generalized Linear Models. | | Draft ; Section 3 of Main Notes | | |
| | | 4/6/2022 | | Problem Set 0 (Due at 11:59 pm PT - Ungraded) | | | |
| | TA Lecture 2 | 4/8/2022 | Probability Review | | Notes ; Slides | | |
| Week 3 | Lecture 5 | 4/11/2022 | Gaussian discriminant analysis. Naive Bayes. | | Section 4.1 of Main Notes | | |
| | Lecture 6 | 4/13/2022 | Naive Bayes, Laplace Smoothing. | | Section 4.2 of Main Notes | | |
| | | 4/15/2022 | | Final Project Proposal (Due at 11:59 pm PT) | | | |
| | TA Lecture 3 | 4/15/2022 | Python/Numpy | | Slides ; Materials | | |
| Week 4 | Lecture 7 | 4/18/2022 | Kernels | | Section 5 of Main Notes | | |
| | Lecture 8 | 4/20/2022 | Neural Networks 1 | | Draft , Template , Notes ; Section 7.1 & 7.2 of Main Notes | | |
| | | 4/20/2022 | | Problem Set 1 (Due at 11:59 pm PT) | | | |
| | TA Lecture 4 | 4/22/2022 | Evaluation Metrics | | Slides | | |
| Week 5 | Lecture 9 | 4/25/2022 | Neural Networks 2 (backprop) | | Section 7.3 of Main Notes | | |
| | Lecture 10 | 4/27/2022 | Bias - Variance. Regularization. | | Section 8 of Main Notes | | |
| | TA Lecture 5 | 4/29/2022 | Deep Learning (Conv Nets) | | Slides | | |
| Week 6 | Lecture 11 | 5/2/2022 | Feature / Model selection. ML Advice. | | Section 9 of Main Notes, slides (only subset of first 40 pages are covered in the lecture) | | |
| | Lecture 12 | 5/4/2022 | K-Means. GMM (non EM). Expectation Maximization. | | Draft ; Section 10 , 11.1 , 11.2 of Main Notes | | |
| | | 5/4/2022 | | Problem Set 2 (Due at 11:59 pm PT) | | | |
| | | 5/6/2022 | | Final Project Milestone (Due at 11:59 pm PT) | | | |
| | TA Lecture 6 | 5/6/2022 | Midterm Review | | Slides | | |
| Week 7 | Lecture 13 | 5/9/2022 | GMM (EM) | | Draft ; Section 11.2-11.4 of Main Notes | | |
| | Lecture 14 | 5/11/2022 | Factor Analysis/PCA | | Draft ; Section 12&13 of Main Notes | | |
| | | 5/12/2022 | | MIDTERM (CEMEX Auditorium, 6 pm - 9 pm PT) | | | |
| | | | No TA Lecture (Midterm Week) | | | | |
| Week 8 | Lecture 15 | 5/16/2022 | PCA/ICA | | Draft ; Draft ; Section 13 of Main Notes | | |
| | Lecture 16 | 5/18/2022 | Self-supervised learning | | Draft ; Section 14 of Main Notes | | |
| | | 5/18/2022 | | Problem Set 3 (Due at 11:59 pm PT) | | | |
| | | 5/20/2022 | No TA Lecture | | | | |
| Week 9 | Lecture 17 | 5/23/2022 | basic concepts in RL, value iteration, policy iteration | | Draft ; Section 15 of Main Notes | | |
| | Lecture 18 | 5/25/2022 | Societal impact of ML (Guest lecture by Prof. James Zou) | | | | |
| | TA Lecture 7 | 5/27/2022 | Decision Trees + Boosting | | Slides (Boosting) , Slides (Decision Trees) | | |
| Week 10 | Lecture 19 | 5/30/2022 | MEMORIAL DAY. NO LECTURE. | | | | |
| | Lecture 20 | 6/1/2022 | Model-based RL, value function approximator | | | | |
| | TA Lecture 8 | 6/3/2022 | Learning Theory | | Notes | | |
| | | 6/1/2022 | | Problem Set 4 (Due at 11:59 pm PT) | | | |
| | | 6/6/2022 | | Final Project Report (Due at 11:59 pm PT) | | | |

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| | | 6/7/2022 | Final Project Poster Session (3:30 pm - 6:30 pm PT) | | |
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Other Resources

(Hover over each cell for hyperlinks)

All lecture videos can be accessed through [Canvas](#).

Advice on applying machine learning: Slides from Andrew Ng's lecture on getting machine learning algorithms to work in practice can be found [here](#).

Previous projects: Projects from previous years can be found in the ["Final Projects" doc](#) on the home page.

Data: Here is the [UCI Machine learning repository](#), which contains a large collection of standard datasets for testing learning algorithms. If you want to see examples of recent work in machine learning, start by taking a look at the conferences [NeurIPS](#) (all old NeurIPS papers are online) and ICML. Some other related conferences include UAI, AAAI, IJCAI.

Viewing PostScript and PDF files: Depending on the computer you are using, you may be able to download a [PostScript](#) viewer or [PDF viewer](#) for it if you don't already have one.

[Machine learning study guides tailored to CS 229](#) by Afshine Amidi and Shervine Amidi.

[The Matrix Cookbook](#): Quick reference for matrix identities, approximations, relations, etc.