

		Date	Lecture Topics	Deliverables	Notes	Slides		
Week 1	Lecture 1	9/27/2023	Introduction					
	TA Lecture 1	9/29/2023	Linear Algebra Review		<a href="#">Notes</a>	<a href="#">Slides</a>		
Week 2	Lecture 2	10/2/2023	Supervised learning setup. LMS.		<a href="#">Sections 1.1, 1.2 of main notes</a>			
	Lecture 3	10/4/2023	Weighted Least Squares. Logistic regression. Newton's Method		<a href="#">Sections 1.3, 1.4, 2.1, 2.3 of main notes</a>			
		10/4/2023		Problem Set 0 (Due at 11:59 pm PT - Ungraded)				
	TA Lecture 2	10/6/2023	Probability Review		<a href="#">Notes</a>	<a href="#">Slides</a>		
Week 3	Lecture 4	10/9/2023	Dataset split; Exponential family. Generalized Linear Models.		<a href="#">Section 2.2 and Chapter 3 of main notes</a>			
		10/9/2023		Final Project Proposal (Due at 11:59 pm PT)	CS229 Final Project Fall 2022-23			
	Lecture 5	10/11/2023	Bias-variance tradeoff, regularization		Sections 8.1, 9.1, 9.3	<a href="#">Bias/variance slides</a> <a href="#">Ridge regression slides</a> <a href="#">Lasso regression slides</a> <a href="#">Bias/variance annotated</a> <a href="#">Ridge annotated</a>		
		10/11/2023		Problem Set 1 (Due at 11:59 pm PT)				
	TA Lecture 3	10/13/2023	Python/Numpy		<a href="#">jupyter notebook</a>	<a href="#">slides</a>		
Week 4	Lecture 6	10/16/2023	Decision trees		<a href="#">Decision trees &amp; ensemble learning</a>	<a href="#">Boosting slides</a> <a href="#">Decision Trees slides</a> <a href="#">Decision Trees annotated</a> <a href="#">Decision Trees Overfitting</a> <a href="#">Lasso annotated</a>		
	Lecture 7	10/18/2023	Boosting		<a href="#">Decision trees &amp; ensemble learning</a>			
	TA Lecture 4	10/20/2023	Evaluation Metrics			<a href="#">slides</a>		
Week 5	Lecture 8	10/23/2023	Gaussian discriminant analysis. Naive Bayes.		<a href="#">Section 4.1, 4.2 of main notes</a>			
	Lecture 9	10/25/2023	Kernels; SVM		Chapter 5			
		10/27/2023		Problem Set 2 (Due at 11:59 pm PT)				
	TA Lecture 5	10/27/2023	Midterm Review			<a href="#">Slides</a>		
		10/27/2023		Final Project Milestone (Due at 11:59 pm PT)	CS229 Final Project Fall 2022-23			
Week 6	Lecture 10	10/30/2023	K-Means. GMM. Expectation Maximization.		Section 10, 11 of main notes	<a href="#">K-means slides</a> <a href="#">EM slides</a> <a href="#">PCA slides</a> <a href="#">K-means annotated</a> <a href="#">EM annotated</a> <a href="#">PCA annotated</a>	<a href="#">GMM slides</a>	<a href="#">GMM annotated</a>
	Lecture 11	11/1/2023	ML Advice			<a href="#">ML advice</a>		
		11/3/2023		MIDTERM: HEWLET200 (Last name A-L) & STLC111 (Last name M-Z) , 6 pm - 9 pm PT				
			No TA Lecture (Midterm Week)					
Week 7	Lecture 12	11/6/2023	Neural Networks 1		Sections 7.1, 7.2			
	Lecture 13	11/8/2023	Neural Networks 2 (backprop)		Section 7.3			
		11/10/2023		Problem Set 3 (Due at 11:59 pm PT)				
	TA Lecture 6	11/10/2023	Deep Learning (Convnets)			<a href="#">Slides</a>		
Week 8	Lecture 14	11/13/2023	Basic concepts in RL, value iteration, policy iteration.					
	Lecture 15	11/15/2023	Model-based RL, value function approximator					
	TA Lecture 7	11/17/2023	GANs					

Week 9	Lecture 16	11/27/2023	EM, PCA				
	Lecture 17	11/29/2023	Other learning settings. Large language models & foundation models			<a href="#">Learning + foundation models</a>	
		<b>12/1/2023</b>		<b>Problem Set 4 (Due at 11:59 pm PT)</b>			
Week 10	Lecture 18	12/4/2023	fairness, algorithmic bias, explainability, privacy			<a href="#">fairness</a> <a href="#">fairness annotated</a> <a href="#">privacy</a> <a href="#">privacy annotated</a> <a href="#">explainability</a> <a href="#">explainability annotated</a>	
	Lecture 19	12/6/2023	fairness, algorithmic bias, explainability, privacy				
		<b>12/8/2023</b>		<b>Final Project Report (Due at 11:59 pm PT)</b>	CS229 Final Project Fall 2022-23		
		<b>12/13/2023</b>		<b>Final Project Poster Session (3:30 pm - 6:30 pm PT)</b>			
<b>Other Resources</b>							
(Hover over each cell for hyperlinks)							
All lecture videos can be accessed through <a href="#">Canvas</a> .							
Advice on applying machine learning: Slides from Andrew Ng's lecture on getting machine learning algorithms to work in practice can be found <a href="#">here</a> .							
Previous projects: Projects from previous years can be found in the " <a href="#">Final Projects</a> " doc on the home page.							
Data: Here is the <a href="#">UCI Machine learning repository</a> , 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 <a href="#">NeurIPS</a> (all old NeurIPS papers are online) and <a href="#">ICML</a> . Some other related conferences include <a href="#">UAI</a> , <a href="#">AAAI</a> , <a href="#">IJCAI</a> .							
Viewing PostScript and PDF files: Depending on the computer you are using, you may be able to download a <a href="#">PostScript</a> viewer or <a href="#">PDF viewer</a> for it if you don't already have one.							
<a href="#">Machine learning study guides tailored to CS 229</a> by Afshine Amidi and Shervine Amidi.							
<a href="#">The Matrix Cookbook</a> : Quick reference for matrix identities, approximations, relations, etc.							