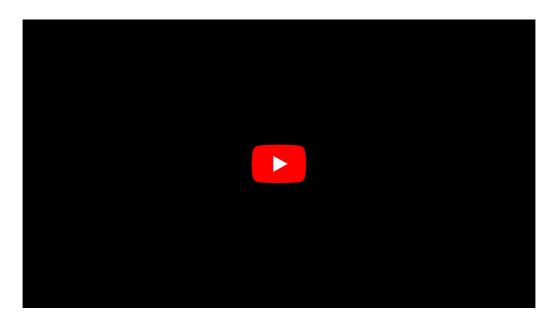
CS229: Machine Learning - The Summer Edition!

Course Description This is the summer edition of CS229 Machine Learning that was offered over 2019 and 2020. CS229 provides a broad introduction to statistical machine learning (at an intermediate / advanced level) and covers supervised learning (generative/discriminative learning, parametric/non-parametric learning, neural networks, support vector machines); unsupervised learning (clustering, dimensionality reduction, kernel methods); learning theory (bias/variance tradeoffs, practical); and reinforcement learning among other topics. The structure of the summer offering enables coverage of additional topics, places stronger emphasis on the mathematical and visual intuitions, and goes deeper into the details of various topics.

Full playlist (YouTube)



Syllabus and Course Schedule

Event	Date	Description	Materials and Assignments				
Introduction and Pre-requisties review (3 lectures)							
Lecture 1 [YouTube]	6/24	Introduction and LogisticsReview of Linear Algebra	Class Notes ■ Introduction [pptx] ■ Linear Algebra (section 1-3) [pdf]				
Lecture 2 [YouTube]	6/26	Review of Matrix CalculusReview of Probability	 Class Notes Linear Algebra (section 4) [pdf] Probability Theory [pdf] Probability Theory Slides [pdf] 				
Lecture 3 [YouTube]	6/28	Review of Probability and StatisticsSetting of Supervised Learning	Class NotesSupervised Learning [pdf]Probability Theory [pdf]				

		Supervised Learning (8 le	
Lecture 4 [YouTube]	7/1	 Linear Regression [Stochastic] Gradient Descent ([S]GD) Normal Equations Probabilistic Interpretation Maximum Likelihood Estimation (MLE) 	Supervised Learning (section 1-3) [pd:
Lecture 5 [YouTube]	7/3	PerceptronLogistic RegressionNewton's Method	Class Notes • Supervised Learning (section 5-7) [pdf
Lecture 6 [YouTube]	7/5	Exponential FamilyGeneralized Linear Models (GLM)	Class Notes • Supervised Learning (section 8-9) [pd
Lecture 7 [YouTube]	7/8	 Gaussian Discriminant Analysis (GDA) Naive Bayes Laplace Smoothing 	Class Notes • Generative Algorithms [pdf]
Lecture 8 [YouTube]	7/10	Kernel MethodsSupport Vector Machine	Class Notes • Kernel Methods and SVM [pdf]
Lecture 9 [YouTube]	7/12	 Bayesian Methods Parametric (Bayesian Linear Regression) Non-parametric (Gaussian process) 	 Class Notes Gaussian Processes [pdf] Optional The Multivariate Gaussian Distribution [pdf] More on Gaussian Distribution [pdf]
Lecture 10 [YouTube]	7/15	 Neural Networks and Deep Learning 	Class Notes • Deep Learning (skip Sec 3.3) [pdf] Optional • Backpropagation [pdf]
Lecture 11 [YouTube]	7/17	Deep Learning (contd)	
		Theory (2 lectures)	
Lecture 12 [YouTube]	7/19	Bias and VarianceRegularization, Bayesian Interpretation	Class Notes • Regularization and Model Selection [p

Lecture 13 [YouTube]	7/22	Bias-Variance tradeoff (wrap-up)Empirical Risk MinimizationUniform Convergence	Class NotesBias Variance Analysis [pdf]Statistical Learning Theory [pdf]
		Reinforcement Learning (2	lectures)
Lecture 14 [YouTube]	7/24	 Reinforcement Learning (RL) Markov Decision Processes (MDP) Value and Policy Iterations 	Class Notes • Reinforcement Learning and Control (Sec 1-2) [pdf]
Lecture 15 [YouTube]	7/26	RL (wrap-up)Learning MDP modelContinuous States	Class Notes • Reinforcement Learning and Control (Sec 3-4) [pdf]
		Unsupervised Learning (3	lectures)
Lecture 16 [YouTube]	7/29	 Unsupervised Learning K-means clustering Mixture of Gaussians (GMM) Expectation Maximization (EM) 	 Class Notes K-means [pdf] Mixture of Gaussians [pdf] Expectation Maximization (Sec 1-2, skip 2.1) [pdf]
Lecture 17 [YouTube]	7/31	EM (wrap-up)Factor Analysis	Class NotesExpectation Maximization (Sec 3) [pdf]Factor Analysis [pdf]
Lecture 18 [YouTube]	8/2	 Factor Analysis (wrap-up) Principal Components Analysis (PCA) Independent Components Analysis (ICA) 	 Class Notes Principal Components Analysis [pdf] Independent Components Analysis [pdf]
		Miscellaneous Topics (3 le	ectures)
Lecture 19	8/5	 Maximum Entropy and Exponential Family KL-Divergence Calibration and Proper Scoring Rules 	Class Notes • Maximum Entropy [pdf]
Lecture 20	8/7	 Variational Inference EM Variants Variational Autoencoder	Class Notes • VAE (Sec 4) [pdf]
Lecture 21	8/9	Evaluation Metrics	Class Notes • Evaluation Metrics [pptx]

Recap and wrap-up (2 lectures)						
Lecture 22	8/12	Practical advice and tipsReview for Finals	Class Notes			
Lecture 23	8/14	Review for Finals	Class Notes			
Final	8/16					

Other Resources

- 1. Advice on applying machine learning: Slides from Andrew's lecture on getting machine learning algorithms to work in practice can be found here.
- 2. Previous projects: A list of last year's final projects can be found here.
- 3. 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.
- 4. 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.
- 5. Machine learning study guides tailored to CS 229 by Afshine Amidi and Shervine Amidi.