

ECE595 / STAT598: Machine Learning I

Course Overview

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Elements of Learning?

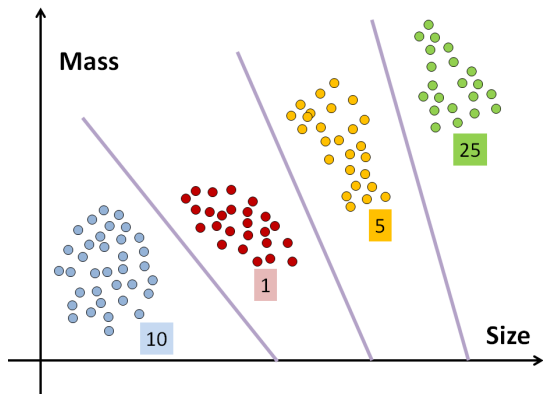
- Data
- Computer
- Algorithm

What is Learning? What is NOT learning?



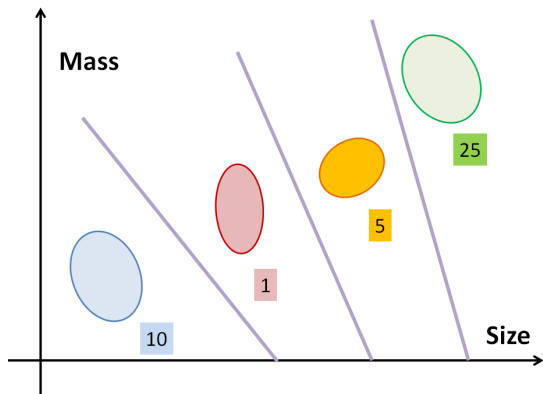
- You are given a bag of US coins.
- Your task: Build a classifier.
- Four classes: Penny, Nickel, Dime, Quarter.

Approach 1: Learning



- You measure mass and size.
- Put each coin to its class. Plot a 2D histogram.
- Create the classifier.

Approach 2: Design



- You go to United States Mint to ask ideal mass and size of the coins.
- You ask them to give you the measurement error. Plot 2D distribution.
- Create the classifier.

Which one fits learning? Which one fits design?

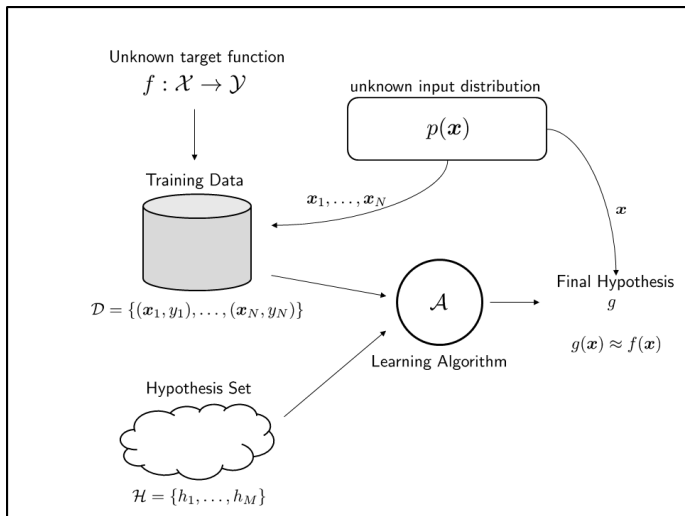
- Determining the age at which a particular medical test should be performed.
- Classifying numbers into primes and non-primes.
- Detecting potential fraud in credit card charges.
- Determining the time it would take a falling object to hit the ground.
- Determining the optimal cycle for traffic lights in a busy intersection.

Machine Learning Model

- Data points $\mathbf{x}_1, \dots, \mathbf{x}_N$.
- Labels y_1, \dots, y_N .

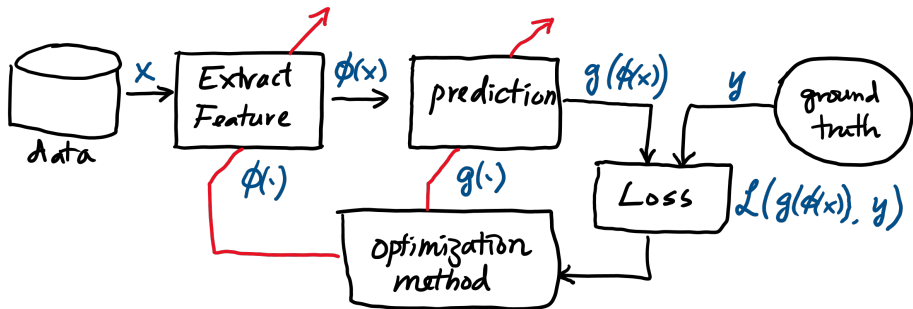
- Where does a data point \mathbf{x}_n come from?
- How is a label y_n defined?
- What do we mean by a learning algorithm?
- What is a classifier?
- How to evaluate a classifier?

Machine Learning Model



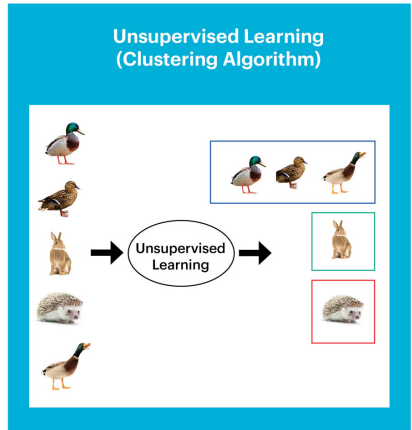
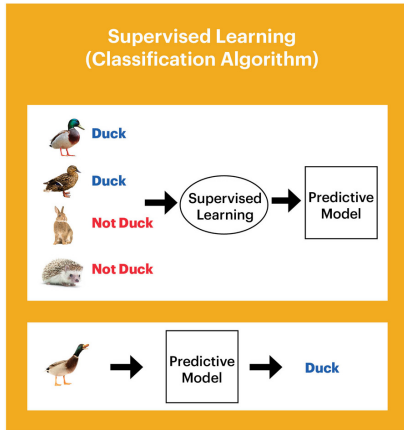
See Learning from Data (Chapter 1).

Learning Algorithm



Types of Learning

- Supervised Learning: Labels available.
- Unsupervised Learning: No label.

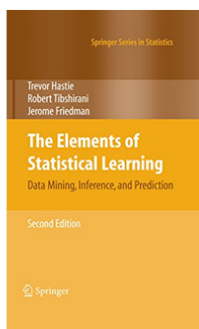
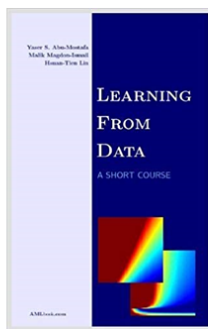
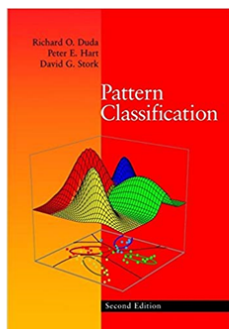
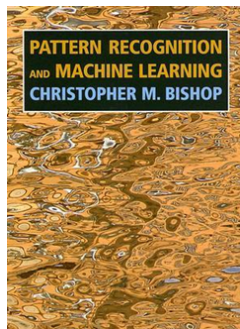


Outline of ECE 595

- **Part 1: Mathematical Background** (2 weeks)
 - Linear Regression and Optimization
 - Please review linear algebra, probability, optimization in the Tutorial Note.
- **Part 2: Classification** (5 weeks)
 - Methods to train linear classifiers
 - Feature analysis, Geometry, Bayesian decision rule, logistic regression, perceptron algorithm, support vector machine
- **Part 3: Handling Uncertainty** (3 weeks)
 - Imperfect data: noisy label, unbalanced data, missing data, knowledge transfer
 - Robustness study: adversarial attack and defense
- **Part 4: Learning Theory** (5 weeks)
 - Evaluation of a classifier.
 - Feasibility of learning, VC dimension, bias-variance, validation

Textbook and References

- **Pattern Classification**, by Duda, Hart and Stork, 2000.
- **Pattern Recognition and Machine Learning**, by Bishop, 2006.
- **Learning from Data**, by Abu-Mostafa, Magdon-Ismail and Lin, 2012.
- **Elements of Statistical Learning**, by Hastie, Tibshirani and Friedman, 2009.



Pre-Requisites

- **Linear Algebra:**

- Matrix-vector multiplication \mathbf{Ax} , transpose \mathbf{A}^T , symmetric matrices $\mathbf{A} = \mathbf{A}^T$, norm $\|\mathbf{x}\|$, trace $\text{Tr}(\mathbf{A})$, inverse \mathbf{A}^{-1} , determinant $|\mathbf{A}|$, eigenvalue and eigenvector $\mathbf{A} = \mathbf{U}\mathbf{\Lambda}\mathbf{U}^T$.
- Gilbert Strang, *Linear Algebra and Its Applications*, 5th Edition

- **Probability:**

- Random variable X , probability density function $p(x)$, cumulative distribution function $F(x)$, expectation $\mathbb{E}[X]$, variance $\text{Var}[X]$, function of random variable $\mathbb{E}[g(X)]$, joint Gaussian, Law of Large Number, Central Limit Theorem.
- Dimitri Bertsekas, *Introduction to Probability*, Athena Scientific, 2008, 2nd Edition.

- **Optimization:**

- Convex function, convex set, operations which preserve convexity, Lagrange multiplier, KKT conditions, primal optimal, dual optimal, complementary slackness, constrained optimization, duality theorem.
- Stephen Boyd and Lieven Vandenberghe, *Convex Optimization*, Cambridge 2004.

Reading List

Tutorials

- Tutorial on Linear Algebra
https://engineering.purdue.edu/ChanGroup/ECE595/files/Tutorial01_handout.pdf
- Tutorial on Probability
https://engineering.purdue.edu/ChanGroup/ECE595/files/Tutorial02_handout.pdf