

CSE 417T

Introduction to Machine Learning

Instructor: Chien-Ju Ho

Plan for Today

- Welcome and introduction
- What's the class about?
- Logistics (including waitlist and enrollments)
- Lecture
 - Setting up the learning problem
 - Perceptron learning algorithm
- Homework 1 will be announced next week

About Me

- Joined WashU in Fall 2017
- Research interests:
 - Design and analysis of human-in-the-loop systems
 - Crowdsourcing and human computation, machine learning, game theory, optimization, online behavioral social science, and human-computer interactions

What is Machine Learning?

What Machine Learning Can Do?

Recommended for You

These recommendations are based on items you own and more.

[All](#) | [New Releases](#) | [Coming Soon](#)



Cybertext: Perspectives on Ergodic Literature

by Espen J. Aarseth (Aug 6, 1997)

Average Customer Review: ★★★★★ (3)

In Stock

List Price: \$22.95

Price: **\$19.55**

29 used & new from \$10.82

Add to cart

Add to

☐ I own it ☐ Not interested ☒ ☆☆☆☆☆ Rate it

Recommended because you added [Hamlet on the Holodeck](#) to your Shopping Cart and more ([Fix this](#))



Narrative as Virtual Reality: Immersion and Interactivity in Lit Media (Parallax: Re-visions of Culture and Society)

by Mark J. P. O'Connell (Oct 3, 2003)

COMPOSE

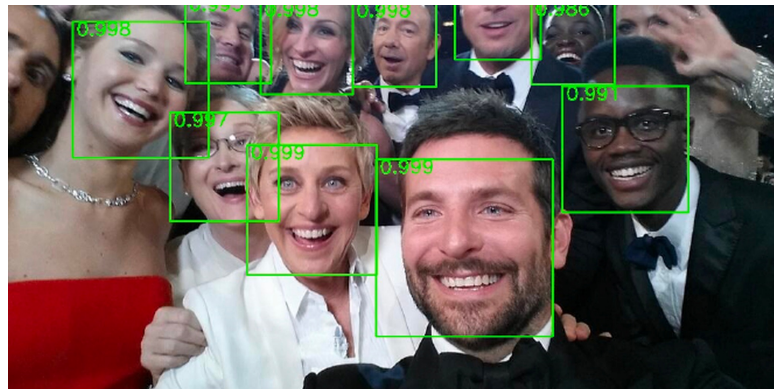
Inbox

Sent Mail

Drafts (6)

Spam (589)

What can I help
you with?



Example: Credit Card Approval

Input: customer information

age	32 years
gender	male
salary	40,000
debt	26,000
years in job	1 year
years at home	3 years
...	...

Output: a prediction

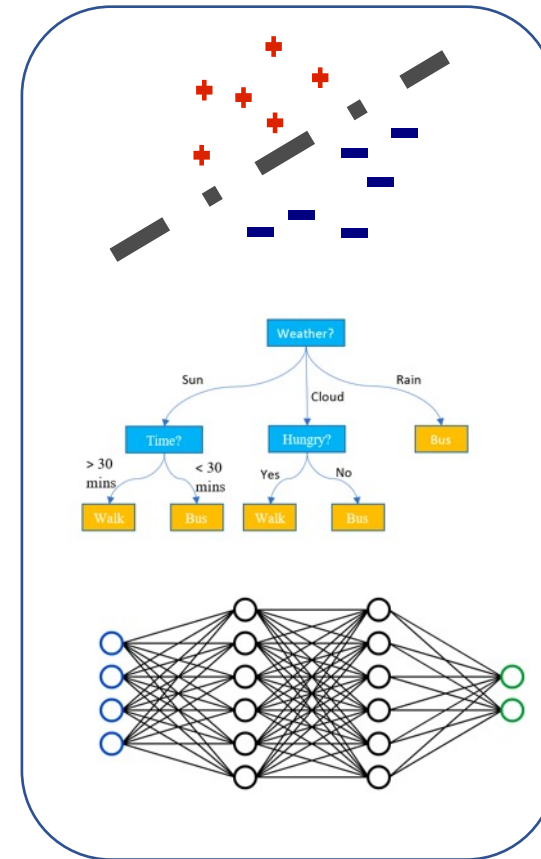
Will the customer be a good customer for the bank?

- A brute-force solution
 - Build a large table that maps all possible attribute combinations to a prediction
- Not really feasible at scale
 - Possible attribute combinations could be infinite
 - Storage, computation, how to come up with the table?

A "Machine Learning" Approach

Hypothesis / Model (Some math function)

Data (Historical
Customer Data)



Find a hypothesis that "fits" the data
(The process requires a lot of computation)

What is Machine Learning?

“learning from data”

“using a set of observations to uncover an underlying pattern”

Use scenarios of machine learning

- A pattern exists
- No analytical solution: We cannot pin it down mathematically
- We have data on it

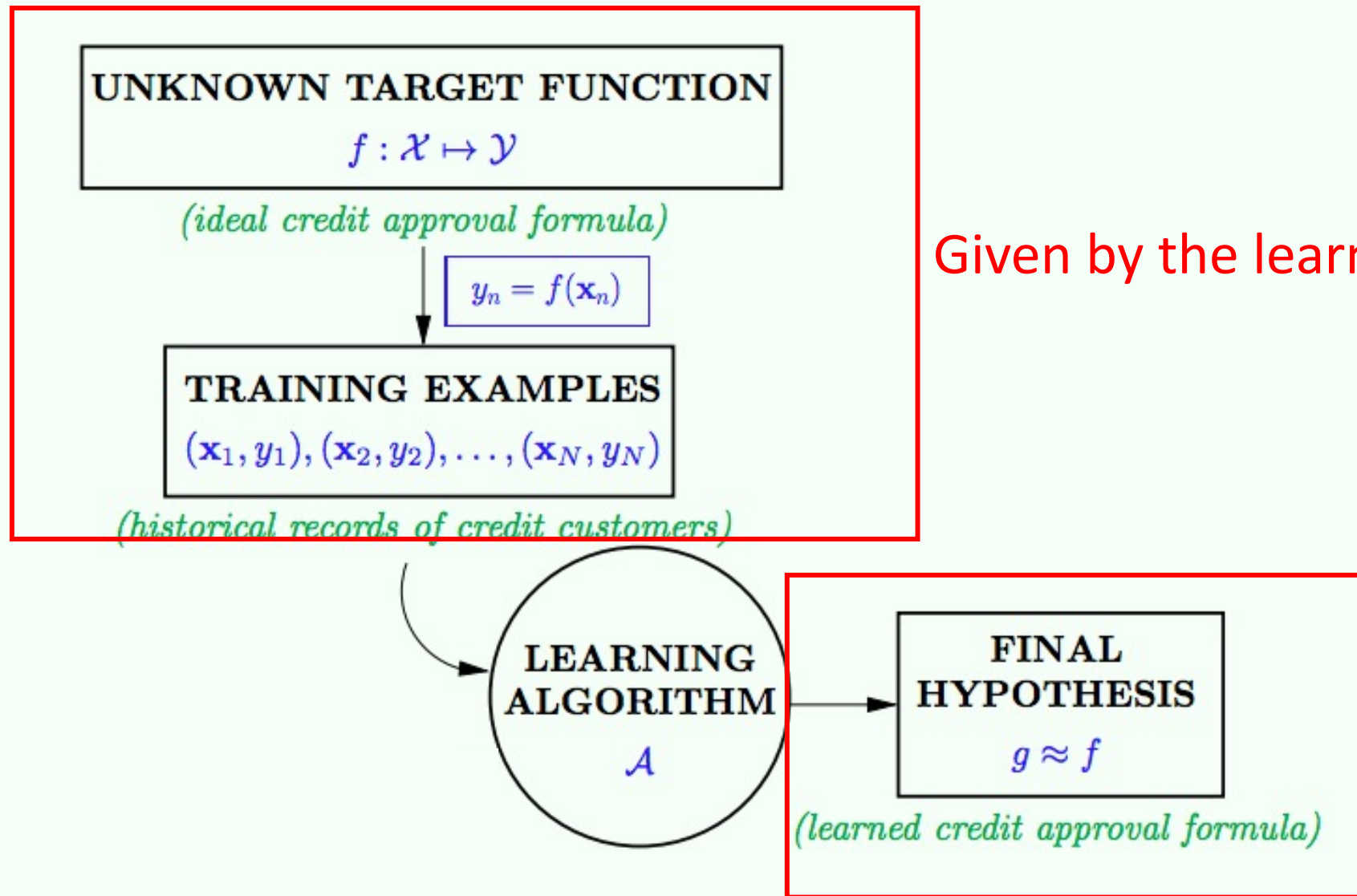
More Formally (For Supervised Learning)

- Formulation: (credit card approval example)
 - input (features): $\vec{x} = (x_1, x_2, \dots, x_d) \in X$ (customer's information)
 - output (label): $y \in Y$ (good/bad customer)
 - **unknown** target function: $f: X \rightarrow Y$ (ideal credit approval formula)
 - data $(\vec{x}_1, y_1), \dots, (\vec{x}_N, y_N)$ (historical records)
 - goal: learn a g close to f (formula to be used)
- Two central questions
 - How do we learn g ?
 - What can we say about how close g is to f ?

Note on notations:

We interchangeably use (bold font) x and \vec{x} to denote a vector in this course.

The former is used in the textbook.
the later is for the convenience of writing.



Given by the learning problem

Goal of learning

UNKNOWN TARGET FUNCTION

$$f : \mathcal{X} \mapsto \mathcal{Y}$$

(ideal credit approval formula)

$$y_n = f(\mathbf{x}_n)$$

TRAINING EXAMPLES

$$(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \dots, (\mathbf{x}_N, y_N)$$

(historical records of credit customers)

Given by the learning problem

**LEARNING
ALGORITHM**

\mathcal{A}

**FINAL
HYPOTHESIS**

$$g \approx f$$

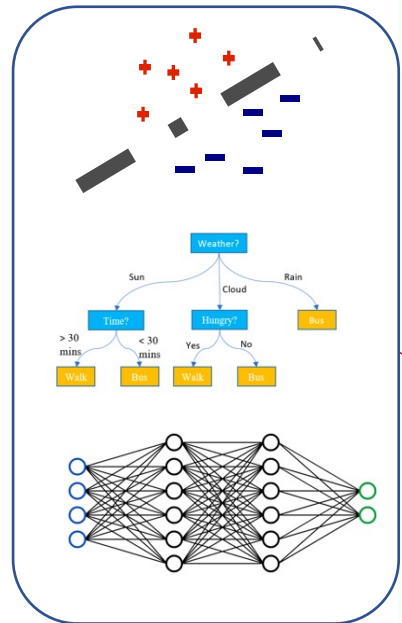
(learned credit approval formula)

Goal of learning

HYPOTHESIS SET

\mathcal{H}

(set of candidate formulas)



UNKNOWN TARGET FUNCTION

$$f : \mathcal{X} \mapsto \mathcal{Y}$$

(ideal credit approval formula)

$$y_n = f(\mathbf{x}_n)$$

TRAINING EXAMPLES

$$(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \dots, (\mathbf{x}_N, y_N)$$

(historical records of credit customers)

**LEARNING
ALGORITHM**

$$\mathcal{A}$$

**FINAL
HYPOTHESIS**

$$g \approx f$$

(learned credit approval formula)

HYPOTHESIS SET

$$\mathcal{H}$$

(set of candidate formulas)

learning model

Course Plan

- First half of the semester: **Foundations**
 - Focus on **linear models**
 - Fundamental components of many other models
 - Discuss the theoretical foundations of machine learning
 - Heavy use of probability, linear algebra, and optimization
- Second half of the semester: **Techniques**
 - Discuss different learning models

Course Plan

- Foundations

- What's machine learning
- Feasibility of learning
- Generalization
- Linear models
- Non-linear transformations
- Overfitting and how to avoid it
 - Regularization
 - Validation

- Techniques

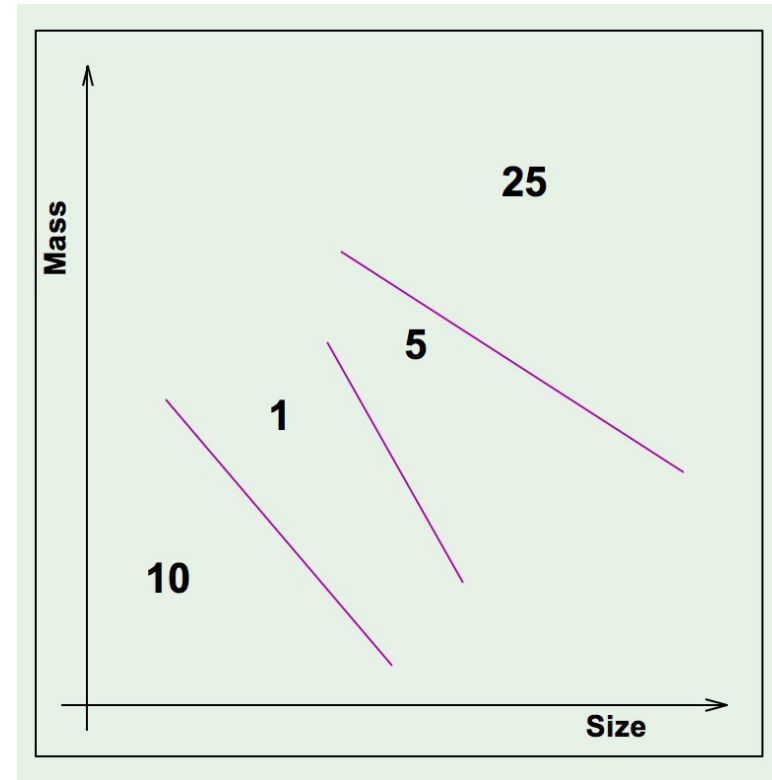
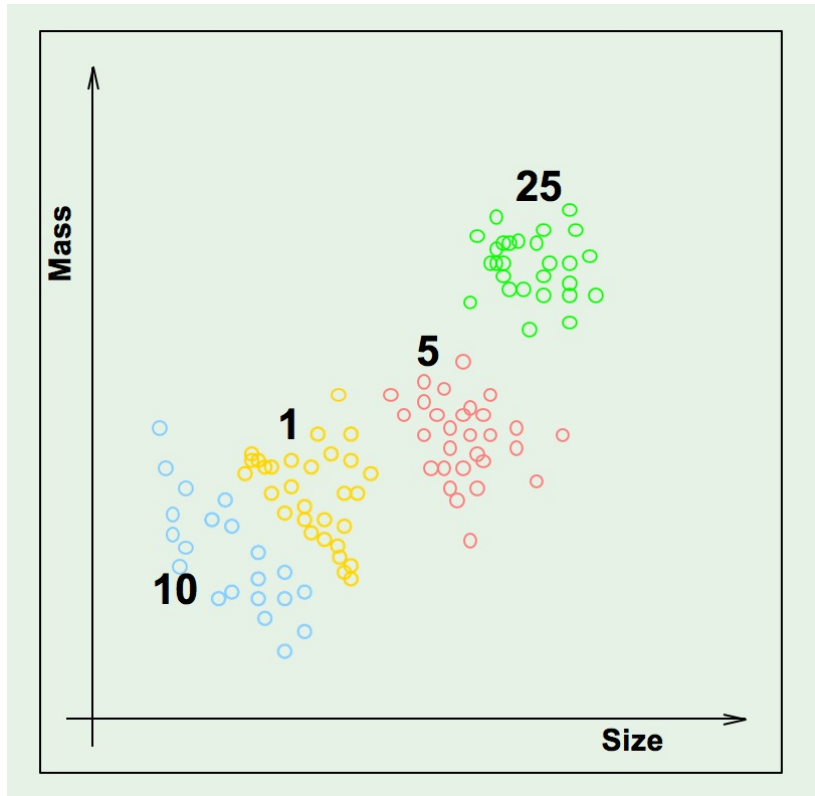
- Nearest neighbors
- Decision tree
- Support vector machine
- Boosting
- Random forest
- Neural networks
- ...

Types of Learning

- Supervised learning (the focus of this course)
 - Given training data (input, correct output)
 - Try to predict the output for data not seen before
- Unsupervised learning
 - Given data in the form of (input)
 - Find patterns in data
- Reinforcement learning
 - Learn how to act, based on rewards for actions

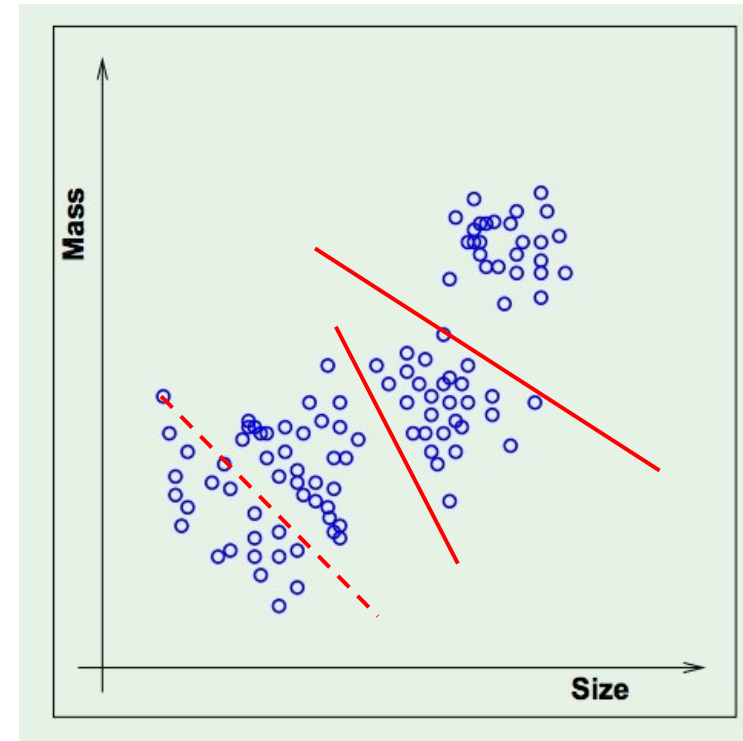
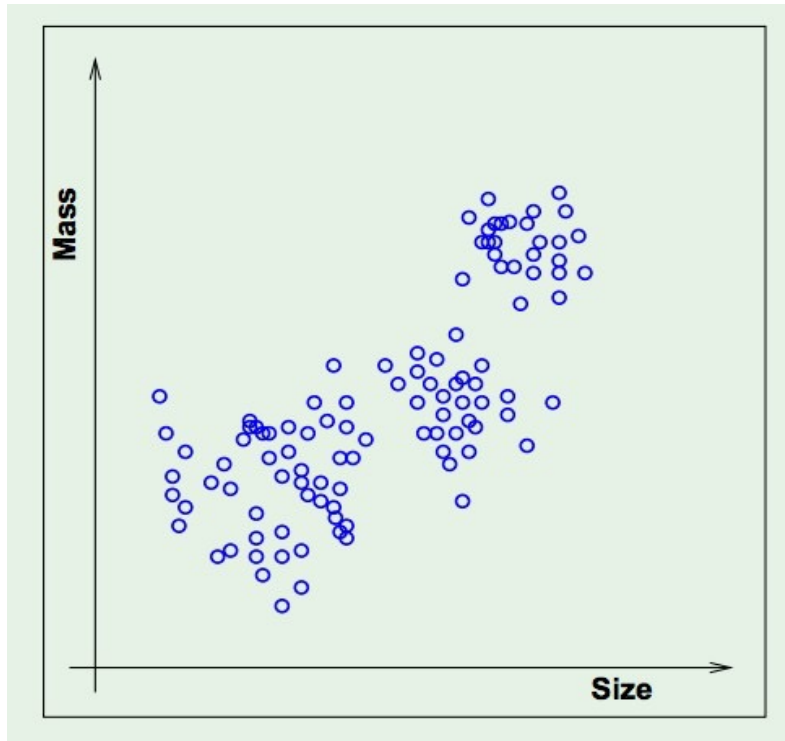
Supervised Learning

- Given data with (input, correct output), learn a pattern that can predict previously unseen data



Unsupervised Learning (more in 517A)

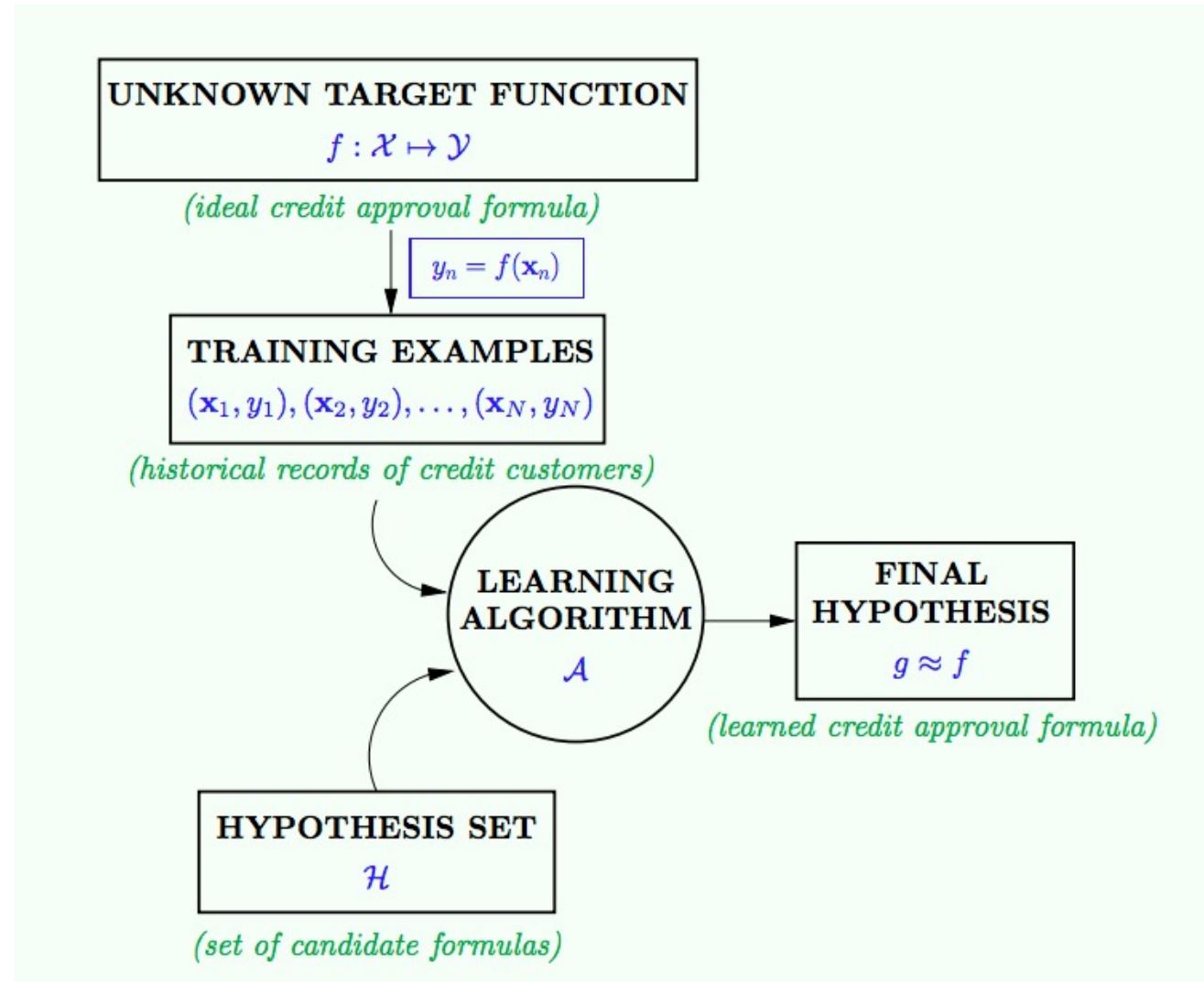
- Suppose you only have the feature vectors but no labels. Still want to describe the data in some useful way.



Reinforcement Learning (more in 412A)

- Agent interacts with the world by taking actions
- Feedback is in the form of rewards (or costs)
- Agent must learn a policy, which maps from the state of the world to an action
- Major issues:
 - Exploration / exploitation
 - Delayed reward / credit assignment

Course plans (focus on supervised learning)



Logistics

- Websites

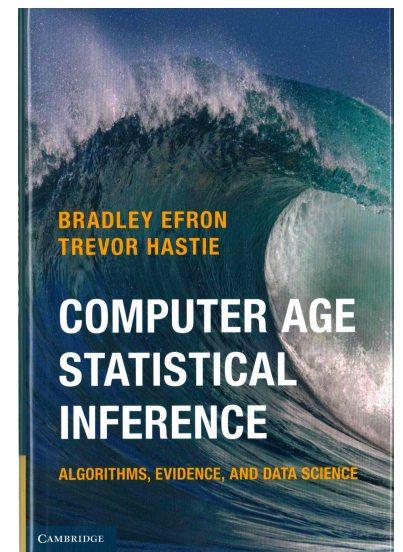
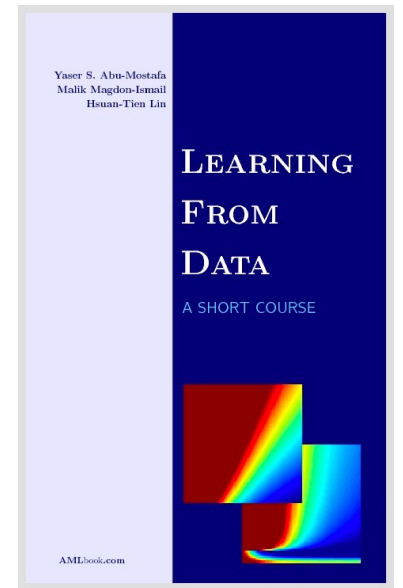
- Course website: <http://chienjuho.com/courses/cse417t>
- Piazza for discussion
- Gradescope for homework submissions
- You are responsible for following the announcements on website and Piazza

- TA and Office Hours

- There will be several graduate/undergraduate TAs
- Office hours will be announced in the 2nd week and start in the 3rd week

Course Information: Textbooks

- *Learning From Data*
 - Y. Abu-Mostafa, M. Magdon-Ismail, and H-T Lin.
 - <http://amlbook.com/>
 - We will go through this book in the first half of the semester.
- *Computer Age Statistical Inference: Algorithms, Evidence, and Data Science.*
 - B. Efron and T. Hastie
 - <https://web.stanford.edu/~hastie/CASI/>
 - We will reference a few sections as the course materials. The PDF file is freely available on their website



Course Information: Grading

- Homework assignments (5 to 6): 50%
 - Mix of programming and pencil-and-paper problems
 - Worst score discounted by 50%
 - Programming language: Python
 - 5 total late days, no more than 2 on any one assignment
- Two exams: 50% (25% each)
 - One in the middle of the semester (sometime in March)
 - One on the last day of the semester
 - Each exam covers around half of the materials. No separate final exam.
 - More details will be announced later

Course Information

- Take a look at the syllabus
 - Collaborations
 - You are encouraged to discuss homework with other students
 - **Must write your own solutions**
 - **Must cite all external sources (including other students)**
 - Other accommodation resources
- Academic integrity
 - **Zero tolerance** on the violation
 - Will be reported to the university
 - There will be permanent record if found guilty

Course Information

- Questions not covered in the syllabus?
 - Ask me!
 - Generally, I don't grant individual exceptions
 - Can I do extra work to get more points?
 - I have another exam this week, can I get an additional day for the assignment?
 - I work really hard but can't finish the assignment on time. Can I get an additional day for the assignment?
 - I work really hard. Can I get higher grades?
 - **No** to all the above.
 - Exceptions: family/medical emergencies
- Rule of thumb:
 - I only say yes if I feel comfortable giving the same treatment to everyone in the class.

Getting in Touch

- Please use **Piazza** as the main communication channel
 - Emails might not be responded
- Very strongly encouraged to use **public** posts
 - Other students might have the same questions
 - Other students might help answer the questions
- When to use private posts
 - The questions involve disclosing your answers to the homework
 - The questions are about your personal matter

Is This Course for You?

- This is going to be a very theory-heavy course
 - The “T” in CSE 417T stands for “Theory”
 - There will be **A LOT OF** math!
- We focus on understanding the foundations of machine learning
- If your main goal is to learn how to apply ML to solve problems, this might not be the best course for that.
 - Check out CSE 217A, CSE 412A, ESE 417, BME 440, INFO 558

Waitlist and Enrollment

- If you are on the waitlist and want to get enrolled:
 - Complete [homework 0](#) (posted on the website)
 - Prove Chebyshev's inequality, one form of the law of large numbers
 - Prove PLA (discussed next) converges in finite steps
 - Implement PLA and observe its empirical performance
 - Explain why you want/need to take this course in this semester
 - Submit via [Gradescope](#) by [11:30am next Tuesday \(Jan 25\)](#)
 - Priorities will be given to students who benefit the most by taking this course this semester (condition on having satisfactory answers to the questions)
- If you are enrolled
 - You should look at homework 0 as well (please don't submit yet)
 - It helps you get a better sense of what this course is about

Questions?

Lecture Today

UNKNOWN TARGET FUNCTION

$$f : \mathcal{X} \mapsto \mathcal{Y}$$

(ideal credit approval formula)

$$y_n = f(\mathbf{x}_n)$$

TRAINING EXAMPLES

$$(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \dots, (\mathbf{x}_N, y_N)$$

(historical records of credit customers)

**LEARNING
ALGORITHM**

$$\mathcal{A}$$

**FINAL
HYPOTHESIS**

$$g \approx f$$

(learned credit approval formula)

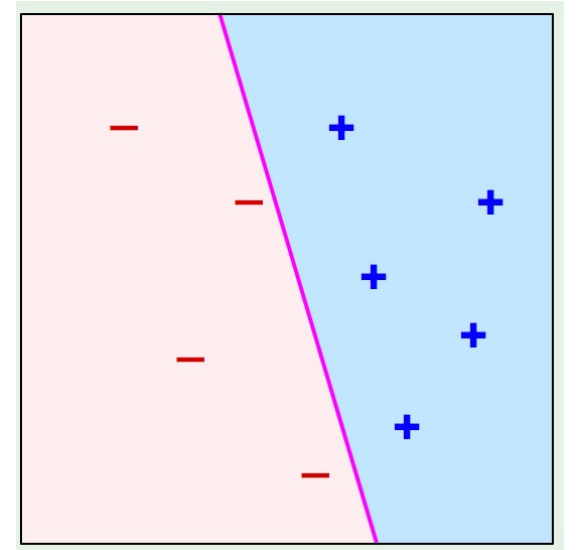
HYPOTHESIS SET

$$\mathcal{H}$$

(set of candidate formulas)

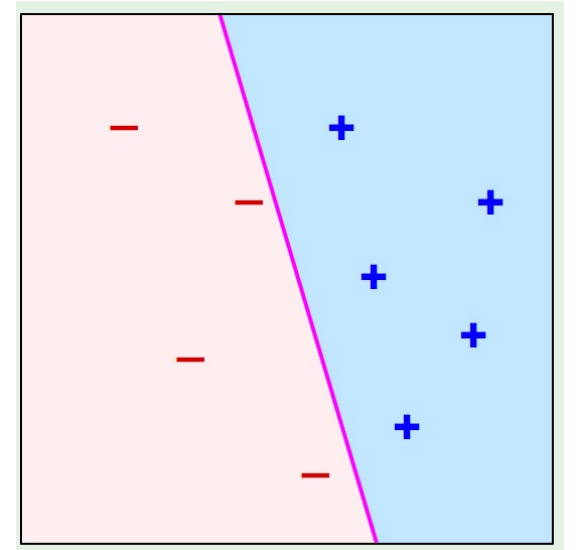
Linear Hypothesis Space (Perceptron)

- Input $\vec{x} = (x_1, x_2, \dots, x_d)$
- Output $y \in \{-1, +1\}$



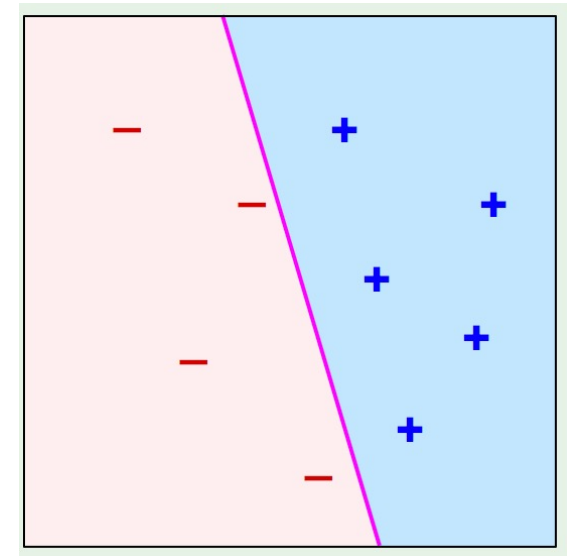
Linear Hypothesis Space (Perceptron)

- Input $\vec{x} = (x_1, x_2, \dots, x_d)$
- Output $y \in \{-1, +1\}$
- A hypothesis h is a linear separator $\vec{w}^T \vec{x} = b$, characterized by
 - weight vector $\vec{w} = (w_1, \dots, w_d)$
 - threshold b
- $h(\vec{x}) = \text{sign}(\sum_{i=1}^d w_i x_i - b) = \text{sign}(\vec{w}^T \vec{x} - b)$
 - Predict $+1$ if $\vec{w}^T \vec{x} > b$
 - Predict -1 if $\vec{w}^T \vec{x} < b$



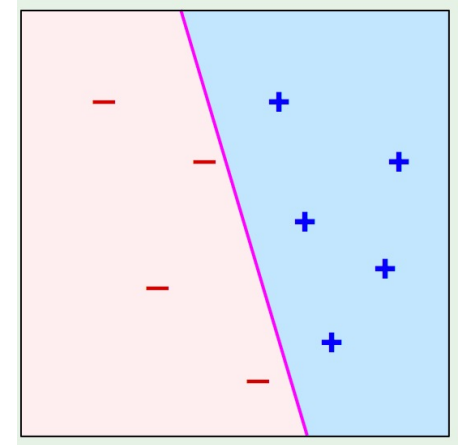
Linear Hypothesis Space (Perceptron)

- To simplify $h(\vec{x}) = \text{sign}(\vec{w}^T \vec{x} - b)$, define
 - $x_0 = 1$
 - $w_0 = -b$
- And we implicitly let
 - $\vec{x} = (x_0, x_1, \dots, x_d)$
 - $\vec{w} = (w_0, w_1, \dots, w_d)$
- A hypothesis can then be written as
 - $h(\vec{x}) = \text{sign}(\vec{w}^T \vec{x})$
 - We will interchangeably use h and \vec{w} to express a hypothesis in Perceptron



Perceptron Learning Algorithm (PLA)

- Given a dataset $D = \{(\vec{x}_1, y_1), \dots, (\vec{x}_N, y_N)\}$
- Assume the dataset is **linearly separable**
- How do we learn a hypothesis that separates the data?
- Perceptron Learning Algorithm
 - Initialize $\vec{w}(0) = \vec{0}$
 - For $t = 0, \dots$
 - Find a misclassified data point $(\vec{x}(t), y(t))$ in D
 - That is, $\text{sign}(\vec{w}(t)^T \vec{x}(t)) \neq y(t)$
 - If no such data point exists
 - Return $\vec{w}(t)$
 - Else
 - $\vec{w}(t + 1) \leftarrow \vec{w}(t) + y(t)\vec{x}(t)$



Notation:

We use $\vec{w}(t)$ to denote the value of \vec{w} at step t of the algorithm.

Similarly, we use $(\vec{x}(t), y(t))$ to denote the data point found at step t .

Some Intuitions

Perceptron Learning Algorithm (PLA)

- Theorem (informal):
 - If a dataset D is linearly separable, PLA find a linear separator that separates the data in D within a finite number of steps.
- HW0:
 - Prove Chebyshev's inequality
 - Prove the above theorem
 - Implement PLA using Python
 - Explain why you want/need to take this course this semester