CSE4088 Introduction to Machine Learning

Components of Learning
Perceptron Learning Algorithm

Slides are adopted from lecture notes of Yaser Abu-Mostafa

Outline

- Components of learning
- The perceptron learning algorithm
- Types of learning

Review: The essence of machine learning

- A pattern exists
- We can not pin it down mathematically
- We have data on it

- Example:
  - Tree images have a pattern.
  - It is difficult to give a mathematical definition of a tree.
  - However, we can learn to classify tree images if we have enough data.

Exercise

Which of the following problems are best suited for Machine Learning?

(i) Classifying numbers into primes and non-primes.
(ii) Detecting potential fraud in credit card charges.
(iii) Determining the time it would take a falling object to hit the ground.
(iv) Determining the optimal cycle for traffic lights in a busy intersection.

[a] (ii) and (iv)
[b] (i) and (ii)
[c] (i), (ii), and (iii)
[d] (iii)
[e] (i) and (iii)

Components of Learning

Application: Credit approval

Applicant information:

age	23 years
gender	male
annual salary	$30,000
years in residence	1 year
years in job	1 year
current debt	$15,000
...	...

Approve credit?

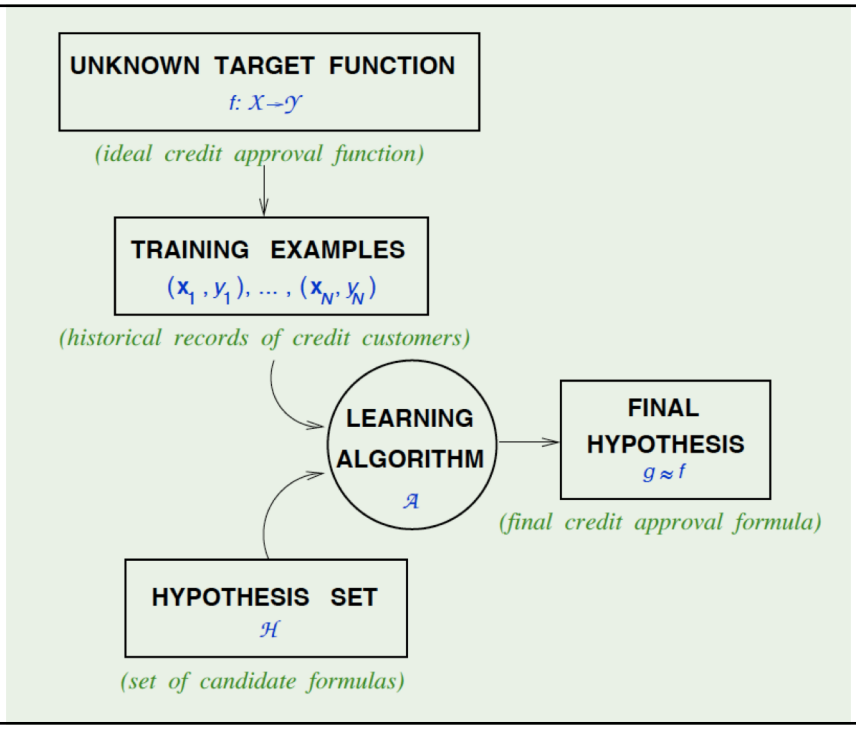
Components of Learning

Formalization:

- Input: $\mathbf{x}$ (*customer application*)
- Output: y (*good/bad customer?*)
- Target function: $f : \mathcal{X} \rightarrow \mathcal{Y}$ (*ideal credit approval formula*)
- Data: $(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \cdots, (\mathbf{x}_N, y_N)$ (*historical records*)

$\downarrow \quad \downarrow \quad \downarrow$

- Hypothesis: $g : \mathcal{X} \rightarrow \mathcal{Y}$ (*formula to be used*)



Solution Components

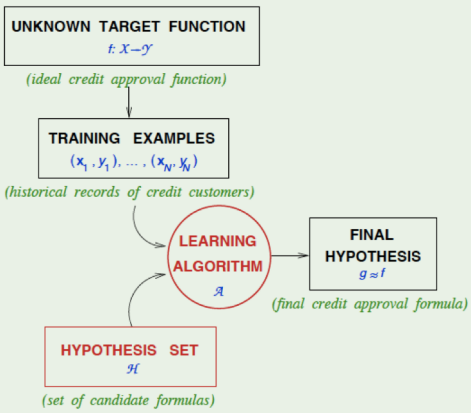
The 2 solution components of the learning problem:

- The Hypothesis Set

  $\mathcal{H} = \{h\}$ $\quad g \in \mathcal{H}$

- The Learning Algorithm

Together, they are referred to as the *learning model*.



A simple hypothesis set – the perceptron

For input $\mathbf{x} = (x_1, \cdots, x_d)$ 'attributes of a customer'

$$\text{Approve credit if} \quad \sum_{i=1}^{d} w_i x_i > \text{threshold},$$

$$\text{Deny credit if} \quad \sum_{i=1}^{d} w_i x_i < \text{threshold}.$$

This linear formula $h \in \mathcal{H}$ can be written as

$$h(\mathbf{x}) = \text{sign}\left(\left(\sum_{i=1}^{d} w_i x_i\right) - \text{threshold}\right)$$

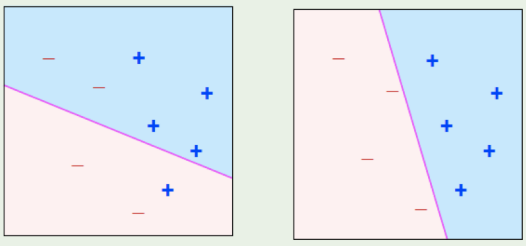
$$h(\mathbf{x}) = \text{sign}\left(\left(\sum_{i=1}^{d} w_i\, x_i\right) + \quad w_0 \quad\right)$$

Introduce an artificial coordinate $x_0 = 1$:

$$h(\mathbf{x}) = \text{sign}\left(\sum_{i=0}^{d} w_i\, x_i\right)$$

In vector form, the perceptron implements

$$h(\mathbf{x}) = \text{sign}(\mathbf{w}^{\mathsf{T}}\mathbf{x})$$



'linearly separable' data

A Simple Learning Algorithm:
Perceptron Learning Algorithm (PLA)

The perceptron implements

$$h(\mathbf{x}) = \text{sign}(\mathbf{w}^\intercal \mathbf{x})$$

Given the training set:

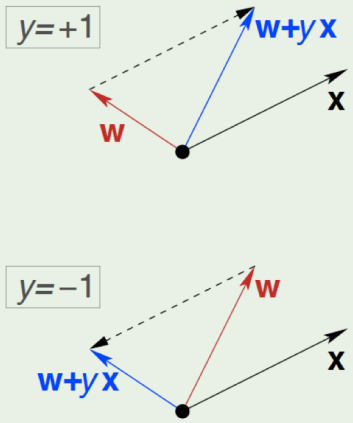
$$(\mathbf{x}_1, y_1), (\mathbf{x}_2, y_2), \cdots, (\mathbf{x}_N, y_N)$$

pick a misclassified point:

$$\text{sign}(\mathbf{w}^\intercal \mathbf{x}_n) \neq y_n$$

and update the weight vector:

$$\mathbf{w} \leftarrow \mathbf{w} + y_n \mathbf{x}_n$$



Iterations of PLA

- One iteration of the PLA:

$$\mathbf{w} \leftarrow \mathbf{w} + y\mathbf{x}$$

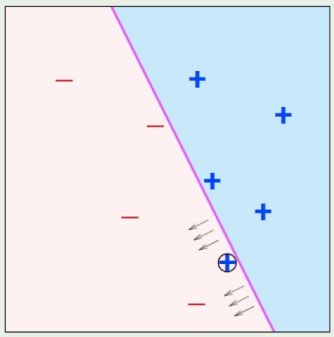
where $(\mathbf{x}, y)$ is a misclassified training point.

- At iteration $t = 1, 2, 3, \cdots$, pick a misclassified point from

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

and run a PLA iteration on it.

- That's it!



Review: Types of Learning

Basic premise of learning

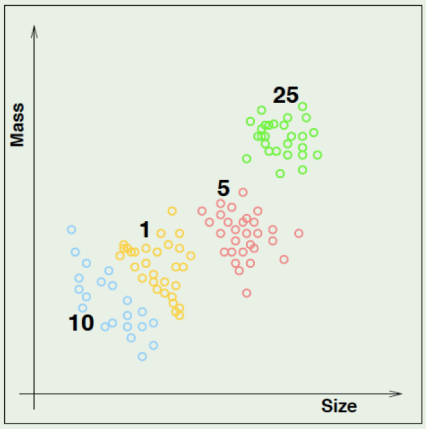
"using a set of observations to uncover an underlying process"

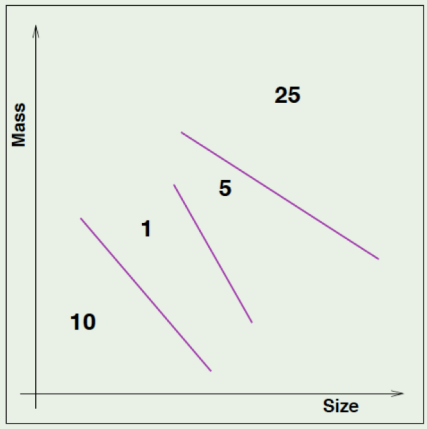
broad premise $\Longrightarrow$ many variations

- Supervised Learning
- Unsupervised Learning
- Reinforcement Learning

Supervised Learning

Example from vending machines – **coin recognition**



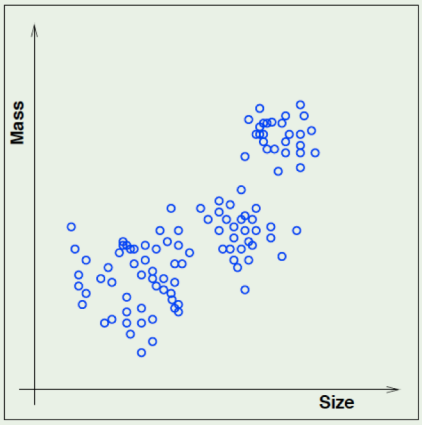


Supervised Learning

- Unknown target function $y = f(\mathbf{x})$
- Data set $(\mathbf{x}_1, y_1), \cdots, (\mathbf{x}_N, y_N)$
- Learning algorithm picks $g \approx f$ from a hypothesis set $\mathcal{H}$

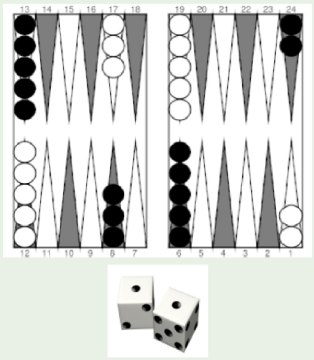
Unsupervised Learning

Instead of (input,correct output), we get (input, ?)



Reinforcement Learning

Instead of (input,correct output),
we get (input,*some* output,grade for this output)



The world champion was a neural network!