

# Machine Learning CS60050



Introduction

# Learning From Data

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Lecture 1: The Learning Problem





# The learning problem - Outline

- Example of machine learning
- Components of Learning
- A simple model
- Types of learning
- Puzzle

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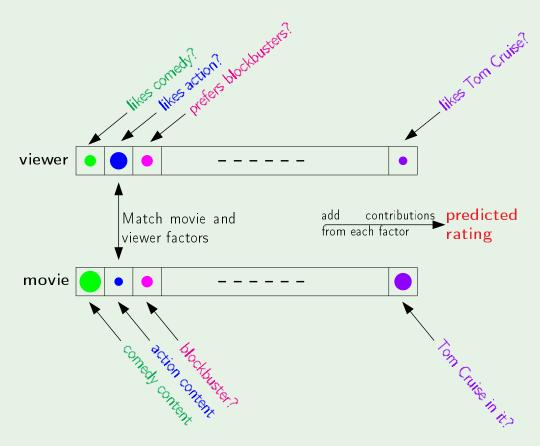
**Example:** Predicting how a viewer will rate a movie

10% improvement = 1 million dollar prize

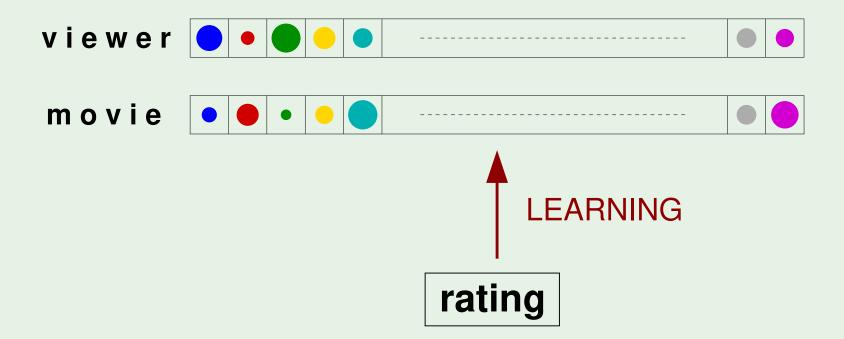
The essence of machine learning:

- A pattern exists.
- We cannot pin it down mathematically.
- We have data on it.

# Movie rating - a solution



# The learning approach



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# Components of learning

Metaphor: Credit approval

Applicant information:

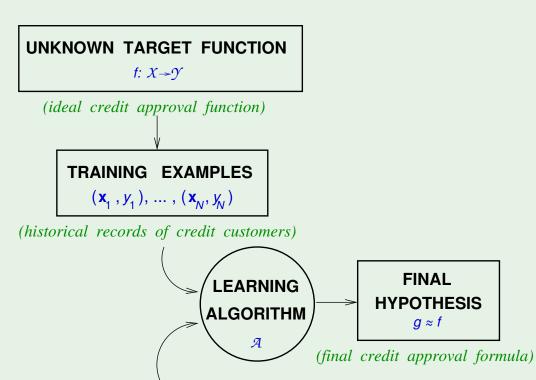
age	23 years
gender	male
annual salary	\$30,000
years in residence	1 year
years in job	1 year
current debt	\$15,000
• • •	

Approve credit?

#### Components of learning

#### Formalization:

- $\bullet$  Input:  $\mathbf{x}$  (customer application)
- Output: y (good/bad customer?)
- ullet Target function:  $f:\mathcal{X} o \mathcal{Y}$  (ideal credit approval formula)
- ullet Data:  $(\mathbf{x}_1,y_1),(\mathbf{x}_2,y_2),\cdots,(\mathbf{x}_N,y_N)$  (historical records)
  - $\downarrow$   $\downarrow$   $\downarrow$
- ullet Hypothesis:  $g:\mathcal{X} \to \mathcal{Y}$  (formula to be used)



(set of candidate formulas)

 $\begin{array}{cc} \text{HYPOTHESIS} & \text{SET} \\ \mathcal{H} \end{array}$ 

#### Solution components

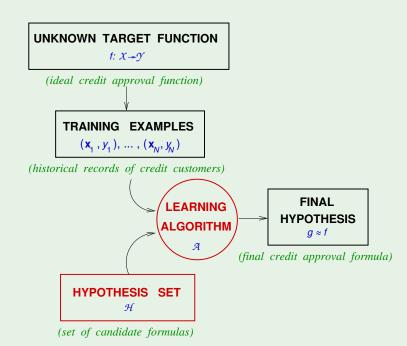
The 2 solution components of the learning problem:

• The Hypothesis Set

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

• 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'

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

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

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

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

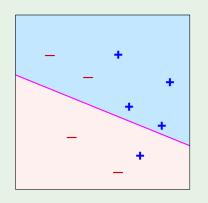
$$h(\mathbf{x}) = \operatorname{sign}\left(\left(\sum_{i=1}^{d} \mathbf{w_i} \ x_i\right) + \mathbf{w_0}\right)$$

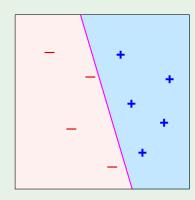
Introduce an artificial coordinate  $x_0=1$ :

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

In vector form, the perceptron implements

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





'linearly separable' data

#### A simple learning algorithm - PLA

The perceptron implements

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

Given the training set:

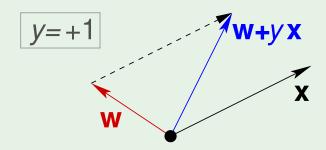
$$(\mathbf{x}_1,y_1),(\mathbf{x}_2,y_2),\cdots,(\mathbf{x}_N,y_N)$$

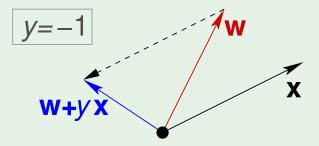
pick a misclassified point:

$$sign(\mathbf{w}^{\mathsf{T}}\mathbf{x}_n) \neq y_n$$

and update the weight vector:

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





#### Iterations of PLA

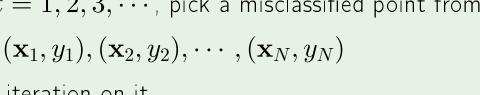
• One iteration of the PLA:

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

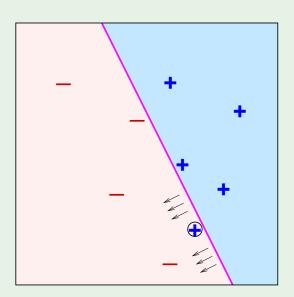
where  $(\mathbf{x}, y)$  is a misclassified training point.

ullet 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!



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#### Basic premise of learning

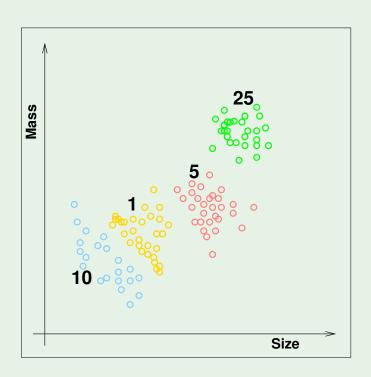
"using a set of observations to uncover an underlying process"

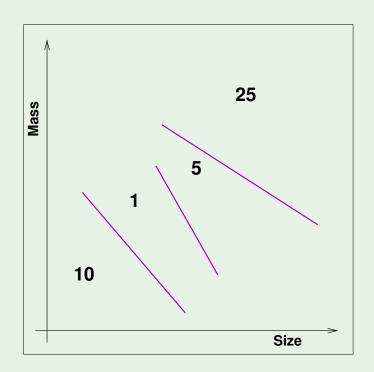
broad premise  $\implies$  many variations

- Supervised Learning
- Unsupervised Learning
- Reinforcement Learning

# Supervised learning

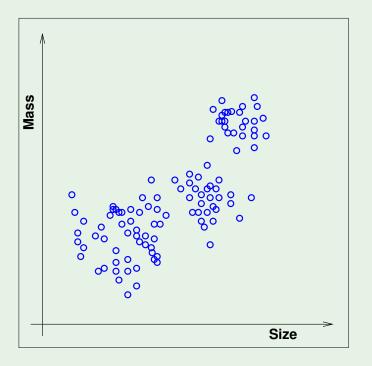
# Example from vending machines - coin recognition





# Unsupervised learning

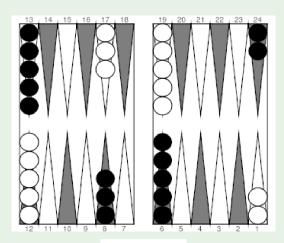
Instead of (input, correct output), we get (input, ? )



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#### Reinforcement learning

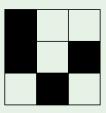
Instead of (input,correct output),
we get (input,some output,grade for this output)

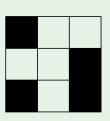


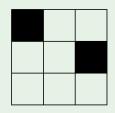
The world champion was a neural network!



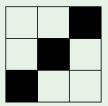
# A Learning puzzle

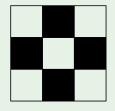


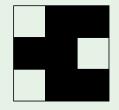




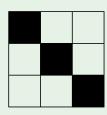
$$f = -1$$







$$f = +1$$



$$f = ?$$

# Thank You!

