



Introduction to machine learning

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This lecture: the learning problem

- ▶ Example of machine learning problem
- ▶ Component of learning
- ▶ A simple model
- ▶ Paradigms in machine learning

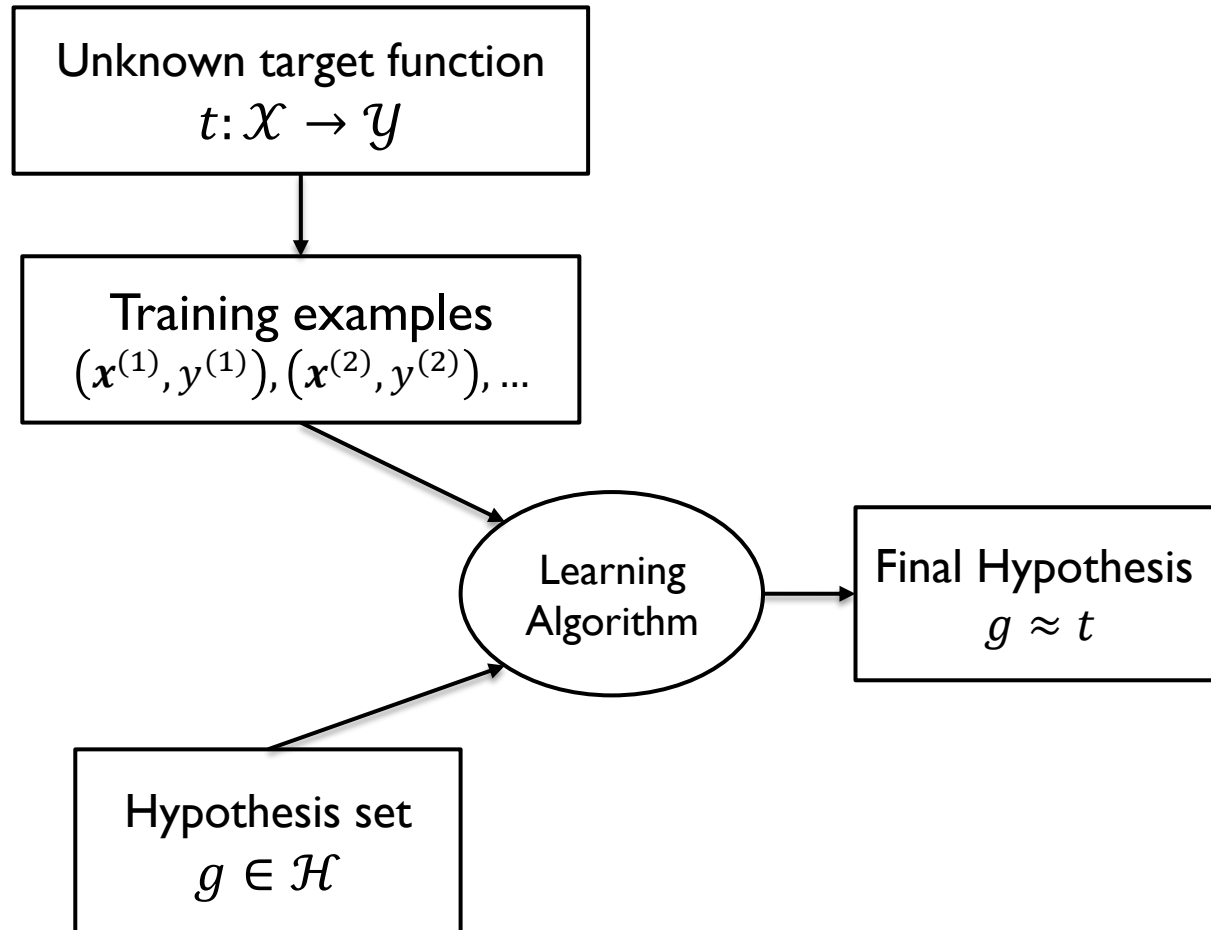
Example

- ▶ Predicting **the risk of heart attack**
 - ▶ Is this a risky person for heart attack? (yes or no)

age	59
gender	Female
diabetes	Yes
weight	90
...	...

- ▶ The essence of machine learning
 - ▶ A pattern exist
 - ▶ We do not know it mathematically
 - ▶ We have data on it

Components of learning



Solution component

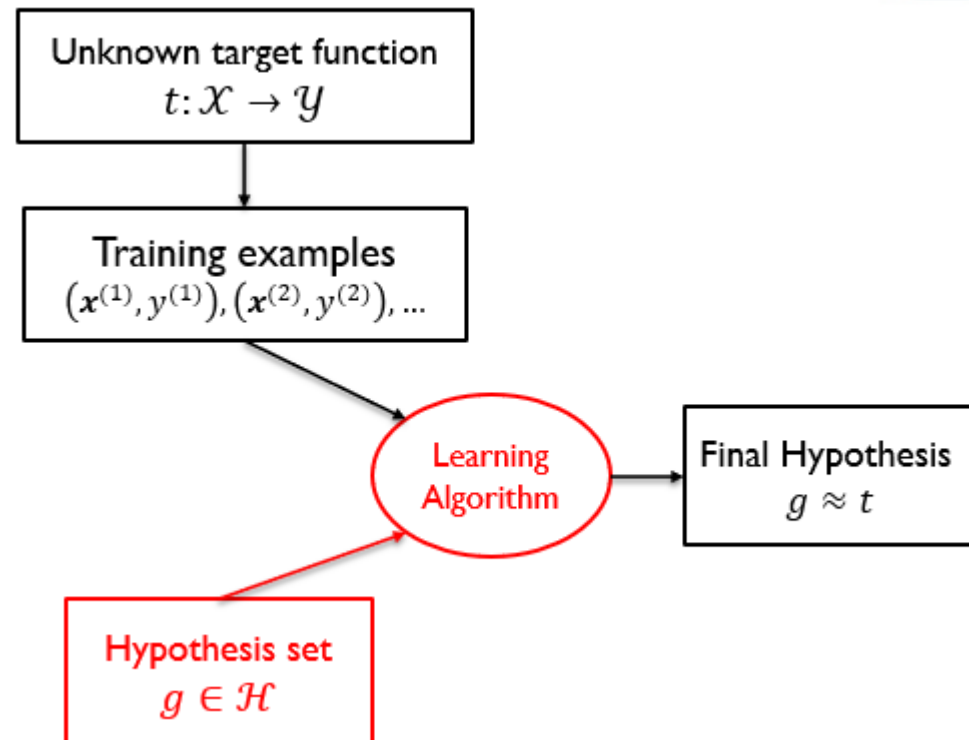
- ▶ The learning model:

- ▶ The hypothesis set

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

- ▶ The learning algorithm

- ▶ Search the hypothesis set to find the best estimate of the target function



A simple hypothesis set

- ▶ Predicting **the risk of heart attack**
 - ▶ Is this a risky person for heart attack? (yes (**+1**) or no (**-1**))
- ▶ For input vector $\mathbf{x} = [x_1, \dots, x_d]$, a person attributes

x_1 :	age	59
x_2 :	gender	Female
x_3 :	diabetes	Yes
x_4 :	weight	90
...		...

- ▶ A simple hypothesis set: The perceptron

A simple hypothesis set

- ▶ A case with a high risk of heart attack

A risky person: if $\sum_{i=1}^d w_i x_i > threshold$

- ▶ Our hypothesis set:

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

A learning algorithm for perceptron

$$h(\mathbf{x}) = \text{sign} \left(\sum_{i=1}^d \mathbf{w}_i x_i - \mathbf{w}_0 \right)$$

- ▶ Considering $x_0 = 1$,

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

- ▶ Given a training set: $(\mathbf{x}^{(1)}, y^{(1)})$, $(\mathbf{x}^{(2)}, y^{(2)})$, ...
 - ▶ **Attributes** of a set of **normal or case of heart attack** persons

A learning algorithm for perceptron

Repeat

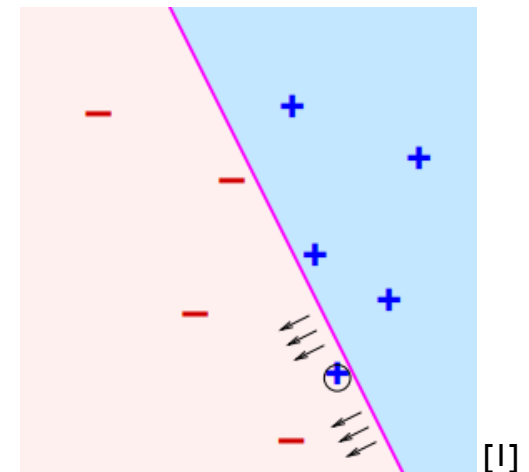
Pick a **misclassified** point $(\mathbf{x}^{(i)}, y^{(i)})$ from training data

$$\text{sign}(\mathbf{w}^T \mathbf{x}^{(i)}) \neq y^{(i)}$$

Update \mathbf{w} :

$$\mathbf{w} = \mathbf{w} + y^{(i)} \mathbf{x}^{(i)}$$

Until all training data points are correctly classified by g



Generalization

- ▶ We don't intend to memorize data but want to distinguish the pattern.
- ▶ A core objective of learning is to generalize from the experience.
 - ▶ Generalization: ability of a learning algorithm to perform accurately on new, unseen examples after having experienced?

Experience in ML

- ▶ Basic premise of learning:
 - ▶ Using a set of observations to uncover an underlying process
- ▶ We have different types of (getting) observations in different types or paradigms of ML methods

A definition of ML

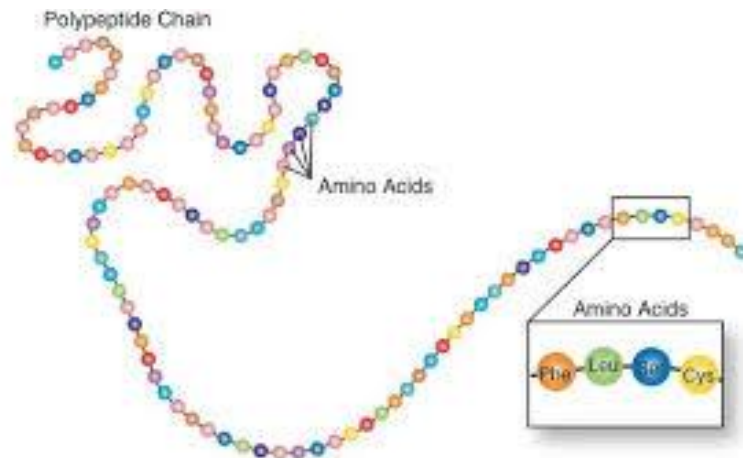
- ▶ Tom Mitchell (1998):
 - ▶ A computer program is said to learn a **task** from **experience** if its **performance** improves with experience
- ▶ Using the observed data to make better decisions
 - ▶ Generalizing from the observed data

Paradigms of machine learning

- ▶ Supervised learning
(input, correct output)
- ▶ Unsupervised learning
(input, ?)
- ▶ Reinforcement learning
(input, some output, grade for this output)
- ▶ Other paradigms: semi-supervised learning, online learning, active learning, etc

Supervised learning

- ▶ Supervised learning
(input, correct output)
- ▶ Our risky heart attack identifier
- ▶ Predicting the function of protein sequences



[<https://medium.com>]

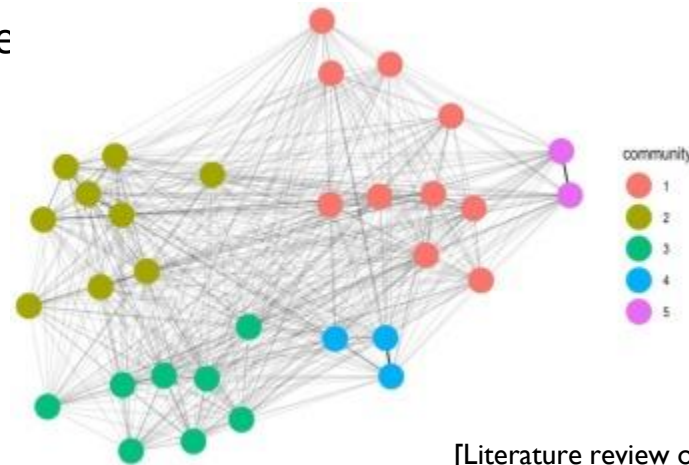
Supervised learning

- ▶ Extract useful information as features
 - ▶ Represent a protein sequence in a vectorized format
 - ▶ Proteins with a length of 1000 amino-acids
 - ▶ Each amino-acid is represented as a one hot vector

x_1	x_2	...	x_{999}	x_{1000}
1	0		0	0
0	1		1	0
...
0	0		0	0
0	0		0	1

Unsupervised learning

- ▶ Revealing structure in the observed data
(input, ?)
- ▶ Clustering: partitioning of data into groups of similar data points.
 - ▶ Customer segmentation in marketing
 - ▶ Community detection in social networks
 - Users are represented with the



[Literature review on data analytics
for social microblogging platforms]

Reinforcement learning

- ▶ Partial (indirect) feedback, no explicit guidance
(input, some output, grade for this output)
 - ▶ AlphaZero
 - ▶ DeepMind chess player
 - ▶ Autonomous driving

Some Learning Application Areas

- ▶ Computer Vision (Photo tagging, face recognition,...)
- ▶ Natural language processing (e.g., machine translation)
- ▶ Robotics
- ▶ Speech recognition
- ▶ Autonomous vehicles
- ▶ Social network analysis
- ▶ Web search engines
- ▶ Medical outcomes analysis
- ▶ Marketing (stock prediction)
- ▶ Computational biology
- ▶ Self-customizing programs (recommender systems)