Introduction to Machine Learning Course Overview

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Course Introduction

Machine Learning

You probably use it dozens of times a day without even knowing it.

Application examples:

- Effective web search.
- Social networks recognize friends from photos or suggest friends.
- Email spam detection.
- Handwriting recognition.
- Understanding the human genome.
- Medical diagnostics.
- Predict possibility for a certain disease on basis of clinical measures.
- Fraud detection.
- Drive vehicles.
- Recommendations (eg, Amazon, Netflix).
- Natural language processing.

The aim of ML is to build computer systems that can adapt to their environments and learn form experience.

Machine Learning

What is Machine Learning?

- ▶ A science of getting computers to learn without being explicitly programmed ¹.
- ➤ Study of algorithms that improve their performance P at some task T with experience E².



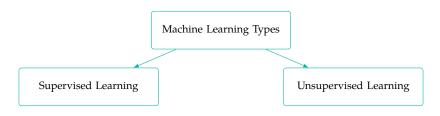
- T: recognition of a handwritten letter "a" from its image.
- E: images of a handwritten "a".
- P: recognition rate.

^{1.} Arthur Samuel.

^{2.} Tom Mitchell.

Types of Machine Learning Problems

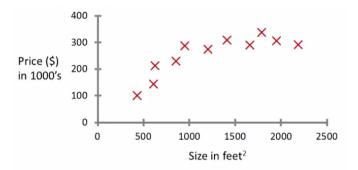
In general, any machine learning problem can be assigned to one of two broad types:



Other: Semi-supervised Learning, Reinforcement learning, Recommender system, etc...

Supervised Learning

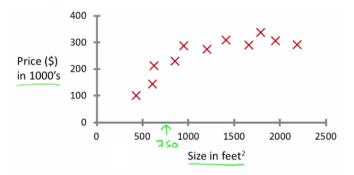
Let's say we want to predict housing prices. We plot a data set and it looks like this:



^{3.} Example from Andrew Ng's MOOC.

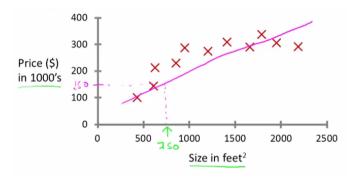
Example: House price prediction

Let's say we own a house that is, say 750 square feet and hoping to sell the house and we want to know how much we can get for the house.



Example: House price prediction

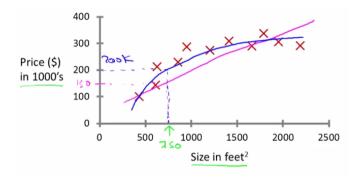
A learning algorithm can for example "fit" a straight line to the data and, based on that, it looks like maybe the house can be sold for maybe about 150 000\$.



Ourse Introduction Supervised Learning Unsupervised Learning Course overview

Example : House price prediction

There might be a better learning algorithm! Maybe a *quadratic function* to this data.



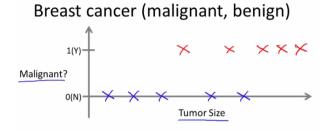
If we do that, and make a prediction here, then it looks like maybe we can sell the house for closer to $200\,000$ \$.

In this example, there is target variable "Price". It is a continuous variable.

Example : Medical diagnosis

Let's say we want to look at medical records and try to predict of a breast cancer as malignant or benign 4 .

A collected data set gave the following:

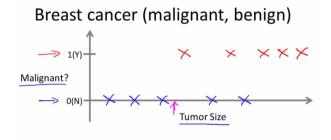


▶ In this example, there is target variable "malignant or benign". It is a discrete variable.

^{4.} A malignant tumor is a tumor that is harmful and dangerous and a benign tumor is a tumor that is harmless.

Example : Medical diagnosis

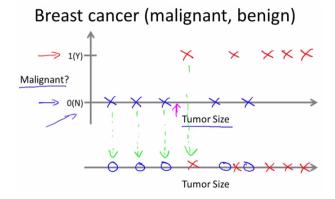
Let's say a person has a breast tumor, and her breast tumor size is known.



► The machine learning question here is, can you estimate what is the **probability** that a tumor is malignant versus benign?

Example : Medical diagnosis

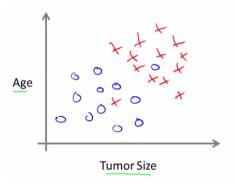
We can take the data set on top and map it down using different symbols.



In this example, there is only one input.

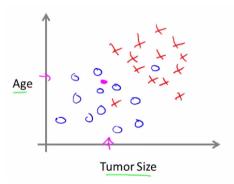
Example: Medical diagnosis

Let's say that we know both the age of the patients and the tumor size. In that case maybe the data set will look like this.



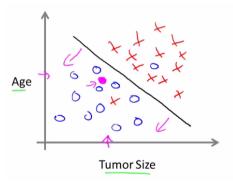
Example: Medical diagnosis

So, let's say a person who tragically has a tumor. And maybe, their tumor size and age falls around there (rose point):



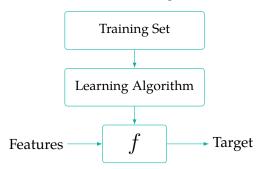
Example: Medical diagnosis

Given a data set like this, a learning algorithm may throw a straight line through the data to try to separate out the malignant tumors from the benign ones.



Supervised Learning: Definition & Model

The term supervised learning refers to the fact that we gave the algorithm a data set in which the "right answers" (known as labels) were given.



- \triangleright Supervised Learning refers to a set of approaches for estimating f.
- *f* is also called *hypothesis* in Machine Learning.

Regression and Classification

Regression

- ▶ The example of the house price prediction is also called a **regression** problem.
- ▶ A regression problem is when we try to predict a **quantitative (continuous)** value output. Namely the price in the example.

Classification

- The process for predicting qualitative (categorical, discrete) responses is known as classification.
- ▶ Methods : Logistic regression, Support Vector Machines, etc..

Supervised Learning: Notations

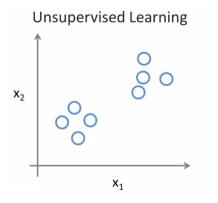
Notations:

- ▶ The size of the house in the first example, tumor size and age in the second example, are the input variables. Typically denoted by *X*.
- ► The inputs go by different names, such as *predictors*, *independent variables*, *features*, *predictor* or sometimes just *variables*.
- ► The house price in the first example and the diagnosis in the second example are the output variables, and are typically denoted using the symbol *Y*.
- ▶ The output variable is often called the *response*, *dependent variable* or *target*.

Unsupervised Learning

Unsupervised Learning : "No labels"

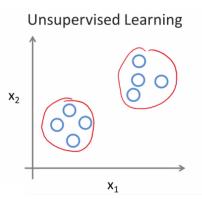
In Unsupervised Learning, we're given data that doesn't have any labels. For example :



Question: Can you find some structure in the data?

Unsupervised Learning: Structure

Given this data set, an Unsupervised Learning algorithm might decide that the data lives in two different clusters.



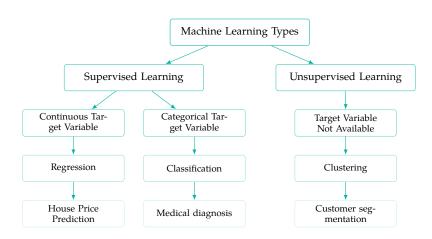
This is called a **clustering** algorithm. Example: Market segmentation.

Unsupervised Learning: Example

One example where clustering is used is in Google News (news.google.com)



Types of Machine Learning Problems



Course overview

OURSE INTRODUCTION SUPERVISED LEARNING UNSUPERVISED LEARNING COURSE OVERVIEW

Aims & Prerequisites

Aims

After the course you should be able to:

- ▶ Recognize which learning method is suitable for a given task.
- ▶ Describe the theory behind the methods.
- ▶ Apply the method to example problems with few data.
- ▶ Undertake an experimental assessment of learning methods and report the results.

Prerequisites

- Math.
- Linear algebra.
- Statistics & Probabilities.
- Programming in R.

Course Information

Format:

▶ 3h per week \times 9 weeks = 27h (3h projects presentations)

Language:

▶ English

Material:

Main course webpage: http://mghassany.com/MLcourse

Additional readings:

- ▶ The Elements of Statistical Learning (by Friedman, Tibshirani and Hastie).
- ▶ Pattern Recognition and Machine Learning (by Bishop).
- Andrew Ng.'s Machine Learning course on Coursera.

Course Information

Evaluation:

- ▶ 15 mins Quiz every week (Quiz i is about week i 1 and week i).
- ▶ Reports at the end of every session.
- ▶ Small project to be presented in the last session (in groups of 3).

Communication:

- ▶ Announcements on Yammer (Group : ESILV Promo 2019 Initiale- A4).
- Announcements via Moodle.
- Email, Desk L521.

Course Contents

Supervised Learning:

- ▶ Week 1 : Simple Linear Regression.
- Week 2 : Multiple Linear Regression.
- ▶ Week 3 : Logistic Regression.
- ▶ Week 4 : Linear Discriminant Analysis.
- Week 5 : Support Vector Machines.

Unsupervised Learning:

- Week 6 : Dimensionality Reduction.
- Week 7 : Clustering k-means.
- Week 8 : Hierarchical Clustering.
- Week 9 : Projects presentations.

About the instructors

MOHAMAD GHASSANY

- Associate Professor at ESILV.
- ▶ PhD in Computer Science from Université Paris 13.
- Master 2 in Applied Mathematics & Statistics from Université Grenoble Alpes.
- Works on Machine Learning research subjects.
- ▶ Teaches Machine Learning, Probability and Statistics.
- Worked as:
 - > Research engineer at Telecom Business School.
 - > Post-doctoral researcher at ENS Cachan.
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ABOUT THE INSTRUCTORS



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- Software product manager for 11 years.
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- Mémorandum Consulting Firm : Partner since 2014.
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Have a nice course!