Multivariate data analysis

R Workshop

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Overview

Machine Learning

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

Application examples:

- effective web search
- social networks recognize friends from photos or suggest friends
- email spam detection
- handwriting recognition
- understanding the human genome
- predict possibility for a certain disease on basis of clinical measures
- fraud detection
- drive vehicles
- recommendations (eg, Amazon, Netflix)

Machine Learning

Automatically learn programs by **generalizing from examples**. As more data becomes available, more ambitious problems can be tackled.

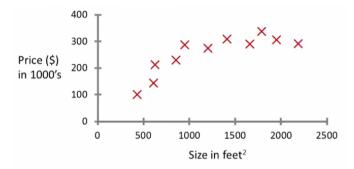
Machine Learning is a branch of artificial intelligence and an interdisciplinary field of CS, statistics, math and engineering.

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

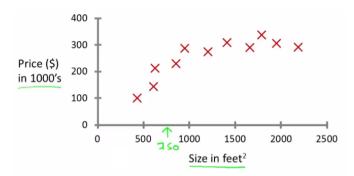
Supervised Learning and Unsupervised Learning

Supervised Learning

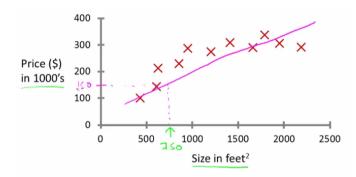
Let's say we want to predict housing prices. We plot a data set and it looks like this



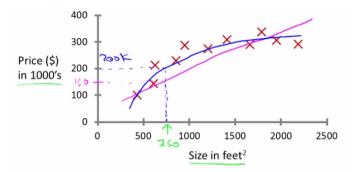
Let's say we own a house that is, say 750 square feet and hoping to sell the house and we want to know how much we can get for the house.



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.



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.

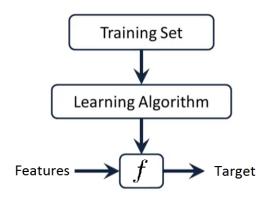
Supervised Learning: Definition

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

Notations:

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

Supervised Learning: Model



- 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:

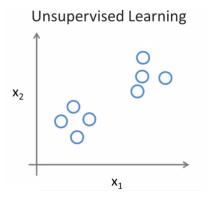
- The process for predicting qualitative (categorical, discrete) responses is known as classification.
- Methods: Logistic regression, Support Vector Machines, etc...

Unsupervised Learning

Unsupervised Learning: "No labels"

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

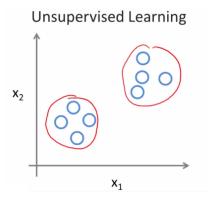
For example:



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.

Unsupervised Learning: Example

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



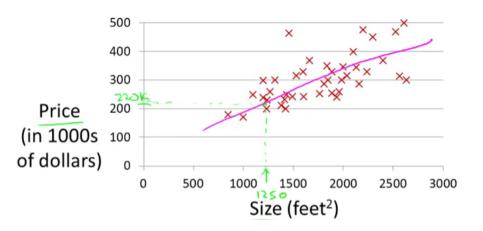
Additional readings

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

Supervised Learning – Predictive Models

Linear Regression



Simple Linear Regression

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- Coefficients:

$$\hat{\beta}_1 = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^{n} (x_i - \bar{x})^2} = \frac{s_{xy}}{s_x^2}$$

and

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

Multiple Linear Regression

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- Coefficients: $\hat{\beta} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{Y}$

Linear Regression - Hypothesis testing

 H_0 : There is no relationship between X and Y

VS

 H_1 : There is some relationship between X and Y

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$$H_0: \beta_i = 0 \quad \forall i$$

VS

$$H_1: \exists i \quad s.t. \quad \beta_i \neq 0$$

Linear Regression – Example

	Coefficient	Std. error	t-statistic	p-value
Constant	2.939	0.3119	9.42	< 0.0001
X_1	0.046	0.0014	32.81	< 0.0001
X_2	0.189	0.0086	21.89	< 0.0001
X_3	-0.001	0.0059	-0.18	0.8599

In this table we have the following model

$$Y = 2.939 + 0.046X_1 + 0.189X_2 - 0.001X_3$$

Note that for each individual predictor a t-statistic and a p-value were reported. These p-values indicate that X_1 and X_2 are related to Y, but that there is no evidence that X_3 is associated with Y, in the presence of these two.

Application on R

Application on the "Boston" data set.