

# Machine Learning (Andrew Ng) Notes #1

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标签（空格分隔）： Coursera ML DL AI Andrew\_Ng Stanford

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## A. Supervised Learning

定义: 已知数据有确定对应标签, 通过学习现有对应关系预测新对应关系。例如: 线性回归

1. Classification: 数据对应discrete 类别, 如鉴别良性肿瘤或恶性肿瘤
2. Regression: 数据对应continuous 数据, 如通过房子大小预测售价

## B. Unsupervised Learning

定义: 已知数据并无确认标签, 需通过分析找出现有数据规律。例如: 将新闻按照内容相似性归类 (clustering)

## C. Class notation:

**m**: number of training examples

**x**'s: 'input' variable / features

**y**'s: 'output' variable / 'target' variable

**(x,y)**: one training example

**(x<sup>i</sup>, y<sup>i</sup>)**: the i<sup>th</sup> training example

## D. Sample cost function

### Hypothesis Function:

$$h_{\theta}(x) = \theta_0 + \theta_1 x$$

### Parameters:

$$\theta_0, \theta_1$$

### Cost Function:

$$J(\theta_0, \theta_1) = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x^i) - y^i)^2$$

### Goal:

Minimize:

$$J(\theta_0, \theta_1)$$

## E. Gradient Descent algorithm

### Definition:

The idea is to use a fixed jumping distance \* slope (  $\frac{\partial}{\partial \theta_j}$  ) to iterate to the local minimum of  $\theta_j$ .

$$\theta_j := \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta_0, \theta_1)$$

$$\alpha = \text{learning\_rate}$$

Notes: Need to simultaneously update both  $\theta_0$  and  $\theta_1$

## F. Applying gradient descent into cost function

$$\theta_0 := \theta_0 - \alpha \frac{1}{m} \sum_{i=1}^m (h_0(x^i) - y^i)$$

$$\theta_1 := \theta_1 - \alpha \frac{1}{m} \sum_{i=1}^m (h_0(x^i) - y^i) * x^i$$

The cost function can also be directly solved with **normal equation method**, however, gradient descent will scale better at large data sets.