

Linear Regression - multivariate linear regression

In regression problems, we are taking input variables and trying to fit the output onto a continuous expected result function.

Variables

- x_j^i = value of feature j in the i^{th} training example
- x^i = the column vector of all the feature inputs of the i^{th} training example
- m = number of training examples
- $n = |x^i|$; the number of features

Hypothesis function

$$h_{\theta}(x) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_3 + \dots + \theta_n x_n$$

- equation of a straight line

Cost function

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

- measuring accuracy of hypothesis
- also called "Square error function"

Gradient Descent

repeat until convergence: {

$$\theta_j := \theta_j - \alpha \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)}) * x_j^i$$

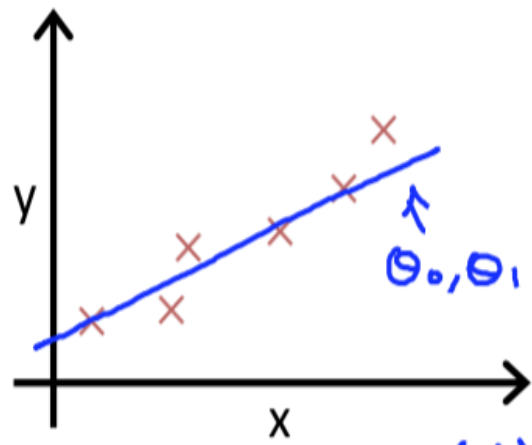
for $j := 0..n$

}

- estimate the parameters in hypothesis function
- start with a guess for our hypothesis and then repeatedly apply these gradient descent equations, the hypothesis will become more and more accurate
- θ_j = a constant that will be changing simultaneously with all other θ_j
- $x^{(i)} y^{(i)}$ = values of the given training set

Example Data & Notes

input x	output y
1	2
2	3
3	4



- must be linear relationship between independent and dependent variables
- can be used with supervised learning
- always separates data with a straight line