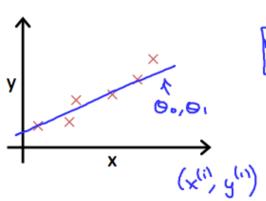
Cost Function

We can measure the accuracy of our hypothesis function by using a **cost function**. This takes an average difference (actually a fancier version of an average) of all the results of the hypothesis with inputs from x's and the actual output y's.

$$J(heta_0, heta_1) = rac{1}{2m} \sum_{i=1}^m (\hat{y}_i - y_i)^2 = rac{1}{2m} \sum_{i=1}^m (h_ heta(x_i) - y_i)^2$$

To break it apart, it is $\frac{1}{2}$ \bar{x} where \bar{x} is the mean of the squares of $h_{\theta}(x_i) - y_i$, or the difference between the predicted value and the actual value.

This function is otherwise called the "Squared error function", or "Mean squared error". The mean is halved $\left(\frac{1}{2}\right)$ as a convenience for the computation of the gradient descent, as the derivative term of the square function will cancel out the $\frac{1}{2}$ term. The following image summarizes what the cost function does:



 $h_{\Theta}(x^{(i)}) = \Theta_{0} + \Theta_{1}x^{(i)}$

J(00,01) = 1 = 2m = (ho(x(1)-y(1)))

Idea: Choose $\underline{\theta_0}, \underline{\theta_1}$ so that $\underline{h_{\theta}(x)}$ is close to \underline{y} for our training examples $(\underline{x}, \underline{y})$

× ,4

queel error faction

Andrew Ng

✓ Complete

Go to next item





