

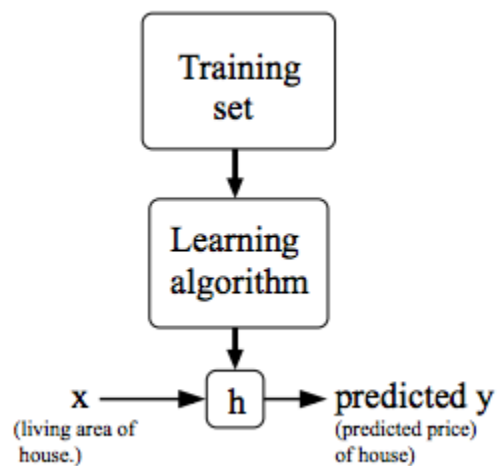
## Linear Regression using Single Variable / Univariate Linear Regression :

In this we will have data set (training set) to build a model that find the relation between our input variable, X and output Variable y, so that our model predicts the value of y for unseen value of X.

**Motivation Example :** To predict housing price based on the size in sq. Ft.

| Size in sq. Ft (X) | Price(\$) in 1000's (y) |
|--------------------|-------------------------|
| 2104               | 460                     |
| 1416               | 232                     |
| 1534               | 315                     |
| 852                | 178                     |
| .....              | .....                   |
| .....              | .....                   |

### Model Representation:



### Notations :

$X \rightarrow$  training data set

$Y \rightarrow$  output

$m \rightarrow$  size of our training data set.

$(x^{(i)}, y^{(i)}) \rightarrow$   $i^{\text{th}}$  training example.

$h \rightarrow$  hypothesis ( $h : X \rightarrow y$ ).

Here hypothesis is the function that show how X and y values are related to each other.

Representation of h:  $h_{\text{theta}}(x) = \text{theta0} + \text{theta1} * x$ .

### Cost Function ( $J$ ) :

In the above equation of  $h$ , the  $\theta_0, \theta_1$  are parameters (learning parameters), now the problem is how to choose these parameters.

**Idea** : Our goal is to have values of the  $\theta_0, \theta_1$  such that our  $h_{\theta_0, \theta_1}(x)$  value should be close to the  $y$  value (we can check that while we are training).

So, the difference  $\rightarrow h_{\theta_0, \theta_1}(x) - y$  should be minimum, to achieve this we can use Error functions like Mean Square Error, Root Mean Square error, ....., etc.

$$\text{Mean Square Error} = J(\theta_0, \theta_1) = \frac{1}{2m} * \sum_{i=1}^m (h_{\theta_0, \theta_1}(x^i) - y^i)^2$$