

When the dependent variable is categorical, then logistic regression is one of the most appropriate regression model. Logistic regression is a binary classifier, but multiple classification is possible using one vs. all model. If a data set  $D = (x, y) : x \in X, y \in Y$  where  $X$  is the independent variables and  $Y$  is the set of category. The hypothesis function  $h_{\theta}(x)$  is given below which is a sigmoid function,

$$h_{\theta}(x) = \frac{1}{1+e^{-\theta^T x}}$$

Using this hypothesis function a curved line is drawn which is used to classify the data. To derive the coefficients a cost function  $J(\theta)$  is used. The value of  $J(\theta)$  is shown in following equation,

$$J(\theta) = -\frac{1}{m} \sum_{i=1}^m [y_i \log h_{\theta}(x_i) + (1-y_i) \log (1 - h_{\theta}(x_i))]$$

Figure 1: Basic Logistic Regression

After minimizing the cost function, the values of the coefficients are derived. The following equation is used to update the values of  $\theta$ ,

$$\theta_j = \theta_j - \frac{1}{m} \sum_{i=1}^m (h_j(x_i) - y_i) x_{ij}$$

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