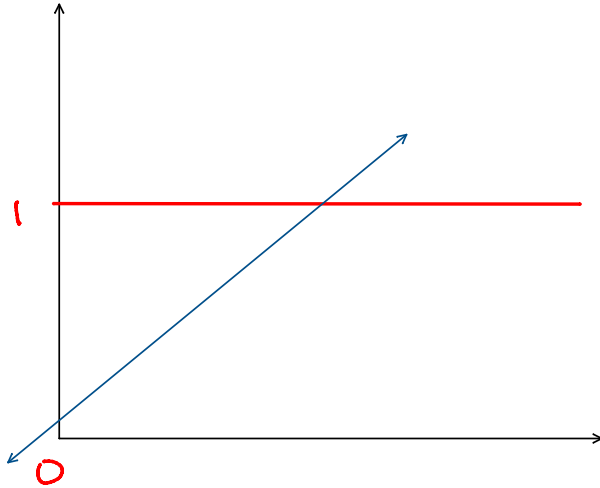


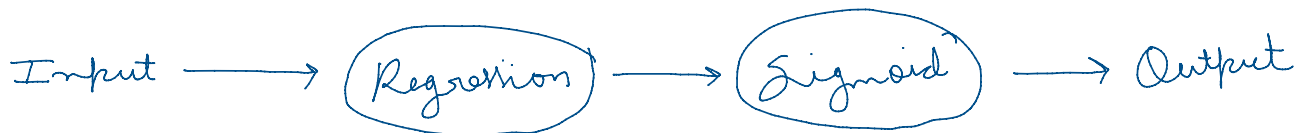
Logistic Regression

27 January 2024 09:16 AM

- Logistic regression is a supervised learning algorithm.
- It is used for binary classification
- It is based on the regression approach



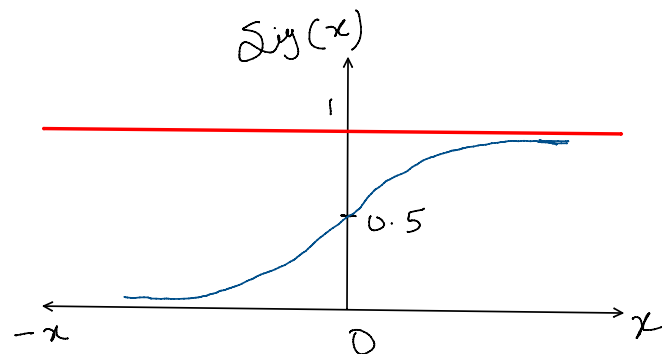
- A Regression approach can be used to predict values 0 to 1.
- But the predictions may go above 1 or below 0.
- In logistic regression the model gives a probability which is converted to labels based on threshold value.
- Probability above 1 or below 0 does not make sense.
- We use sigmoid function to limit the output between 0 to 1.



* SIGMOID FUNCTION! -

$$\text{Sig}(x) = \frac{1}{1 + e^{-x}}$$

$$e = 2.71..$$



CASE 1: x is very large +ve value

$$\Rightarrow e^{-x} \approx 0$$

$$\Rightarrow 1 + e^{-x} \approx 1$$

$$\Rightarrow \text{Sig}(x) \approx 1 \quad (\text{still loss})$$

CASE 2: x is very large -ve value

$$\Rightarrow e^{-x} \approx +\infty \quad (\text{very large value})$$

$$\Rightarrow 1 + e^{-x} \approx +\infty$$

$$\Rightarrow \text{Sig}(x) \approx 0$$

CASE 3: x is 0

$$\Rightarrow e^{-x} = e^0 = 1$$

$$\Rightarrow 1 + e^{-x} = 1 + 1 = 2$$

$$\Rightarrow \text{Sig}(x) = 0.5$$

* LOGISTIC REGRESSION:

$$Z = \beta_0 + \beta_1 x_1 + \beta_2 x_2 \dots \beta_n x_n$$

$$\hat{y} = \text{Sig}(z)$$

$$\hat{y} = \text{Sig}(\beta_0 + \beta_1 x_1 + \beta_2 x_2 \dots \beta_n x_n)$$

* The Cost function used for logistic regression is log loss.

$$\text{Logloss} = -\frac{1}{n} \sum \log(P(Y_i))$$

ACTUAL LABEL	PRED. PROB.	CORRECTED PROB.
1	0.9	0.9
1	0.8	0.8
0	0.1	0.9
0	0.3	0.7
1	0.2	0.2
0	0.7	0.3

$$P_c = \begin{cases} p, & y=1 \\ 1-p, & y=0 \end{cases}$$

$$\begin{aligned} \text{Log Loss} &= -\frac{1}{6} [\log(0.9) + \log(0.8) + \log(0.9) + \\ &\quad \log(0.7) + \log(0.2) + \log(0.3)] \\ &= 0.87 \end{aligned}$$

$$\text{Log Loss} = -\frac{1}{n} \sum [y_i \cdot \log(p) + (1 - y_i) \log(1 - p)]$$