

Linear Regression

- Linear Regression is one of the supervised algorithms that is used to predict the continuous values based on the given inputs.

- It follows the mathematical equation given as:

$$\mathbf{y} = \mathbf{w}_1\mathbf{x}_1 + \mathbf{w}_2\mathbf{x}_2 + \dots + \mathbf{w}_n.\mathbf{x}_n + \mathbf{b}$$

x is the given input features.

y is the value to be predicted.

w describes the importance of each feature.

b represents the bias. (intercept)

n is the number of input features.

- Graphically it represents a straight line.

WHY? Linear Regression

- To determine the best values of weight and biases such that the difference between the actual value and the predicted value remains low.
- To measure how good, the model performs, we use loss function, Mean Square Error(MSE) used for Linear Regression:

$$\text{MSE} = 1/m \left[\sum_{i=1}^m (y_i - y_p)^2 \right]$$

Where , y_i =actual values

y_p =predicted values

- It follows linear relationship between the target and the features.

Limitations

- Only suitable for continuous datas.
- Fails to work for the categorical datas or classification purpose.
- Sensitive to external noise, may effect output .

Logistic Regression

- It is other supervised learning ML algorithm, that overcomes the above mentioned limitations of Linear Regression.
- It is used to calculate the probability between two classes, either 0 or 1.
- It follows the mathematical equation:

$$y = w_1x_1 + w_2x_2 + \dots + w_nx_n + b$$

- It then applies sigmoid function so the equation becomes:
- $\sigma(z) = 1/(1 + e^{-z})$
- $P(y=1) = 1/(1 + e^{-(w_1x_1 + w_2x_2 + \dots + b)})$

If $y > 0.5$ the predicted class is **1**.

If $y < 0.5$ the predicted class is **0**.

Sigmoid function is also known as activation function for

binary class where as **Softmax**

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- **Sigmoid function** is also known as **activation function**. It is used for binary class whereas for multiple classes, the activation function used is **Softmax Function**.
- It first predicts the log-odds before predicting the probability.
- Log odds refers to the ratio of an event occurring to that of an event not occurring.
- Mathematically:

$\log\text{-odds}=\log(p/1-p)$

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