

## Logistic Regression

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### Logistic Regression

- \* It comes under Supervised Learning [both x and y]
- \* It is used to solve classification problems. [target is categorical or discrete]
- Eg: Diabetic data  
Cancerous data  
Email classification, Covid-19, Heart disease.
- \* It is used to predict the probability of Categorical dependent variables.

### Assumptions of Logistic Regression:

- \* The target variable is binary i.e Yes/No, True/False, 0/1,
- \* No or little multicollinearity  $\rightarrow$  less correlation among input variables.
- \* Outliers should be handled.
- \* Sample size should be sufficiently large.
- \* Data is linearly separable.

Linearly separable  $\rightarrow$  There should be clear separation among the classes



Non-linear  $\rightarrow$  Overlapping of classes.



### How Logistic Regression works?

Problem Statement:

I want to predict whether student will pass or not in exam.

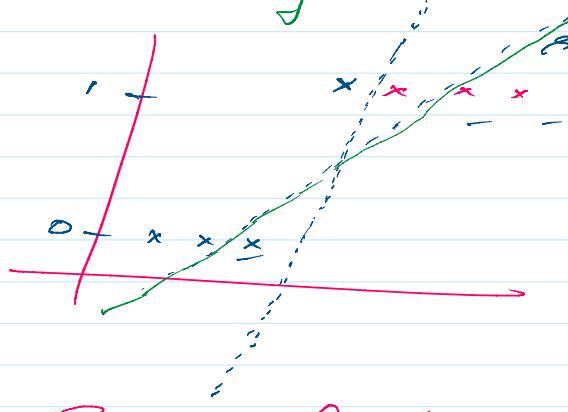
Data

<u>No of hours studied (x)</u>	<u>Result (y)</u>
1	Fail
4	Fail
3	Fail
8	Pass
10	Pass
12	Pass
5	?

What happens if we try to use Linear Regression?

\* Linear regression will fit a straight line which separates two classes.

\* If we use linear regression to make predictions it will use straight line and gives prediction other than 0 and 1.



\* If data has outliers then line gets shifted, it results in wrong classification

## Sigmoid Function

Sigmoid function is mathematical function that we use to transform continuous values into probability values which lies within a range 0 to 1.

$$h(x) = g(x) = \frac{1}{1 + e^{-x}} \rightarrow [0, 1]$$

where  $x = \alpha x + c$   
 $\alpha$  is slope  
 $c$  is intercept

$x = \beta_0 + \beta_1 x$  Just like

<u>x</u>	<u>Result</u>	<u><math>\alpha x + c</math></u>	<u><math>h(x)</math></u>	<u>Predictions</u>
2	Fail(0)	-50	0.2	Fail
3	Fail(0)	-40	0.4	Fail
5	Fail(0)	-20	0.3	Fail
8	Pass(1)	70	0.6	Pass
10	Pass(1)	80	0.8	Pass
12	Pass(1)	120	0.9	Pass

## Set a Threshold

$$\begin{aligned} h(x) \geq 0.5 &\rightarrow y = 1 / \text{Pass} \\ h(x) < 0.5 &\rightarrow y = 0 / \text{Fail} \end{aligned}$$

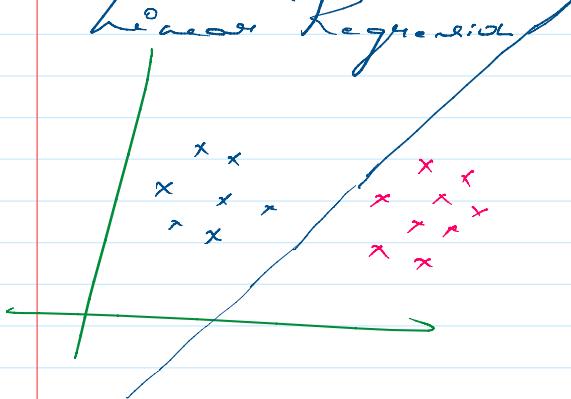
Break

$$0.5 \rightarrow \underline{\text{Lo Pm}}$$

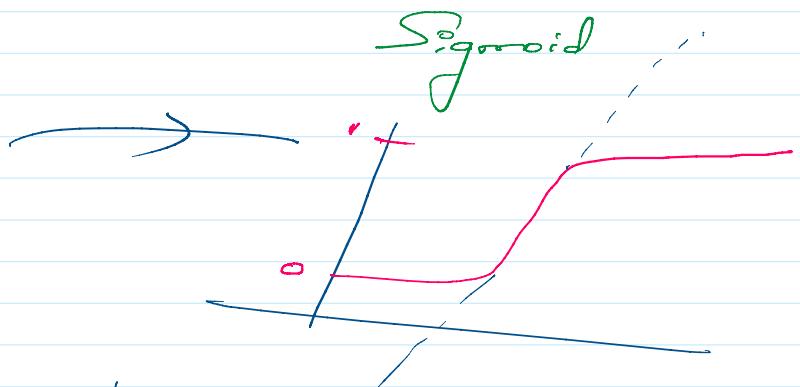
## How it works?

- It will fit line
- Make predictions
- Transform continuous values using sigmoid function
- Set threshold
- If threshold is greater than 0.5 classify as pass  
and if threshold is less than 0.5 classify as Fail

## Linear Regression



## Sigmoid



## How classification works?

$$y = w_0 + w_1 x_1 + w_2 x_2 + w_3 x_3 + C$$

$$\rightarrow y = \underline{w^T x + C}$$

where  $w^T = \begin{bmatrix} w_0 \\ w_1 \\ w_2 \\ w_3 \end{bmatrix}$

$$x = \begin{bmatrix} x_0 & x_1 & x_2 & x_3 \end{bmatrix}$$

Let's say dimension through origin,  $C=0$ .

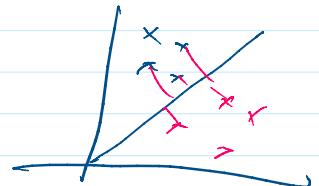
$$y = \underline{w^T x}$$

## How to find distance b/w point and plane?

$$d = \frac{|w^T x|}{\|w\|}$$

$$\boxed{d = \underline{w^T x}} \quad y = -$$

$$y \times d = \underline{y \times w^T x} \geq 0, \text{ then } \Rightarrow \text{the model has}$$



$y \times d = y \times w^T x \geq 0$ , then we say the model has done correct classification.

Pass  $\Rightarrow d > 0$



- \* Any point above the line/plane will have positive distance
- \* Any point below the line/plane will have negative distance.
- \* All positive classes are denoted by  $1$
- \* All negative classes are denoted by  $-1$

Case ①

$x_1$  belongs to positive class.

$$y=1, w^T x > 0$$

$$\begin{aligned} y \times w^T x &\geq (+ve)(+ve) = +ve \\ y \times w^T x &> 0 \\ x_1 \text{ is } &\text{Correctly classified} \end{aligned}$$

Case ②

$x_2$  belongs to negative class

$$y=-1, w^T x > 0$$

$$y \times w^T x = (-ve)(+ve) = -ve > 0$$

$x_2$  is  $\text{misclassified}$

Case ③

$x_3$  belongs to negative class

$$y=-1, w^T x < 0$$

$$y \times w^T x = (-ve)(-ve) = +ve$$

$$y \times w^T x < 0$$

$x_3$  is  $\text{misclassified}$

Case ④

$x_4$  belongs to positive class

$$y=1, w^T x < 0 ..$$

$$y \times w^T x = (+ve)(-ve) = -ve$$

$$y \times w^T x < 0 ..$$

$x_4$  is  $\text{misclassified}$