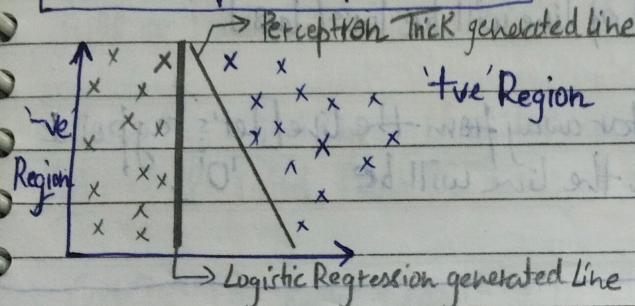


Logistic Expression

Firstly, I will talk about that, what is the problem coming in the Perception Trick. So that, we moved to Logistic Regression.

Results as per Perception Trick



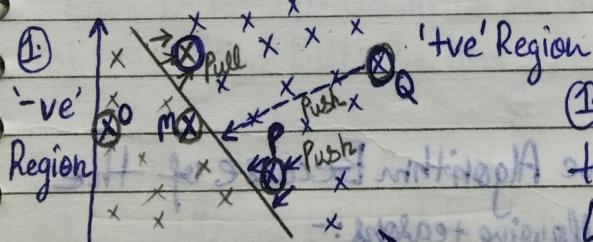
Scale:-

 $x \Rightarrow$ Selected Candidates $x \Rightarrow$ Not Selected Candidates

We can analyse that the Perception Trick had generated the line which is more closer to the points that are selected Candidates. But, when the new data will come it will move more closely to the Positive region.

So, Applying Logistic Regression :-

Algorithm of Logistic Regression



Scale:-

 $x \Rightarrow$ Selected Candidates $x \Rightarrow$ Not Selected Candidates

(1) How or What changes we have to make to make Perception line to Logistic Regression line. $x: (P - Y) / (1 + b_0 w) = w w$

Step 1:- Misclassified points \Rightarrow Line \Rightarrow Pull

Right Classified points \Rightarrow Line \Rightarrow Push

Step 2:- The Magnitude also matters in Push and Pull of point's line.

Case(a) [Classified / Right Classified] Points.

The right classified point is too much closer to the line. [let's say]
 \therefore So, the Magnitude will be height of pushing the line. Point P

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Case (b) The right classified point is far away from the line [let's say point 'P'].
∴ So in that case, Magnitude of pushing the line will be low.

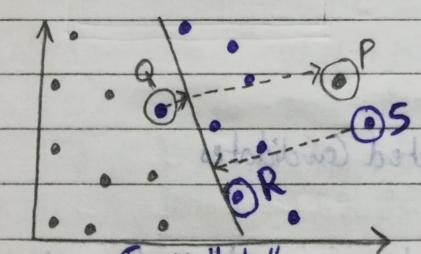
Wrong Classified Points

Case (a) When the misclassified point is more closer to the line. [let's say point 'Q'].
∴ So in that case the Magnitude of pull of line will be low [less].

Case (b) When the misclassified point is far away from the line [let's say point 'R'].
∴ So, in that case the Magnitude of pull the line will be '0'.

Sigmoid Function

Sigmoid Algorithm



Scale:-
• → Selected
• → Not Selected

In Sigmoid Function we are changing the Algorithm Because of the
 $W_{old} + \gamma (Y_i - \hat{Y}_i) \cdot X_i$ following reasons :-

- (i) In Perception Trick we only talk about the Misclassified points like point 'P' and 'Q', and move the line accordingly so that they can properly classified.
- (ii) But, in this Algorithm include both Classified and Misclassified points and the concept is used as: Right Classified Points \rightarrow Push the Line.
Mis Classified Points \rightarrow Pull the Line.

[P, Q]

Spiral

Q What strategy is used in Sigmoid Function?

Ans According to the Graph "abc", the strategy of Sigmoid Function will be:-

(i) In this Algorithm we include both the points Classified [Right] [R,S] points and Misclassified Points [P,Q].

(ii) The Classified Points [R,S] \Rightarrow Push the line.

(iii) The Misclassified Points [P,Q] \Rightarrow Pull the line.

(iv) One more important things, "Magnitude", in which Magnitude they Pull or Push the line. This includes 4 cases:-

Case 1:- For Right Classified which is nearer to the line. i.e, Point 'R'.

Sol: \Rightarrow 'R' Point Push the Line at a higher Magnitude.

Case 2:- For Right Classified which is far away to the line. i.e, Point 'S'.

Sol: \Rightarrow 'S' Point Push the line at a lower Magnitude.

Case 3:- For Mis Classified which is nearer to the line, i.e, Point 'Q'.

Sol: \Rightarrow 'Q' Point Pulls the line at a lower Magnitude.

Case 4:- For Mis Classified which is far away to the line. i.e, Point 'P'.

Sol: \Rightarrow 'P' Point Pulls the line at a higher Magnitude.

Q Working to the Algorithm:-

$$\text{sol: } W_{\text{new}} = W_{\text{old}} + \eta (Y_i - \hat{Y}_i) X_i$$

According to Case 1 and Case 2:-

$$\text{sol: } W_{\text{new}} = W_{\text{old}} + \eta (Y_i - \hat{Y}_i) X_i$$

$$\text{sol: } W_{\text{new}} = W_{\text{old}} + \eta (0 - \hat{Y}_i) X_i$$

\Rightarrow No Change in the line.

According to Case 3:-

$$\boxed{\text{sol: } W_{\text{new}} = W_{\text{old}} + \eta X_i}$$

$$\boxed{\text{Becoz } (Y_i - \hat{Y}_i) \geq 1}$$

According to Case 4:-

$$\boxed{\text{sol: } W_{\text{new}} = W_{\text{old}} - \eta X_i}$$

$$\boxed{\text{Becoz } (Y_i - \hat{Y}_i) = -1}$$

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So we can see that in Case 1 and Case 2, there is no change in the line. But we want to shift the line even the points are tightly classified.

Q Why this line is not shifting [in Case 1 and Case 2]?

Sol: Because, $(Y - \hat{Y}) = 0$ it becomes zero, $\Rightarrow [W_{\text{new}} = W_{\text{old}}]$

To stop this, firstly stop " $Y - \hat{Y} \neq 0$ " i.e., we want that $Y - \hat{Y} \neq 0$ should not equal to zero.

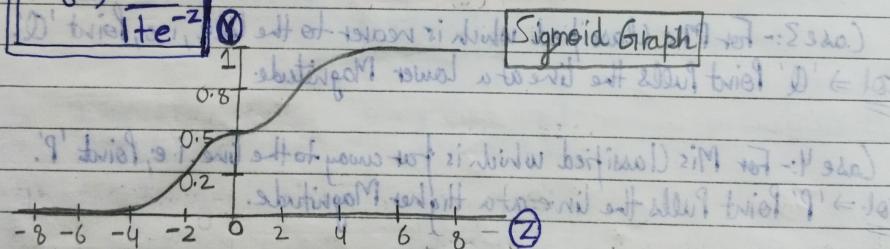
We can make changes only in the Predicted Value [\hat{Y}]. Assume how to do it?

Sol: Firstly, we calculate the $\hat{Y} \rightarrow \sum W_i X_i = [0, 1] \rightarrow \text{Step Function}$

To overcome this, we use the Sigmoid Function.

Q What is Sigmoid Function?

A Ans: $\sigma(z) = \frac{1}{1 + e^{-z}}$



According to the Graph :-

Y lies between $\{-\infty, \infty\}$

i.e., $Z \in \{-\infty, \infty\}$

Y lies between $[0, 1]$

and when $Z = 0$

$$Y = 0.5$$

It means at a Positive Value, z 's Value becomes the positive value.

So, this function solves our problem that $(Y - \hat{Y}) \neq 0$ because it provides the Smallest Value also which is between $0 \rightarrow 1$.

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$$Y_i = \sum w_i x_i$$

So, in Perception Trick, we generally calculate the Y_i as:- $w_1 x_1 + w_2 x_2 + \dots + w_n x_n$

$$w_0 = \boxed{z}$$

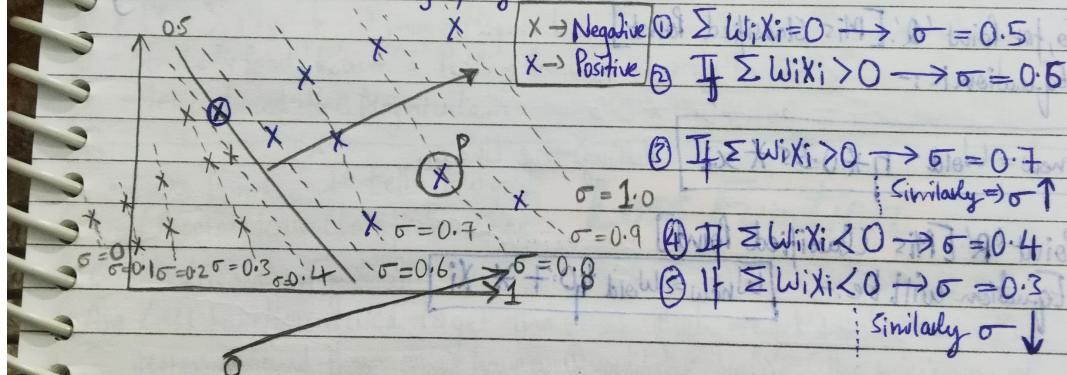
If this Value i.e. $\boxed{z} > 0 \rightarrow$ Selected
and If this Value $\boxed{z} < 0 \rightarrow$ Not Selected.

Now, We use the Sigmoid Function to calculate the Y_i .

$$\hat{Y}_i = \sum w_i x_i = \boxed{z}$$

- This Value which comes, i.e. ' z ' is passed to Sigmoid Function.
- If the Value after the Sigmoid Function is Positive. $\Rightarrow P > 0.5 \rightarrow 1$
If the Sigmoid Function is Negative. $\Rightarrow P < 0.5 \rightarrow 0$

① Additional Functionality of Sigmoid Function :-



$\sigma \rightarrow P$ Probability [Positive $\rightarrow P$, Negative $\rightarrow N$]

We can say that the $P(P) = 0.5$ or above.

Sigmoid Function is beneficial because of following:-

- We can say that Jetha Hamn like Sei Dutt Jayenge $\Rightarrow P(P)$ will increase
- We can also say that, Jetha Hamn like Sei Dutt Jayenge [Negative Side] $\Rightarrow P(N)$ will decrease.

- i.e., With the help of Sigmoid Function, we can calculate the Probability of both Selected and Not Selected. let's say 'P' point Selected Probability $\rightarrow 0.8$

$$P(\text{Not Selected}) = 1 - 0.8 = 0.2$$

① Impact of Sigmoid Function

$$W_{new} = W_{old} + \eta (Y - \hat{Y}) X_i$$

Task-1 :- Not Making the $(Y - \hat{Y}) = 0$

$$Y_i = \sum w_i x_i \Rightarrow [z]$$

Now we can see that $(Y_i - \hat{Y}_i)$ is never '0'.

So for Point 'P'. [Right Classified Point]

Now, Equation will be :-

$$W_{new} = W_{old} + \eta * 0.2 * X_i$$

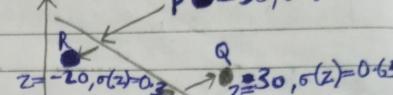
Pushing the Line...

Date
Scale:-

- \rightarrow Not Selected
- \bullet \rightarrow Selected

$$z = \sum w_i x_i$$

$$P \bullet z = 50, \sigma(z) = 0.8$$



Y_i	\hat{Y}_i	$Y_i - \hat{Y}_i$
1	0.8	0.2
0	0.65	-0.65
1	0.3	0.7
0	0.15	-0.15

Note:- $\sigma(z)$ these all values are Assuring.

So, for Point 'Q'. [Mis Classified Point]

Equation is :- $0 < \hat{Y}_i < 1$

$$W_{new} = W_{old} - \eta * 0.65 * X_i$$

Pulling the Line....

Point 'R' [Mis Classified Point]

$$\text{Equation will be :- } W_{new} = W_{old} + \eta * 0.7 * X_i$$

Pulling the Line....

Point 'S' [Right Classified Point]

$$\text{Equation will be :- } W_{new} = W_{old} - \eta * 0.15 * X_i$$

Pushing the Line....

Task-2 :- Finding How is the Magnitude Working on the Points.

For this, let's take an example:-

$$P: z = 10, \sigma(z) \Rightarrow 0.6$$

$$Q: z = 20, \sigma(z) \Rightarrow 0.8$$

- \rightarrow Selected
- \rightarrow Not Selected

$$R: z = 5, \sigma(z) \Rightarrow 0.2$$

$$S: z = 15, \sigma(z) \Rightarrow 0.5$$

Let's say I take, the point 'P'. Assuming that $z [\sum w_i x_i] = 10$. So, its value after Sigmoid function will be :- 0.6.

$$\text{Equation} \Rightarrow w_h = w_{old} + \eta * 0.4 * x_i = x_1 - ① \quad \text{Pushing the line....}$$

At Point $\Rightarrow z = 20, \sigma(z) = 0.8$

$$\text{Equation} \Rightarrow w_h = w_{old} + \eta * 0.2 * x_i = x_2 - ②$$

Pushing the line....

From ① and ②

Let's say Magnitude = "M". it means Magnitude of x_1 is more than Magnitude of x_2).

$M_{x_1} > M_{x_2}$, it means Magnitude of x_1 is more than Magnitude of x_2).

Similarly Point 'R' and 'S' Pulling the line where $M_R < M_S$.

Hence proved the Magnitude...

$$L_0 * P_0 * P_0 * F_0 \leftarrow L_{\text{leben}}$$