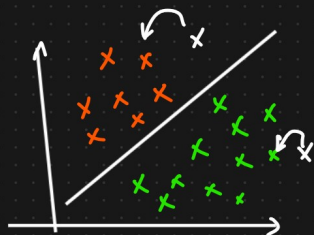


# Performance Metrics, Accuracy, Precision, Recall And F-Beta

## Topics to be covered

- ① Confusion matrix
- ② Accuracy
- ③ Precision
- ④ Recall
- ⑤ F-Beta Score



Logistic Regression

In case of Regression

R squared

Adjusted R squared

### ① Confusion Matrix

		1	0	→ Actual Values
1	→	3	2	
0	↓	1	1	←
		↓ Predicted Values		

Dataset		o/p	
$x_1$	$x_2$	$y$	$\hat{y}$ ← pred by model
-	-	0	1
-	-	1	1
-	-	0	0
-	-	1	1
-	-	1	1
-	-	0	1
-	-	1	0

A confusion matrix (2x2) is a performance measurement tool used for classification models

TP :- True Positive	FP :- False Positive
FN :- False Negative	TN :- True Negative

		1	0	→ Actual
1	→	TP	FP	
0	↓	FN	TN	←
		↓ Predicted		

$$\begin{aligned}
 \text{Accuracy} &= \frac{TP + TN}{TP + FP + FN + TN} \\
 &= \frac{3 + 1}{3 + 2 + 1 + 1} \\
 &= \frac{4}{7}
 \end{aligned}$$

E.g.

④ Dataset Binary classification  
 ↳ 1000 datapoints  $\left\{ \begin{array}{l} 900 \rightarrow 1 \\ 100 \rightarrow 0 \end{array} \right\}$  Imbalanced Dataset

↓  
90% Accuracy

But in the real it is not true. In the imbalanced data set, it can give you a completely biased result.

④ Precision =  $\frac{TP}{TP+FP}$  } Out of all the actual value  
 how many are correctly predicted

	1	0	Actual
1	TP	FP	
0	FN	TN	
Predicted			

- Focuses on reducing False Positives (FP)
- Useful when False Positives (FP) are costly (e.g., fraud detection).

② Recall =  $\frac{TP}{TP+FN}$  } Out of all the predicted value  
 how many are correctly predicted

- Focuses on reducing False Negative (FN)
- Important when False Negatives (FN) are costly (e.g., medical diagnosis).

Usecase ①

Spam classification

Mail → Spam } Good  
 Model → Spam }

	Spam	Not Spam	
	1	0	Actual
Spam	TP	FP ✓	
Not Spam	FN	TN	

Mail → Not Spam } Blunder  
 Model → Spam }

↓ ↓

$$\text{Precision} = \frac{TP}{TP+FP}$$

Usecase 2

To predict whether person has diabetes or not

✓ Truth → diabetes }  
 ✓ Model → Doesn't diabetes } Blunder

	Diab	No Diab	
	1	0	
Diab	TP	FP	
No Diab	FN	TN	

Recall =  $\frac{TP}{TP+FN}$

Truth → diabetes } good  
 Model → " }

↓ ↓ ↓  
 usecase of diseases

Truth  $\rightarrow$  Not diabetes }  $\Rightarrow$  2nd opinion Not blunder  
 Model  $\rightarrow$  Diabetes }  $\Downarrow$   
 check

Assignment

④ Tomorrow the stock market will crash or not

Reducing FPR or FN

Ans :- Reduce FN, bcoz if predicting Not-crash, actually crash it will be blunder.

④  $\boxed{F\text{-Beta Score}} = (1 + \beta^2) * \frac{\text{Precision} * \text{Recall}}{\text{Precision} + \text{Recall}}$

① If FP & FN are both important

$\beta = 1$

$F1 \text{ Score} = 2 * \frac{\text{Precision} * \text{Recall}}{\text{Precision} + \text{Recall}}$

$\Rightarrow$  Harmonic Mean

② If FP is more important FN

$\beta = 0.5$

$F_{0.5} \text{ Score} = (1 + 0.25) \frac{P * R}{P + R}$

③ If  $FN \gg FP$

$\beta = 2$

$\boxed{F_2 \text{ Score} = (1 + 4) \frac{P * R}{P + R}}$