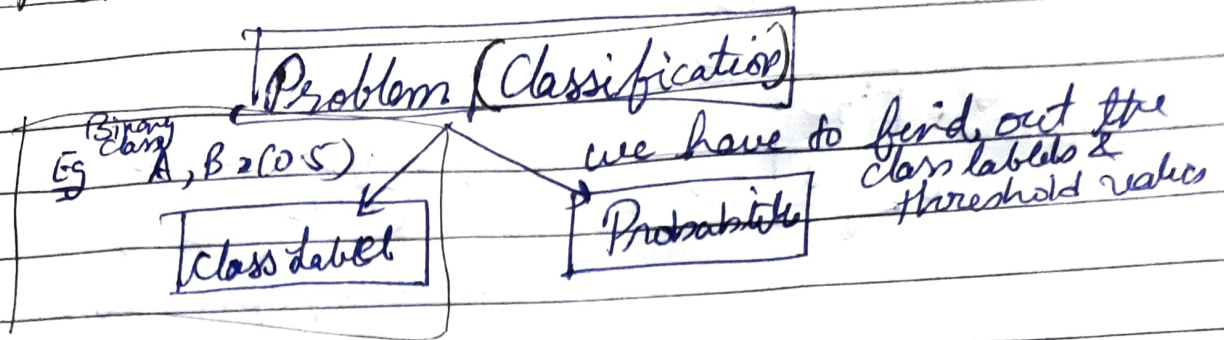


# Metrics In Classification



Eg A person has cancer or not, threshold may or may not be 0.5.

1000 records

$\left\{ \begin{array}{l} 500 Y, 500 N \\ 600 Y, 400 N \\ 400 Y, 600 N \end{array} \right\}$ 
 Balanced Dataset

Use: Accuracy

Imbalanced

$\left\{ \begin{array}{l} 950 Y, 50 N \\ 900 Y, 50 N \end{array} \right\}$

Use: Precision, Recall  
F-beta

Don't use: Accuracy

If these data points are given to ML algo, then ML algo won't get biased on max no of output

Actual Values

		1	0
Predicted Values	1	TP	FP
	0	FN	TN

Type 1 Error  $\Rightarrow$  FPR

$$FPR = \frac{FP}{FP + TN}$$

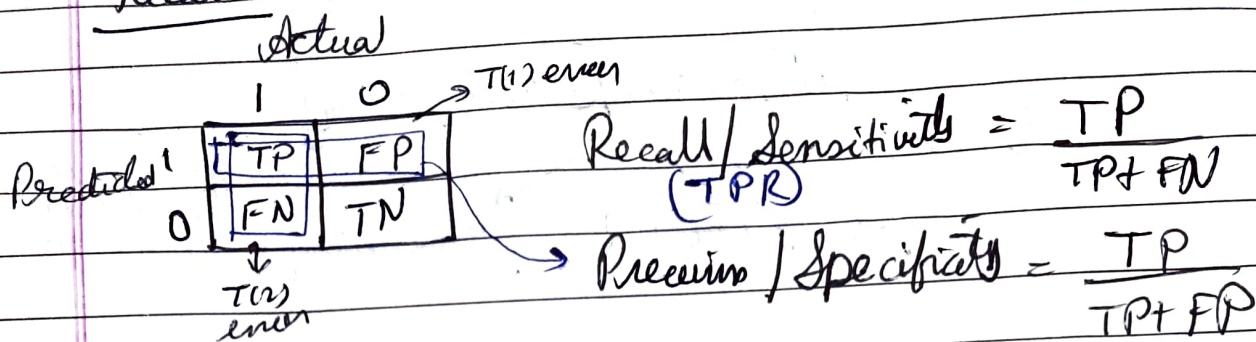
Type 2 Error  $\Rightarrow$  FNR

$$FNR = \frac{FN}{FP + TN}$$

$$Accuracy = \frac{TP + TN}{TP + FP + FN + TN}$$

Any classification problem, try reducing Type 1, Type 2 error.

## Recall



**Recall:** Out of total actual +ve values (1) / True, how many (ve) did we predict correctly

**Precision:** Out of total predicted +ve results, how many results were actually true.

When to use what?

① Spam Detection: Precision (FP ↓)

If a mail is not a spam but the model predicted that it is a spam, try to reduce false positive value because if a mail is not a spam & it is predicted as spam, customer is gonna miss that mail, which may be very imp to him/her.  
(FP ↓)

② Cancer Detection: Recall (FN ↓)

If a person is having a cancer but he is predicted as not having a cancer, it will be disaster.



If FP is much more impact  $\Rightarrow$  Precision  
If FN is much more impact  $\Rightarrow$  Recall

If FP, FN both are impact  $\Rightarrow$  F-B score.

F-B score

$$F_B = (1 + B^2) \frac{P \times R}{B^2 (P + R)}$$

$B = 1$  F-1 score

$B = 0.5$  F 0.5 score

$B = 2$  F 2 score.

$B = 1$   $F_1 = 2 \frac{P \cdot R}{P + R}$  Harmonic mean  $= \left( \frac{2xy}{x+y} \right)$

If FP is having more impact than FN  $\Rightarrow$  (T-2) need reduce your  $\beta$  value ( $< 1$ )

If FN is having ~~more~~ <sup>more</sup> impact  $\Rightarrow$  (T-1) need increase your  $\beta$  value ( $> 1$ )

# ROC and AUC <sup>generally used</sup> (Binary Classification)

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

$y \rightarrow \hat{y}$

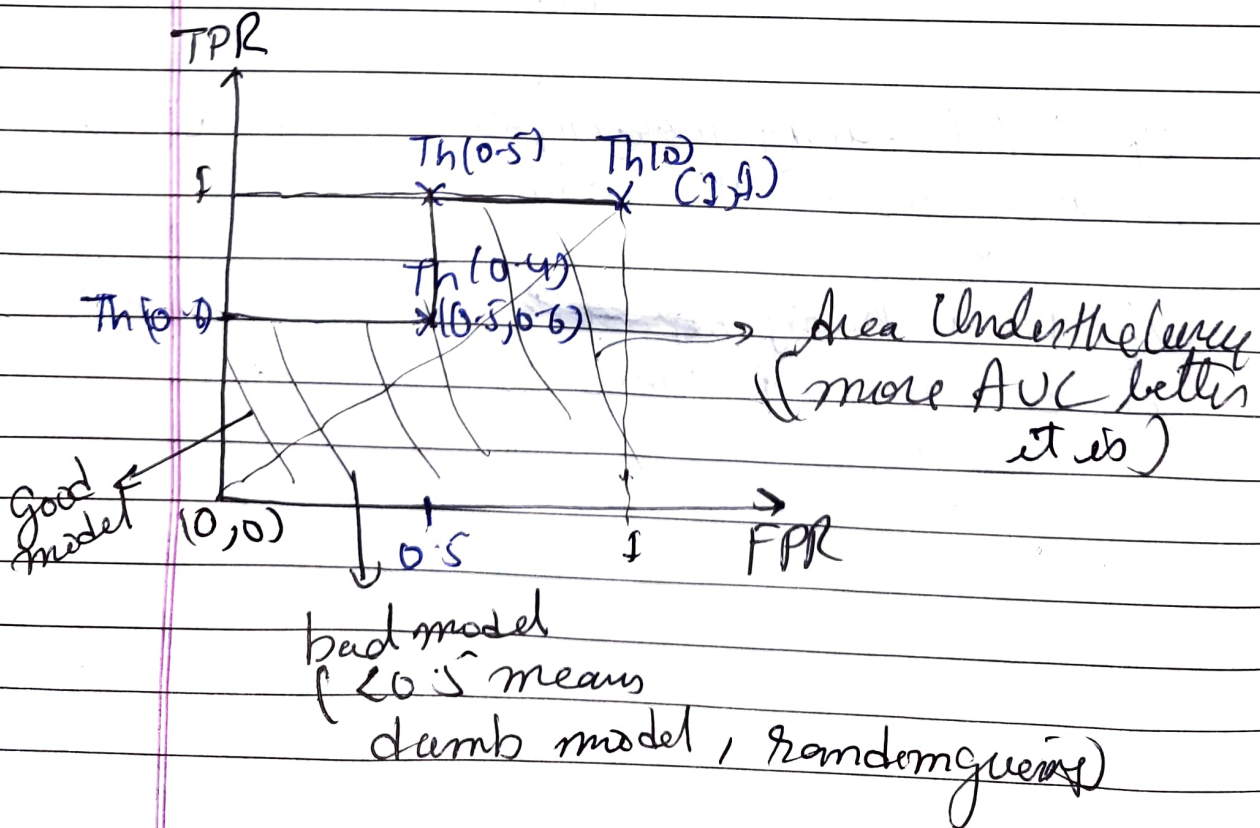
$$FPR = \frac{FP}{FP + TN}$$

Threshold: [0, 0.2, 0.4, 0.6, 0.8, 1]

$y$	$\hat{y}$	$\hat{y}(0)$	$\hat{y}(0.2)$
1	0.8	1	1
0	0.9	1	1
1	0.4	1	1
1	0.3	1	1
0	0.2	1	0
1	0.7	1	1

$$TPR = \frac{4}{4+0} = 1$$

$$FPR = \frac{2}{2+0} = 0.5$$



Higher TP

①  $T_H(0.6)$   
 ②  $T_H(0.2)$   
 ③  $T_H(0.4)$

<del>TP</del> TPR	FPR
<del>TP</del> 0.6	0
<del>0.6</del> 1.0	0.5
0.5	0.6

- ① Only focus on FPR
- ② FPR doesn't matter as long as TP is high
- ③ trade off b/w the two