

Metrics in Classification.

Performance metrics in classification.

- ① Confusion matrix
- ② FP (Type I Error)
- ③ FN (Type II Error)
- ④ Recall [True positive rate]
- ⑤ Precision [Positive Prediction value.] ~~or~~
- ⑥ Accuracy
- ⑦ F Beta.
- ⑧ Cohen Kappa
- ⑨ ROC Curve, AUC Curve
- ⑩ PR Curve

(Precision-Recall)

"How much Good Your model is?"

Classification

Problem Statement

Probabilities

Class labels

ROC Curve.

AUC Curve.

PR Curve.

Suppose Binary pr.
and it has only 2
outcome.

A, B

1000 records.

500 / 500

balance dataset

⇒ Default threshold
Value = 0.5

Here ML algorithm
not biast.

600 / 400

same bias

200 / 300

model or algo.

can understand.

output < 0.5

800 / 200

class A or
class B

Dataset

Imbalanced

Provide in ML Algorithm.

Some of the Algorithm get biast.

Balance dataset - (used) → Accuracy

Imbalance dataset → Precision, Recall, F1 Measure or
F Beta...

Let's an classification problem.

Confusion Matrix.

		Actual	
		Has Heart Disease	Does not have heart disease
Predicted	Has heart Disease	True Positive	False Positive
	Does not have heart disease	False Negative	True Negative

Type I Error

Type II Error

Always focus on Reduce FP ↓ and FN ↓

Type I Error

$$\textcircled{3} \text{ FPR} = \text{false Positive Rate} = \frac{\text{FP}}{\text{FP} + \text{TN}}$$

$$\textcircled{4} \text{ FNR} = \text{false Negative Rate} = \frac{\text{FN}}{\text{FN} + \text{TP}}$$

$$\textcircled{1} \text{ Sensitivity} = \frac{\text{True Positive}}{\text{True Positive} + \text{False Negative}} \quad [\text{Recall}]$$

$$\textcircled{2} \text{ Specificity} = \frac{\text{True Negative}}{\text{True Negative} + \text{False Positive}} \quad [\text{Precision}]$$

Considering it's a balance dataset.

$$\therefore \text{Accuracy} = \frac{\text{Right Result}}{\text{Total Result}}$$

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

Imbalance dataset.

		Test	
		900, 100	→ B
A	→	800, 200	

→ max max model Predicted all are class A

$$\therefore \text{Accuracy} = \frac{90+0}{100} = 90\%$$

So there is a Problem.

→ we can't use Accuracy for imbalance dataset.

→ Recall
 → Precision
 → F1 Beta.

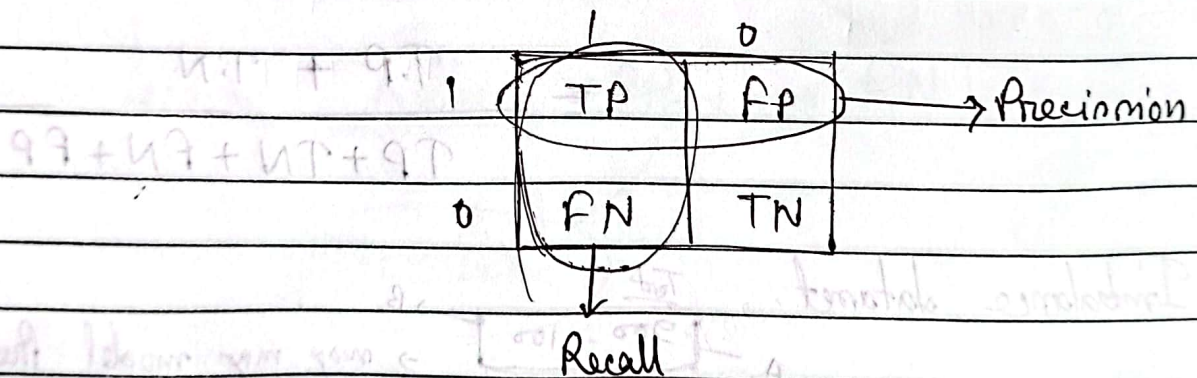
$$\therefore \text{Recall} = \frac{\text{True Positive}}{\text{True Positive} + \text{False Negative}} \quad [\text{Sensitivity}]$$

$$\text{Precision} = \frac{\text{True Positive}}{\text{True Positive} + \text{False Positive}}$$

[use Spam detection]

→ If FP Greater impact on role, go to Precision.

FN Greater impact on role → use Recall



→ When FP and FN That mean's Type I and Type II

Error both are important or greater

impact, we use "F1 Measure" or F-Beta.

→ Both are Equally Important $\leftarrow \beta = 1$

$$\Rightarrow F_{\beta} = \frac{(1 + \beta^2) \text{Precision} * \text{Recall}}{\beta^2 * \text{Precision} + \text{Recall}}$$

$$F_1 = 2 * \frac{\text{Precision} * \text{Recall}}{\text{Precision} + \text{Recall}}$$

$$= \text{Harmonic Mean } \frac{2xy}{x+y}$$

$\beta = 1$ F_1 score

$\beta = 0.5$ $F_{1.5}$ score

$\beta = 2$ F_2 score

$\beta = 1$, when FN and FP both are equally important.

$\beta = 0.5$, when FP > FN more important.

$\beta = 2$, when FN > FP more important.

β value (1-10)

ROC and AUC

Consider a binary Problem.

Suppose

$0.5 > = 0$

$0.5 < = 1$

Y	\hat{Y} (Predict)	$\hat{Y}_{(0)}$	$\hat{Y}_{(0.2)}$	$\hat{Y}_{(0.4)}$
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	0.2	1	1	

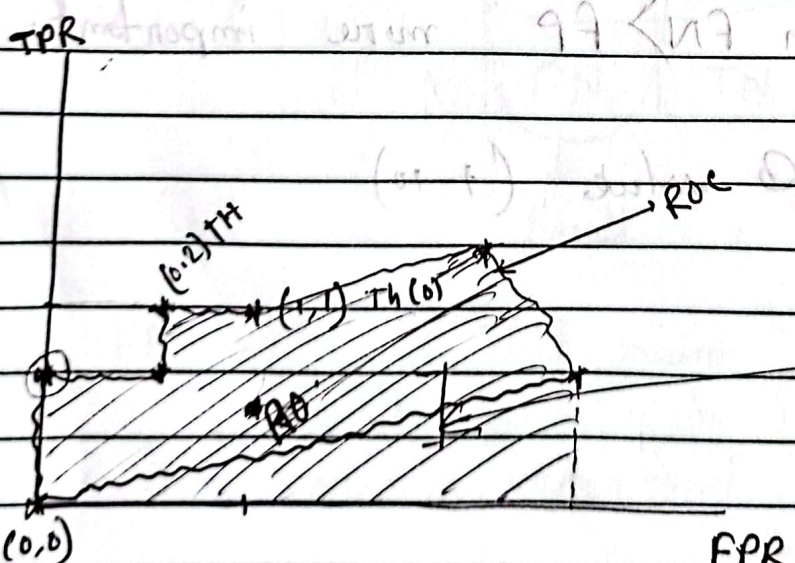
Threshold value

Consider.

0, 0.2, 0.4, 0.6, 0.8

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

$$FPR = \frac{FP}{FP + TN} = \frac{2}{2 + 0} = 1$$



Area Under Curve

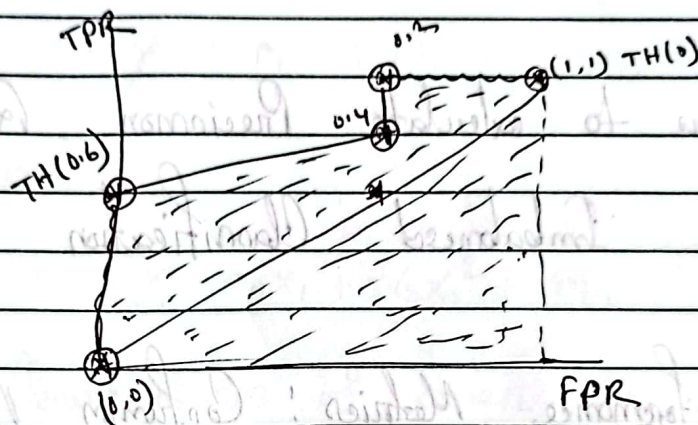
The more the area under the curve the better model.

Depend on Domain Expert Person

what threshold value will select.

→ Hei monu i need higher TPR so i selected

TH(0.6)



Based on Considered Threshold value, we will get

TPR and FPR value, and we

→ value diya jayega curve par (ROC) Point sukhi Place

→ curve ko under a ya bhagdo, or AUC.

Kiavay TH value Select karva, Domain expert chotav hoga,

1. Matrices to Evaluate your Machine learning Algorithm.

(towards datascience)

2. Confusion Matrix, Accuracy, Precision, Recall, F1 Score.

(medium.)

3. How to calculate Precision, Recall, and F Measure for Imbalanced Classification (machinelearningmastery)

4. Performance Metrics: Confusion Matrix, Precision, Recall, F1 Score. (Towardsdatascience)

You should Read more on those article from website.

The answer you get depend upon the Question you ask.

Thomas Kuhn.