

Confusion Matrix

→ is a mechanism to measure the performance of classifier.

↳ It contains information about actual and predicted classification

		Predicted Class		
		Positive	Negative	
Actual class	Positive	True Positive (TP) Type II errors	False Negative (FN) Type I errors	Recall (R) or Sensitivity $\frac{TP}{TP+FN}$
	Negative	False Positive (FP) Type I errors	True Negative (TN)	Specificity $\frac{TN}{TN+FP}$
		Precision (P) $\frac{TP}{TP+FP}$ +ve predicted value	-ve Predictive Value $\frac{TN}{TN+FN}$	

Precision or +ve predicted value — The proportion of +ve cases that were correctly identified

-ve predicted value — The proportion of -ve cases that were correctly identified

Sensitivity or Recall — The proportion of actual +ve cases which are correctly identified

Specificity — The proportion of actual -ve cases which are correctly identified

Accuracy — The proportion of the total number of predictions that were correct.

TP → +ve class correctly identified as +ve class

FN → +ve class incorrectly identified as -ve class.

FP → -ve class incorrectly identified as +ve class

TN → -ve class correctly identified as -ve class

Example: Distinguish Spam emails from non-spam emails

→ 100^{email} examples Predicted class

→ 100 ^{email} examples

		Predicted class	
		Spam	Non-Spam
Actual class	Spam	TP 45	FN 20
	Non-Spam	FP 5	TN 30

$$\textcircled{1} \text{ Recall (R)} = \frac{TP}{TP+FN} = \frac{45}{45+20} = 69.23\%$$

The 69.23% Spam-emails are correctly classified and excluded from all non-spam emails

$$\textcircled{2} \text{ Precision} = \frac{TP}{TP+FP} = \frac{45}{45+5} = 90\%$$

The 90% of examples are classified as spam are actually spam

$\textcircled{3}$ Specificity — It is also known True Negative Rate

$$= \frac{TN}{TN+FP} = \frac{30}{30+5} = 85.71\%$$

The 85.71% non-spam emails are accurately classified and excluded from all spam emails

$\textcircled{4}$ Accuracy

$$= \frac{TP+TN}{TP+FP+TN+FN} = \frac{45+30}{45+5+30+20} = 75\%$$

The 75% of examples are classified correctly by the classifier.

$\textcircled{5}$ F1-Score → is a weighted average of the recall and precision

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

Problem 1

		Predicted class	
		SPam	non - Spam
Actual class	Spam	TP 20	FN 10
	non-Spam	FP 30	TN 40

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

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

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

$$\text{Recall} = 66.6\%$$

$$\text{Precision} = 40\%$$

$$\boxed{\text{Accuracy} = 60\%}$$

Why higher accuracy does not always imply a good classified.

$$\textcircled{6} \text{ Classification error} = \frac{\text{error total}}{\text{total}} = \frac{FP + FN}{TP + TN + FP + FN}$$

$$\textcircled{7} \text{ Miss classification error} = 1 - \text{Accuracy}$$

$$\textcircled{8} \text{ False +ve Rate} = \frac{FP}{FP + FN} \rightarrow \% \text{ of -ve we misclassified as +ve}$$

or
False Alarm Rate

$$\textcircled{9} \text{ False -ve Rate} = \frac{FN}{TP + FN} \rightarrow \% \text{ of +ve we classified as -ve}$$

or
Miss Rate

Problem 2:

		Predicted class		
		A	B	C
Actual class	A	15	2	3
	B	7	15	8
	C	2	3	45

True Positive

Note:

① The total no. of FN \rightarrow the sum of values in corresponding row (Excluding TP)

② The total no. of FP \rightarrow The sum of values in the corresponding column (excluding the TP)

③ The total no. of TN \rightarrow for certain class will be the sum of all columns and rows excluding the class column and row

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

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

$$P_A = \frac{15}{15+7+2} = \frac{15}{24} = 62.5\% \quad R_A = \frac{15}{15+2+3} = \frac{15}{20} = 0.75 = 75\%$$

$$P_B = \frac{15}{15+2+3} = \frac{15}{20} = 75\% \quad R_B = \frac{15}{15+7+8} = \frac{15}{30} = 0.5 = 50\%$$

$$P_C = \frac{45}{45+3+8} = \frac{45}{56} = 80\% \quad R_C = \frac{45}{45+2+3} = \frac{45}{50} = 0.9 = 90\%$$

✓ Overall Accuracy

✓ wtd avg. precision

✓ wkd avg. recall

$$A_A =$$

$$A_B =$$

$$A_C =$$

Programming \rightarrow Result. Question
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