PRECISION, RECALL, ACCURACY, ERROR RATE

		<u>True Class</u>	
		T	F
Acquired Class	>	True Positives (TP)	False Positives (FP)
Acquire	z	False Negatives (FN)	True Negatives (TN)

True Positive Rate (TPR) =
$$\frac{TP}{TP + FN}$$

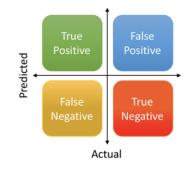
False Positive Rate (FPR) = $\frac{FP}{FP + TN}$
Accuracy (ACC) = $\frac{TP + TN}{TP + FP + TN + FN}$

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

$$FNR = \frac{FN}{Actual\ Positive} = \frac{FN}{TP + FN}$$

$$TNR = \frac{TN}{Actual\ Negative} = \frac{TN}{TN + FP}$$

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



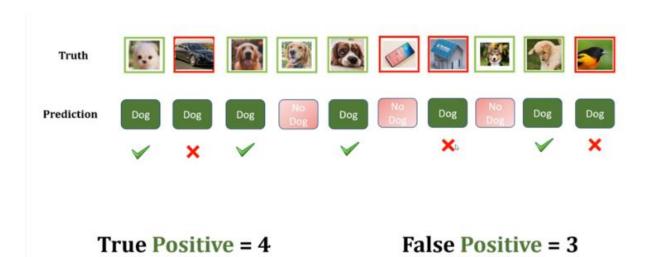
A	_	True Positive + True Negative	
Accuracy	=	Total	

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1	Prediction	Actual value	Туре	Explanation
2	1	1	True Positive	Predicted Positive and was Positive
3	0	0	True Negative	Predicted Negative and was Negative
4	1	0	False Positive	Predicted Positive but was Negative
5	0	1	False Negative	Predicted Negative but was Positive

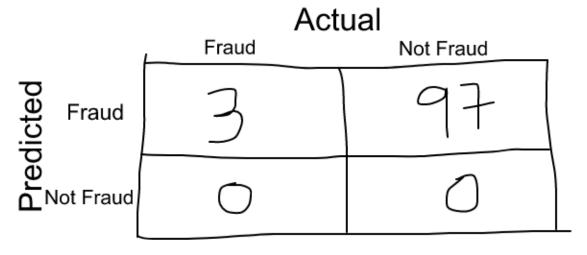
		Predi	icted Class	
		Positive	Negative	
Actual Class	Positive	True Positive (TP)	False Negative (FN) Type II Error	Sensitivity $\frac{TP}{(TP+FN)}$
Actual Class	Negative	False Positive (FP) Type I Error	True Negative (TN)	Specificity $\frac{TN}{(TN+FP)}$
	$\frac{Precision}{TP}$ $\frac{TP}{(TP+FP)}$	Negative Predictive Value $\frac{TN}{(TN + FN)}$	Accuracy $\frac{TP + TN}{(TP + TN + FP + FN)}$	

- True Positive: the truth is positive, and the test predicts a positive.
 The person is sick, and the test accurately reports this.
- True Negative: the truth is negative, and the test predicts a negative. The person is not sick, and the test accurately reports this.
- False Negative: the truth is positive, but the test predicts a negative. The person is sick, but the test inaccurately reports that they are not. Also called a Type II error in statistics.
- False Positive: the truth is negative, but the test predicts a
 positive. The person is not sick, but the test inaccurately reports
 that they are. Also called a Type I error in statistics.



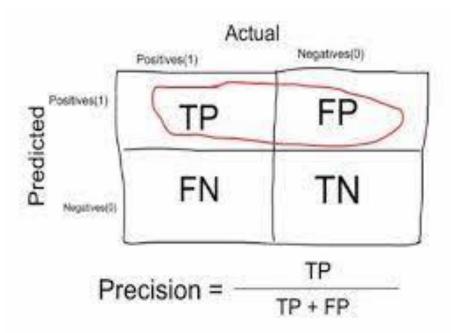


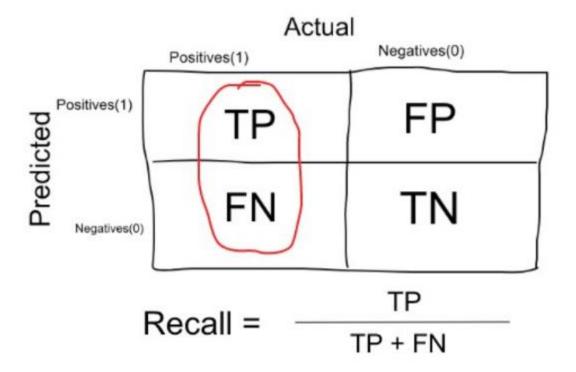
True Negative = 1 False Negative = 2



Precision =
$$\frac{3}{100}$$
 = 3%

Recall =
$$\frac{3}{3}$$
 = 100%





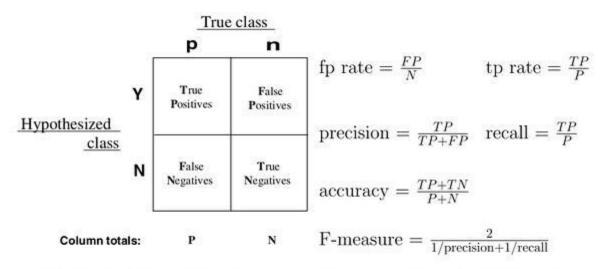


Fig. 1. Confusion matrix and common performance metrics calculated from it.

Main metrics — The following metrics are commonly used to assess the performance of classification models:

Metric Formula		Interpretation	
Accuracy	$\frac{TP+TN}{TP+TN+FP+FN}$	Overall performance of model	
Precision	$\frac{\mathrm{TP}}{\mathrm{TP} + \mathrm{FP}}$	How accurate the positive predictions are	
Recall $\frac{TP}{TP+FN}$		Coverage of actual positive sample	
Specificity	$\frac{\rm TN}{\rm TN + FP}$	Coverage of actual negative sample	
F1 score	$\frac{\text{2TP}}{\text{2TP} + \text{FP} + \text{FN}}$	Hybrid metric useful for unbalanced classes	

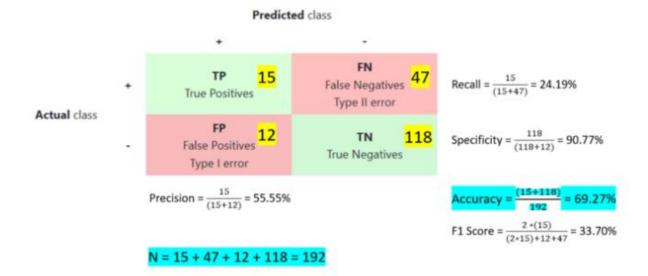
Training set

Age	Heart rate	Blood pressure	Heart problem
65	78	150/70	Yes
37	83	112/76	No
71	67	108/65	No

Prediction set

Age	Heart rate	Blood pressure	Heart problem
43	98	147/89	?
65	58	106/63	?
84	77	150/65	?

TABLE 1 – TRAINING AND PREDICTION SETS FOR MEDICAL DATABASE



 $\underline{https://towardsdatascience.com/a-look-at-precision-recall-and-f1-score-36b5fd0dd3ec}$