

Metrics for classification and regression

ML Instruction Team, Fall 2022

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MNIST Dataset

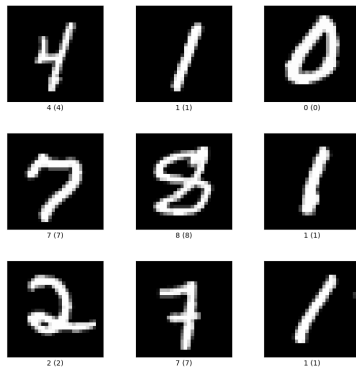


Figure: MNIST dataset

Never5 Classifier

- Suppose a very dumb classifier that just classifies every single image in the “not-5” class.
- It will achieve a accuracy around **90%**.
- Accuracy is not always a good measurment.

Confusion Matrix

		Predicted	
		Negative	Positive
Actual	Negative	8 3 9 7 2	6
	Positive	5 5 5	5 5 5

TN (True Negative) is indicated for the top-left cell (Actual Negative, Predicted Negative).
 FP (False Positive) is indicated for the top-right cell (Actual Negative, Predicted Positive).
 FN (False Negative) is indicated for the bottom-left cell (Actual Positive, Predicted Negative).
 TP (True Positive) is indicated for the bottom-right cell (Actual Positive, Predicted Positive).

Figure: Confusion Matrix For a Classifier

$$ERR = \frac{FP + FN}{FP + FN + TP + TN} = 1 - ACC \quad (1)$$

$$ACC = \frac{TP + TN}{FP + FN + TP + TN} = 1 - ERR \quad (2)$$

False Positive Rate and False Negative Rate

$$TPR = \frac{TP}{P} = \frac{TP}{TP + FN} = 1 - FNR \quad (3)$$

$$FPR = \frac{FP}{N} = \frac{FP}{FP + TN} = 1 - TNR \quad (4)$$

$$FNR = \frac{FN}{N} = \frac{FN}{FN + TP} = 1 - TPR \quad (5)$$

$$TNR = \frac{TN}{N} = \frac{TN}{TN + FP} = 1 - FPR \quad (6)$$

Precision, Recall, F_1 Score

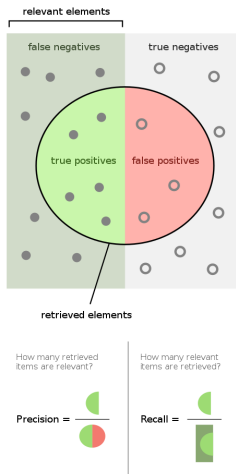


Figure: source

Precision, Recall, F_1 Score

$$PRE = \frac{TP}{TP + FP} \quad (7)$$

$$REC = TPR = \frac{TP}{FN + TP} \quad (8)$$

$$F_1 = 2 \cdot \frac{PRE \cdot REC}{PRE + REC} \quad (9)$$

Decision Threshold

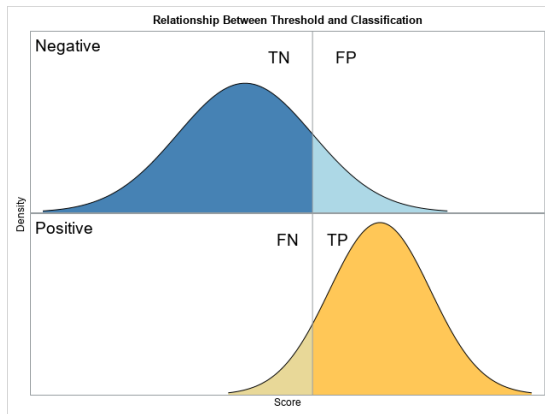


Figure: [source](#)

Precision/Recall Trade-off

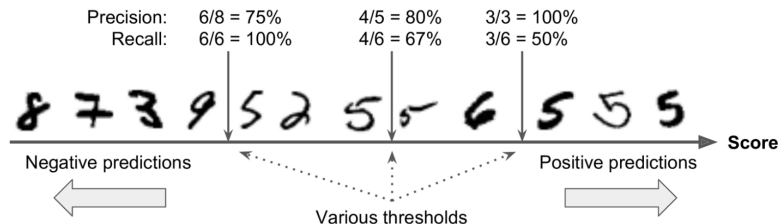


Figure: the higher the threshold, the lower the recall, but (in general) the higher the precision

Precision/Recall Trade-off

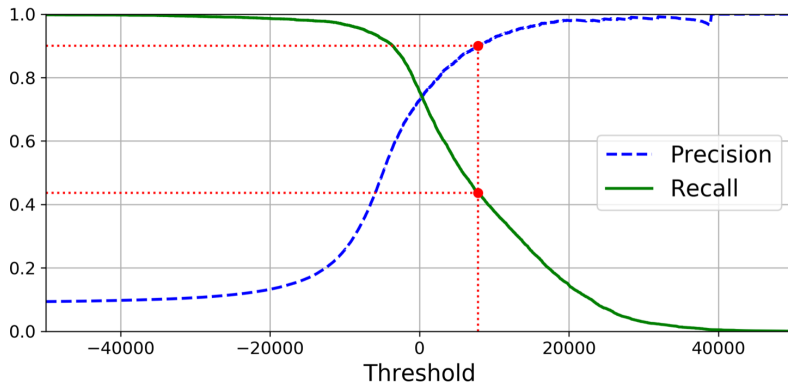


Figure: Precision and recall versus the decision threshold

Sensitivity and Specificity

$$SEN = TPR = \frac{TP}{P} = \frac{TP}{TP + FN} \quad (10)$$

$$SPC = TNR = \frac{TN}{N} = \frac{TN}{TN + FP} \quad (11)$$

- Sensitivity (SEN) measures the recovery rate of the Positives and complimentary.
- Specificity (SPC) measures the recovery rate of negatives.

ROC Curve

- plots the true positive rate against false positive rate.
- used with binary classifiers.

ROC Curve

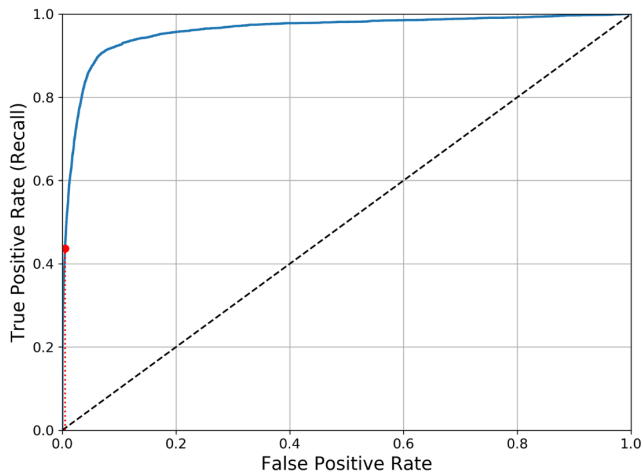


Figure: ROC Curve

Area Under The Curve (AUC)

- another way to compare classifiers.
- perfect classifier has a ROC AUC equal to 1.

Area Under The Curve (AUC)

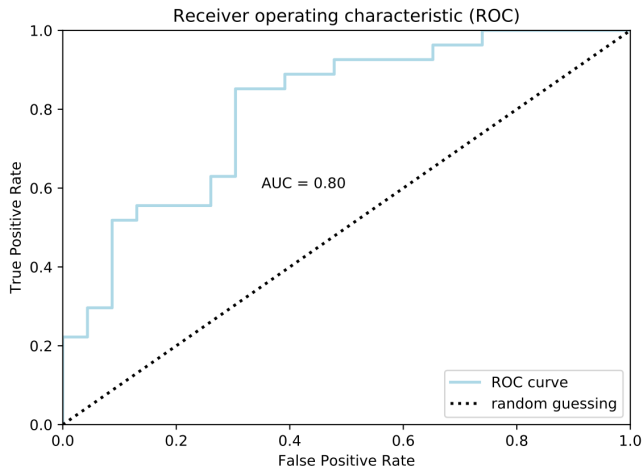


Figure: ROC AUC

Root Mean Squared Error (RMSE)

- a metric for regressors.
- perfect regressor has a RMSE equal to 0.

$$RMSE = \sqrt{\left(\frac{1}{n}\right) \sum_{i=1}^n (y_i - \hat{y}_i)^2} \quad (12)$$

Thank You!

Any Question?