

Evaluation Metrics Guide

Evaluation Metrics Definitions

IoU grouping parameter is used to discard linkage of wrong label between ground truth and prediction having IoU less than IoU Grouping

Generally, positive/negative defines whether we have a prediction or not and True/False defines whether our prediction is True/False

	Prediction	Reality
True Positive (TP)	Positive	Positive
True Negative (TN)	Negative	Negative
False Positive (FP)	Positive	Negative
False Negative (FN)	Negative	Positive

$$\text{Accuracy: } \frac{\text{Correct answers}}{\text{All Possibilities}} = \frac{TP + TN}{TP + TN + FP + FN}$$

- Measure of how correct our Data is.
- Best = 1 ----- Worst = 0




$$\text{Precision: } \frac{\text{Predicted Correctly}}{\text{All Predicted}} = \frac{TP}{TP + FP}$$

- Measure the fraction of relevant examples among all of the examples predicted that belong to a certain class.
- Best = 1 ----- Worst = 0

$$\text{Recall: } \frac{\text{Predicted Correctly}}{\text{All that should be Predicted}} = \frac{TP}{TP + FN}$$

- Measure the fraction of relevant examples which were predicted to belong to a class with respect to all examples that truly belong to a class.
- Best = 1 ----- Worst = 0

$$\text{F-Score: } \frac{(1 + \beta^2) \times \text{Precision} \times \text{Recall}}{(\beta^2 \times \text{Precision}) + \text{Recall}}$$

- Evaluate machine learning model in one number by combining precision and recall
- A good f-score means that the model is effective in terms of how precisely it classifies the data and that it covers a good proportion of the cases that it should have classified correctly.
- Best = 1 ----- Worst = 0
 - o If $\beta = 1$  neutral
 - o If $\beta < 1$  focuses more on precision
 - o If $\beta > 1$  focuses more on recall

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- Before Continuing, we need to give an example:

- Let us take a case where we have 95% of the population does not have cancer and 5% of the population have cancer.
 - If over 100 person we detect 100 persons that do not have cancer, then we will have a case where accuracy = 95%.
 - But in reality, we need this model to predict if a person has cancer or not so these results are unacceptable, because we cannot find any of the persons that has cancer.
 - Another example is the case where 95% of data should be detected but we detect 100% of the data True. In this case we will have precision = 95% and Recall = 100% but the purpose of finding the 5% makes the model be very bad.
 - This is why a new metric will be introduced (balanced accuracy), in which we are taking into consideration how the dataset is distributed or balanced
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True Negative Rate: $\frac{TN}{TN + FP}$

True Positive Rate: $\frac{TP}{TP + FN}$

Balanced Accuracy: $\frac{TNR + TPR}{2}$

- This parameter takes into consideration how balanced is the data.