

Pytorch course

02. Custom metrics



Metrics to evaluate your machine learning models

Generally when deep learning models are calculated, a metric is needed to weight the model. Choosing the right metric **is crucial** while **evaluating** machine learning (**ML**) models.

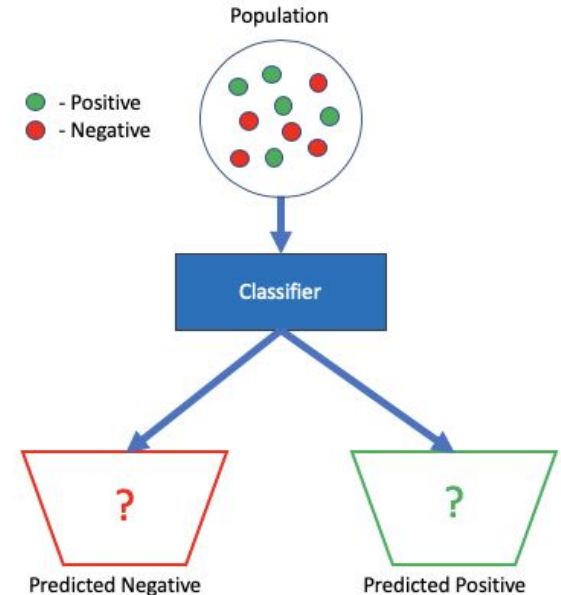
In some applications looking at a single metric may not give you the whole picture of the problem you are solving, and you **may want to use a subset of the metrics** to have a concrete evaluation of your models.

Accuracy, precision, sensitivity, AUC, recall, etc...

Nevertheless, each of the metrics have a set of true and false positives and negatives samples, that is needed to know what it means

Metrics to evaluate your machine learning models

- ★ **True positive samples (TP):** are **positive outcomes** that the model **predicted correctly**.
Example: this means that **patients who were predicted** to have cancer by the model **indeed does have cancer**
- ★ **True negative samples (TN):** are **negative outcomes** that the model **predicted correctly**.
Example: this means that **patients who were predicted** to be healthy **indeed does not have cancer**.
- ★ **False positive samples (FP):** are **positive outcomes** that the model **predicted incorrectly**. This is also known as Type I error.
Example: this means that patients who were predicted to have cancer were actually healthy.
- ★ **False negative samples (FN):** are **negative outcomes** that the model **predicted incorrectly**. This is also known as Type II error.
Example: this means that **patients who were predicted** to be healthy **actually had cancer**.



Metrics to evaluate your machine learning models

		Real label	
		Positive	Negative
Predicted label	Positive	True positive (TP)	False positive (FP)
	Negative	False negative (FN)	True negative (TN)

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

Accuracy is the metric that counts the proportion of correct predictions.

Metrics to evaluate your machine learning models

		Real label	
		Positive	Negative
Predicted label	Positive	True positive (TP)	False positive (FP)
	Negative	False negative (FN)	True negative (TN)

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

Precision is the metric that questions out of all the examples that predicted as positive, how many are really positive?

Metrics to evaluate your machine learning models

		Real label	
		Positive	Negative
Predicted label	Positive	True positive (TP)	False positive (FP)
	Negative	False negative (FN)	True negative (TN)

$$\text{Sensitivity} = \frac{\text{TP}}{\text{(TP + FN)}}$$

Sensitivity is the metric that questions out of all the positive examples, how many are predicted as positive? Also known as positive accuracy

Metrics to evaluate your machine learning models

		Real label	
		Positive	Negative
Predicted label	Positive	True positive (TP)	False positive (FP)
	Negative	False negative (FN)	True negative (TN)

$$\text{Specificity} = \frac{\text{TN}}{\text{TN} + \text{FP}}$$

Specificity is the metric that questions out of all the people that do not have the disease, how many got negative results? Also known as negative accuracy