

Al for Product Managers: Model Evaluation for Classification

Classification

In classification, every item in a dataset is assigned a **class** or **label**.

Training a Classification Model

- Requires a training dataset of (input, actual label) pairs.
- Model is fed these pairs.
- Model processes the *inputs*.
- Model produces *predicted labels*.
- Model updates its weights to minimize the mismatch between actual labels and predicted labels.

Binary Classification

- Actual label has only two values representing the presence or absence of something.
- Example: in diagnosing pneumonia, the labels are *pneumonia* or *normal*.

Positive and Negative Labels

- Positive (actual) labels represent the presence of a condition/pattern/quality.
- Negative (actual) labels represent the absence of the condition/pattern/quality.

 #Dataset = #Actual positives + #Actual negatives[†]

Actual Positives	
Actual Negatives	

Evaluating a Classification Model

- Requires a validation or test dataset of (input, actual label) pairs.
- The model is fed only inputs, the actual labels are withheld: (input, actual label).
- Model processes the *inputs*.
- Model produces *predicted labels*.
- Evaluation compares predicted labels against withheld actual labels.

True Positives and False Negatives

- When a model classifies an actual positive:
 - It either correctly classifies the input as a positive (true positives).
 - Or misclassifies it as a negative (false negative).

	Predicted as Positives	Predicted as Negative
Actual Positives	True Positives(TP)	False Negatives(FN)

^{† #} represents count

^{**} Some software may produce a rotated table, in which actual positives and negatives are in the columns and predicted positives and negatives are in the rows.



#Actual positives = #True positives + #False negatives*

True Negatives and False Positives

- When a model classifies an actual negative:
 - It either correctly classifies the input as a negative (true negative).
 - Or misclassifies it as a positive (false positive).

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_	Predicted as Negatives	as Positive	
Actual Negatives	True Negatives (TN)	False Positives (FP)	

#Actual negatives = #True negatives + #False positives[†]

Remembering the Correspondence

- Actual label == Predicted label => True (Positive/Negative)
- Actual label != Predicted label => False (Positive/Negative)

The Confusion Matrix

- Represents the outcome of a model's evaluation.
- For binary classifiers, it is a 2x2 matrix or table.

**	Predicted Positives	Predicted Negatives
Actual Positives	TP	FN
Actual Negatives	FP	TN

TP+FN = total number of positive inputs in the dataset TN+FP = total number of negative inputs in the dataset TP+FN+TN+FP = size of the dataset

Evaluation Metrics

Accuracy: fraction of correct predictions.

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

 Precision: fraction of predicted positives that are also actual positives.

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

 Recall: fraction of actual positives that are also predicted positives.

$$Recall = \frac{TP}{(TP + FN)}$$

 F1 Score combines precision and recall in to a single number, which makes comparing two models easier.

$$F1 = \frac{2*Precision*Recall}{(Precision+Recall)}$$

Multiclass Classifiers

- Actual labels have more than 2 values.
- E.g. Car, Bus, Truck.
- Metrics are calculated for each of the classes and then their average is taken.

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