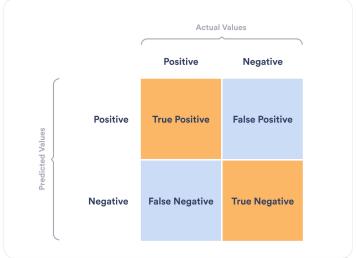
Confusion Matrix and Metrics

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What is a Confusion Matrix?

- ➤ A confusion matrix is a tool used to assess the performance of a classification model.
- It compares the actual and predicted classifications.



Confusion Matrix Structure

	Predicted Positive	Predicted Negative
Actual Positive	True Positive (TP)	False Negative (FN)
Actual Negative	False Positive (FP)	True Negative (TN)

- ► **TP**: Correct positive predictions
- ► TN: Correct negative predictions
- ▶ **FP**: Incorrect positive predictions (Type I Error)
- ► **FN**: Incorrect negative predictions (Type II Error)

Accuracy

Formula:

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

Example:

Suppose we have a confusion matrix:

	Predicted Positive	Predicted Negative
Actual Positive	50	10
Actual Negative	5	35

Solution:

Accuracy =
$$\frac{50+35}{50+35+5+10} = \frac{85}{100} = 0.85$$

Therefore, the accuracy is 85%.

Precision (Positive Predictive Value)

Formula:

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

Example:

From the previous confusion matrix:

$$Precision = \frac{50}{50+5} = \frac{50}{55} = 0.91$$

Therefore, the precision is 91%.

Recall (Sensitivity or True Positive Rate)

Formula:

$$\mathsf{Recall} = \frac{\mathit{TP}}{\mathit{TP} + \mathit{FN}}$$

Example:

From the previous confusion matrix:

$$\mathsf{Recall} = \frac{50}{50+10} = \frac{50}{60} = 0.83$$

Therefore, the recall is 83%.

F1 Score

Formula:

F1 Score =
$$2 \times \frac{\text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$

Example:

• Using precision = 0.91 and recall = 0.83:

F1 Score =
$$2 \times \frac{0.91 \times 0.83}{0.91 + 0.83} = 2 \times \frac{0.7553}{1.74} = 0.87$$

Therefore, the F1 Score is 87%.

Specificity (True Negative Rate)

Formula:

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

Example:

From the confusion matrix:

Specificity =
$$\frac{35}{35+5} = \frac{35}{40} = 0.875$$

Therefore, the specificity is 87.5%.

Matthews Correlation Coefficient (MCC)

Formula:

$$\mathsf{MCC} = \frac{(\mathit{TP} \times \mathit{TN}) - (\mathit{FP} \times \mathit{FN})}{\sqrt{(\mathit{TP} + \mathit{FP})(\mathit{TP} + \mathit{FN})(\mathit{TN} + \mathit{FP})(\mathit{TN} + \mathit{FN})}}$$

Example:

$$MCC = \frac{(50 \times 35) - (5 \times 10)}{\sqrt{(50 + 5)(50 + 10)(35 + 5)(35 + 10)}}$$

Solving step-by-step:

$$MCC = \frac{1750 - 50}{\sqrt{(55)(60)(40)(45)}} = \frac{1700}{\sqrt{594000}} = \frac{1700}{771.67} = 0.88$$

Therefore, the MCC is 0.88.

Conclusion

- ➤ The confusion matrix provides a powerful way to evaluate classification models.
- ▶ Different metrics like accuracy, precision, recall, F1 score, and MCC offer insights into model performance, especially when dealing with imbalanced datasets.
- ► These metrics help in making decisions about trade-offs in model performance (e.g., precision vs recall).