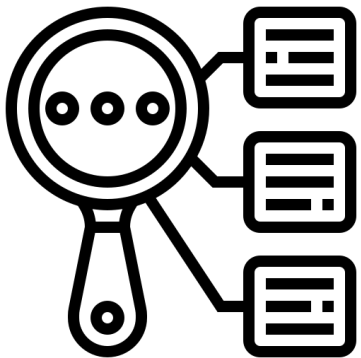


# Evaluation Metrics

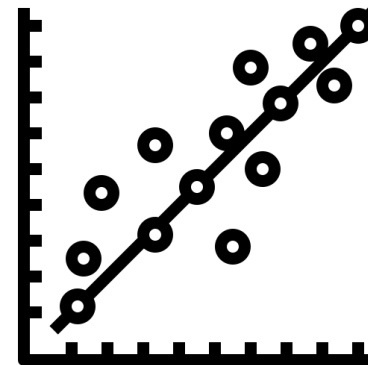
**Why?** Model evaluation is a critical step to assess the predictive performance of a model, ensuring its robustness and reliability.

classification metrics



- Accuracy
- Precision
- Recall
- F-score
- ROC-AUC

regression metrics



- Mean Absolute Error (MAE)
- Root Mean Square Error (RMSE)
- $R^2$  score

# Classification Metrics

- **Accuracy**

$$\text{Accuracy} = \frac{\text{Number of correct predictions}}{\text{Total number of predictions}}$$

Measures the proportion of correct predictions over all instances.

**Range:**  $0 \leq \text{Accuracy} \leq 1$

- A value of 0 indicates that none of the predictions are correct.
- A value of 1 indicates that all predictions are correct.

- **Precision**

$$\text{Precision} = \frac{\text{True Positive}}{\text{True Positive} + \text{False Positive}}$$

Measures the proportion of true positives among instances classified as positive.

**Range:**  $0 \leq \text{Precision} \leq 1$

- A value of 0 means that none of the positive predictions are correct.
- A value of 1 means that all positive predictions are correct.

- **Recall**

$$\text{Recall} = \frac{\text{True Positive}}{\text{True Positive} + \text{False Negative}}$$

Measures the proportion of true positives among actual positive instances

**Range:**  $0 \leq \text{Recall} \leq 1$

- A value of 0 indicates that none of the actual positive instances are identified.
- A value of 1 indicates that all actual positive instances are identified.

- **F-score**

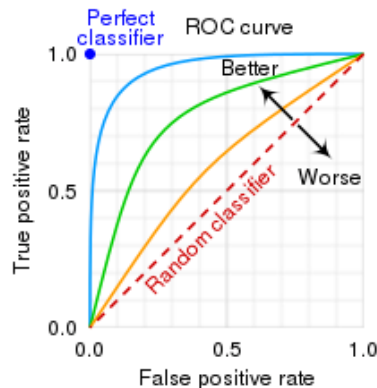
$$F_1\text{-score} = 2 \times \frac{\text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}} = \frac{2\text{TP}}{2\text{TP} + \text{FP} + \text{FN}}$$

Harmonic mean of Precision and Recall, balances the two metrics

**Range:**  $0 \leq F_1\text{-score} \leq 1$

- A value of 0 suggests that either the precision or the recall is zero.
- A value of 1 suggests perfect precision and recall.

- **ROC-AUC:**



Quantifies the model's ability to discriminate between positive and negative classes

**Range:**  $0 \leq \text{ROC-AUC} \leq 1$

- A value of 0.5 suggests no discrimination (akin to random guessing).
- A value of 0 indicates that the model is making all predictions incorrectly.
- A value of 1 indicates perfect discrimination between the positive and negative classes.

# Confusion Matrix

		Predicted	
		Negative (N) -	Positive (P) +
Actual	Negative -	True Negative (TN)	False Positive (FP) Type I Error
	Positive +	False Negative (FN) Type II Error	True Positive (TP)

# Regression Metrics

- **Mean Absolute Error (MAE):** The average of the absolute differences between the predicted and actual values.
- **Root Mean Square Error (RMSE):** The square root of the average of the squared differences between the predicted and actual values.
- **$R^2$  score:** Represents the proportion of the variance for the dependent variable that's explained by the independent variables.