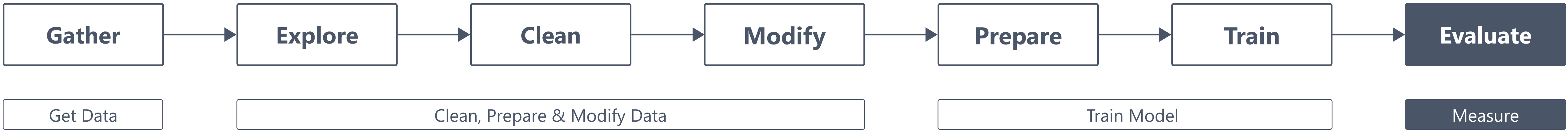


# AI Workflow

## Model Evaluation - Classification



### CONFUSION MATRIX

		ACTUAL VALUE	
		POSITIVE	NEGATIVE
PREDICTED VALUE	POSITIVE	TP (True Positive)	FP (False Positive)
	NEGATIVE	FN (False Negative)	TN (True Negative)

Recall is associated with TP and FN. Precision is associated with TP and FP.

**TP (True Positive) & TN (True Negative):**  
The prediction matches the Truth correctly  
*E.g. we correctly predicted an object as a car*

**FP (False Positive) & FN (False Negative):**  
The prediction does not match the Truth  
*E.g. we incorrectly predicted an object as a car, while it was something else*

### METRICS

#### ACCURACY

Formula:  $\frac{(TP + TN)}{(TP + TN + FP + FN)}$  OR  $\frac{\#CORRECT\_PREDICTIONS}{\#TOTAL}$

Summary: How well does the model perform?

Example: Our model is 95% accurate

#### RECALL

Formula:  $\frac{(TP)}{(TP + FN)}$  OR  $\frac{\#CORRECT\_POSITIVE\_PREDICTIONS}{\#TRUE\_TRUTH\_VALUES}$

Summary: How often did we wrongly classify something as not true (= false?)

Example: 5% of the time we said it was not a car, while it was a car (we could've hit it)

#### PRECISION

Formula:  $\frac{(TP)}{(TP + FP)}$  OR  $\frac{\#CORRECT\_POSITIVE\_PREDICTIONS}{\#POSITIVE\_SAMPLES}$

Summary: How often are we correct in our positive prediction? (or how much are we being wrong?)

Example: 5% of the time we said an object was a car, while it actually was not. (wrong action will be taken - e.g. increasing speed)

#### F-SCORE (AND F1 SCORE)

Formula:  $F_{\beta} = (1 + \beta^2) \frac{(PRECISION * RECALL)}{(\beta^2 * PRECISION) + RECALL}$

$F_1 = 2 * \frac{(PRECISION * RECALL)}{(PRECISION + RECALL)}$

Summary: Utilize the precision and recall to create a test's accuracy through the "harmonic mean". Also known as the Sørensen–Dice Coefficient

Example: Our model is 88% accurate based on high-business impact markers (#wrong detections and #false positives)

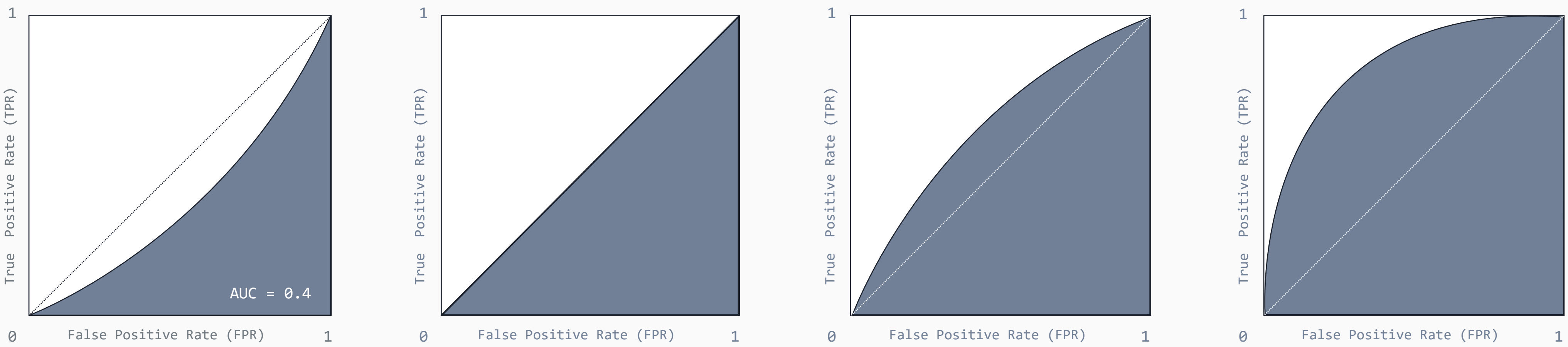
### ROC CURVE

Formula: False Positive Rate (FPR) = X-Axis  
True Positive Rate (TPR) = Y-Axis  
 $TPR = \frac{(TP)}{(TP + FN)}$        $FPR = \frac{(FP)}{(FP + TN)}$

Summary: The ROC curve allows us to select the optimal model and discard suboptimal ones.

Method: 1. Discretize the threshold for the confidence score (e.g. confidence score of [0, 1] becomes [0.0, ..., 0.9, 1.0])  
2. Calculate the confusion matrix for the given threshold  
3. Determine the TPR and FPR and plot them

Examples:



### READ MORE

Performance Markers (<https://xaviergeerinck.com/ai-performance-markers>)

F1 Score ([https://scikit-learn.org/stable/modules/generated/sklearn.metrics.f1\\_score.html#sklearn.metrics.f1\\_score](https://scikit-learn.org/stable/modules/generated/sklearn.metrics.f1_score.html#sklearn.metrics.f1_score))