### **Model Evaluation in Data Science Methodology**

Model evaluation is a crucial step in the **Data Science Methodology** because it ensures that the model **meets the project's goals and performs well** before deployment. Evaluation helps answer:

- $\checkmark$  Does the model work as expected?
- $\checkmark$  Does it solve the original problem?
- $\varnothing$  Should it be improved?

## Two Main Phases of Model Evaluation

# **Diagnostic Measures (Checking Model Performance)**

- Ensures that the model is making **meaningful predictions**.
- Uses techniques like **to** measure the model's effectiveness:
  - o decision trees,
  - confusion matrices,
  - o accuracy scores,
  - o and error analysis.
- Helps find where adjustments are needed.

### **\*** Example from the Case Study:

- The decision tree model was tested at different **relative misclassification costs** (1:1, 9:1, 4:1).
- The goal was to balance **sensitivity** (correctly predicting "Yes" readmissions) and **specificity** (correctly predicting "No" readmissions).

## **2** \$tatistical Significance Testing

- Ensures that the data is handled correctly and the model's findings are **not random**.
- Helps avoid second-guessing the model's results.
- Can use **p-values**, **confidence intervals**, **and hypothesis testing** to verify results.

#### **\*** Example from the Case Study:

- If the data scientist only looked at **overall accuracy**, the model might seem **good but** misleading.
- Instead, **specific metrics like sensitivity and specificity** were analyzed to ensure the model actually works.

# Finding the Optimal Model with the ROC Curve

#### **★** What is the ROC Curve?

- The Receiver Operating Characteristic (ROC) curve is a tool for evaluating classification models.
- It plots the **true positive rate** (sensitivity) against the **false positive rate** for different model settings.
- The best model has the largest separation from the baseline (random guessing).

#### **Example from the Case Study:**

• Model 3 (4:1 relative cost) had the best ROC curve, meaning it provided the best trade-off between sensitivity and specificity.

# **Final Takeaways from Model Evaluation**

- ✓ Evaluation is an iterative process: You keep testing and improving the model.
- ✓ The goal is not just high accuracy but a balanced trade-off between false positives and false negatives.
- ✓ **ROC curves help find the best model** by comparing different versions.
- ✓ The best model for the problem depends on the real-world impact (e.g., in healthcare, missing a high-risk patient is worse than an unnecessary intervention).