

# Fall 2024 Case Competition

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# Introduction

## Overview

Ocean Crest Hotel faces revenue loss and disruptions due to high booking cancellations, needing a data-driven solution to predict and reduce them.

## Business Problem

Ocean Crest Hotel faces a 37% booking cancellation rate, disrupting revenue forecasting, resource planning, and operational efficiency.

## Solution

Using predictive analytics to improve processes, optimize distribution, and enhance customer management.

# Key Findings



**Non-Refundable  
Deposits (Highest  
Impact)**



**Online marketing  
channel**



**Average Daily Rate  
(ADR)**



**Cancellations: Previous  
and Non-Cancellations**



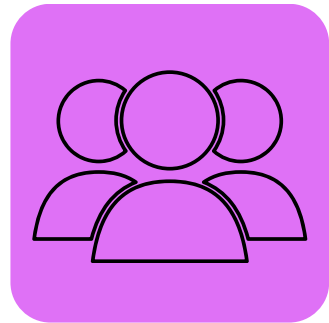
**Lead Time – Short and  
Long**



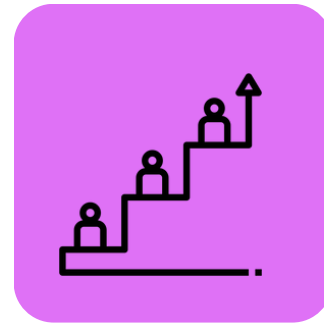
**Customer Requests:  
Special Requests, Parking  
requests etc.**



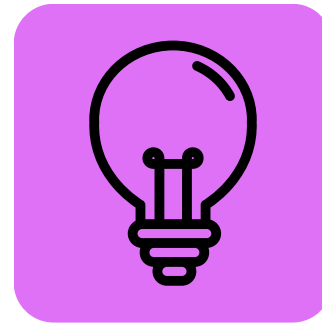
# Recommendations



**Loyalty Program  
2.0**



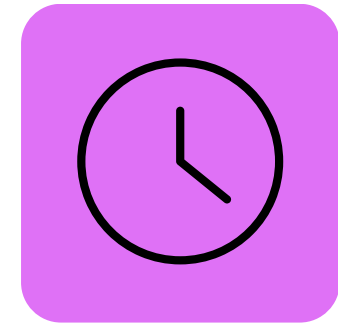
**Tiered Deposit  
System**



**Dynamic Pricing  
Strategy**



**OTA Booking  
Enhancement**



**Lead Time**

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# Business Improvement

01

## Revenue

Projected 28% year-over-year increase

02

## Operations

Improved efficiency and resource management

03

## Guest experience

Enhanced satisfaction and personalization

04

## Distribution

Reduced dependence on high-cancellation platforms

05

## Cancellation Rate

A reduction in the cancellation by 15% in the first year

# Implementation Roadmap

## Short-term (0–3 months)

- Launch a tiered deposit system

## Medium-term (3–6 months)

- Introduce dynamic pricing strategy

## Long-term (6–12 months)

- Optimize OTA partnerships and direct booking channels



# Conclusion

- A data-driven approach to tackle cancellation challenge
- Holistic strategy addressing multiple factors
- Potential for significant revenue growth and operational improvement



# What's Next?

- Prioritize recommendations
  - Allocate resources
  - Set implementation timeline
-



# The Engine Driving the Solution


# Data Overview & Preparation

## Data Overview

- **Total Observations:**  
112,941 rows
- **Total Features:** 31  
columns
- **Period:** Data covers  
bookings from 2015 to  
2017.

## Key Column

- **Is Canceled - Response**
- **ADR** (Average Daily Rate)
- **Lead Time**
- **Special Request**
- **Booking Channel**
- **Country**



# Data Preparation

## Data Cleaning

- Handling of Missing Values
- Handling of Outliers
- Improving data categorization

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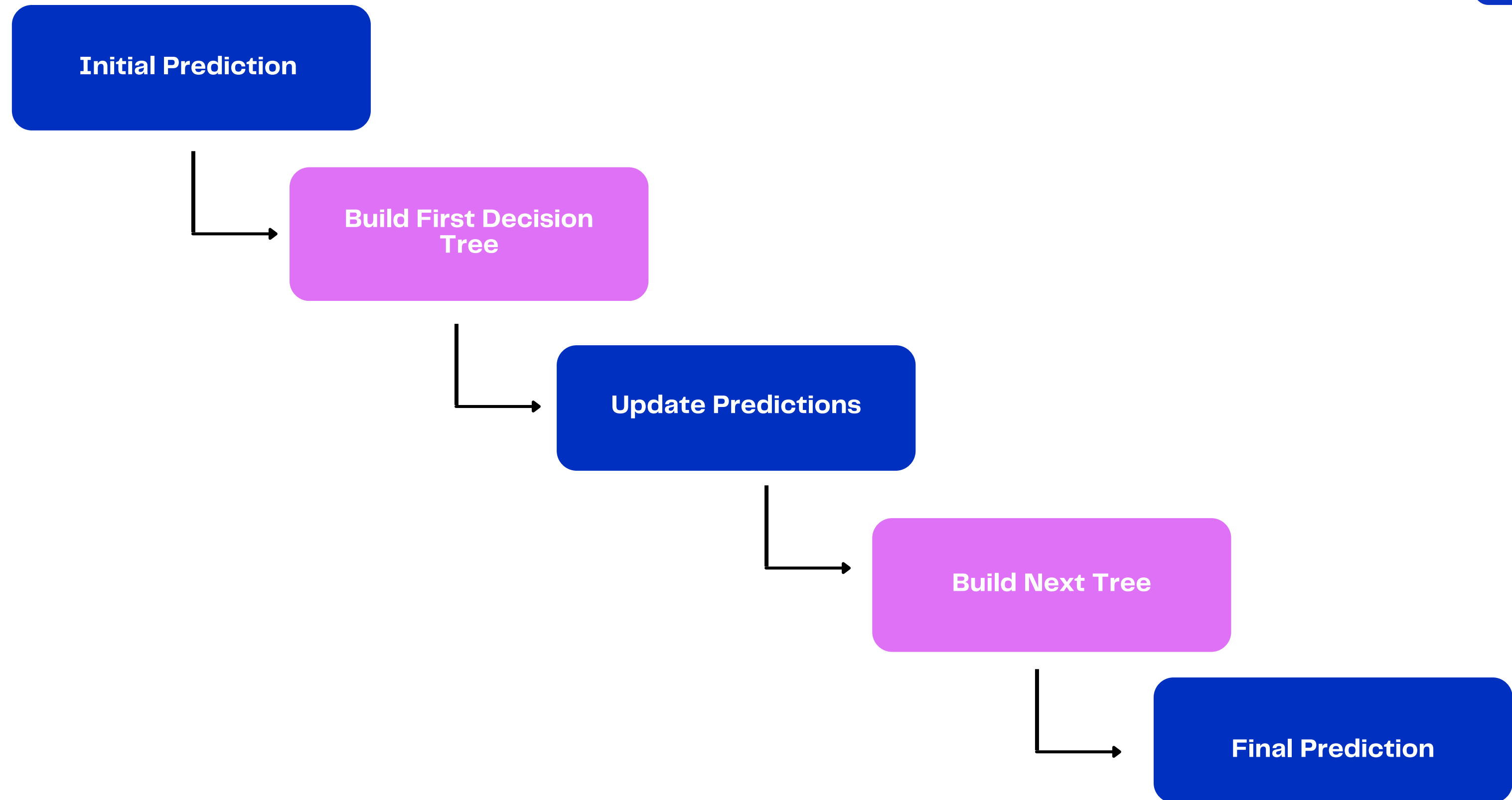
## Feature Engineering

- Geographical Aggregation
- Variable Aggregation
- Dimensionality Reduction
- New Variables

# Model choice

Models	Training AUC	Testing AUC	Pro	Cons
Decision Tree	0.660	0.73	Interpretable, handles non-linear relationships	Prone to overfitting
Random Forest	0.8980	0.825	Good performance, handles feature interactions	Less interpretable than single trees
XGBoost	0.9260	0.925	Best performance, handles complex relationships	Requires careful tuning, less interpretable
LightGBM	0.9000	0.825	Fast training, handles large datasets	Slightly lower performance than XGBoost

# How XGBoost Works



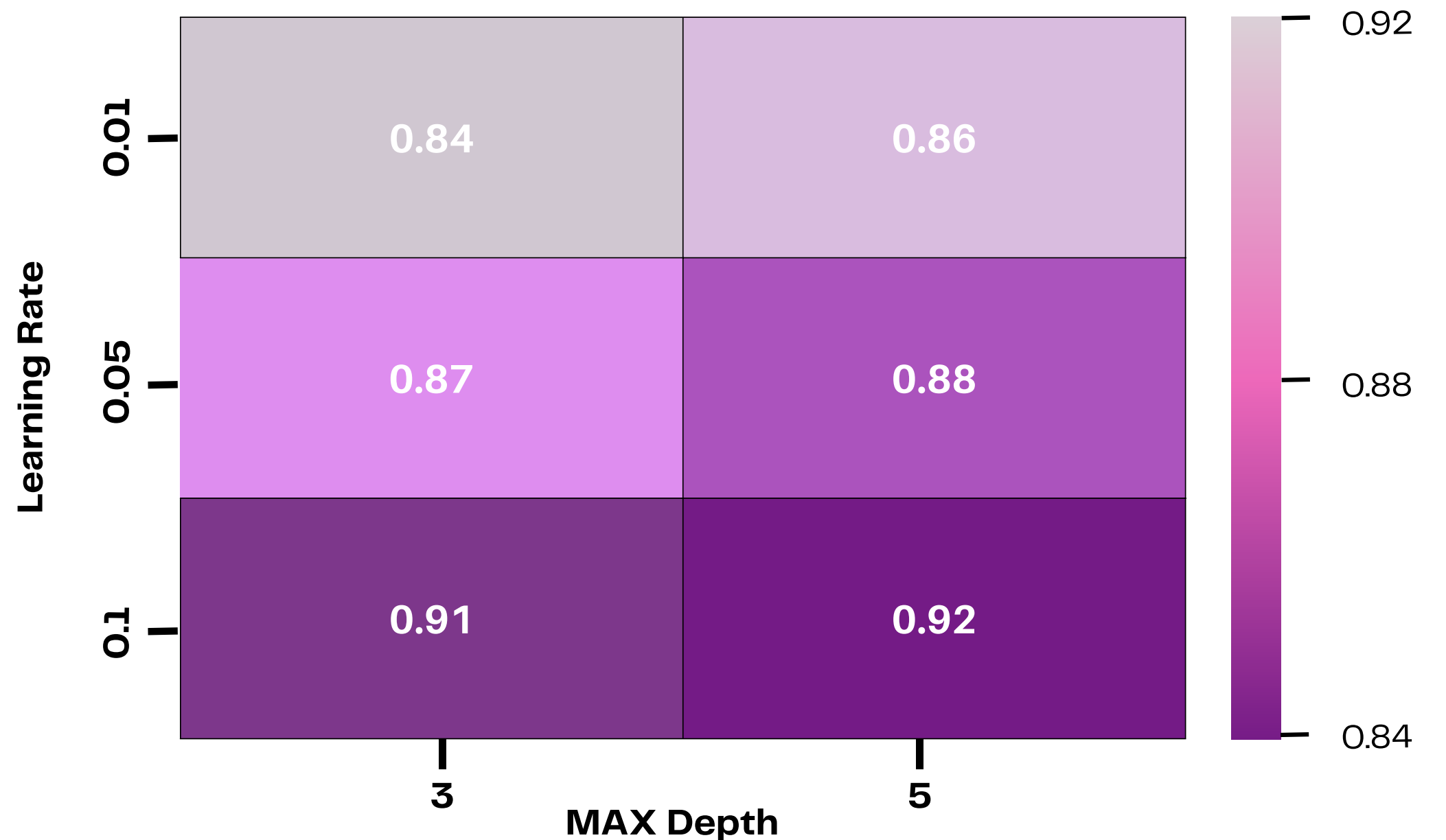
# Avoiding Overfitting

Use of ensemble  
methods

Cross-validation

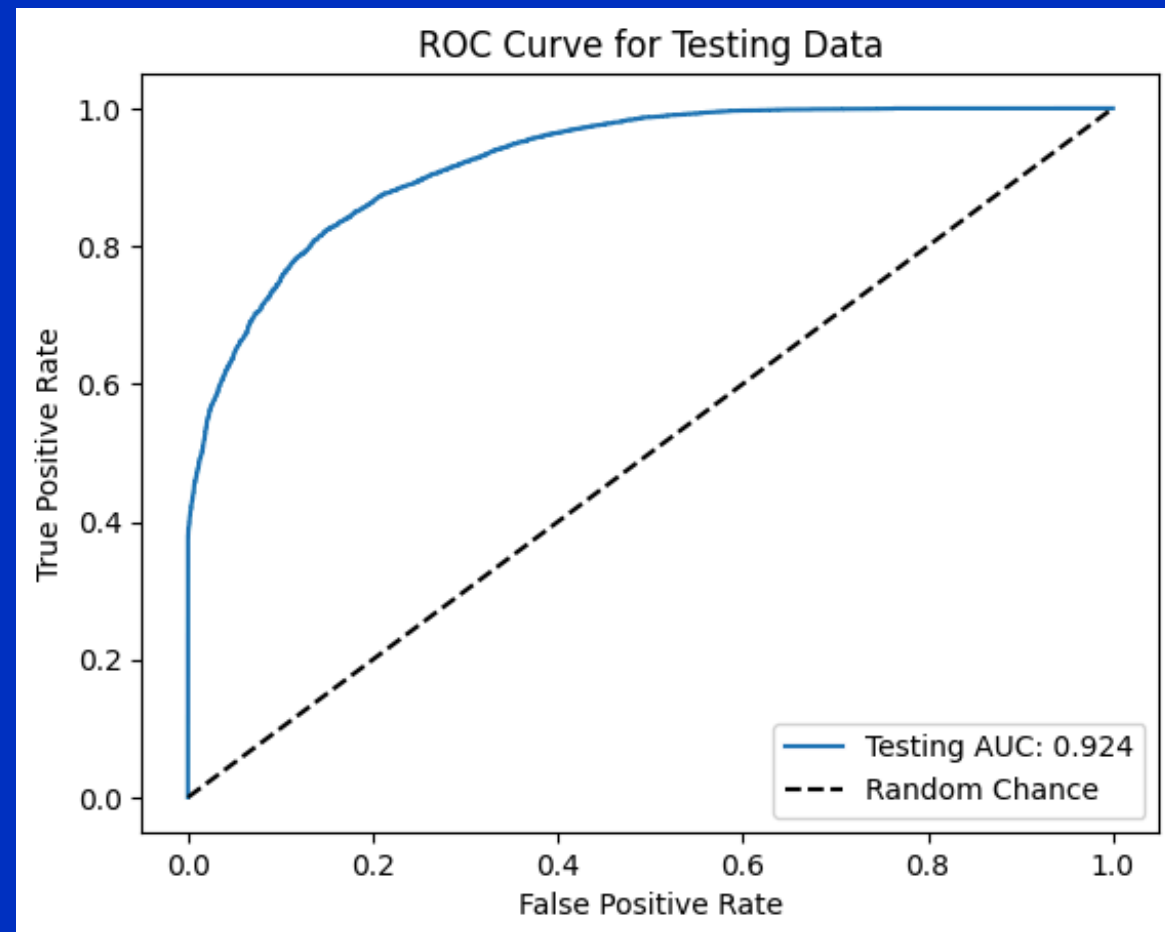
Feature Engineering and  
Selection

XGBoost In-Built  
Regularization

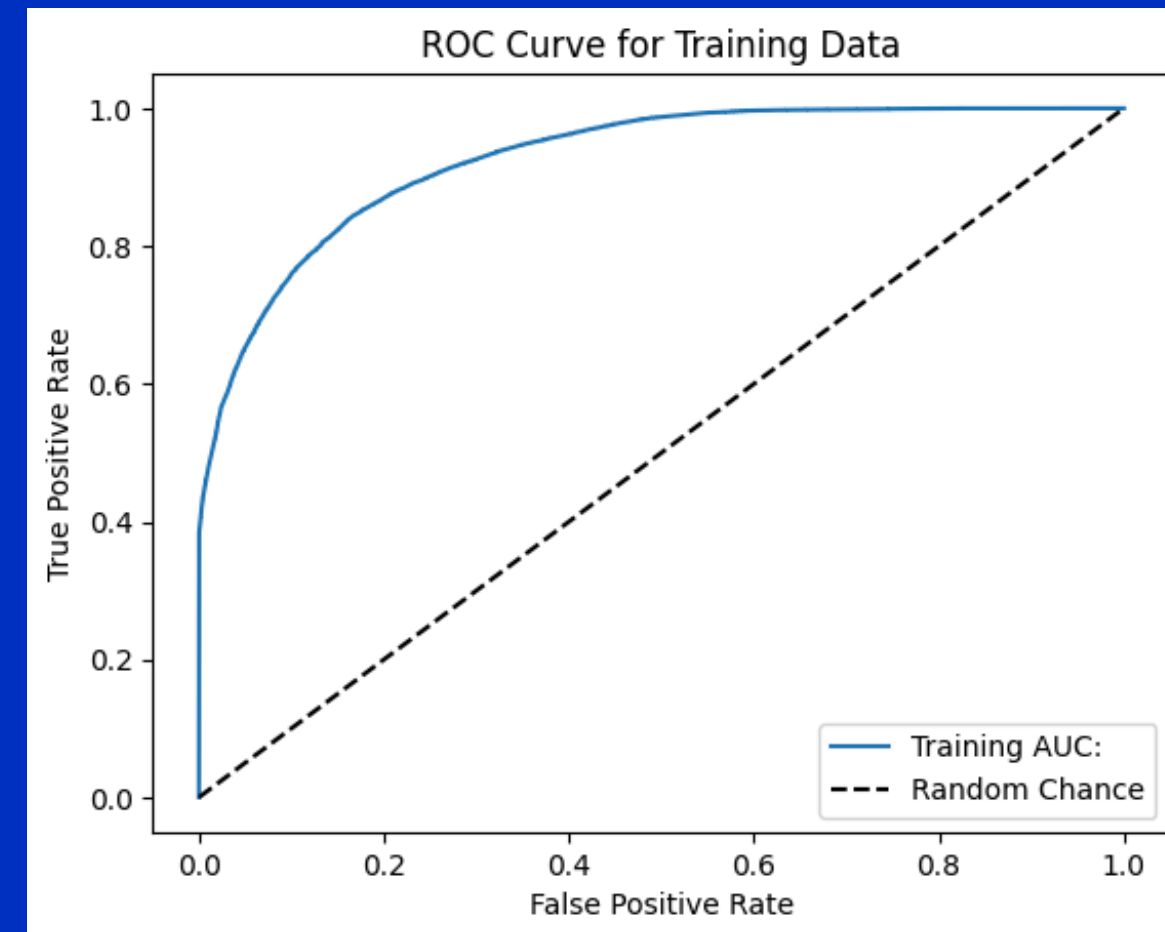




# Model Evaluation – AUC Scores

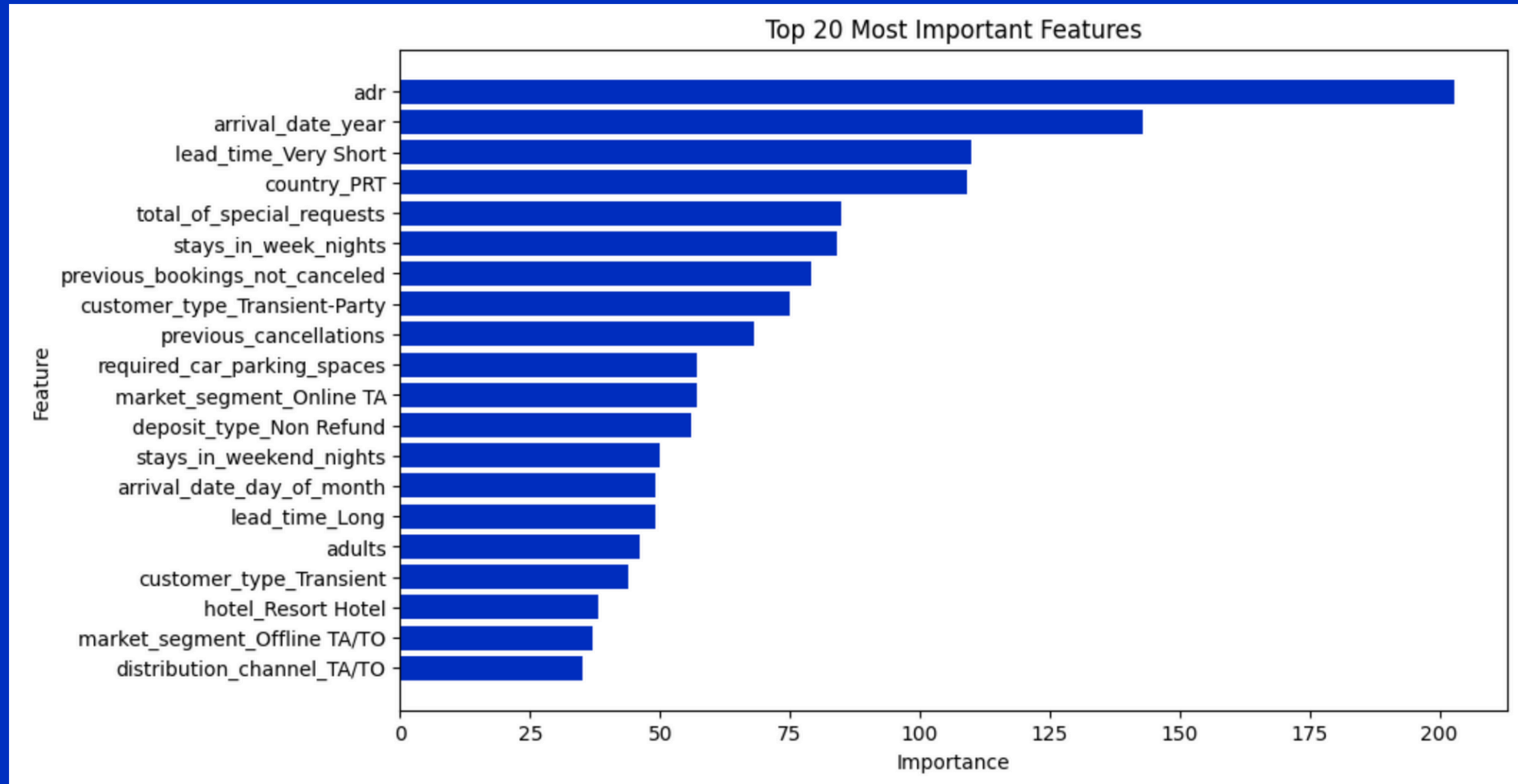


**AUC for Testing Data: 0.924**



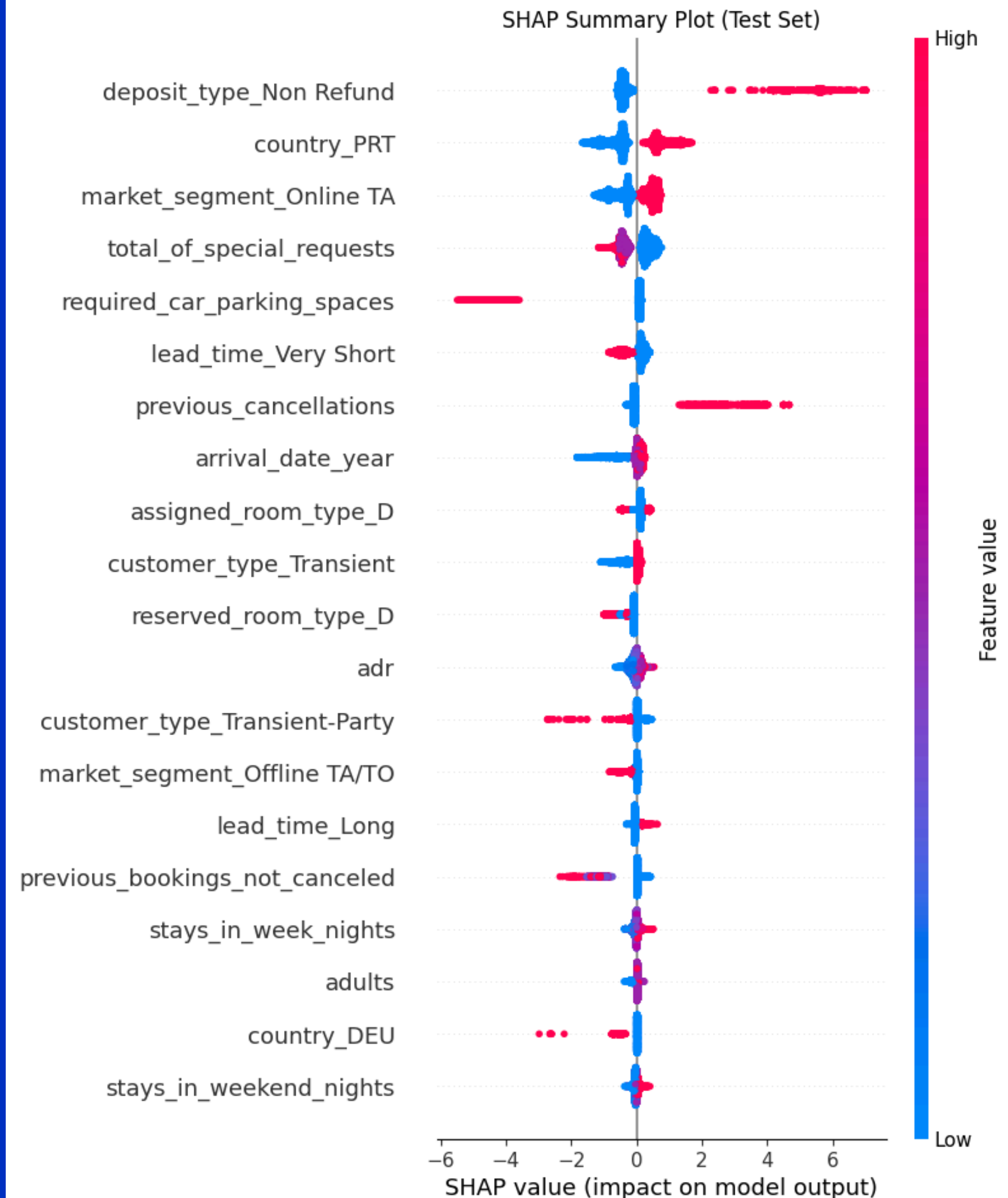
**AUC for Training Data: 0.926**

# Feature Importance



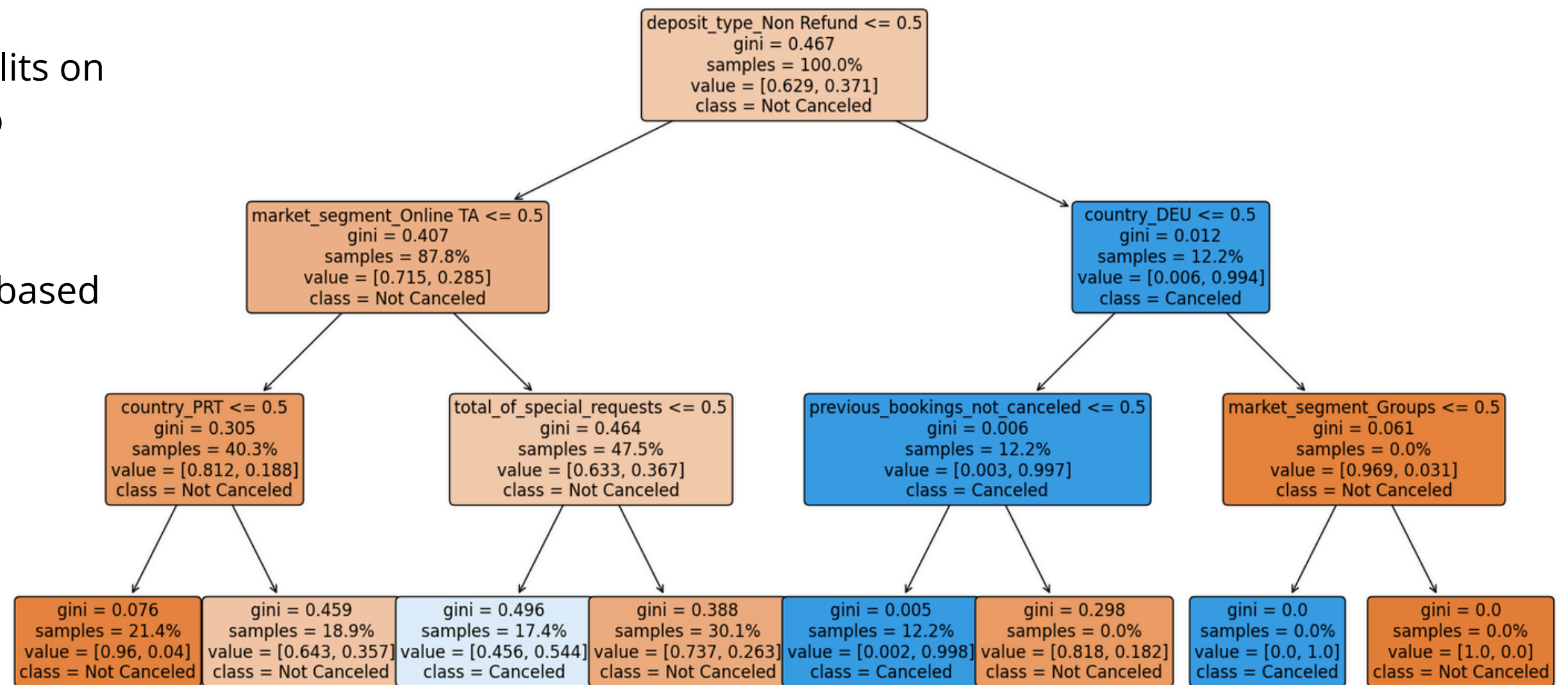
# SHAP Value Analysis

- SHAP values provide insight into how individual predictions are made.
- ADR and Lead Time are the most influential in pushing bookings toward cancellation or non-cancellation.
- SHAP analysis ensures transparency in model decisions.



# Decision Tree Visualization

- A decision tree in the XGBoost model splits on features like ADR and Booking Source to predict cancellation.
- Each split further refines the prediction based on the most important factors.



# Question?

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