

HOTEL BOOKING DEMAND

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OVERVIEW & OBJECTIVES



PROBLEM STATEMENT

Building a machine learning model that classify booking statuses accurately can help hotels plan for:

- Refund policies
- Staffing schedules
- Targeting customers with offers and discounts



DATASET

The dataset consists of 119,390 observations with 32 features.



| | hotel | is_canceled | lead_time | arrival_date_year | arrival_date_month | arrival_date_week_number | arrival_date_day_of_month | stays_in_weekend_nights | stays_ |
|---|-----------------|-------------|-----------|-------------------|--------------------|--------------------------|---------------------------|-------------------------|--------|
| 0 | Resort Hotel | 0 | 342 | 2015 | July | 27 | 1 | 0 | |
| 1 | Resort Hotel | 0 | 737 | 2015 | July | 27 | 1 | 0 | |
| 2 | Resort Hotel | 0 | 7 | 2015 | July | 27 | 1 | 0 | |
| 3 | Resort Hotel | 0 | 13 | 2015 | July | 27 | 1 | 0 | |
| 4 | Resort Hotel | 0 | 14 | 2015 | July | 27 | 1 | 0 | |

5 rows × 32 columns

EXPLANATORY DATA ANALYSIS



DATA CLEANING

Checking features Types

- Features with Incorrect Types:4
- Handling techniques:
 - Change to object
 - Change to integer

- Features with Missing Values:4
- Handling techniques:
 - Column dropping
 - Rows dropping
 - Imputing with mean
 - and mode

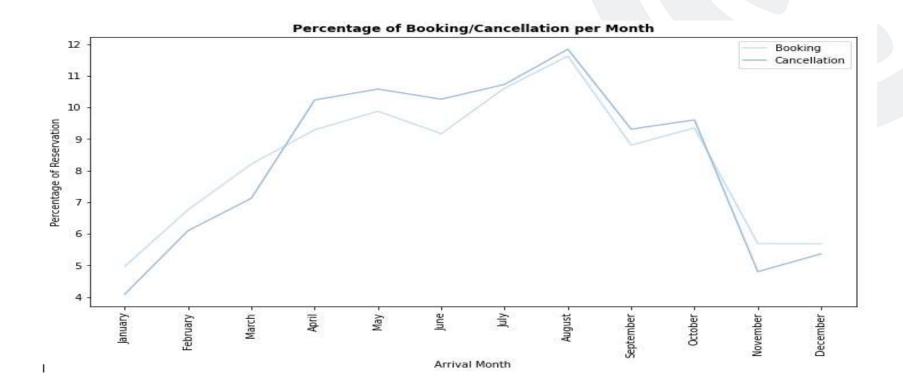
Exploring & Handling the Missing values

HANDLING OUTLIERS

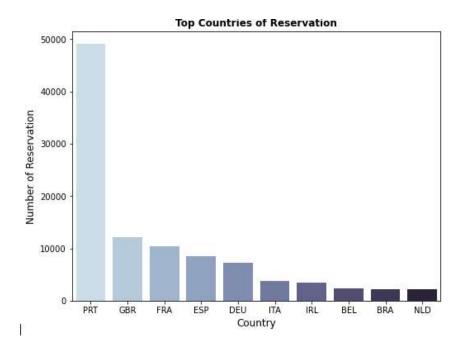
Features with outlier: 1 Handling techniques:

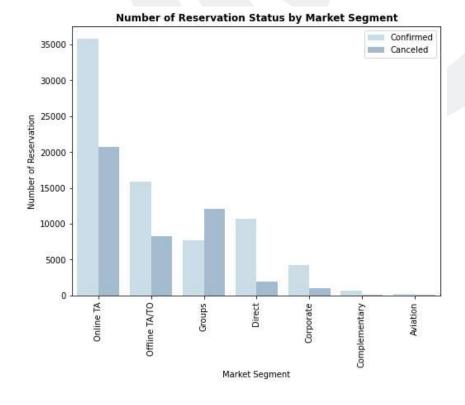
• Dropping of rows with outlier

DATA EXPLORATION

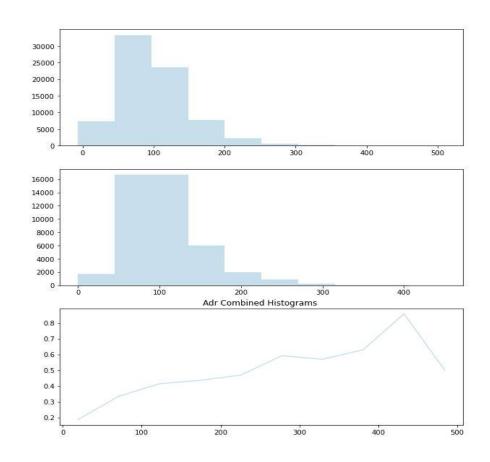


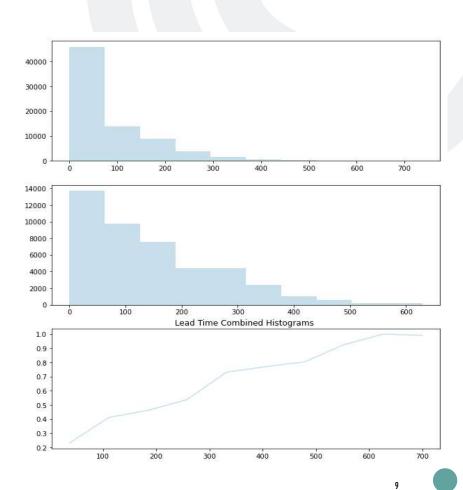
DATA EXPLORATION





DATA EXPLORATION





MODEL BUILDING AND EVALUATION



FEATURE ENGINEERING AND SELECTION

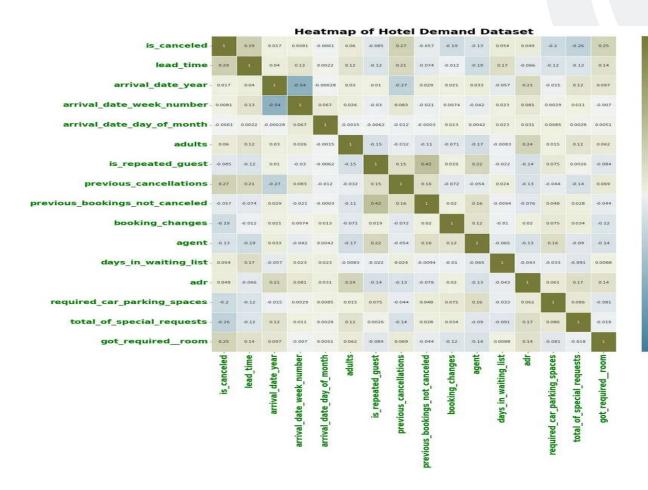
FEATURES ENGINEERING

Adding feature
Changing feature
Numerical Features
Scaling
Encoding Categorical
Features

Select feature based on their importance

FEATURES SELECTION

FEATURE CORRELATION MATRIX



0.75

0.25

-0.25

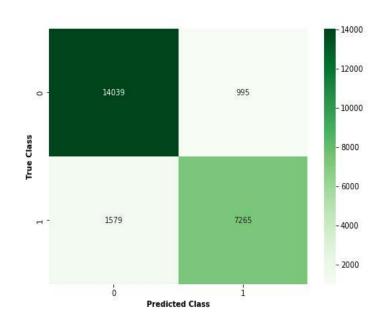
MODEL SELECTION

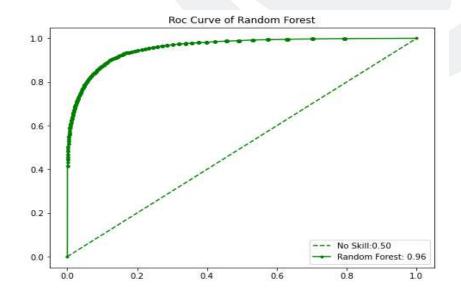
| Model | Precision | Recall | F-Macro (Cross Validation) | F-Macro (Holdout) |
|---------------------------------|-----------|--------|-------------------------------|----------------------|
| Logistic Regression (LR) | 0.81 | 0.68 | 0.80 | 0.80 |
| Naïve Base (NB) | 0.80 | 0.54 | 0.74 | 0.75 |
| K Nearest Neighbor (KNN) | 0.87 | 0.68 | 0.82 | 0.83 |
| Support Vector Machine (SVM) | 0.82 | 0.67 | 0.80 | 0.80 |
| Random Forest (RF) | 0.88 | 0.81 | 0.88 | 0.89 |

Model Performance Using 5 Fold Cross Validation & Holdout

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RANDOM FOREST ANALYSIS





CONCLUSION



CONCLUSION

- Dataset requires cleaning and preparation
- The most important features are:
 - lead_time
 - total_of_special_requests
 - required_car_parking_spaces
 - booking_changes
- Best Model: Random Forest with F-Macro: 0.89



THANK YOU



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