

Data Analysis of Hotel Booking

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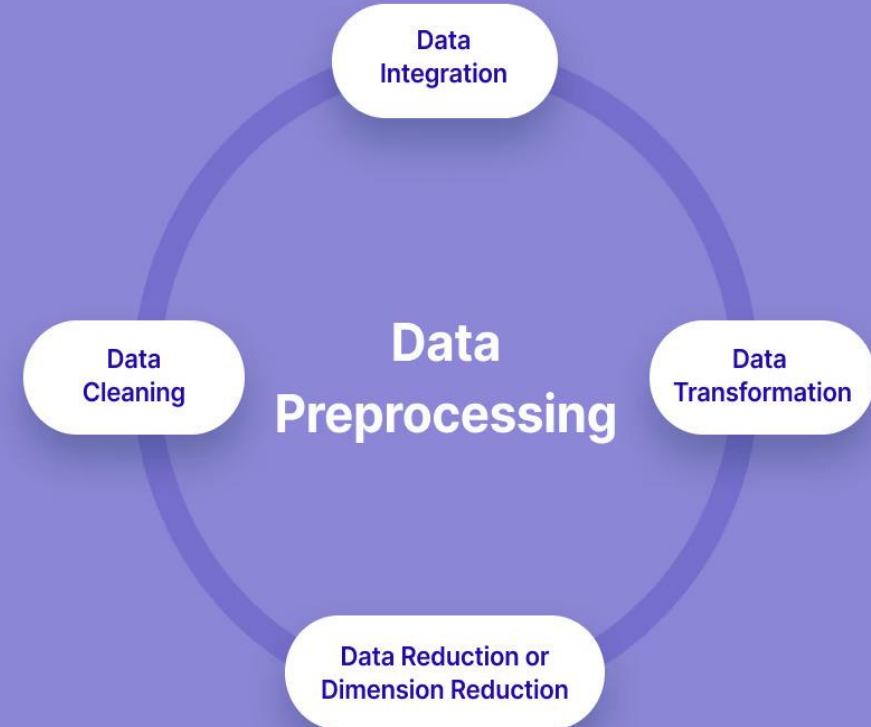
Problems Addressed



- Demand Forecasting
- Customer Segmentation
- Cancellation Analysis
- Optimal Room Allocation

Pre-Processing Data

- Data:** Downloaded hotel dataset (119k records, 33 attributes).
- Database:** Imported into PostgreSQL.
- Cleaning:** Handled missing data, standardized values.
- Transformation:** Normalized, filtered, and removed outliers.
- Validation:** Verified with Weka.
- Result:** Clean, ready dataset.



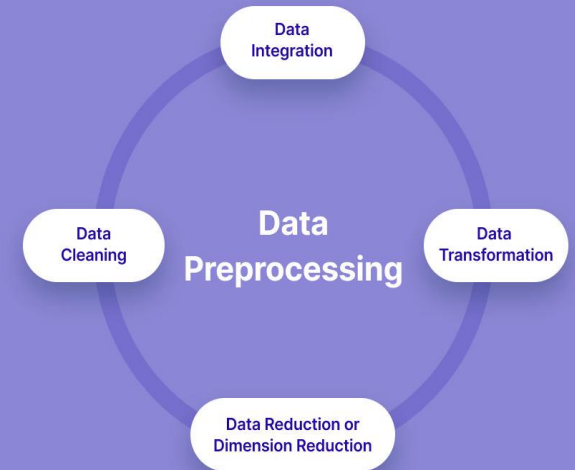
Pre-Processing Data

Approach:

- **"Company" Column:** Removed due to **112,593 (94%)** missing values, which could lead to biased or unreliable results.
- **"Children" Column:** Imputed missing values (4, 0%) using the **"ReplaceMissingValues"** filter in Weka, as the missing data was minimal, and imputation helped maintain the dataset's distribution.

Justification:

- Removing the **"company"** column eliminated unnecessary noise from the dataset, ensuring a more reliable analysis.
- Imputing the 'children' column with minimal missing values preserved the dataset's integrity, while the 'agent' and 'country' columns were processed similarly by imputing missing values with the mode, as they were categorical features with minimal missing data.



Data Analytics

K-Means Clustering

•Why K-Means Clustering ?

- ✓ Efficient for large datasets
- ✓ Scalable to multiple attributes
- ✓ Clear and interpretable clusters
- ✓ Simple and flexible algorithm

K-Means Clusters for the Iris Dataset

Attribute	Full Data (119390.0)	0 (20627.0)	1 (11706.0)	2 (3810.0)	3 (3615.0)	4 (20624.0)	5 (38319.0)	6 (20689.0)
lead_time	104.0114	177.185	288.7355	30.7864	422.2268	46.2948	18.6988	99.9681
stays_in_weekend_nights	0.9276	1.3959	0.9593	0.4478	0.574	2.0971	0.3339	0.5268
stays_in_week_nights	2.5003	3.3621	2.8574	1.4782	2.1046	3.2965	1.754	2.2849
adults	1.8564	1.9283	1.9583	1.3892	1.9001	1.9206	1.7635	1.9135
is_repeated_guest	0.0319	0	0	1	0	0	0	0
previous_cancellations	0.0871	0.0414	0.4934	0.4698	0.0537	0.0128	0.0176	0.0409
previous_bookings_not_canceled	0.1371	0.0139	0.007	3.585	0.0047	0.0057	0.0501	0.0139
days_in_waiting_list	2.3211	3.7288	1.9979	0.1664	32.2515	0.2898	0.1284	2.354
adr	101.8311	108.3132	89.2317	64.446	79.4557	107.7783	102.4016	106.3065

Time taken to build model (full training data) : 3.29 seconds

=== Model and evaluation on training set ===

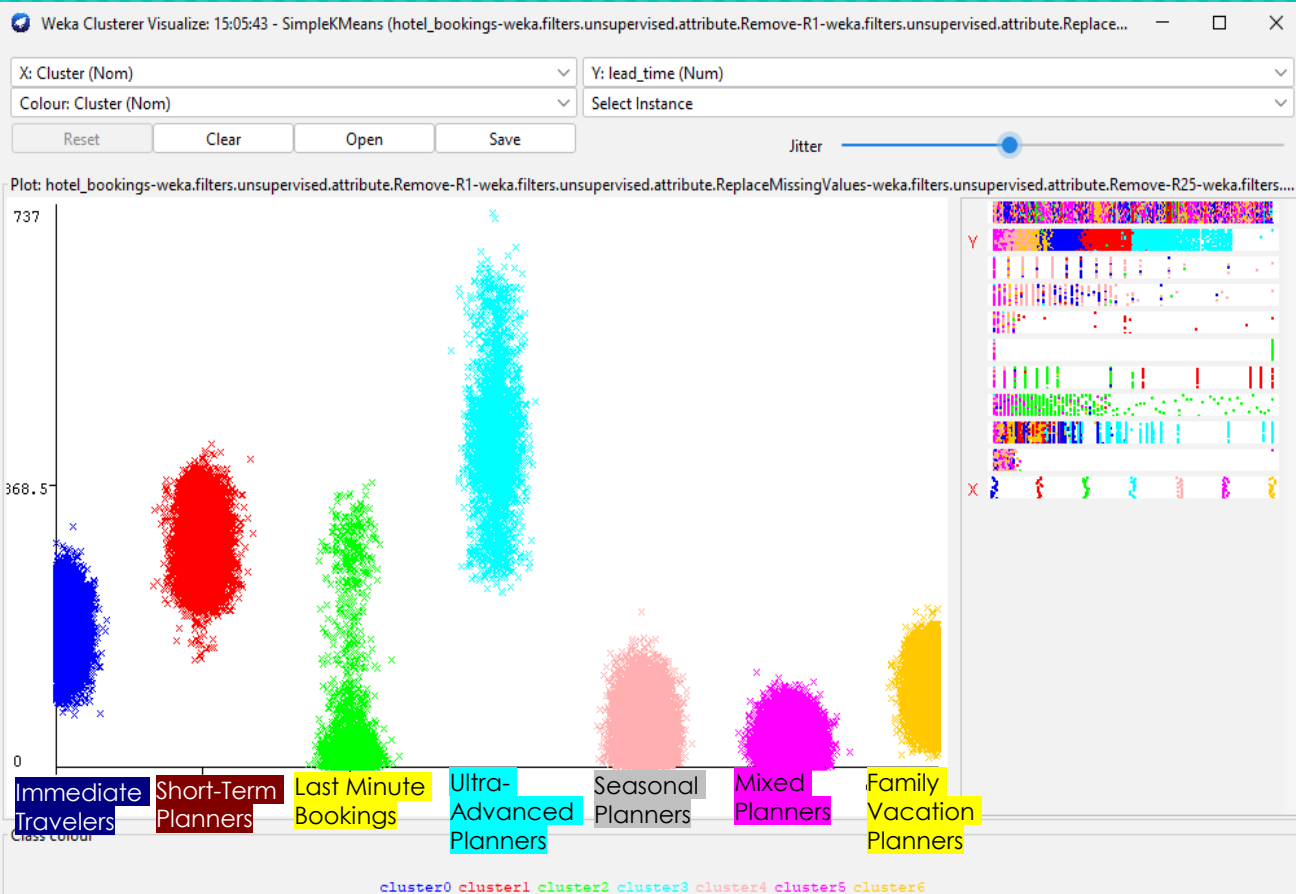
Clustered Instances

0	20627 (17%)
1	11706 (10%)
2	3810 (3%)
3	3615 (3%)
4	20624 (17%)
5	38319 (32%)
6	20689 (17%)

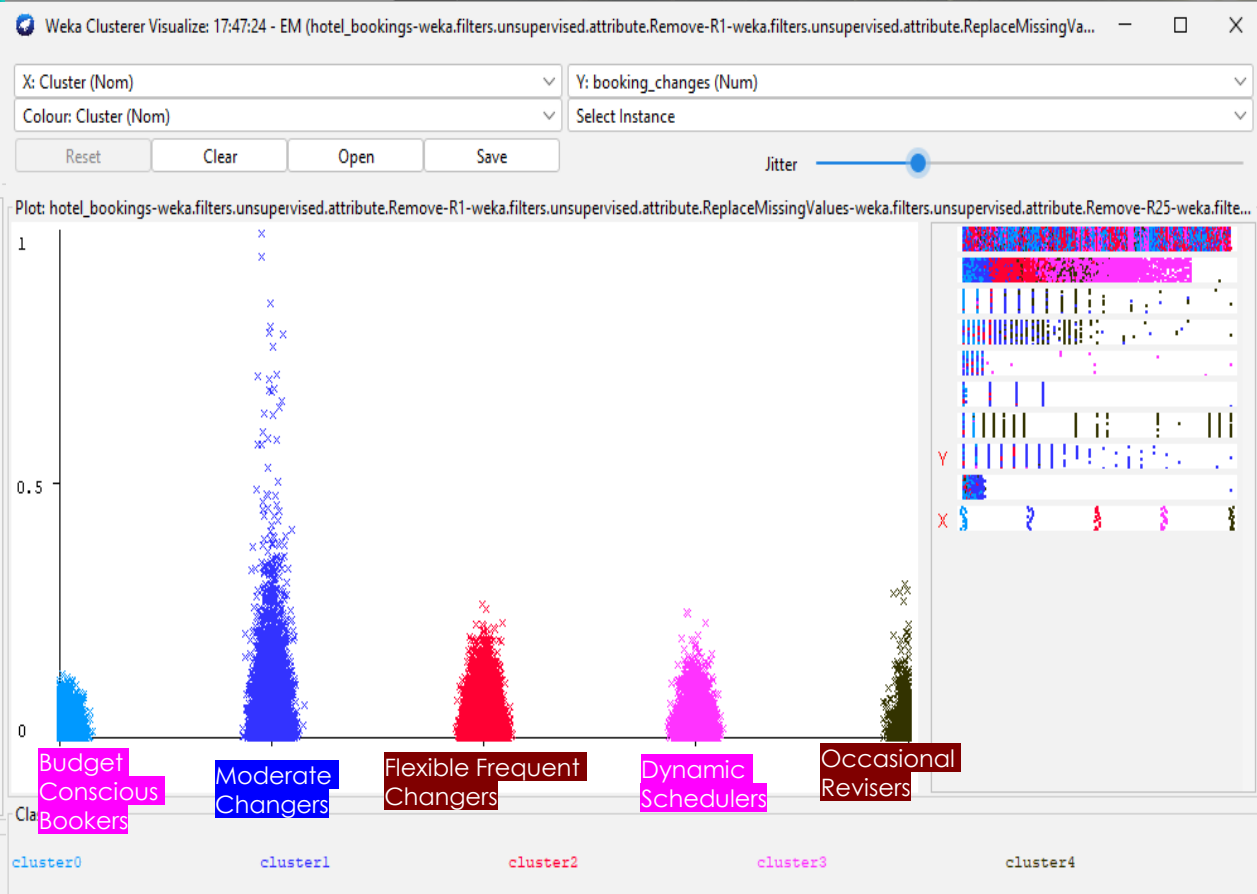
- Key patterns identified in **ADR (Average Daily Rate)**, **booking changes**, and **stay duration**.
- Actionable insights:** Tailor pricing and promotions for each segment.

K-Means Clustering

Clustering of Hotel Bookings Based on Lead Time



Clustering of Hotel Bookings Based on Booking Changes



Data Analytics

Linear Regression

The screenshot shows the Weka Explorer interface with the 'Classify' tab selected. The 'SimpleLinearRegression' classifier is chosen. The 'Test options' section shows 'Percentage split' at 70%. The 'Classifier output' pane displays the model equation and evaluation metrics.

Classifier output

```
-0.0019 * meal=SC +  
-0.0047 * is_repeated_guest +  
-0.0053 * previous_cancellations +  
-0.0211 * reserved_room_type=P,H,A,G,F,B,C,D,E +  
0.0292 * reserved_room_type=H,A,G,F,B,C,D,E +  
-0.0141 * reserved_room_type=A,G,F,B,C,D,E +  
0.0113 * reserved_room_type=G,F,B,C,D,E +  
-0.0022 * reserved_room_type=F,B,C,D,E +  
-0.0106 * reserved_room_type=B,C,D,E +  
0.0084 * reserved_room_type=C,D,E +  
-0.0034 * reserved_room_type=D,E +  
0.0022 * reserved_room_type=E +  
0.0034 * booking_changes +  
0.0027 * deposit_type=Refundable,No Deposit +  
-0.0032 * deposit_type=No Deposit +  
0.0038 * days_in_waiting_list +  
0.0013 * customer_type=Transient,Group,Contract +  
-0.0018 * customer_type=Group,Contract +  
0.0009 * total_of_special_requests +  
0.0274
```

Time taken to build model: 1.44 seconds

=== Evaluation on test split ===

Time taken to test model on test split: 0.16 seconds

=== Summary ===

Correlation coefficient	0.7391
Mean absolute error	0.0043
Root mean squared error	0.006
Relative absolute error	64.1577 %
Root relative squared error	67.3731 %
Total Number of Instances	35817

- **Predicted ADR** using hotel factors
- **Key trends** city hotels, lead time, seasonality
- **Moderate performance** 0.7391 correlation
- **Improvement needed** error of 0.0043 units

Data Analytics

J-48 Decision Tree

Why J48 Decision Tree?

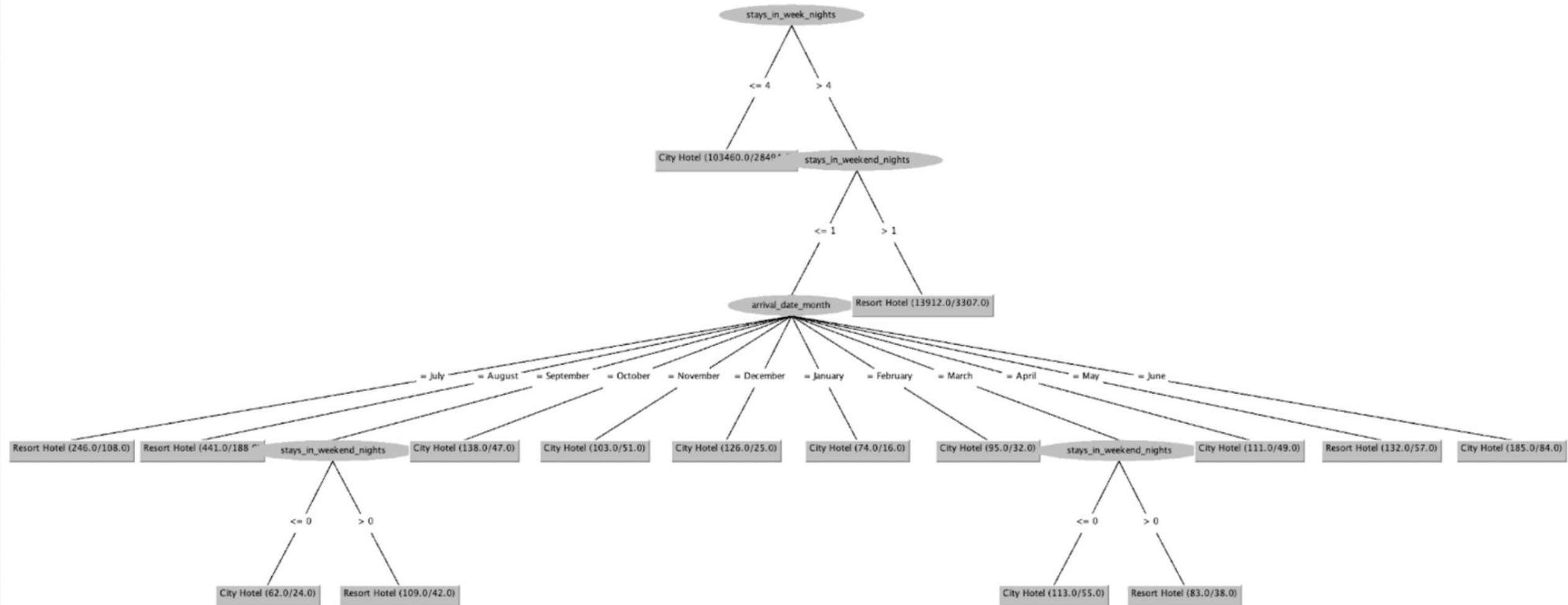
This decision tree will help:

- Predict the type of hotel booking based on stay patterns and seasonality.
- Understand patterns in guest behavior, such as which hotel is more popular during specific months or for specific durations of stay.
- Aid in marketing or operational decisions for hotels, like focusing offers on certain guest segments.

This decision tree focuses on predicting whether a hotel booking will be for a **City Hotel** or a **Resort Hotel** based on three main factors:

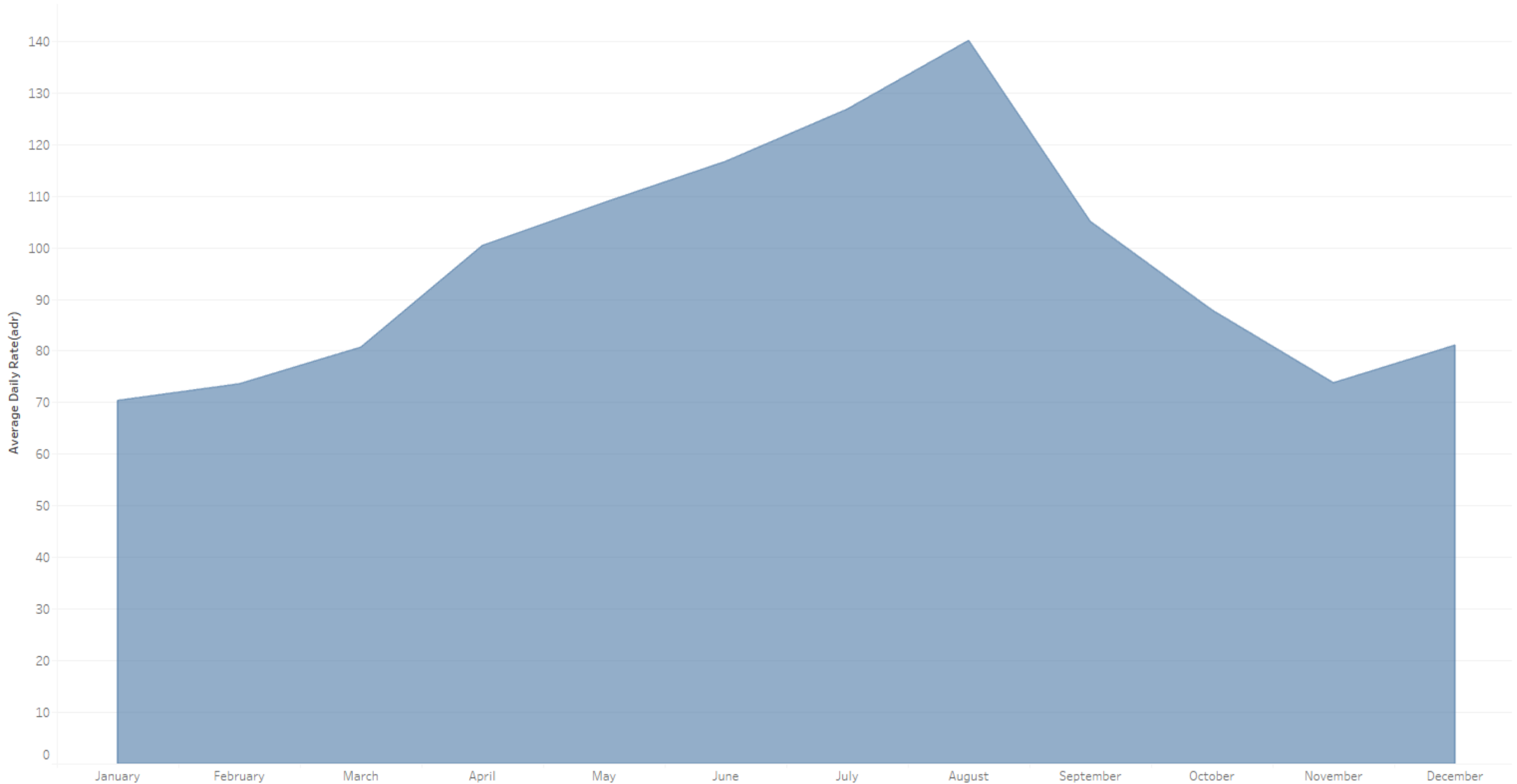
- 1.Length of Stay (Weekday Nights)
- 2.Length of Stay (Weekend Nights)
- 3.Arrival Month

J-48 Decision Tree



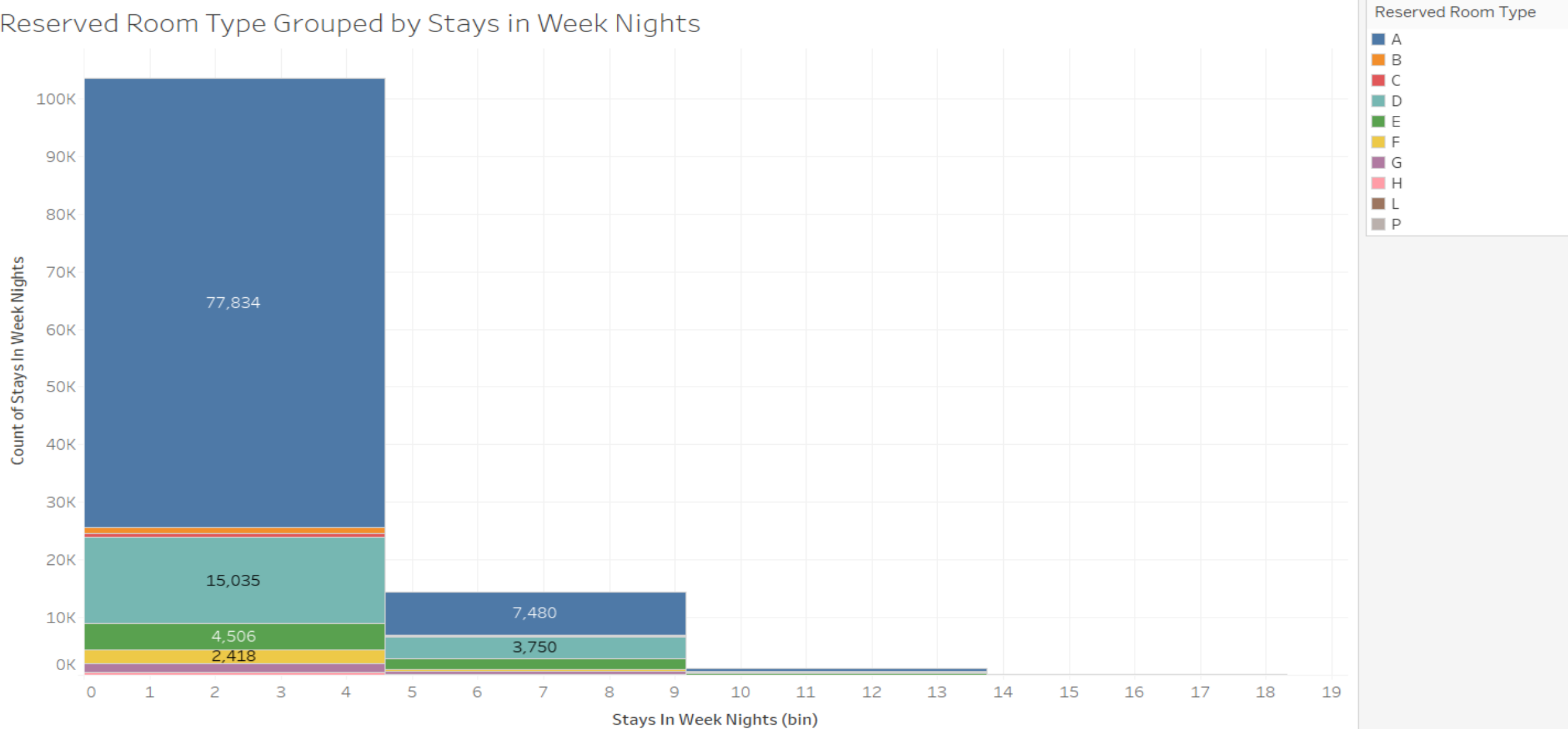
Monthly Trends in Average Daily Rate (ADR)

Arrival Date Month



Histogram

Reserved Room Type Grouped by Stays in Week Nights



Bar Chart

- City Hotels vs Resort Hotels:** City hotels consistently showed higher cancellation rates across all market segments. This could be due to factors like more frequent booking changes or higher business traveler volumes, where last-minute cancellations are more common.

- Bar charts allow us to effectively compare the cancellations of both city and resort hotels while also observing each market segment.



Impact of Lead Time on Average Daily Rate (ADR) Over Time

