



Hotel Demand and Cancellation Forecast

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Machine Learning and the Hotel Industry



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- Hotels have daily limited perishable inventory
 - Unsold rooms = Lost Revenue
 - Overselling rooms:
 - Pay for a comparable hotel
 - Transportation costs
 - Damage to reputation and loyalty



How Machine Learning Can Help

Revenue Management: “sell the right product to the right customer at the right time and at the right price”

Metrics of Success:

- Maximize Revenue Per Available Room (RevPAR)
- Reaching a “perfect sell”

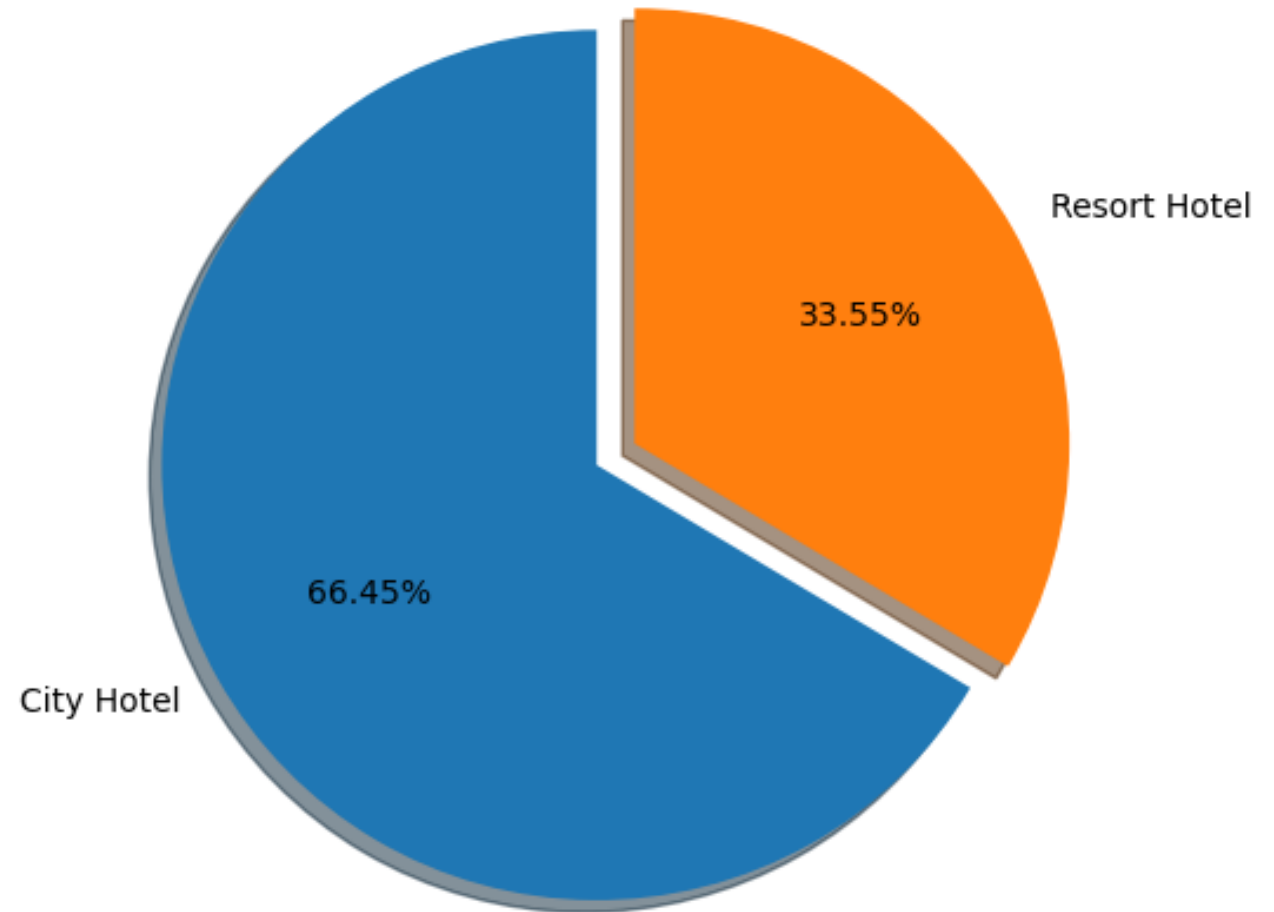
Predicting demand and cancellations to optimize pricing strategies

The Dataset

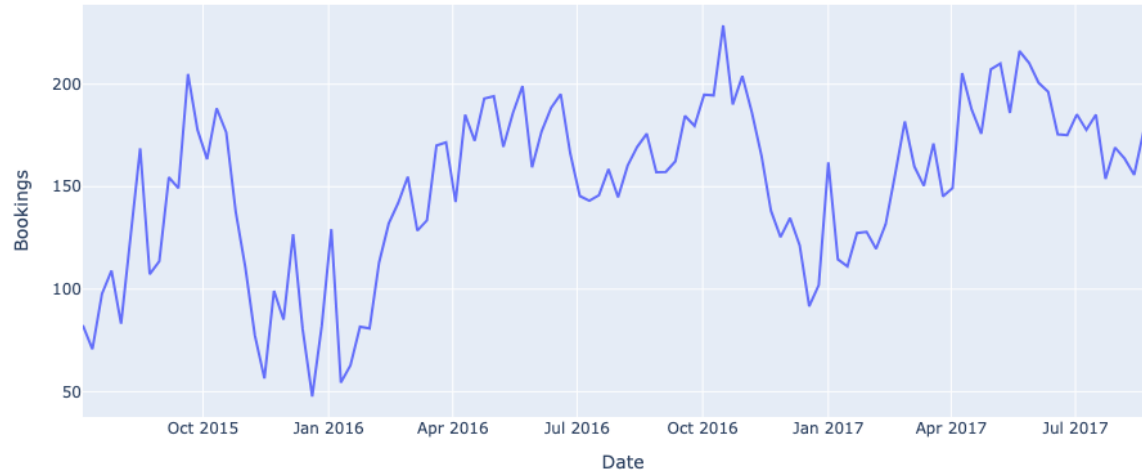
Important Columns:

- *is_canceled*
- *Arrival_date (year, month, day of month)*
- *Market_segment*
- *Lead_time*
- *Customer_type*

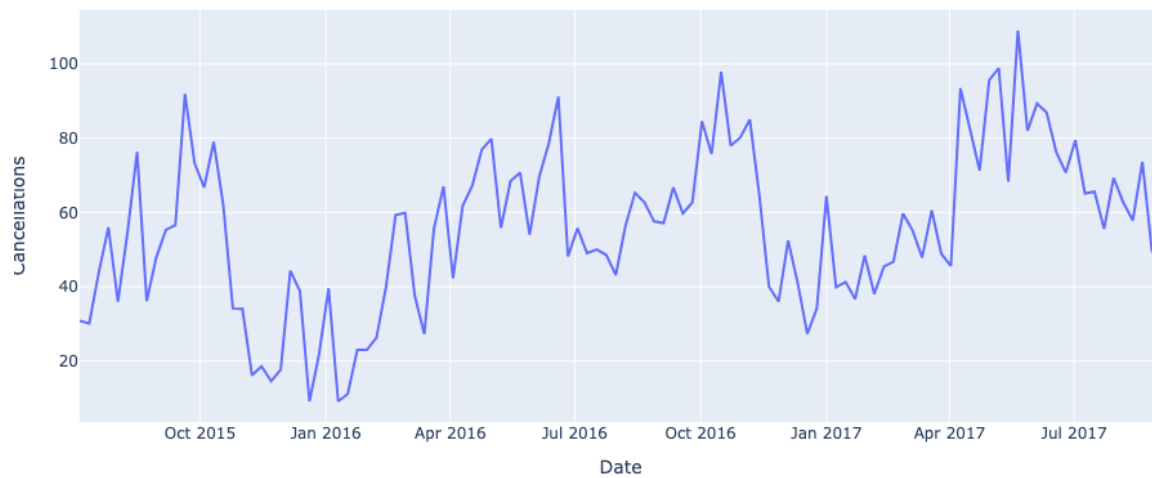
Percentage of Reservations by Hotel



Weekly Demand Over Time



Weekly Cancellations Over Time



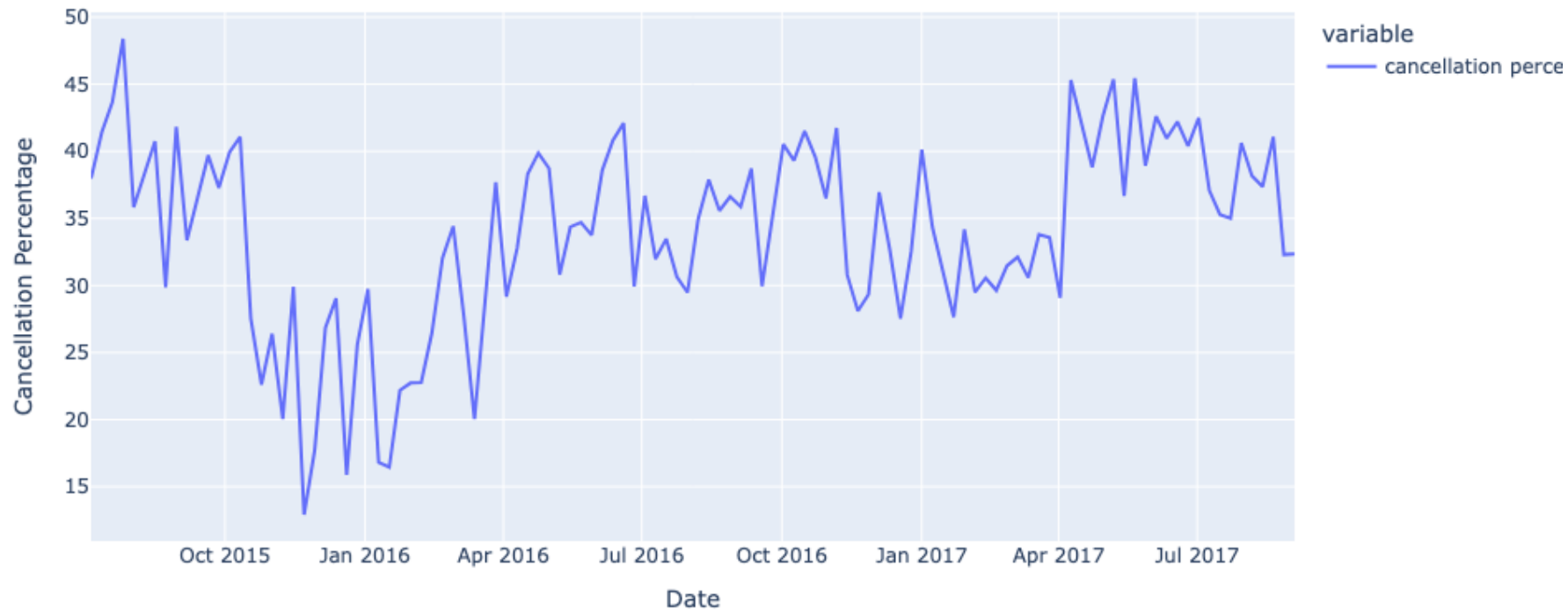
Initial EDA: Cancellation and Demand over Time

Seasonality seen between Demand
and Cancellations

- Peaks in October
- Valleys in January
- Increase in April

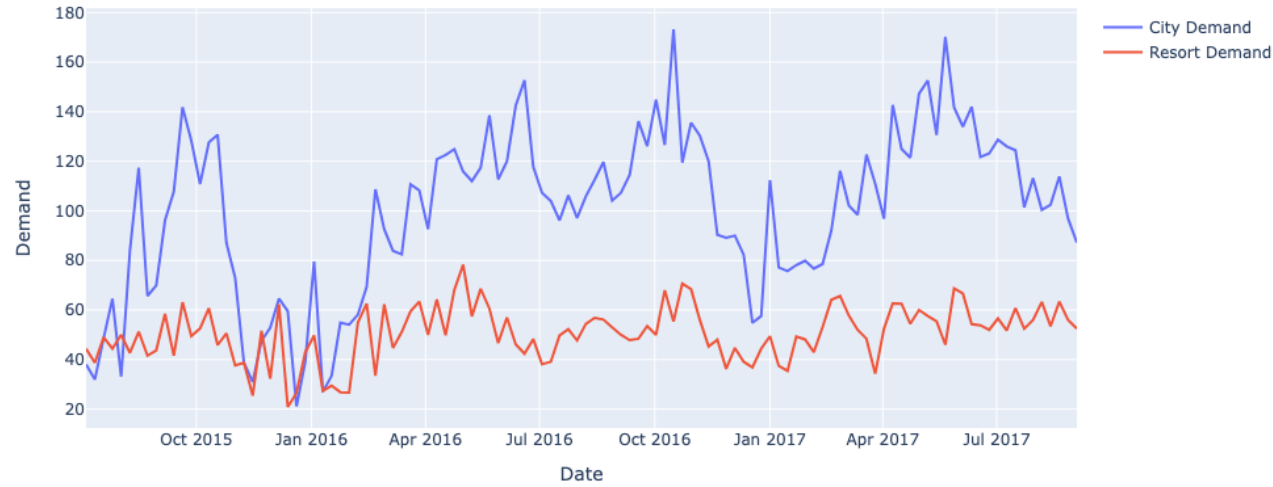
Cancellation Percentages

Weekly Cancellation Percentages Over Time

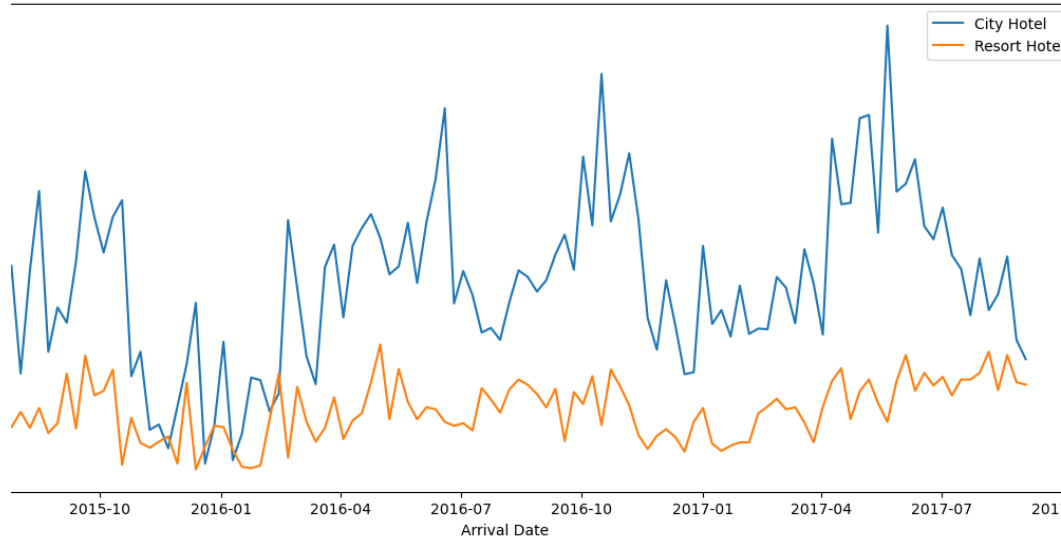


- High cancellation % overall
- Between 30-40% throughout the dataset
- Identified large cancelled group bookings during EDA

Demand by Hotel Over Time



Cancellations by Hotel Over Time



Demand and Cancellation by Hotel Type

City Hotel:

- Higher demand and cancellations
- Larger peaks and valleys
- Likely influences the overall datasets trends

Cancellation Trends:

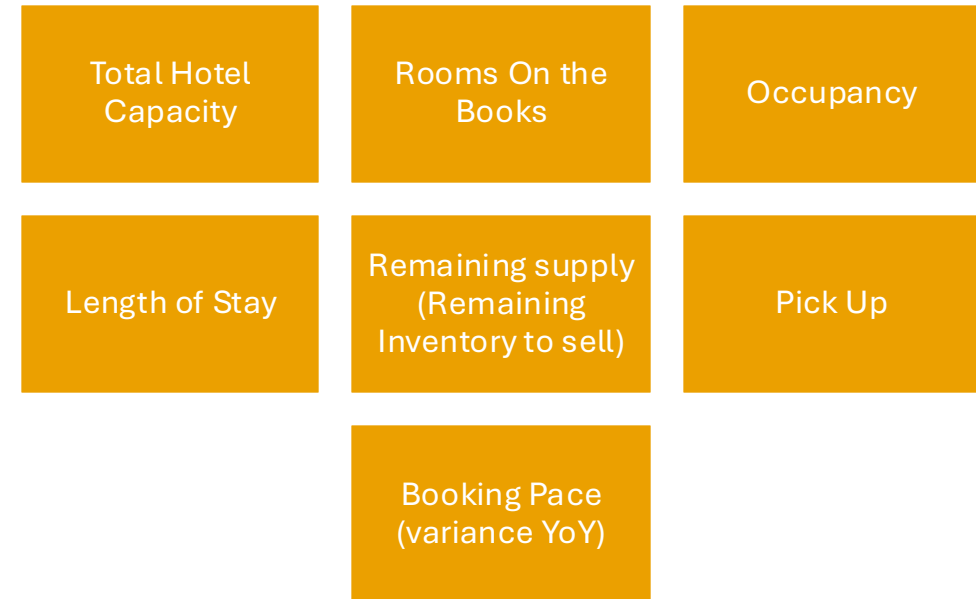
- City hotels have corporate travelers
- Last minute cancellations and bookings
- Resort hotels have mainly leisure guests

Next Steps

Further EDA

- Customer Type
 - Group
 - Contract
 - Transient
- Market Segments
 - Corporate
 - OTA's (Booking.com, Expedia)
 - Direct
- Trends with lead times

Feature Engineering



Thank You

