

**Dataset Link:** <https://www.kaggle.com/datasets/mojtaba142/hotel-booking>

## Business Problem

In recent years, City Hotel and Resort Hotel have seen high cancellation rates. Each hotel is now dealing with a number of issues as a result, including fewer revenues and less than ideal hotel room use. Consequently, lowering cancellation rates is both hotels' primary goal in order to increase their efficiency in generating revenue, and for us to offer thorough business advice to address this problem.

The analysis of hotel booking cancellations as well as other factors that have no bearing on their business and yearly revenue generation are the main topics of this report.

## Assumptions

1. No unusual occurrences between 2015 and 2017 will have a substantial impact on the data used.
2. The information is still current and can be used to analyze a hotel's possible plans in an efficient manner.
3. The biggest factor affecting the effectiveness of earning income is booking cancellations.
4. Cancellations result in vacant room for the booked length of time.
5. Clients make hotel reservations the same year they make cancellations.

## Hypothesis

1. More cancellations occur when prices are higher.
2. When there is a longer waiting list, customers tend to cancel more frequently
3. The majority of clients are coming from offline travel agents to make their reservations.

## Exploratory Data Analysis

### 1. Importing Libraries

```
[1] import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

### 2. Loading The Dataset

```
[2] path = '/content/drive/MyDrive/Dataset/hotel_booking.csv'
df = pd.read_csv(path)
```

### 3. Exploring and Cleaning Data

```
[3] df.head()
```

	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_in_week_nights	adults	...
0	Resort Hotel	0	342	2015	July	27	1	0	0	2	...
1	Resort Hotel	0	737	2015	July	27	1	0	0	2	...
2	Resort Hotel	0	7	2015	July	27	1	0	1	1	...
3	Resort Hotel	0	13	2015	July	27	1	0	1	1	...
4	Resort Hotel	0	14	2015	July	27	1	0	2	2	...

5 rows × 36 columns

```
[4] df.tail()
```

	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_in_week_nights	adults	...
119385	City Hotel	0	23	2017	August	35	30	2	5	2	...
119386	City Hotel	0	102	2017	August	35	31	2	5	3	...
119387	City Hotel	0	34	2017	August	35	31	2	5	2	...
119388	City Hotel	0	109	2017	August	35	31	2	5	2	...
119389	City Hotel	0	205	2017	August	35	29	2	7	2	...

5 rows × 36 columns

#### Dataset Rows and Columns

```
[5] df.shape
```

```
(119390, 36)
```

#### Dataset Information

```
[6] df.columns
```

```
Index(['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_in_week_nights', 'adults', 'children', 'babies', 'meal',  
      'country', 'market_segment', 'distribution_channel',  
      'is_repeated_guest', 'previous_cancellations',  
      'previous_bookings_not_canceled', 'reserved_room_type',  
      'assigned_room_type', 'booking_changes', 'deposit_type', 'agent',  
      'company', 'days_in_waiting_list', 'customer_type', 'adr',  
      'required_car_parking_spaces', 'total_of_special_requests',  
      'reservation_status', 'reservation_status_date', 'name', 'email',  
      'phone-number', 'credit_card'],  
      dtype='object')
```

```
[7] df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 119390 entries, 0 to 119389
Data columns (total 36 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   hotel                                119390 non-null  object
1   is_canceled                          119390 non-null  int64
2   lead_time                           119390 non-null  int64
3   arrival_date_year                    119390 non-null  int64
4   arrival_date_month                   119390 non-null  object
5   arrival_date_week_number             119390 non-null  int64
6   arrival_date_day_of_month            119390 non-null  int64
7   stays_in_weekend_nights              119390 non-null  int64
8   stays_in_week_nights                 119390 non-null  int64
9   adults                               119390 non-null  int64
10  children                             119386 non-null  float64
11  babies                              119390 non-null  int64
12  meal                                 119390 non-null  object
13  country                             118902 non-null  object
14  market_segment                       119390 non-null  object
15  distribution_channel                  119390 non-null  object
16  is_repeated_guest                     119390 non-null  int64
17  previous_cancellations                119390 non-null  int64
18  previous_bookings_not_canceled        119390 non-null  int64
19  reserved_room_type                    119390 non-null  object
20  assigned_room_type                    119390 non-null  object
21  booking_changes                       119390 non-null  int64
22  deposit_type                          119390 non-null  object
23  agent                                103050 non-null  float64
24  company                              6797 non-null   float64
25  days_in_waiting_list                  119390 non-null  int64
26  customer_type                         119390 non-null  object
27  adr                                   119390 non-null  float64
28  required_car_parking_spaces           119390 non-null  int64
29  total_of_special_requests             119390 non-null  int64
30  reservation_status                    119390 non-null  object
31  reservation_status_date               119390 non-null  object
32  name                                  119390 non-null  object
33  email                                 119390 non-null  object
34  phone-number                          119390 non-null  object
35  credit_card                           119390 non-null  object
dtypes: float64(4), int64(16), object(16)
memory usage: 32.8+ MB
```

```
[8] df.describe(include = 'object')
```

	hotel	arrival_date_month	meal	country	market_segment	distribution_channel	reserved_room_type	assigned_room_type	deposit_type	customer_type	reservation_status	reser
count	119390	119390	119390	118902	119390	119390	119390	119390	119390	119390	119390	
unique	2	12	5	177	8	5	10	12	3	4	3	
top	City Hotel	August	BB	PRT	Online TA	TA/TO	A	A	No Deposit	Transient	Check-Out	
freq	79330	13877	92310	48590	56477	97870	85994	74053	104641	89613	75166	

## Checking Duplicates Values

```
[9] len(df[df.duplicated()])
```

```
0
```

## Changing Data Type

```
[10] df['reservation_status_date'] = pd.to_datetime(df['reservation_status_date'])
```

```
[11] df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 119390 entries, 0 to 119389
Data columns (total 36 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   hotel                                119390 non-null  object
1   is_canceled                          119390 non-null  int64
2   lead_time                            119390 non-null  int64
3   arrival_date_year                    119390 non-null  int64
4   arrival_date_month                  119390 non-null  object
5   arrival_date_week_number            119390 non-null  int64
6   arrival_date_day_of_month            119390 non-null  int64
7   stays_in_weekend_nights              119390 non-null  int64
8   stays_in_week_nights                 119390 non-null  int64
9   adults                               119390 non-null  int64
10  children                             119386 non-null  float64
11  babies                               119390 non-null  int64
12  meal                                 119390 non-null  object
13  country                              118902 non-null  object
14  market_segment                       119390 non-null  object
15  distribution_channel                  119390 non-null  object
16  is_repeated_guest                     119390 non-null  int64
17  previous_cancellations                 119390 non-null  int64
18  previous_bookings_not_canceled         119390 non-null  int64
19  reserved_room_type                    119390 non-null  object
20  assigned_room_type                    119390 non-null  object
21  booking_changes                       119390 non-null  int64
22  deposit_type                          119390 non-null  object
23  agent                                 103050 non-null  float64
24  company                               6797 non-null   float64
25  days_in_waiting_list                  119390 non-null  int64
26  customer type                         119390 non-null  object
27  adr                                   119390 non-null  float64
28  required_car_parking_spaces           119390 non-null  int64
29  total_of_special_requests              119390 non-null  int64
30  reservation_status                    119390 non-null  object
31  reservation_status_date                119390 non-null  datetime64[ns]
32  name                                  119390 non-null  object
33  email                                 119390 non-null  object
34  phone-number                          119390 non-null  object
35  credit_card                           119390 non-null  object
dtypes: datetime64[ns](1), float64(4), int64(16), object(15)
memory usage: 32.8+ MB
```

## Getting Unique Values

```
[12] for col in df.describe(include = 'object').columns:
      print(col)
      print(df[col].unique)
```

```
hotel
<bound method Series.unique of 0      Resort Hotel
1      Resort Hotel
2      Resort Hotel
3      Resort Hotel
4      Resort Hotel
...
119385      City Hotel
119386      City Hotel
119387      City Hotel
119388      City Hotel
119389      City Hotel
Name: hotel, Length: 119390, dtype: object>
arrival_date_month
<bound method Series.unique of 0      July
1      July
2      July
3      July
4      July
...
119385      August
119386      August
119387      August
119388      August
119389      August
Name: arrival_date_month, Length: 119390, dtype: object>
meal
<bound method Series.unique of 0      BB
1      BB
2      BB
3      BB
4      BB
..
```

## Missing Values or Null Values

```
[13] df.isnull().sum()
```

```
hotel                                0
is_canceled                          0
lead_time                            0
arrival_date_year                     0
arrival_date_month                    0
arrival_date_week_number              0
arrival_date_day_of_month             0
stays_in_weekend_nights               0
stays_in_week_nights                 0
adults                               0
children                              4
babies                                0
meal                                  0
country                             488
market_segment                       0
distribution_channel                  0
is_repeated_guest                     0
previous_cancellations                0
previous_bookings_not_canceled        0
reserved_room_type                    0
assigned_room_type                    0
booking_changes                       0
deposit_type                          0
agent                               16340
company                             112593
days_in_waiting_list                 0
customer_type                         0
adr                                  0
required_car_parking_spaces           0
total_of_special_requests              0
reservation_status                    0
reservation_status_date               0
name                                  0
```

```
email 0
phone-number 0
credit_card 0
dtype: int64
```

## Handling Missing Values

```
[14] df.drop(['agent', 'company'], axis = 1, inplace = True)
      df.dropna(inplace = True)
```

```
[15] df.isnull().sum()
```

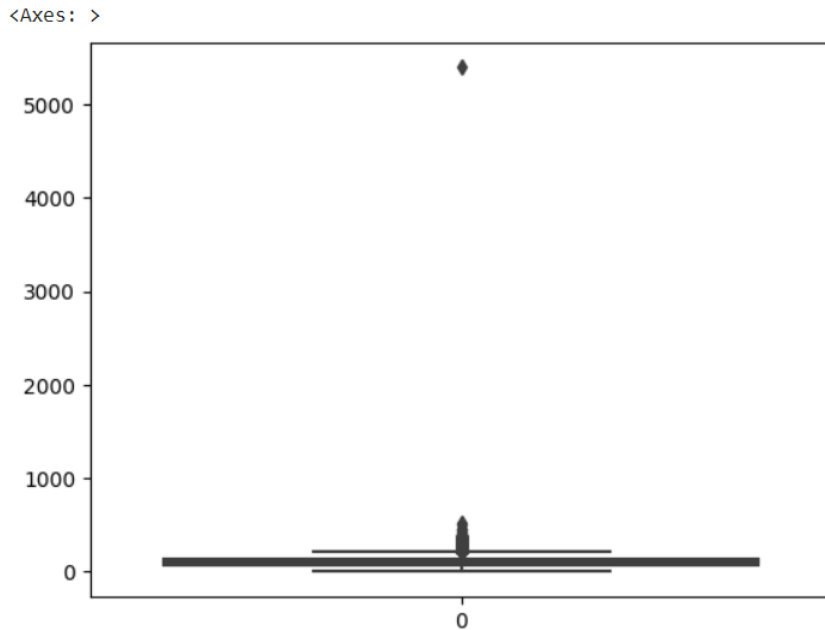
```
hotel 0
is_canceled 0
lead_time 0
arrival_date_year 0
arrival_date_month 0
arrival_date_week_number 0
arrival_date_day_of_month 0
stays_in_weekend_nights 0
stays_in_week_nights 0
adults 0
children 0
babies 0
meal 0
country 0
market_segment 0
distribution_channel 0
is_repeated_guest 0
previous_cancellations 0
previous_bookings_not_canceled 0
reserved_room_type 0
assigned_room_type 0
booking_changes 0
deposit_type 0
days_in_waiting_list 0
customer_type 0
adr 0
required_car_parking_spaces 0
total_of_special_requests 0
reservation_status 0
reservation_status_date 0
name 0
email 0
phone-number 0
credit_card 0
dtype: int64
```

## Identifying and Removing Outliers

```
[16] df.describe()
```

	is_canceled	lead_time	arrival_date_year	arrival_date_week_number	arrival_date_day_of_month	stays_in_weekend_nights	stays_in_week_nights	adults	children
count	118898.000000	118898.000000	118898.000000	118898.000000	118898.000000	118898.000000	118898.000000	118898.000000	118898.000000
mean	0.371352	104.311435	2016.157656	27.166555	15.800880	0.928897	2.502145	1.858391	0.104207
std	0.483168	106.903309	0.707459	13.589971	8.780324	0.996216	1.900168	0.578576	0.399172
min	0.000000	0.000000	2015.000000	1.000000	1.000000	0.000000	0.000000	0.000000	0.000000
25%	0.000000	18.000000	2016.000000	16.000000	8.000000	0.000000	1.000000	2.000000	0.000000
50%	0.000000	69.000000	2016.000000	28.000000	16.000000	1.000000	2.000000	2.000000	0.000000
75%	1.000000	161.000000	2017.000000	38.000000	23.000000	2.000000	3.000000	2.000000	0.000000
max	1.000000	737.000000	2017.000000	53.000000	31.000000	16.000000	41.000000	55.000000	10.000000

```
[17] sns.boxplot(df.adr)
```



```
[18] Q1 = df['adr'].quantile(0.25)
      Q3 = df['adr'].quantile(0.75)
      IQR = Q3-Q1
```

```
print('Q1:', Q1)
print('Q3:', Q3)
print('IQR:', IQR)
```

```
Q1: 70.0
Q3: 126.0
IQR: 56.0
```

```
[19] max_IQR = Q3 + 1.5 * IQR
      min_IQR = Q1 - 1.5 * IQR
```

```
print('min_IQR:', min_IQR)
print('max_IQR:', max_IQR)
```

```
min_IQR: -14.0
max_IQR: 210.0
```

```
[20] outliers = df.loc[(df['adr'] > max_IQR) | (df['adr'] < min_IQR)]
```

```
[21] new_df = len(df['adr']) - len(outliers)
print(new_df)
```

115015

```
[22] new_df = df.loc[(df['adr'] <= max_IQR) & (df['adr'] >= min_IQR)]
print(new_df)
```

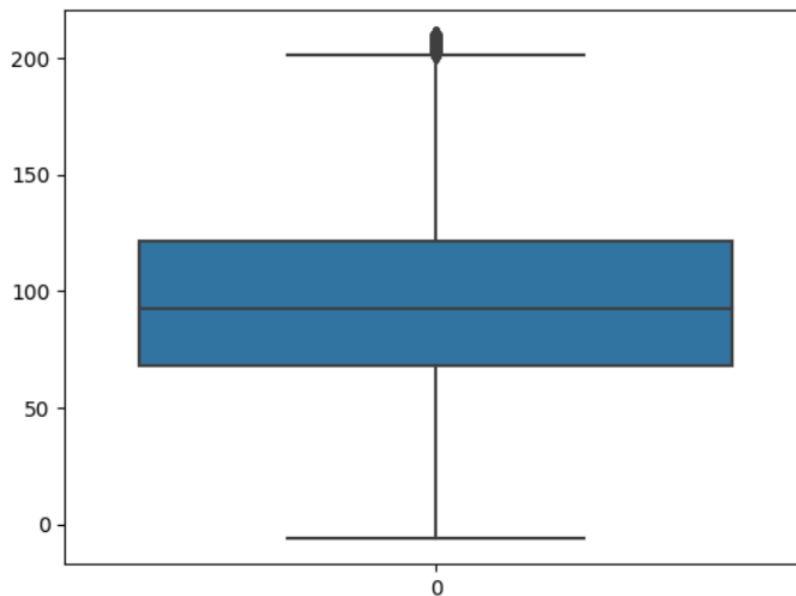
```
   hotel  is_canceled  lead_time  arrival_date_year \
0  Resort Hotel      0        342          2015
1  Resort Hotel      0        737          2015
2  Resort Hotel      0         7          2015
3  Resort Hotel      0        13          2015
4  Resort Hotel      0        14          2015
...
119384  City Hotel      0         21          2017
119385  City Hotel      0         23          2017
119387  City Hotel      0         34          2017
119388  City Hotel      0        109          2017
119389  City Hotel      0        205          2017

   arrival_date_month  arrival_date_week_number \
0          July          27
1          July          27
2          July          27
3          July          27
4          July          27
...
119384      August          35
119385      August          35
119387      August          35
119388      August          35
119389      August          35

   arrival_date_day_of_month  stays_in_weekend_nights \
0              1              0
1              1              0
2              1              0
3              1              0
4              1              0
```

```
[23] sns.boxplot(new_df.adr)
```

<Axes: >





```
[24] new_df.describe()
```

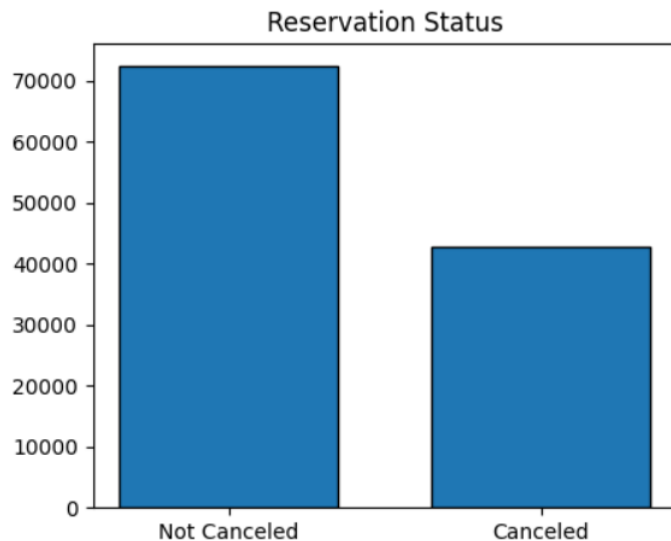
	is_canceled	lead_time	arrival_date_year	arrival_date_week_number	arrival_date_day_of_month	stays_in_weekend_nights	stays_in_week_nights	adults	children
count	115015.000000	115015.000000	115015.000000	115015.000000	115015.000000	115015.000000	115015.000000	115015.000000	115015.000000
mean	0.370795	105.368326	2016.146051	27.045220	15.775247	0.921845	2.482720	1.847794	0.082607
std	0.483020	107.759534	0.706932	13.741209	8.783105	0.995165	1.897364	0.578715	0.349738
min	0.000000	0.000000	2015.000000	1.000000	1.000000	0.000000	0.000000	0.000000	0.000000
25%	0.000000	18.000000	2016.000000	16.000000	8.000000	0.000000	1.000000	2.000000	0.000000
50%	0.000000	70.000000	2016.000000	27.000000	16.000000	1.000000	2.000000	2.000000	0.000000
75%	1.000000	163.000000	2017.000000	38.000000	23.000000	2.000000	3.000000	2.000000	0.000000
max	1.000000	737.000000	2017.000000	53.000000	31.000000	16.000000	41.000000	55.000000	10.000000

## 4. Data Analysis and Visualizations

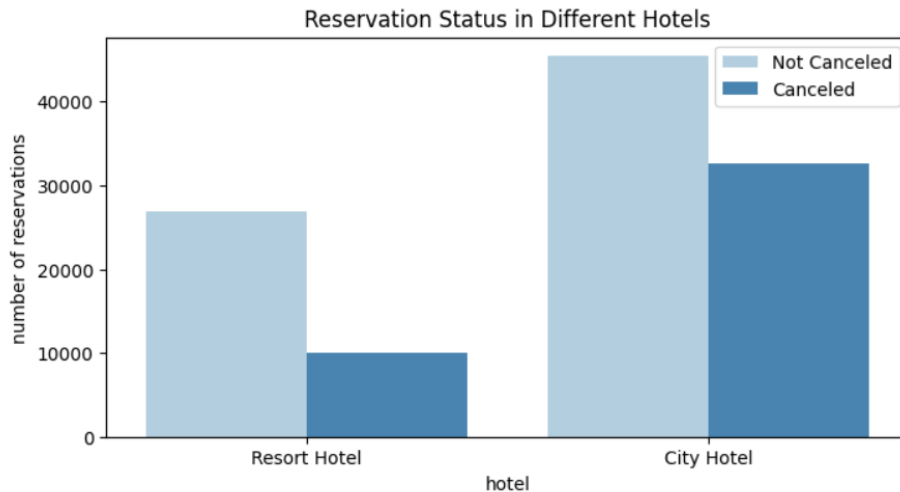
```
[25] cancelled_perc = new_df['is_canceled'].value_counts(normalize = True)
print(cancelled_perc)
```

```
plt.figure(figsize = (5,4))
plt.title('Reservation Status')
plt.bar(['Not Canceled', 'Canceled'], new_df['is_canceled'].value_counts(), edgecolor = 'k', width = 0.7)
plt.show()
```

```
0    0.629205
1    0.370795
Name: is_canceled, dtype: float64
```



```
[26] plt.figure(figsize = (8,4))
ax1=sns.countplot(x = 'hotel', hue = 'is_canceled', data = new_df, palette = 'Blues')
legend_labels,_ = ax1.get_legend_handles_labels()
ax1.legend(bbox_to_anchor=(1,1))
plt.title('Reservation Status in Different Hotels')
plt.xlabel('hotel')
plt.ylabel('number of reservations')
plt.legend(['Not Canceled', 'Canceled'])
plt.show()
```



```
[27] resort_hotel = new_df[new_df['hotel'] == 'Resort Hotel']
      resort_hotel['is_canceled'].value_counts(normalize = True)
```

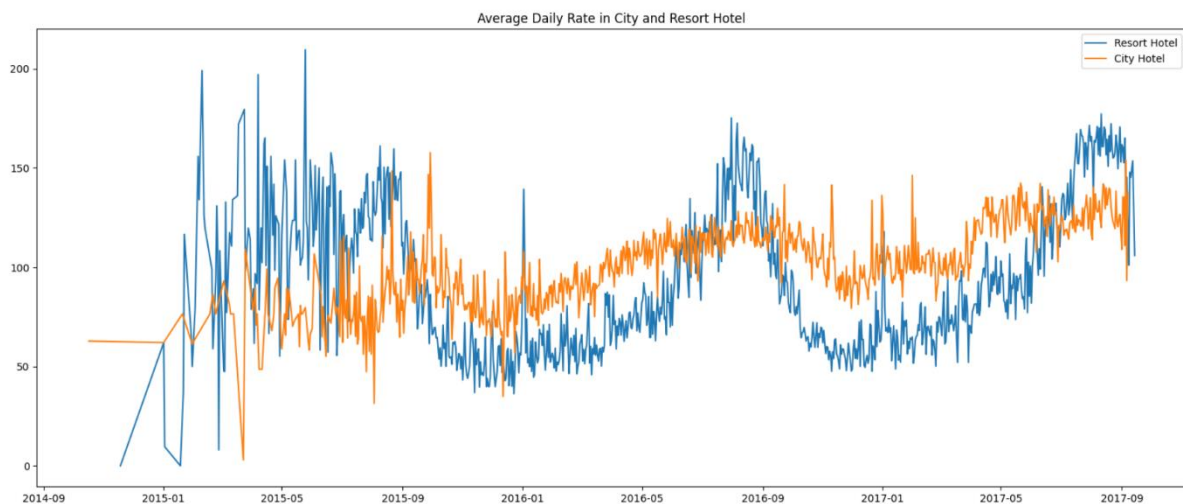
```
0    0.727997
1    0.272003
Name: is_canceled, dtype: float64
```

```
[28] city_hotel = new_df[new_df['hotel'] == 'City Hotel']
      city_hotel['is_canceled'].value_counts(normalize = True)
```

```
0    0.582297
1    0.417703
Name: is_canceled, dtype: float64
```

```
[29] resort_hotel = resort_hotel.groupby('reservation_status_date')[['adr']].mean()
      city_hotel = city_hotel.groupby('reservation_status_date')[['adr']].mean()
```

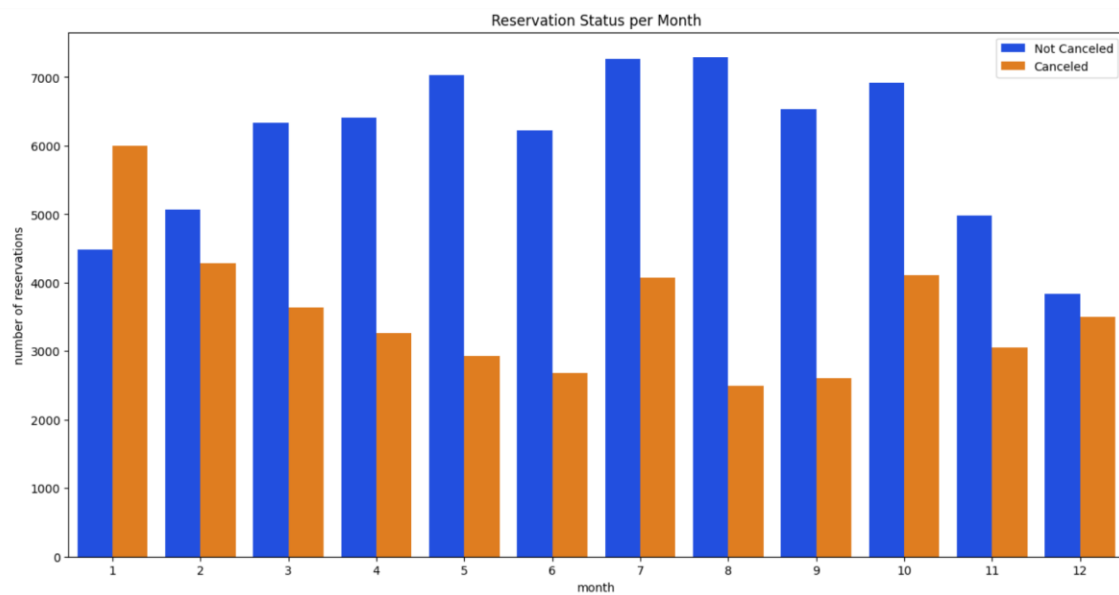
```
[30] plt.figure(figsize = (20,8))
      plt.title('Average Daily Rate in City and Resort Hotel')
      plt.plot(resort_hotel.index, resort_hotel['adr'], label = 'Resort Hotel')
      plt.plot(city_hotel.index, city_hotel['adr'], label = 'City Hotel')
      plt.legend()
      plt.show()
```



- There are price hikes in dataset. This can be due to price hikes during weekends or on holidays.
- Prices hikes in resort hotel are much higher than those in city hotel.

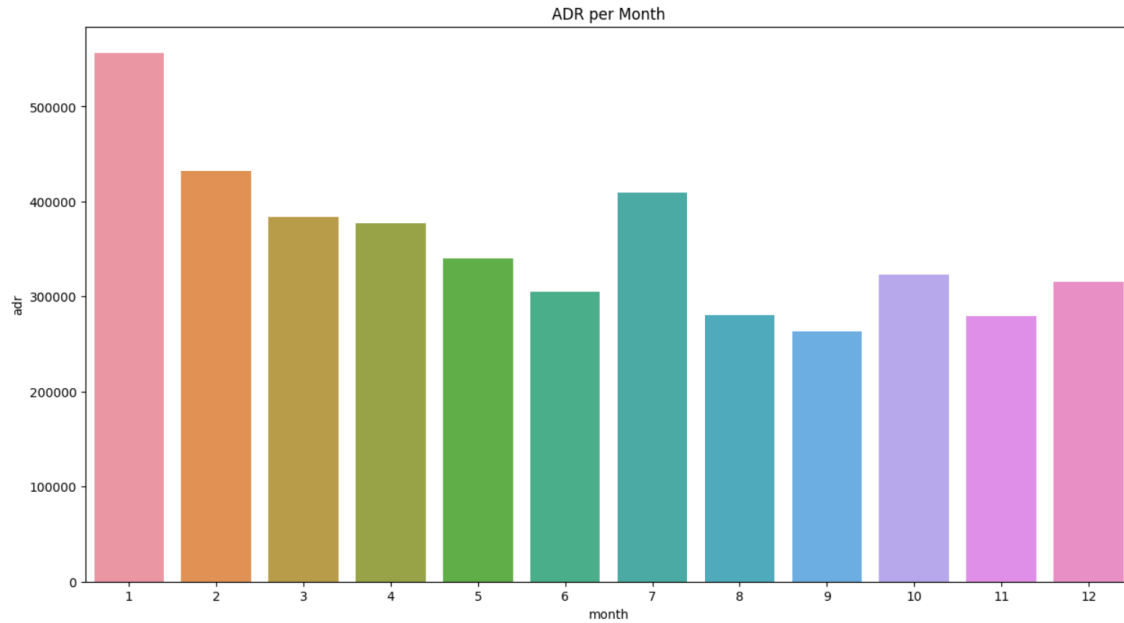
```
[31] new_df = new_df.loc[:]
      new_df['month'] = new_df['reservation_status_date'].dt.month

      plt.figure(figsize = (16,8))
      ax1=sns.countplot(x = 'month', hue = 'is_canceled', data = new_df, palette = 'bright')
      legend_labels,_ = ax1. get_legend_handles_labels()
      ax1.legend(bbox_to_anchor=(1,1))
      plt.title('Reservation Status per Month')
      plt.xlabel('month')
      plt.ylabel('number of reservations')
      plt.legend(['Not Canceled','Canceled'])
      plt.show()
```



- January has the most cancellations and August has the least cancellations.

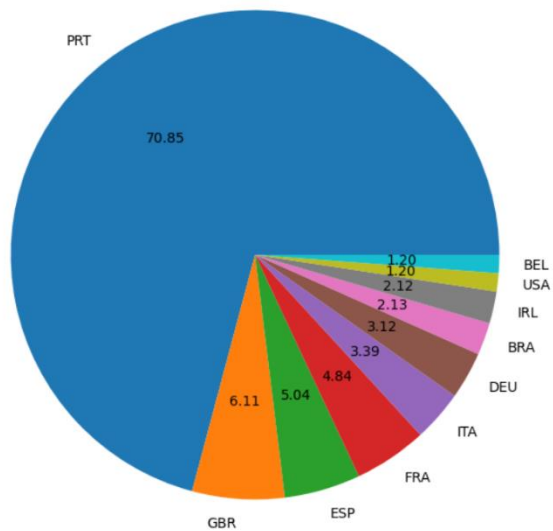
```
[32] plt.figure(figsize = (15,8))
      plt.title('ADR per Month')
      sns.barplot(x = 'month', y = 'adr', data = new_df[new_df['is_canceled'] == 1].groupby('month')[['adr']].sum().reset_index())
      plt.show()
```



- One of the hypothesis is proved where when price is high, the number of cancellations is more

```
[33] cancelled_data = new_df[new_df['is_canceled'] == 1]
top_10_country = cancelled_data['country'].value_counts()[:10]
plt.figure(figsize = (8,8))
plt.title('Top 10 Countries with Reservation Canceled')
plt.pie(top_10_country, autopct = '%.2f', labels = top_10_country.index)
plt.show()
```

Top 10 Countries with Reservation Canceled



- Portugal has the highest cancellations.

```
[34] new_df['market_segment'].value_counts()
```

```
Online TA      53638
Offline TA/TO  24052
Groups         19709
Direct         11541
Corporate       5104
Complementary   734
Aviation        237
Name: market_segment, dtype: int64
```

```
[35] new_df['market_segment'].value_counts(normalize = True)
```

```
Online TA      0.466357
Offline TA/TO  0.209121
Groups         0.171360
Direct         0.100343
Corporate       0.044377
Complementary   0.006382
Aviation        0.002061
Name: market_segment, dtype: float64
```

```
[36] cancelled_data['market_segment'].value_counts(normalize = True)
```

```
Online TA      0.456679
Groups         0.282763
Offline TA/TO  0.193636
Direct         0.040706
Corporate       0.022886
Complementary   0.002110
Aviation        0.001219
Name: market_segment, dtype: float64
```

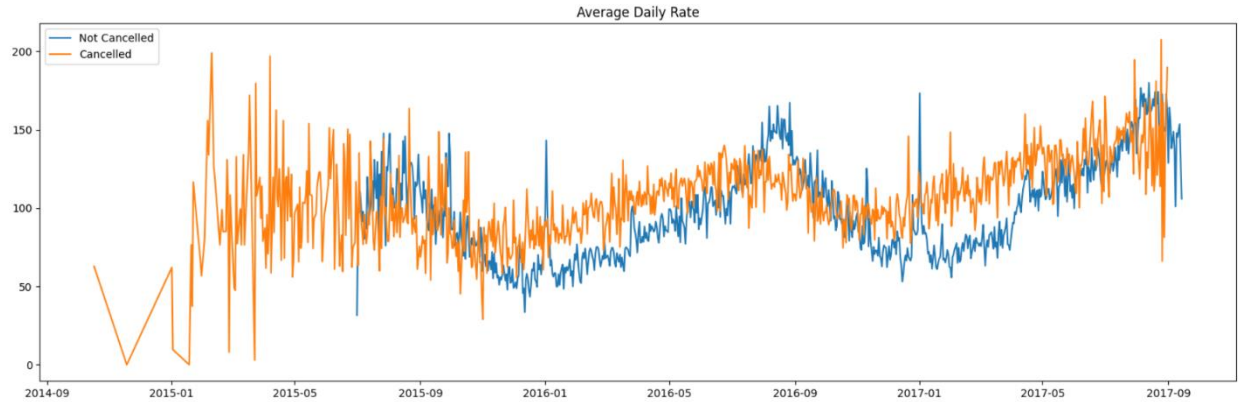
- A whopping 47% of cancellations are from users who booked through online travel agents.

- A possible reason could be that travel agents posted pictures of hotels that don't match reality.

```
[37] cancelled_df_adr = cancelled_data.groupby('reservation_status_date')[['adr']].mean()
cancelled_df_adr.reset_index(inplace = True)
cancelled_df_adr.sort_values('reservation_status_date', inplace = True)

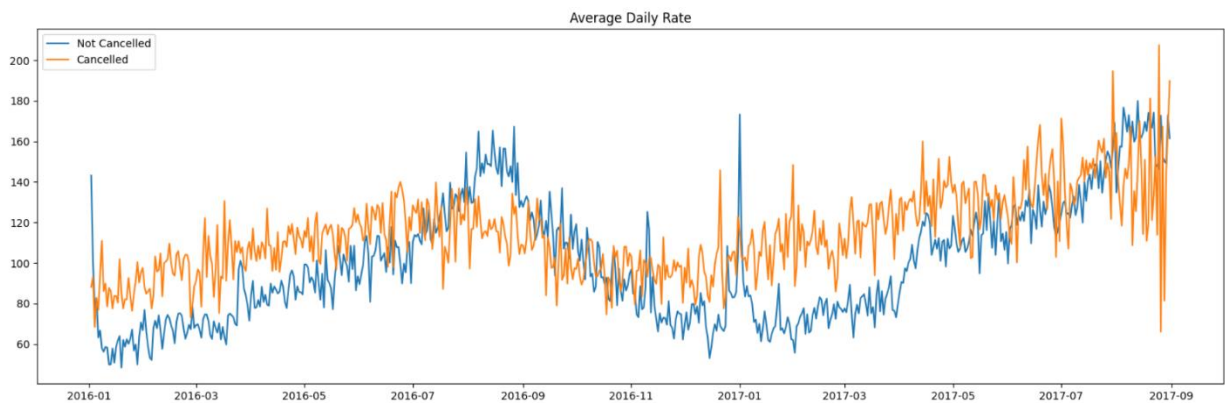
not_cancelled_data = df[df['is_cancelled'] == 0]
not_cancelled_df_adr = not_cancelled_data.groupby('reservation_status_date')[['adr']].mean()
not_cancelled_df_adr.reset_index(inplace = True)
not_cancelled_df_adr.sort_values('reservation_status_date', inplace = True)
```

```
[38] plt.figure(figsize = (20,6))
plt.title('Average Daily Rate')
plt.plot(not_cancelled_df_adr['reservation_status_date'], not_cancelled_df_adr['adr'], label = 'Not Cancelled')
plt.plot(cancelled_df_adr['reservation_status_date'], cancelled_df_adr['adr'], label = 'Cancelled')
plt.legend()
plt.show()
```



```
[39] cancelled_df_adr = cancelled_df_adr[(cancelled_df_adr['reservation_status_date']>'2016') & (cancelled_df_adr['reservation_status_date']<'2017-09')]
not_cancelled_df_adr = not_cancelled_df_adr[(not_cancelled_df_adr['reservation_status_date']>'2016') & (not_cancelled_df_adr['reservation_status_date']<'2017-09')]
```

```
[40] plt.figure(figsize = (20,6))
plt.title('Average Daily Rate')
plt.plot(not_cancelled_df_adr['reservation_status_date'], not_cancelled_df_adr['adr'], label = 'Not Cancelled')
plt.plot(cancelled_df_adr['reservation_status_date'], cancelled_df_adr['adr'], label = 'Cancelled')
plt.legend()
plt.show()
```



## Suggestions

1. Cancellation rates rise as the prices does. In order to prevent cancellations of reservations, hotel could work on their pricing strategies and try to lower the rates for specific hotels based on location. They can also provide some discounts to the customers.
2. As the ratio of the cancellation and not cancellation of the city hotel is higher than the resort hotel. So the hotels should provide a reasonable discount on the room prices on weekends or on holidays.
3. In the month of January, hotels can start campaigns or marketing with a reasonable amount to increase their revenue as the cancellations is the highest in this month.
4. They can also increase the quality of their hotels and their services mainly in Portugal to reduce the cancellation rates.