

Data Description

The dataset consists of hotel booking data with 36 columns and 119,390 rows. It contains all the necessary details required for booking a room in a hotel, including information such as guest demographics, booking dates, room type, number of adults and children, special requests, and more.

Additionally, the dataset includes the cancellation status of each booking, indicating whether the reservation was canceled or not. This information can provide insights into booking trends and cancellation rates within the hotel industry.

Furthermore, the dataset also captures the source of each booking, which indicates the channel or platform through which the reservation was made. This data can help analyze the effectiveness of different booking sources and identify the most popular platforms for hotel bookings.

Overall, the dataset provides a comprehensive view of hotel booking activities, including reservation details, cancellation status, and the sources through which bookings are made. It can be used for various analytical purposes, such as understanding booking patterns, predicting cancellations, and evaluating the performance of different booking channels.

Column descriptions

- hotel**: The name of the hotel (non-null object) **is_canceled**: Indicates whether the booking was canceled (non-null int64) **lead_time**: Number of days between booking and arrival (non-null int64) **arrival_date_year**: Year of arrival (non-null int64) **arrival_date_month**: Month of arrival (non-null object) **arrival_date_week_number**: Week number of arrival (non-null int64) **arrival_date_day_of_month**: Day of the month of arrival (non-null int64) **stays_in_weekend_nights**: Number of weekend nights stayed (non-null int64) **adults**: Number of adults (non-null int64) **children**: Number of children (non-null object) **distribution_channel**: Booking distribution channel (non-null object) **is_repeated_guest**: Indicates if the guest is a repeated guest (non-null int64) **previous_cancellations**: Number of previous cancellations by the guest (non-null int64) **previous_bookings_not_canceled**: Number of previous bookings not canceled by the guest (non-null int64) **reserved_room_type**: Type of room reserved (non-null object) **assigned_room_type**: Type of room assigned (non-null object) **booking_changes**: Number of changes made to the booking (non-null int64) **deposit_type**: Type of deposit made (non-null object) **agent_id**: ID of the travel agency that made the booking (non-null float64) **company**: ID of the company that made the booking (non-null int64) **days_in_waiting_list**: Number of days the booking was on the waiting list (non-null int64) **customer_type**: Type of customer (non-null object) **adr**: Average Daily Rate (non-null float64) **required_car_parking_spaces**: Number of car parking spaces requested (non-null int64) **total_of_special_requests**: Number of special requests made (non-null int64) **reservation_status**: Current status of the reservation (non-null object) **reservation_status_date**: Date at which the reservation status was last updated (non-null object) **name**: Name of the guest (non-null object) **email**: Email address of the guest (non-null object) **phone-number**: Phone number of the guest (non-null object) **credit_card**: Credit card information of the guest (non-null object)

Importing Libraries

```
In [30]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings

Out[30]: <function warnings.filterwarnings(action, message='', category=<class 'Warning'>, module='', lineno=0, append=False)>
```

Loading Dataset

```
In [31]: df=pd.read_csv('hotel_booking.csv')

Out[32]: 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	...	customer_type	
0	Resort Hotel	0	342	2015	July	27	1	0	0	2	...	Transient	
1	Resort Hotel	0	737	2015	July	27	1	0	0	2	...	Transient	
2	Resort Hotel	0	7	2015	July	27	1	0	1	1	...	Transient	
3	Resort Hotel	0	13	2015	July	27	1	0	1	1	...	Transient	
4	Resort Hotel	0	14	2015	July	27	1	0	2	2	...	Transient	

5 rows × 36 columns

Data Analysis

```
In [33]: df.shape
Out[33]: (119390, 36)

In [34]: df.columns
Out[34]: 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')
```

```
In [35]: 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 119390 non-null float64
11 babies 119390 non-null int64
12 meal 119390 non-null object
13 country 119390 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 103850 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
```

```
In [36]: df['reservation_status_date'] =pd.to_datetime(df['reservation_status_date' ])

In [37]: 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	name
count	119390	119390	119390	118902	119390	119390	119390	119390	119390	119390	119390	119390
unique	2	12	5	177	8	5	10	12	3	4	3	81503
top	City Hotel	August	BB	PRT	Online TA	TA/TO	A	A	No Deposit	Transient	Check-Out	Michael Johnson
freq	79330	13877	92310	48590	56477	97870	85994	74053	104641	89613	75166	48

```
In [38]: df.isnull().sum()
Out[38]: 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
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
```

```
In [39]: df.drop(['company','agent'], axis=1 , inplace=True )

In [40]: df.dropna(inplace=True)
```

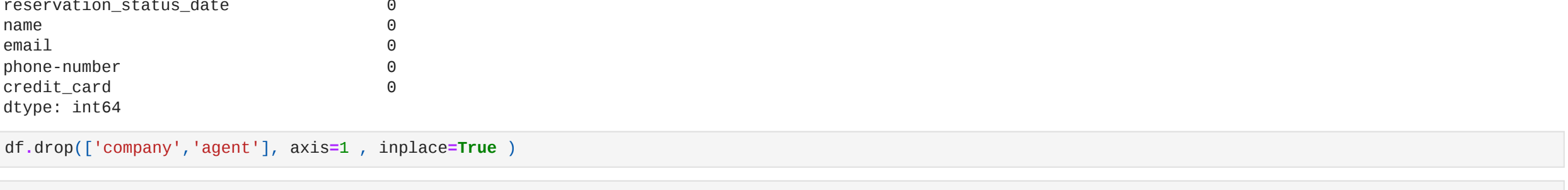
```
In [41]: df.isnull().sum()
Out[41]: hotel 0
is_canceled 0
lead_time 0
arrival_date_year 0
arrival_date_month 0
arrival_date_week_number 0
arrival_date_day_of_month 0
stays_in_weekend_nights 0
stays_in_week_nights 0
adults 0
children 0
babies 0
meal 0
country 0
distribution_channel 0
is_repeated_guest 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
```

Data Visualization

```
In [42]: cancelled_perc =df['is_canceled'].value_counts(normalize=True)
cancelled_perc
```

```
Out[42]: 0    0.628648
1    0.371352
Name: is_canceled, dtype: float64
```

```
In [43]: plt.figure(figsize=(4,4))
plt.title('Reservation Status Count')
plt.bar(['not canceled','canceled'],df['is_canceled'].value_counts(),edgecolor='b')
plt.show()
```



```
In [44]: plt.figure(figsize=(8,4))
sns.countplot(x='hotel', hue='is_canceled', data=df, palette='Blues')
plt.title('Reservation Status in differnt city',size=20)
plt.xlabel('hotel')
plt.ylabel('Num of reservation')
```



```
In [45]: resort_hotel=df[df['hotel']=='Resort Hotel']
resort_hotel['is_canceled'].value_counts(normalize = True)
```

```
Out[45]: 0    0.72925
1    0.27975
Name: is_canceled, dtype: float64
```

```
In [46]: city_hotel=df[df['hotel']=='City Hotel']
city_hotel['is_canceled'].value_counts(normalize = True)
```

```
Out[46]: 0    0.582911
1    0.417089
Name: is_canceled, dtype: float64
```

```
In [47]: resort_hotel=resort_hotel.groupby('reservation_status_date')['adr'].mean()
city_hotel=city_hotel.groupby('reservation_status_date')['adr'].mean()
```

```
In [48]: plt.figure(figsize=(15,8))
plt.title('Avg daily rate in city and resort hotel',size=20)
plt.plot(resort_hotel.index,resort_hotel['adr'],label='Resort Hotel')
plt.plot(city_hotel.index,city_hotel['adr'],label='City Hotel')
plt.legend(fontsize=20)
plt.show
```

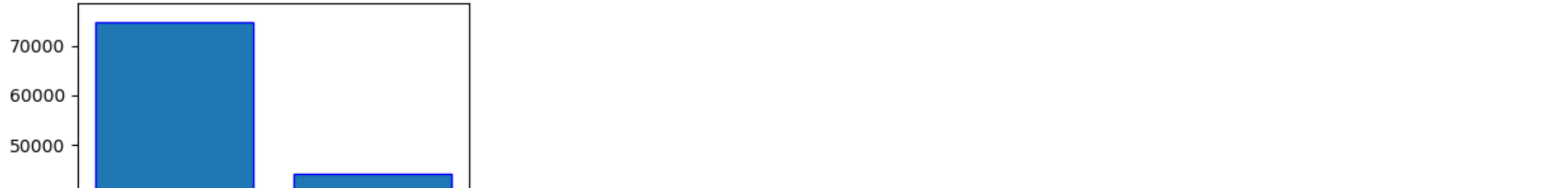
```
Out[48]: <function matplotlib.pyplot.show(close=None, block=None)>
```



```
In [76]: df['month']=df['reservation_status_date'].dt.month
plt.figure(figsize=(10,6))
plt.title('Adr per month', hue='is_canceled', data=df, palette='Blues')
ax=sns.countplot(x='month', hue='is_canceled', data=df, palette='Blues')
plt.title('Reservation status per month',size=15)
plt.xlabel('month')
plt.ylabel('Num of reservation')
plt.legend(['not canceled','canceled'])
plt.show
```

```
-----
TypeError                                Traceback (most recent call last)
~AppData\Local\Temp\ipykernel_7432\3856318725.py in <module>
      3 ax=sns.countplot(x='month', hue='is_canceled', data=df, palette='Blues')
----> 4 plt.title('Reservation status per month',size=15)
      5 plt.xlabel('month')
      6 plt.ylabel('Num of reservation')
      7 plt.legend(['not canceled','canceled'])

TypeError: 'str' object is not callable
```



```
In [52]: plt.figure(figsize=(18,6))
plt.title('Adr per month', fontsize=15)
sns.barplot('month', 'adr', data=df[df['is_canceled']==1].groupby('month')['adr'].sum().reset_index())
plt.show()
```

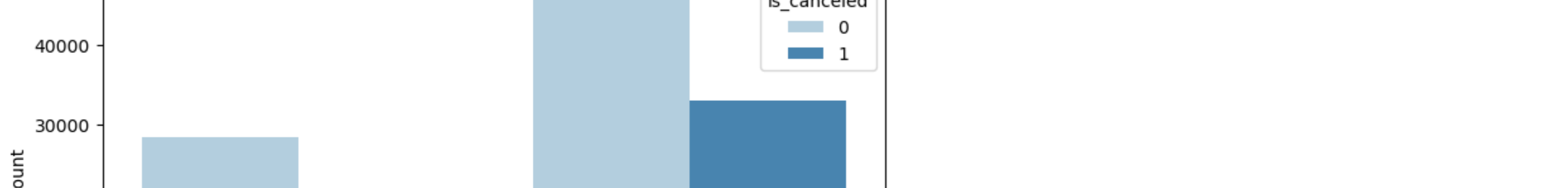
C:\Users\Admin\anaconda3\lib\site-packages\seaborn\decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be 'data', and passing other variables without an explicit binding will result in an error or misinterpretation.

```
warnings.warn()
```

```
Out[52]:
```

```
In [59]: canceled_data=df[df['is_canceled']==1]
top_10_country=canceled_data['country'].value_counts()[0:10]
plt.figure(figsize=(7,7))
plt.title('Top 10 countries with reservation canceled', fontsize=15)
plt.pie(top_10_country, autopct='%2f', labels=top_10_country.index)
plt.show()
```

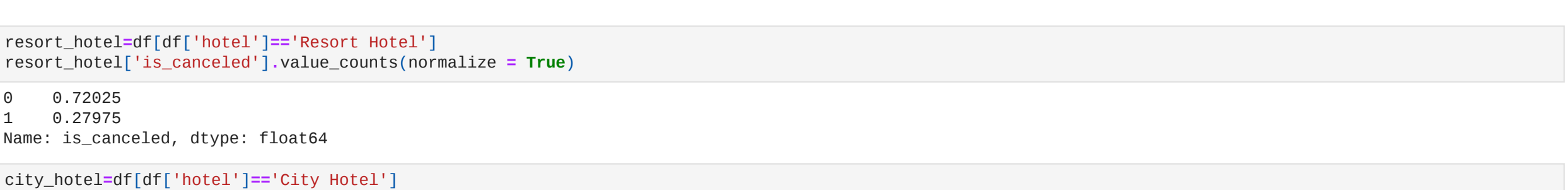
Top 10 countries with reservation canceled



```
In [63]: market_segment=df['market_segment'].value_counts()
market_segment
```

```
Out[63]: Online TA    56462
Offline TA/TO    24168
Groups           19806
Direct           12448
Corporate         5111
Complementary     734
Aviation          237
Name: market_segment, dtype: int64
```

```
In [77]: plt.figure(figsize=(10,6))
ax=sns.countplot(x='market_segment', hue='market_segment', data=df, palette='Reds')
plt.title('Segment',size=15)
plt.xlabel('market_segment')
plt.ylabel('Num of reservation')
```



Conclusion

Based on our analysis, we have found that the cancellation rate in resort hotels is higher compared to city hotels. This indicates that guests booking stays at resort properties are more likely to cancel their reservations.

One of the key factors contributing to the higher cancellation rate is the prevalence of online bookings. The ease of making reservations online has led to an increase in impulsive bookings, where guests might make a reservation without fully considering their plans or commitments. As a result, they are more likely to cancel these bookings later on.

Additionally, the absence of detailed information provided to customers during the booking process contributes to the cancellation trend. When guests do not have a clear understanding of the amenities, services, or specific features of the resort, they may feel compelled to cancel if their expectations are not met upon arrival. Enhancing the clarity and transparency of the booking information can help set appropriate expectations and reduce cancellations.

To address the high cancellation rate, we recommend taking several actions. Firstly, improving the pre-booking experience by providing comprehensive details about the resort, including facilities, activities, and any limitations, can help guests make more informed decisions. Additionally, implementing a stricter cancellation policy or offering incentives for non-refundable bookings can discourage last-minute cancellations and ensure more committed reservations.

Furthermore, enhancing communication with guests after the booking, such as sending reminders closer to the arrival date and providing personalized recommendations, can help reinforce their commitment to the reservation and increase guest satisfaction.

By addressing these factors and implementing measures to reduce cancellations, we can improve the overall occupancy rates, revenue, and guest experience at our resort properties.

```
In [ ]:
```