## **Import Libraries**

### **Load Dataset**

	Booking_ID	no_of_adults	no_of_children	no_of_weekend_nights	no_of_week_nights	type_of_meal_plan	required
0	INN00001	2	0	1	2	Meal Plan 1	
1	INN00002	2	0	2	3	Not Selected	
2	INN00003	1	0	2	1	Meal Plan 1	
3	INN00004	2	0	0	2	Meal Plan 1	
4	INN00005	2	0	1	1	Not Selected	
4							<b>&gt;</b>

	Booking_ID	no_of_adults	no_of_children	no_of_weekend_nights	no_of_week_nights	type_of_meal_plan	req
36270	INN36271	3	0	2	6	Meal Plan 1	
36271	INN36272	2	0	1	3	Meal Plan 1	
36272	INN36273	2	0	2	6	Meal Plan 1	
36273	INN36274	2	0	0	3	Not Selected	
36274	INN36275	2	0	1	2	Meal Plan 1	
4							•

(36275, 19)

<class 'pandas.core.frame.DataFrame'> RangeIndex: 36275 entries, 0 to 36274 Data columns (total 19 columns):

Data	columns (cocal is columns):		
#	Column	Non-Null Count	Dtype
0	Booking_ID	36275 non-null	object
1	no_of_adults	36275 non-null	int64
2	no_of_children	36275 non-null	int64
3	no_of_weekend_nights	36275 non-null	int64
4	no_of_week_nights	36275 non-null	int64
5	type_of_meal_plan	36275 non-null	object
6	required_car_parking_space	36275 non-null	int64
7	room_type_reserved	36275 non-null	object
8	<pre>lead_time</pre>	36275 non-null	int64
9	arrival_year	36275 non-null	int64
10	arrival_month	36275 non-null	int64
11	arrival_date	36275 non-null	int64
12	market_segment_type	36275 non-null	object
13	repeated_guest	36275 non-null	int64
14	no_of_previous_cancellations	36275 non-null	int64
15	<pre>no_of_previous_bookings_not_canceled</pre>	36275 non-null	int64
16	avg_price_per_room	36275 non-null	float64
17	no_of_special_requests	36275 non-null	int64
18	booking_status	36275 non-null	object
dtype	es: float64(1), int64(13), object(5)		-

dtypes: float64(1), int64(13), object(5)

memory usage: 5.3+ MB

	Booking_ID	type_of_meal_plan	room_type_reserved	market_segment_type	booking_status
count	36275	36275	36275	36275	36275
unique	36275	4	7	5	2
top	INN00001	Meal Plan 1	Room_Type 1	Online	Not_Canceled
freq	1	27835	28130	23214	24390

```
Booking_ID
['INN00001' 'INN00002' 'INN00003' ... 'INN36273' 'INN36274' 'INN36275']
type_of_meal_plan
['Meal Plan 1' 'Not Selected' 'Meal Plan 2' 'Meal Plan 3']
room_type_reserved
['Room_Type 1' 'Room_Type 4' 'Room_Type 2' 'Room_Type 6' 'Room_Type 5'
'Room_Type 7' 'Room_Type 3']
market_segment_type
['Offline' 'Online' 'Corporate' 'Aviation' 'Complementary']
booking_status
['Not_Canceled' 'Canceled']
```

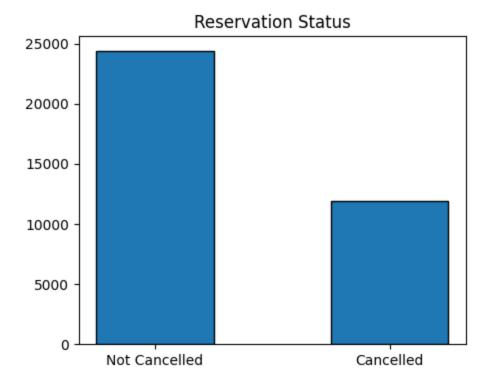
Booking_ID	0
no_of_adults	0
no_of_children	0
<pre>no_of_weekend_nights</pre>	0
no_of_week_nights	0
type_of_meal_plan	0
required_car_parking_space	0
room_type_reserved	0
<pre>lead_time</pre>	0
arrival_year	0
arrival_month	0
arrival_date	0
market_segment_type	0
repeated_guest	0
no_of_previous_cancellations	0
<pre>no_of_previous_bookings_not_canceled</pre>	0
avg_price_per_room	0
<pre>no_of_special_requests</pre>	0
booking_status	0
dtype: int64	

	no_of_adults	no_of_children	no_of_weekend_nights	no_of_week_nights	required_car_parking_space	lead
count	36275.000000	36275.000000	36275.000000	36275.000000	36275.000000	36275.0
mean	1.844962	0.105279	0.810724	2.204300	0.030986	85.2
std	0.518715	0.402648	0.870644	1.410905	0.173281	85.9
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.0
25%	2.000000	0.000000	0.000000	1.000000	0.000000	17.0
50%	2.000000	0.000000	1.000000	2.000000	0.000000	57.0
75%	2.000000	0.000000	2.000000	3.000000	0.000000	126.0
max	4.000000	10.000000	7.000000	17.000000	1.000000	443.0
4						<b>&gt;</b>

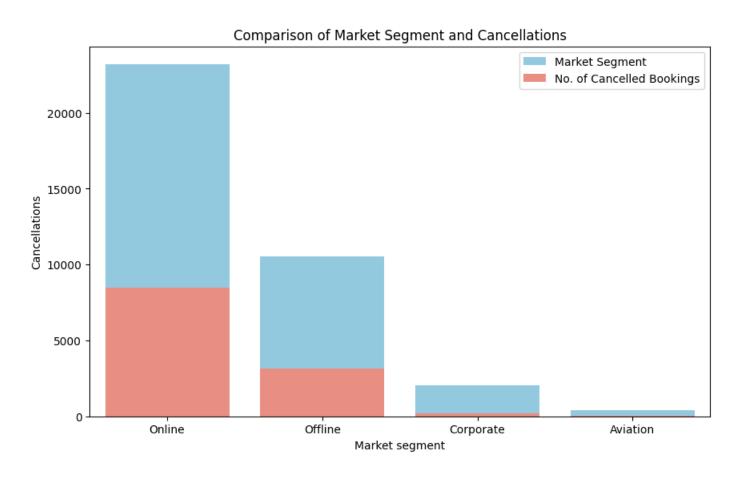
Not\_Canceled 0.672364 Canceled 0.327636

Name: booking\_status, dtype: float64

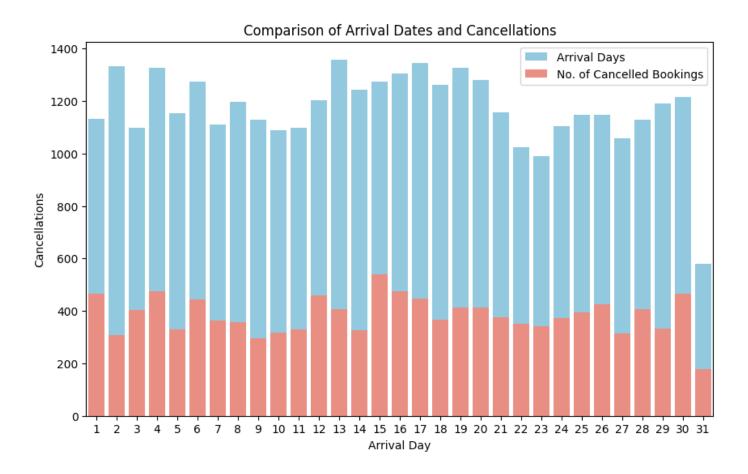
## Analysis of Cancelled and Not\_Cancelled Bookings



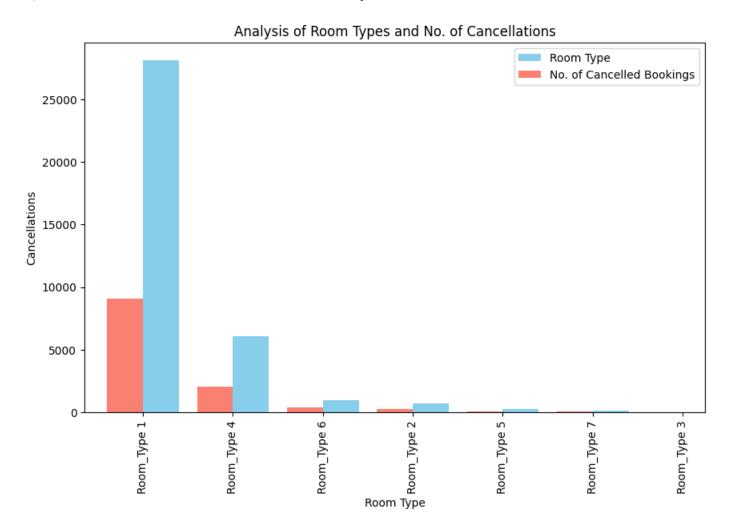
# Analysis of Market Segment Type and Cancelled Bookings



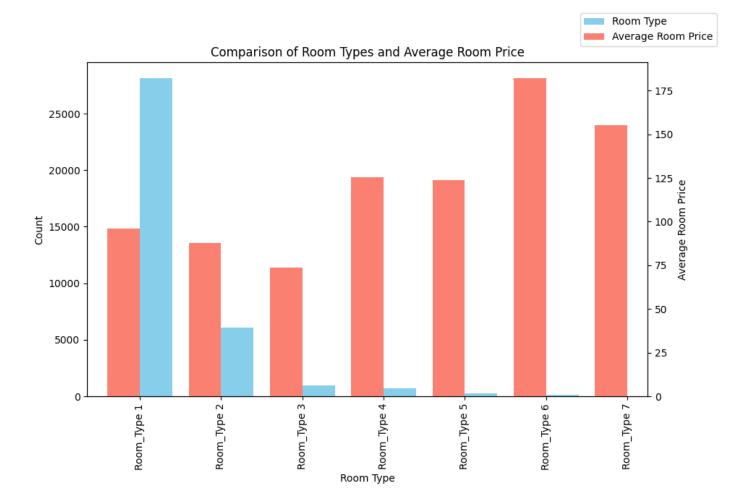
## **Analysis of Arrival Date and Booking Cancellations**



Analysis of Room Type and No. of Cancellations



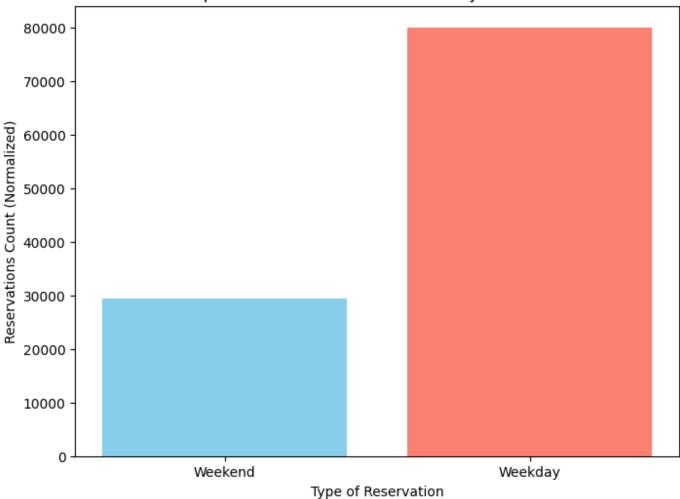
Analysis of Room Type and Average Room Price



# Analyzing No. of Weekend Night Reservations to No. of Weekday Night Reservations

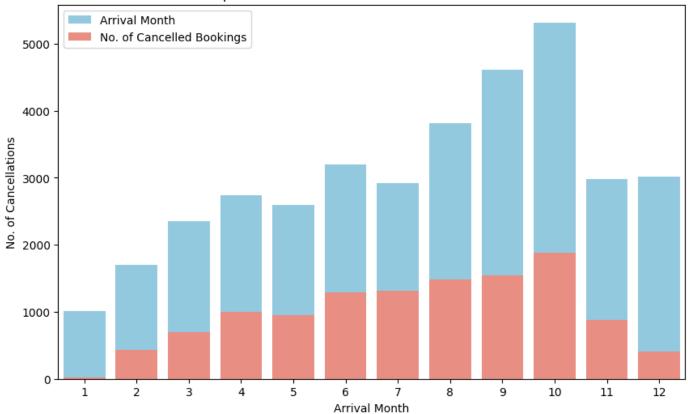
Weekend Reservations: 29409 Weekday Reservations: 79961

### Comparison of Weekend and Weekday Reservations



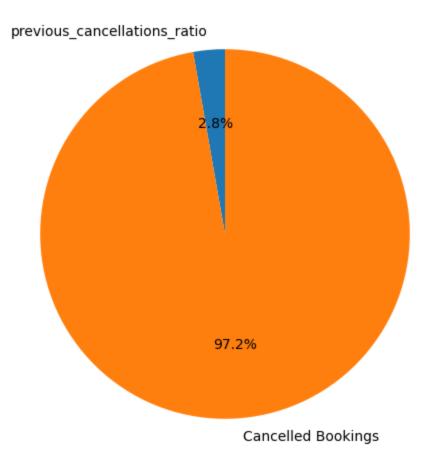
## **Analysis of Arrival Months and Cancellations**

#### Comparison of Arrival Month and No. of Cancellations



Analysis of Previous Cancellations Ratio to Overall Cancellations

#### Ratio of Previous Cancellations to Cancelled Bookings



# Regression Analysis of 'Cancelled Bookings' and 'Average Price per Room'

T-Statistic: 27.433239020374042 P-Value: 5.2303189247282316e-164

T-tests and P-test analysis carried out suggest a very large T-statistic and an extremely small p-value, indicating that the average room price has a statistically significant impact on the cancellation rate. With such a small p-value (close to 0), you can confidently reject the null hypothesis and conclude that there is indeed a significant relationship between the average room price and the cancellation rate. We investigate further.

#### OLS Regression Results

=======================================		====			==========	
Dep. Variable:	booking_sta	tus	R-squared:		0.020	
Model:		0LS	Adj. R-squared	d:	0.020	
Method:	Least Squa	res	F-statistic:		752.6	
Date:	Mon, 23 Oct 2	023	Prob (F-statis	stic):	5.23e-164	
Time:	11:11	:23	Log-Likelihood	d:	-23661.	
No. Observations:	36	275	AIC:		4.733e+04	
Df Residuals:	36	273	BIC:		4.734e+04	
Df Model:		1				
Covariance Type:	nonrob	ust				
=======================================		=====				======
	coef s	td er	r t	P> t	[0.025	0.975]
const	0.1304	0.00	08 17 <b>.</b> 177	0.000	0.116	0.145
<pre>avg_price_per_room</pre>	0.0019 6.	95e-0	27.433	0.000	0.002	0.002
Omnibus:	312722.	===== 067	Durbin-Watson:	:======== :	2.010	
Prob(Omnibus):	0.	000	Jarque-Bera (	JB):	6013.016	
Skew:	0.	704	Prob(JB):		0.00	
Kurtosis:	1.	588	Cond. No.		340.	
=======================================	=========	=====	==========		=========	

#### Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

0.020 is observed to be the R-Squared value, indicating a 2% variance in cancellation rate. This means there is no real relationship between Average Room Prices and Cancellation rate. Other Factors should be considered.

## Summary

### **Observations**

- High Price does NOT Affect Cancellation rates
- Most Bookings and Cancellations occur from ONLINE bookings type
- October always has most Bookings and has most Cancellations also
- Room 1 has the Highest number of Booking Cancellations

## Suggestions

- Prices for Online Bookings should be reduced
- Set Policy to ensure Customers that make bookings are not encouraged to Cancel
- Promotions can be carried out to improve online bookings
- Promotions and discounts can be given for rooms (except room 1)