

Capstone Project - 1

EDA on Hotel Booking Analysis

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Problem Statement

- ❖ For this project we will be analyzing Hotel Booking data. This data set contains booking information for a city hotel and a resort hotel, and includes information such as when the booking was made, length of stay, the number of adults, children, and/or babies, and the number of available parking spaces.
- ❖ Hotel industry is a very volatile industry and the bookings depends on above factors and many more.
- ❖ The main objective behind this project is to explore and analyze data to discover important factors that govern the bookings and give insights to hotel management, which can perform various campaigns to boost the business and performance.

Work Flow

I am dividing this work flow into following 3 steps.

1. Data Collection and Understanding

2. Data Cleaning and Manipulation

3. Exploratory Data Analysis(EDA)...

EDA will be divided into following 3 analysis.

- 1)**Univariate analysis:** Univariate analysis is the simplest of the three analyses where the data you are analyzing is only one variable.
- 2)**Bivariate analysis:** Bivariate analysis is where you are comparing two variables to study their relationships.
- 3)**Multivariate analysis:** Multivariate analysis is similar to Bivariate analysis but you are comparing more than two variables.

Data Collection and Understanding:

- ❖ After collecting data it's very important to understand your data. So we had hotel Booking analysis data. Which had 119390 rows and 32 columns. So let's understand this 32 columns.

Data Description:

hotel :Resort Hotel or City Hotel

is_canceled : Value indicating if the booking was canceled (1) or not (0)

lead_time : Number of days that elapsed between the entering date of the booking and the arrival date

arrival_date_year : Year of arrival date

arrival_date_month : Month of arrival date

arrival_date_week_number : Week number of year for arrival date

arrival_date_day_of_month : Day of arrival date

stays_in_weekend_nights : Number of weekend nights

stays_in_week_nights : Number of week nights.

adults : Number of adults

children : Number of children

babies : Number of babies

meal : Type of meal booked.

country : Country of origin.

market_segment : Market segment designation. (TA/TO)

distribution_channel : Booking distribution channel.(T/A/TO)

is_repeated_guest : is a repeated guest (1) or not (0)

previous_cancellations : Number of previous bookings that were cancelled by the customer prior to the current booking

previous_bookings_not_canceled : Number of previous bookings not cancelled by the customer prior to the current booking

reserved_room_type : Code of room type reserved.

assigned_room_type : Code for the type of room assigned to the booking.

booking_changes : Number of changes made to the booking from the moment the booking was entered on the PMS until the moment of check-in or cancellation

deposit_type : No Deposit, Non Refund , Refundable.

agent : ID of the travel agency that made the booking

company : ID of the company/entity that made the booking .

days_in_waiting_list : Number of days the booking was in the waiting list before it was confirmed to the customer

customer_type : type of customer. Contract,Group,transient,Transient party.

adr : Average Daily Rate as defined by dividing the sum of all lodging transactions by the total number of staying nights

required_car_parking_spaces : Number of car parking spaces required by the customer

total_of_special_requests : Number of special requests made by the customer (e.g. twin bed or high floor)

reservation_status : Reservation last status.

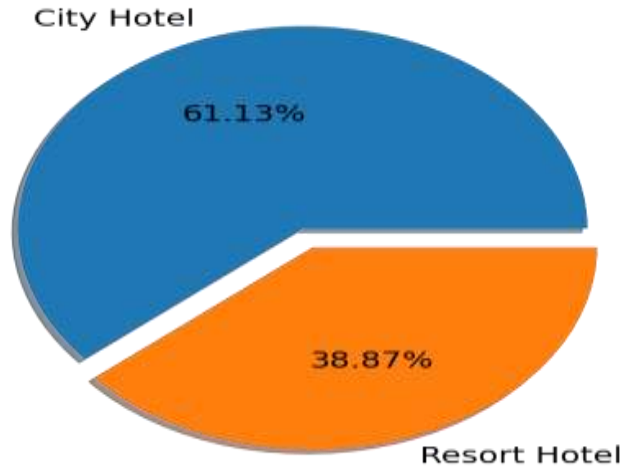
Data Cleaning and Manipulation:

1. company, agent, country and children columns with missing values. I replaced missing values as per requirement.
2. Dropping company column because more than 90% data is missing
3. Data had 31994 duplicates values. So I dropped it from the data.
4. I created 2 new columns
 - A) 'total_People' = from the Children, adults, babies.
 - B) 'total_stay ' = From weekend nights and weekdays night.

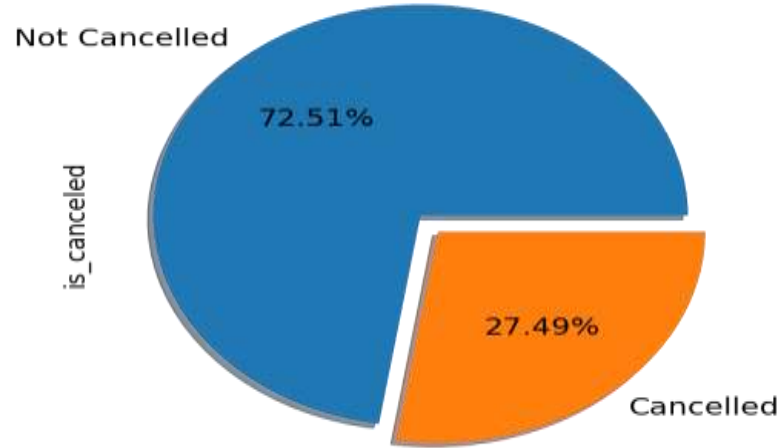
Exploratory Data Analysis (EDA) :

Univariate Analysis

Most Preferred Hotel



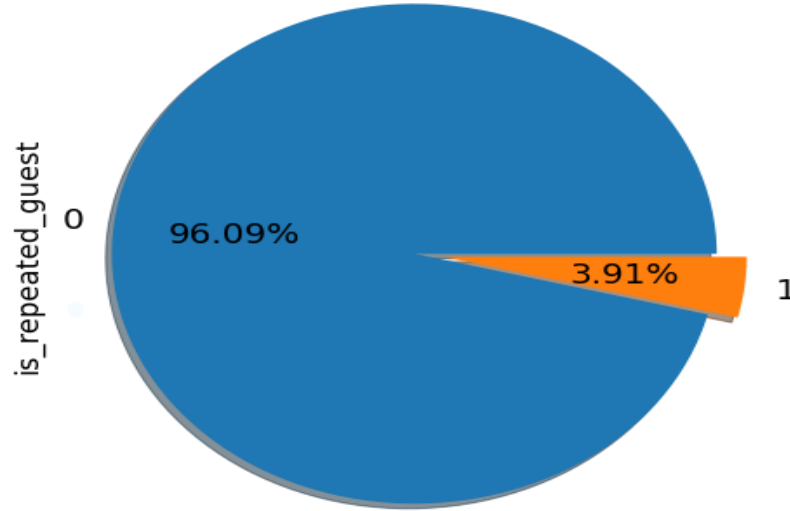
Cancellation and Non Cancellation Rates



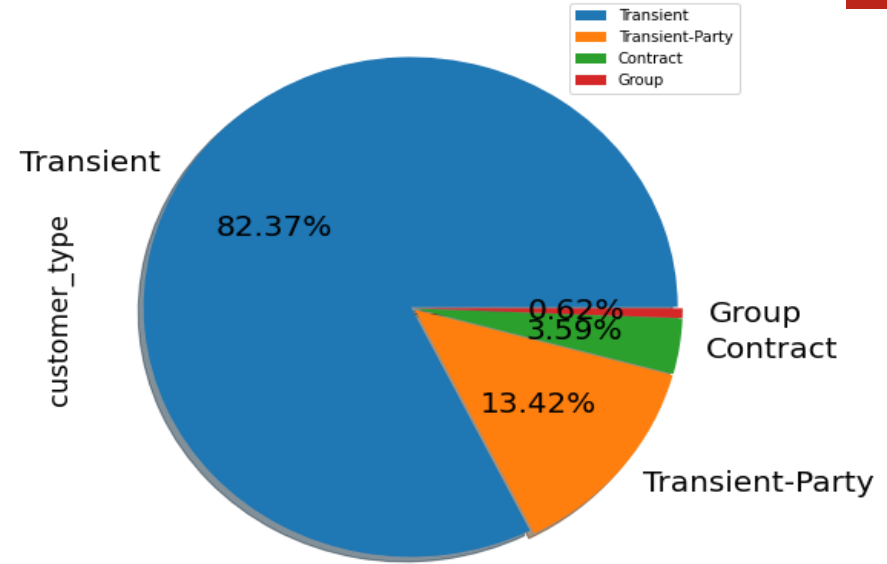
Conclusions:

- City hotels is the most preferred hotel type by the guests. We can say City hotel is the busiest hotel.
- 27.49 % bookings were got cancelled out of all the bookings

Percentage of repeated guest



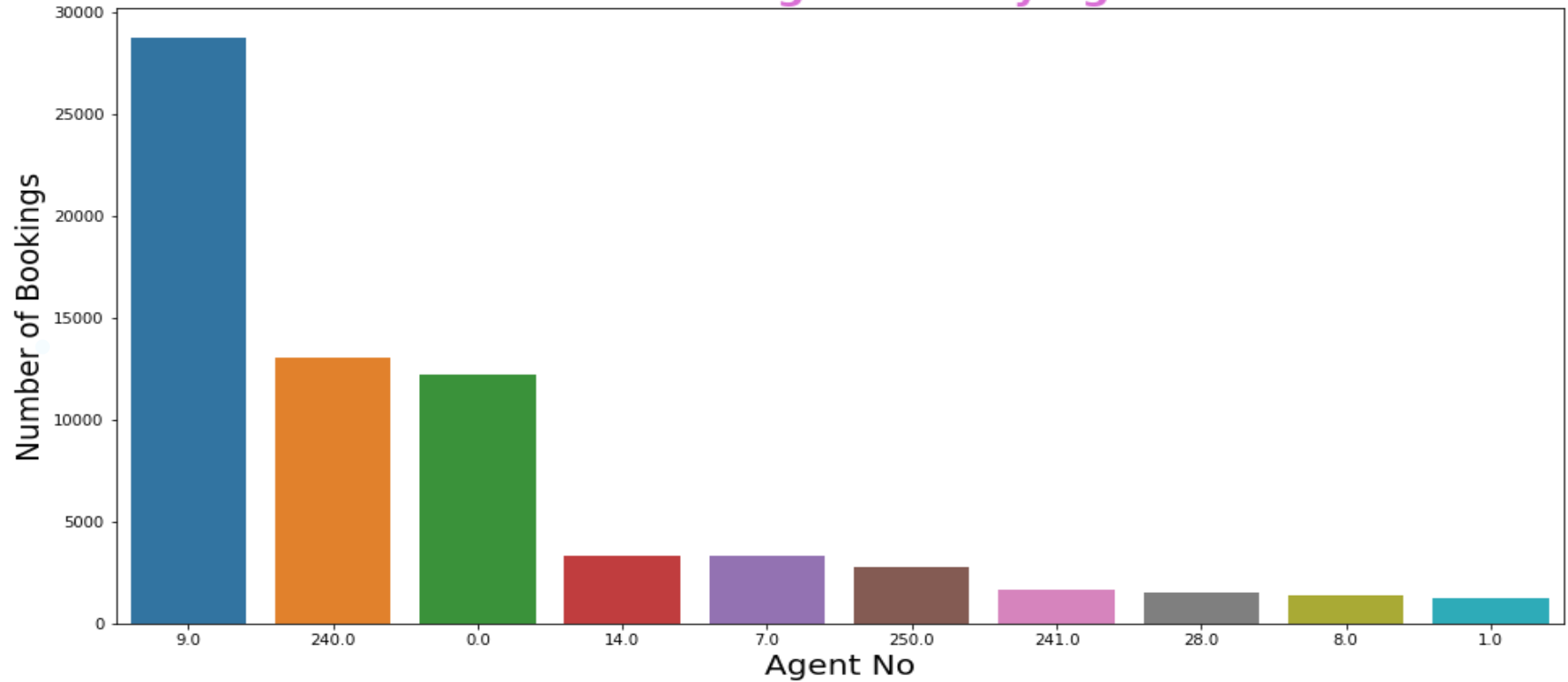
% Distribution of Customer Type



Conclusions:

- Only 3.9 % people were revisited the hotels. Rest 96.09 % were new guests. Thus retention rate is low.
- Most of the customers/guests were Transient type(82.37%). And transient party were 13.42% and 0.62% belongs to group. Remaining guests belongs to Contract type.

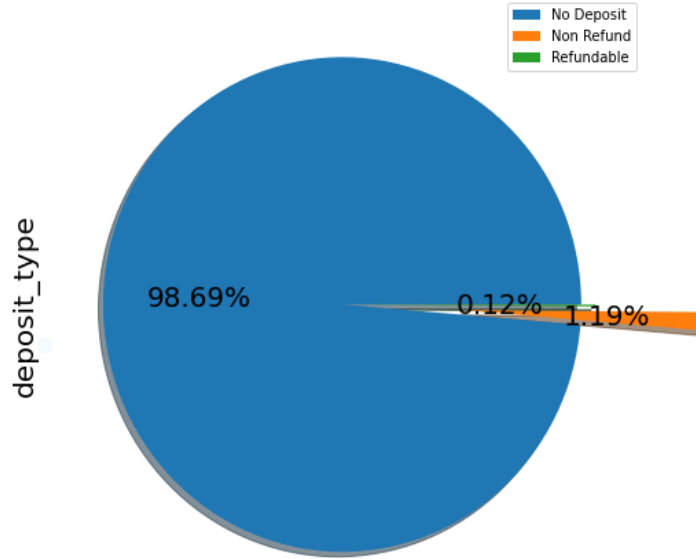
Most Bookings made by agent



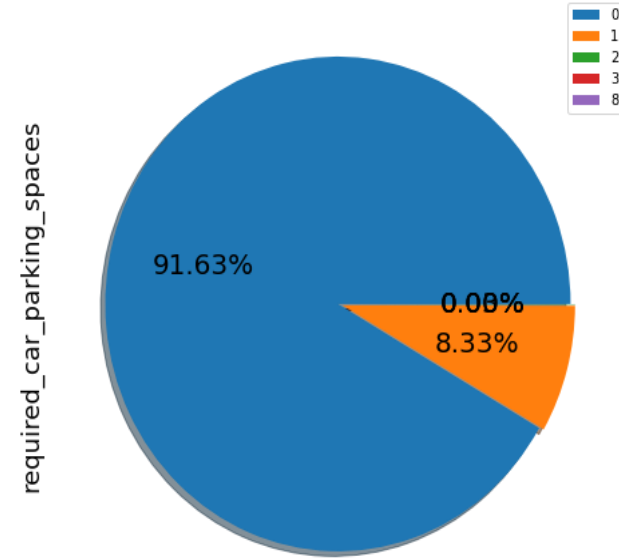
Conclusions:

- . Agent ID no: 9.0 made most of the bookings

% Distribution of deposit type



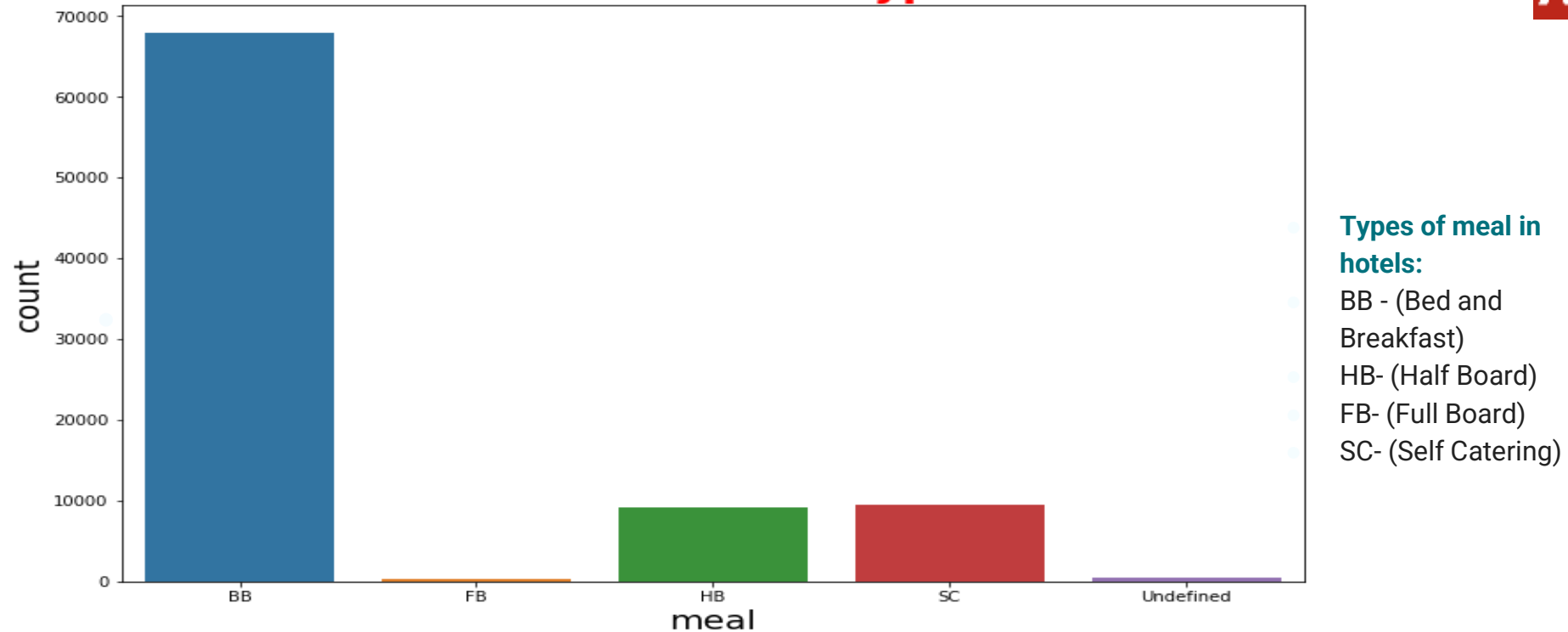
% Distribution of required car parking spaces



Conclusions:

- 98.69 % of the guests prefer "No deposit" type of deposit.
- 91.63 % guests did not required the parking space. only 8.33 % guests required only 1 parking space.

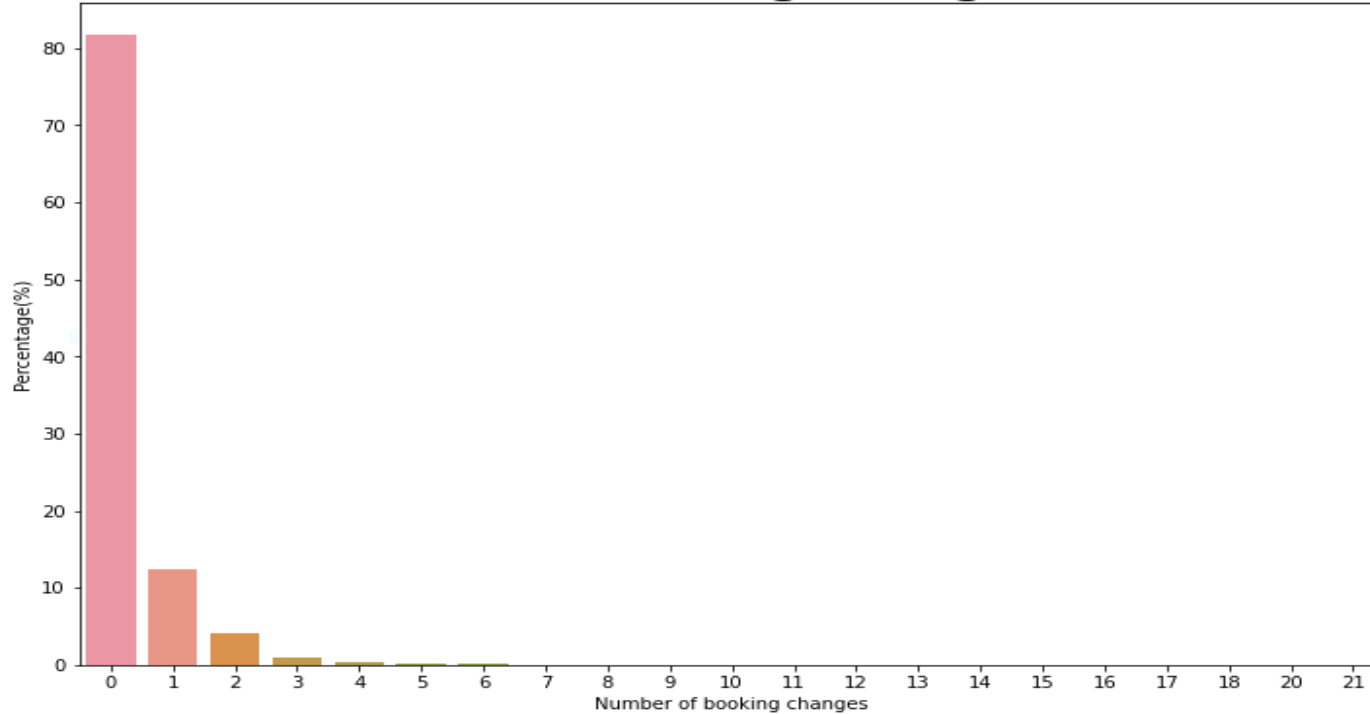
Preferred meal type



Conclusions:

- most preferred meal type by the guests is BB(Bed and Breakfast) HB- (Half Board) and SC- (Self Catering) are equally preferred.

% of Booking change



0 = 0 changes made in the bookings

1 = 1 changes made in the bookings

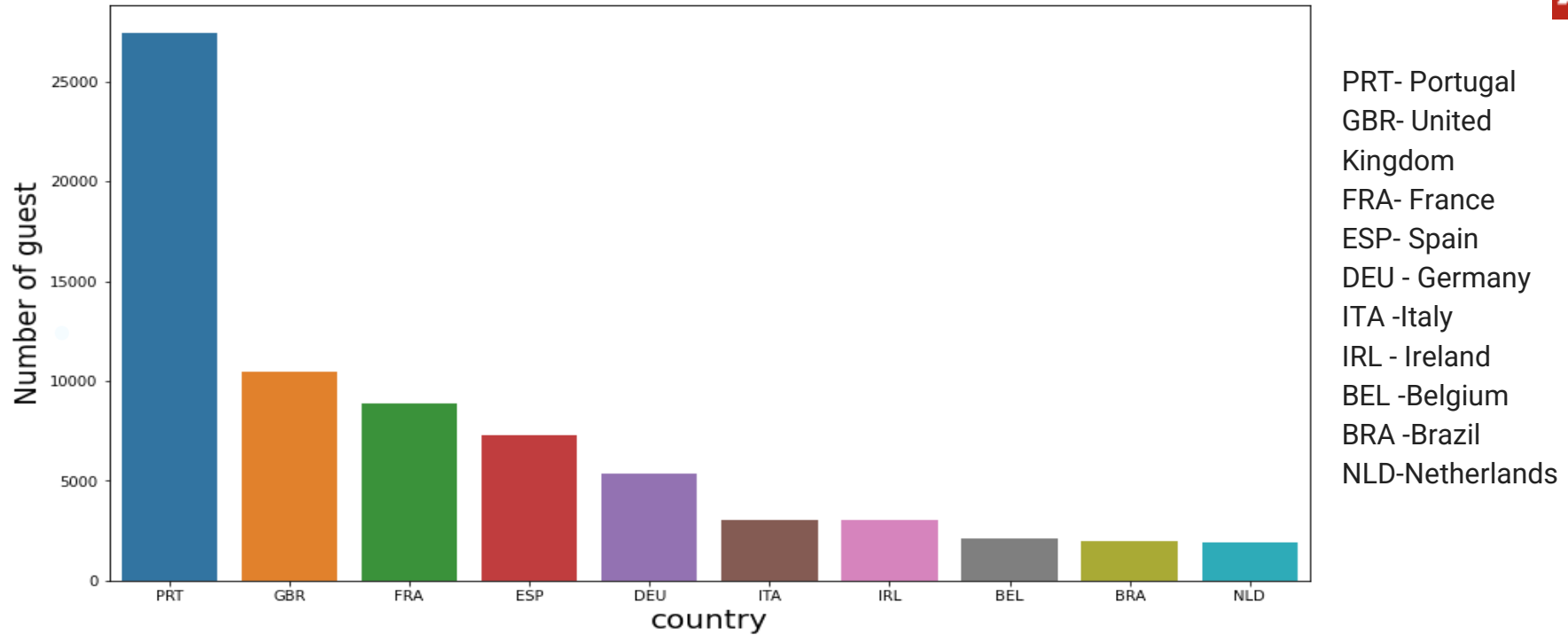
2 = 2 changes made in the bookings

3 = 3 changes made in the bookings

Conclusions:

- . 80% -83% of the bookings were not changed by the customer.

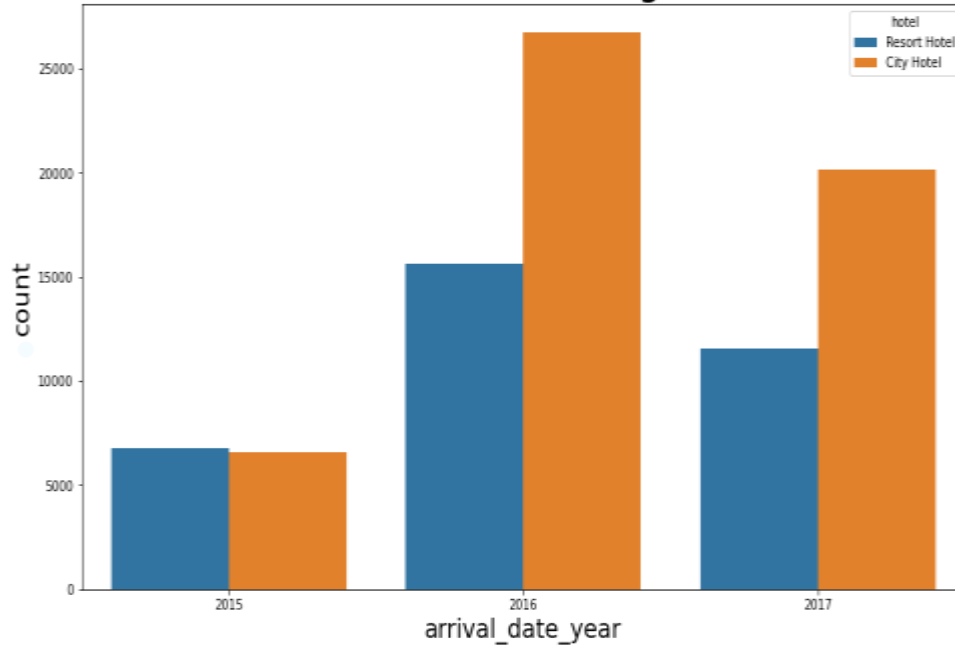
Number of Guests From Different Countries



Conclusions:

- Most of the guests are coming from Portugal i.e. more 25000 guests are from Portugal

Year wise bookings



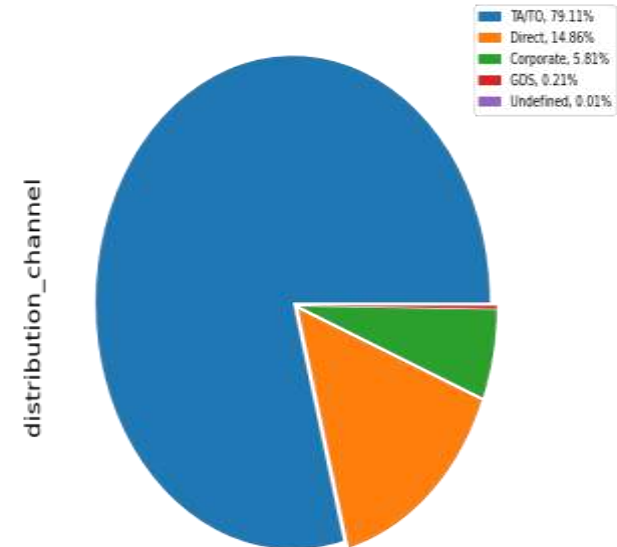
Conclusions:

- 'TA/TO' is mostly(79.11%) used for booking hotels

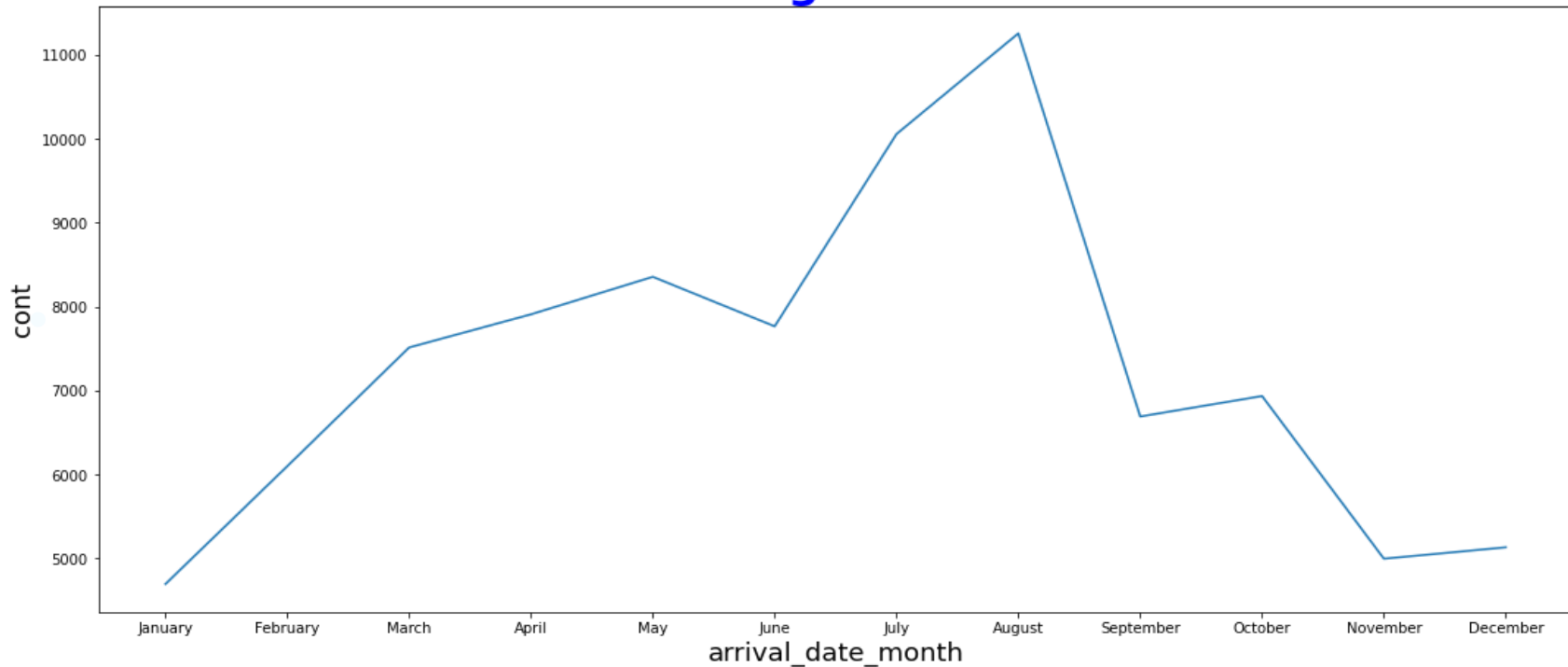
Conclusions:

- 2016 had the highest bookings.
- 2015 had less bookings.
- City hotels had the most of the bookings.

Mostly Used Distribution Channel for Hotel Bookings



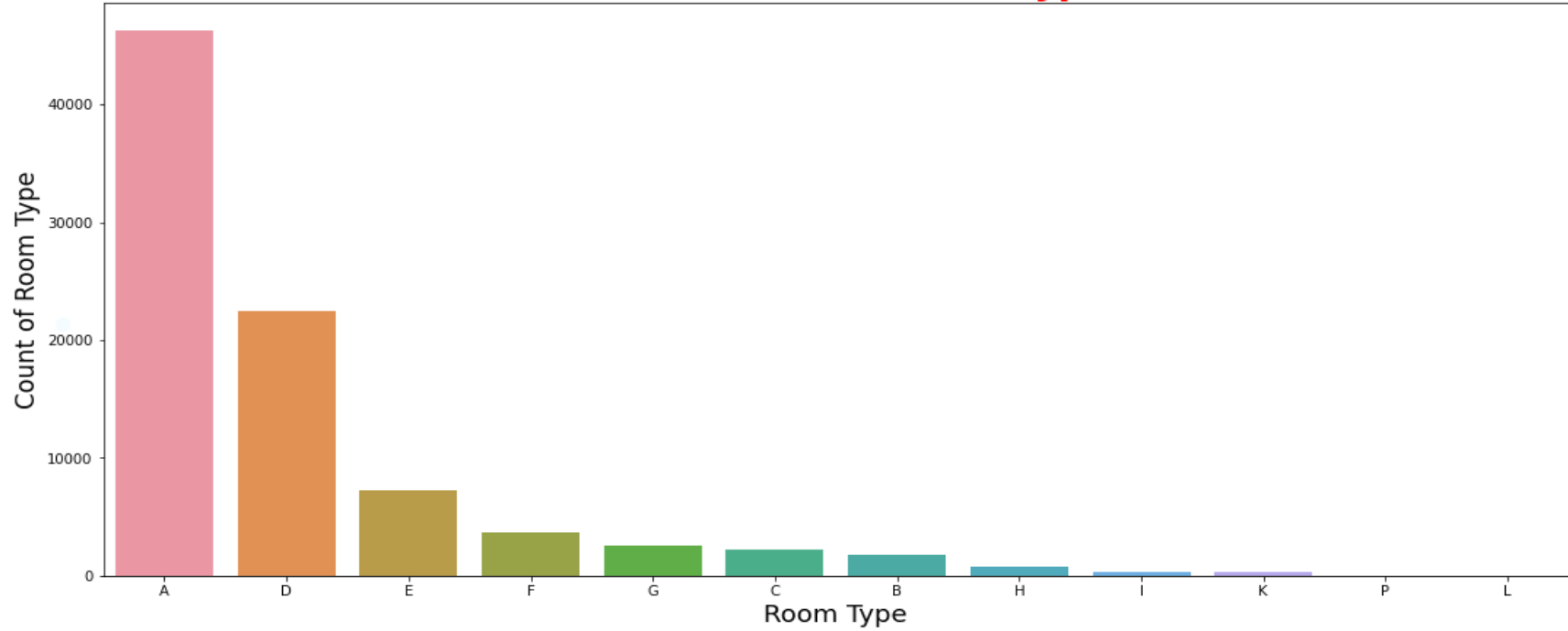
Number of bookings across each month



Conclusions:

- July and August months had the most Bookings. Summer vacation can be the reason for the bookings.

Most Preferred Room Type

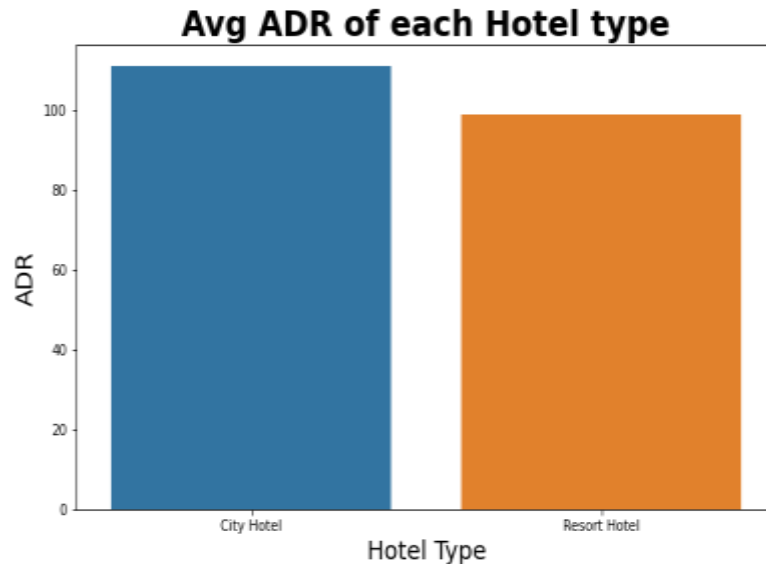
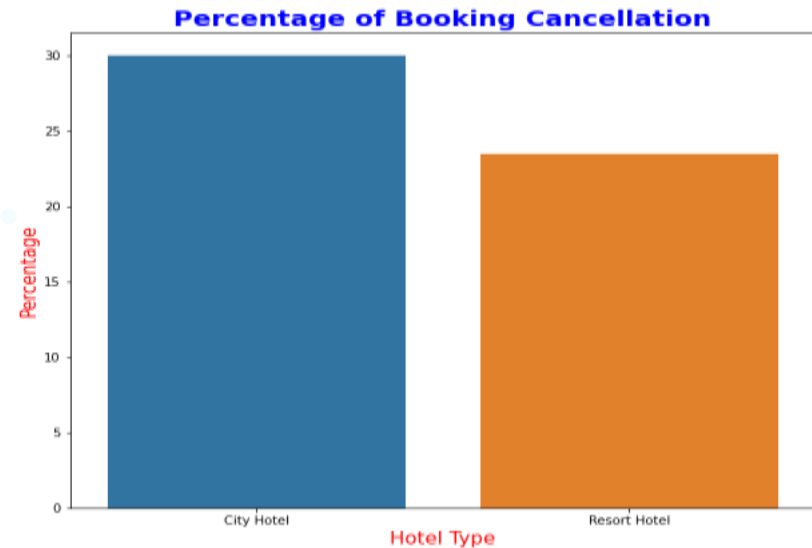


Conclusions:

- The most preferred Room type is "A".

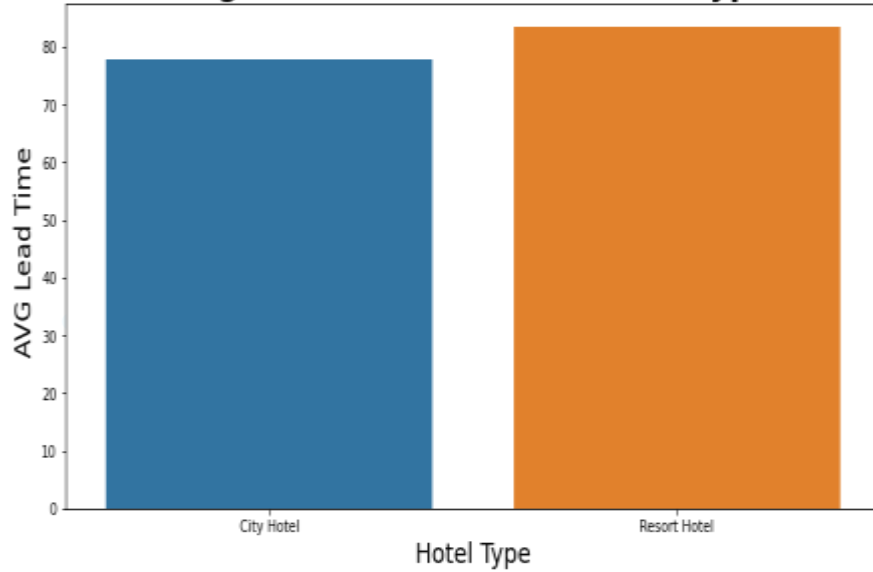
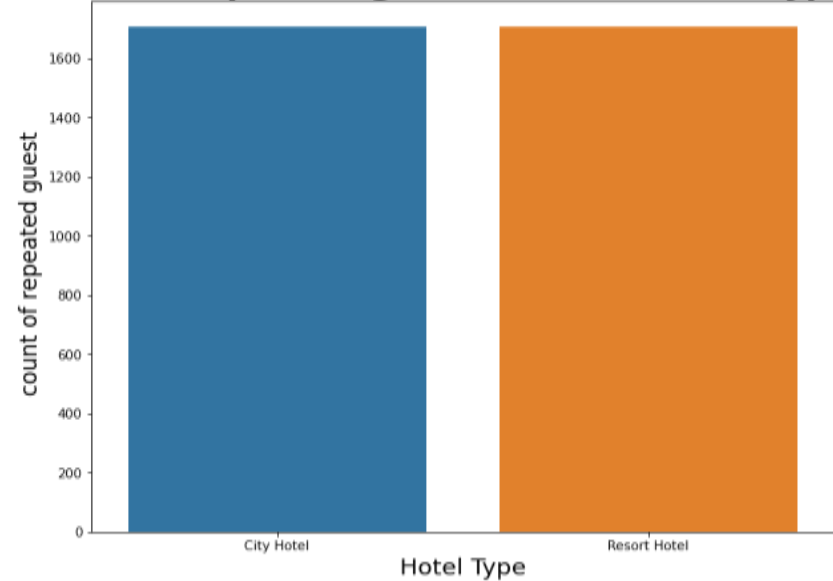
Exploratory Data Analysis (EDA) :

Bivariate and Multivariate Analysis



Conclusions:

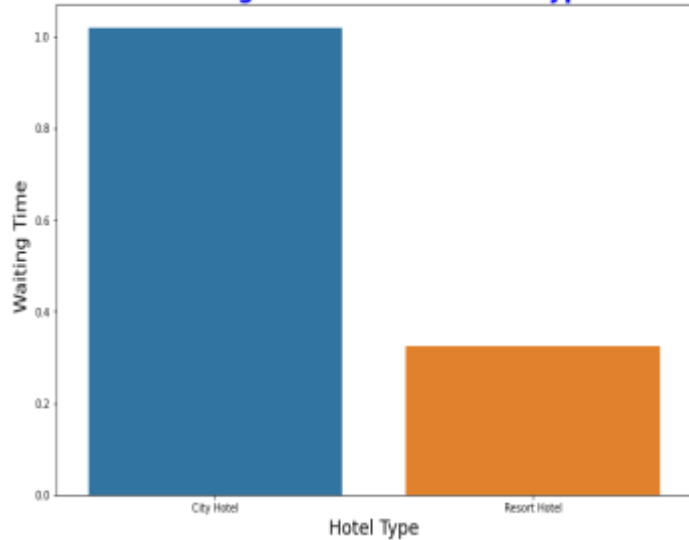
- City Hotel has highest percentage of cancellation which is almost 30%
- City hotel has the highest ADR. That means city hotels are generating more revenues than the resort hotels. More the ADR more is the revenue.

Avg Lead Time for each Hotel Type**Most Repeated guest for each Hotel Type**

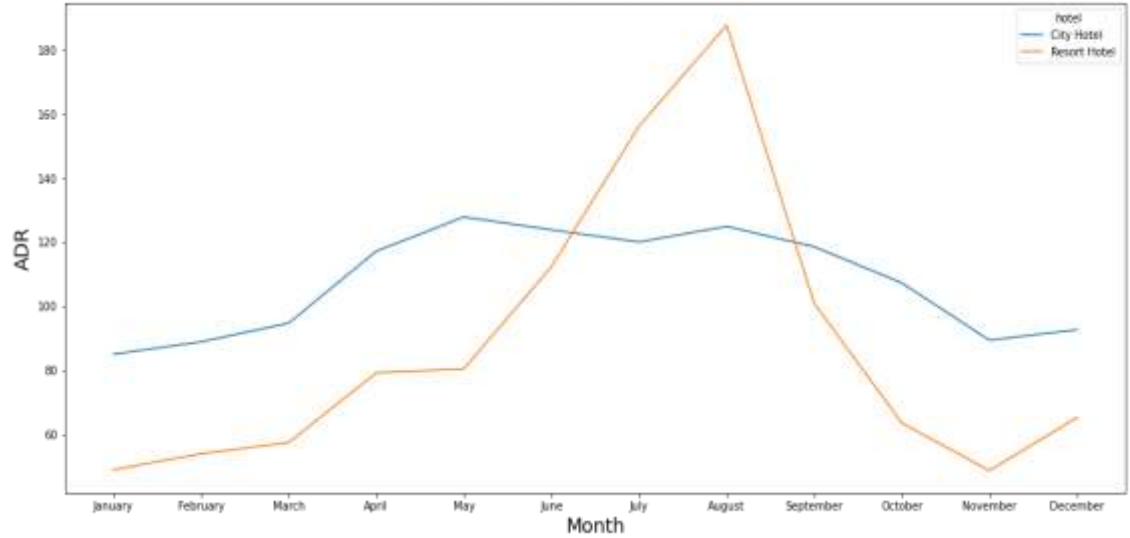
Conclusions:

- Resort hotels has slightly high avg lead time. That means customers plan their trips very early.
- The repeated guest is almost similar for both hotels.

Waiting Time for each Hotel Type



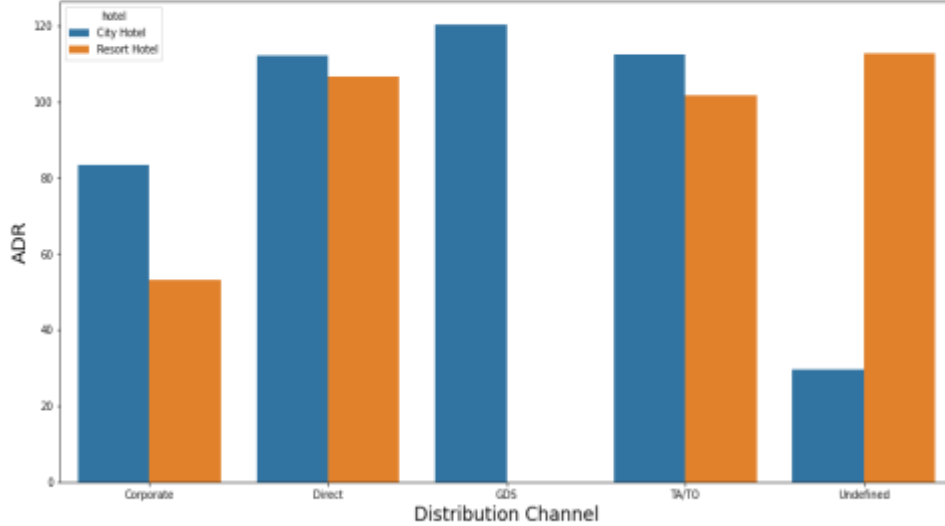
ADR across each month



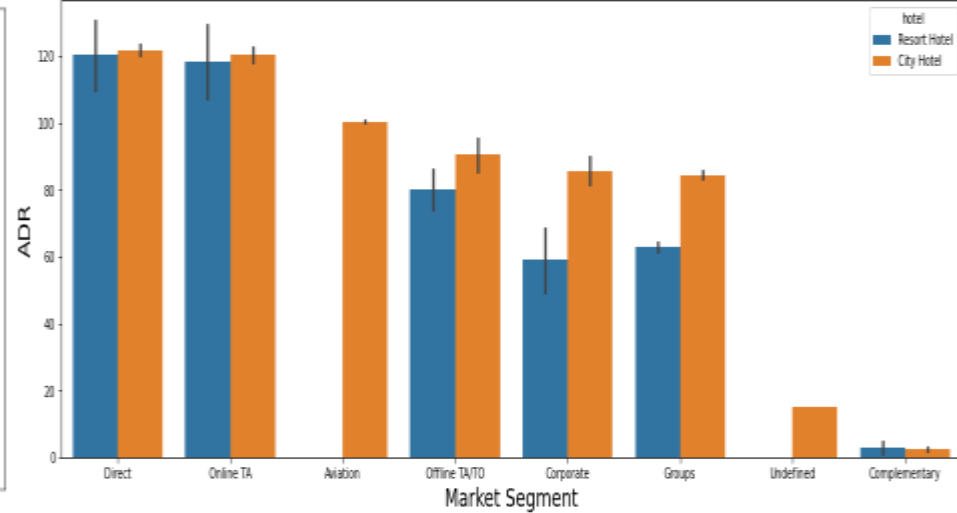
Conclusions:

- So the City Hotels has longer waiting period than the Resort Hotels. Thus we can say that City Hotels are much busier than the Resort Hotels.
- For Resort hotel ADR is high in the months June, July, August as compared to City Hotels. May be Customers/People wants to spend their Summer vacation in Resorts Hotels.
- The best time for guests to visit Resort or City hotels is January, February, March, April, October, November and December as the average daily rate in this month is very low.

ADR across Distribution channel



Adr across market segment



Conclusions:

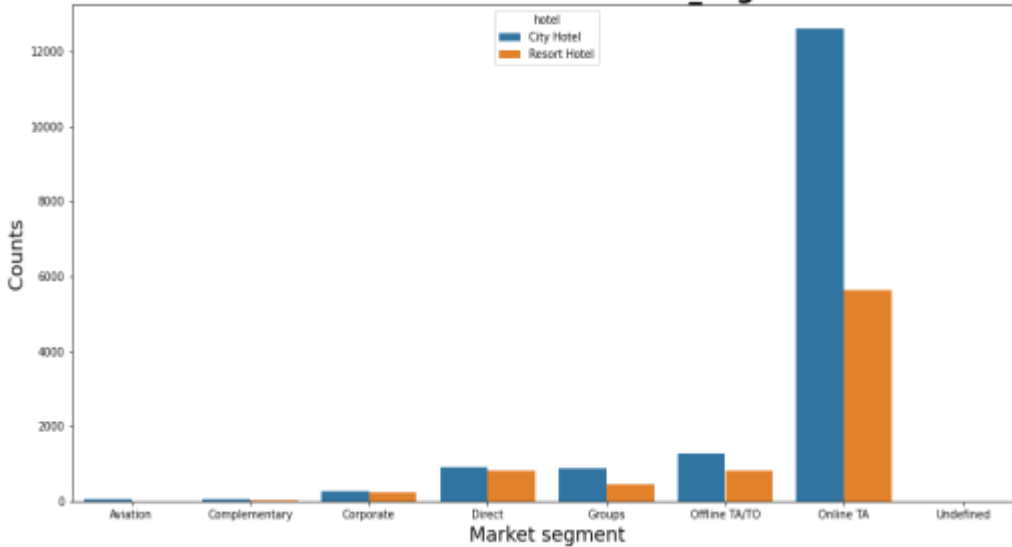
Distribution channel:

- 'Direct' and 'TA/TO' has almost equal adr in both type of hotels which is high among other channels.
- GDS has high adr in 'City Hotel' type. GDS needs to increase Resort Hotel bookings. From this we can say that "Direct" and 'TA/TO' are generating more revenue than the other channels.

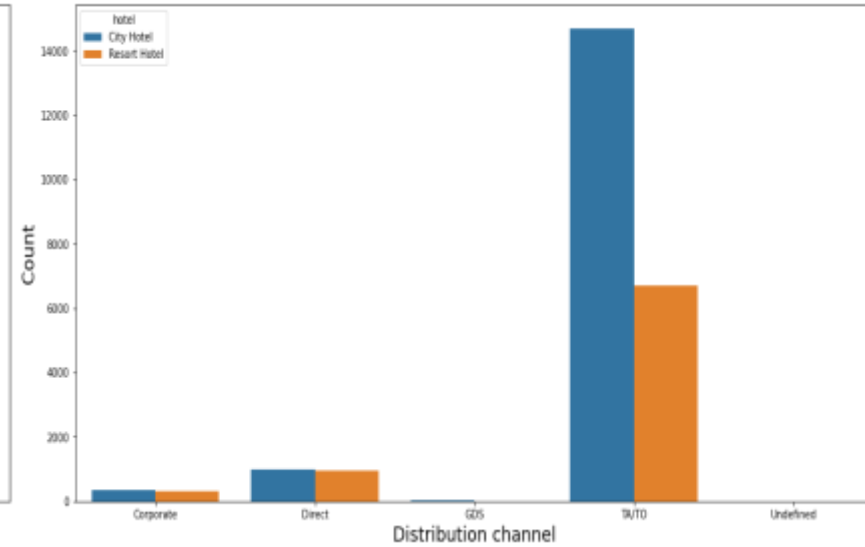
Market Segment:

- Here "Direct" and 'Online Travel Agency' has high adr for both hotel types. Aviation segment needs to increase Resort hotel bookings.

Cancellation Rate Vs market_segment



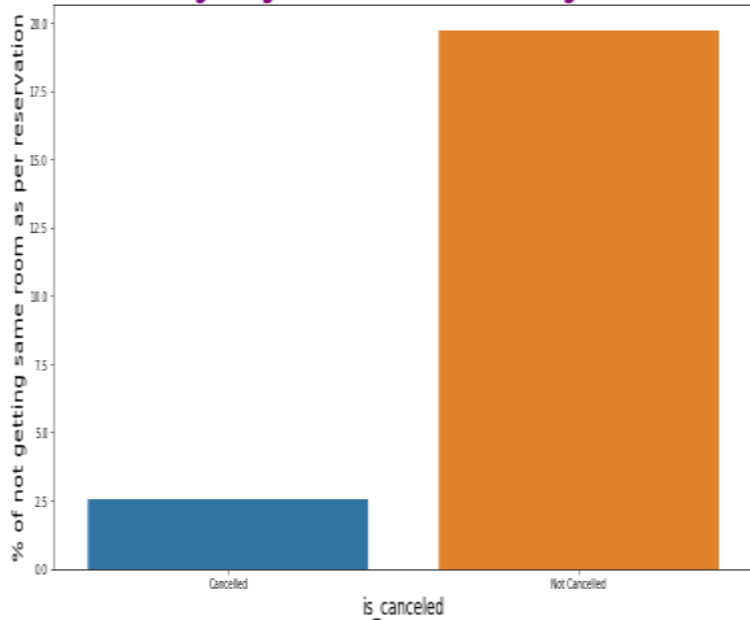
Cancellation Rate Vs Distribution channel



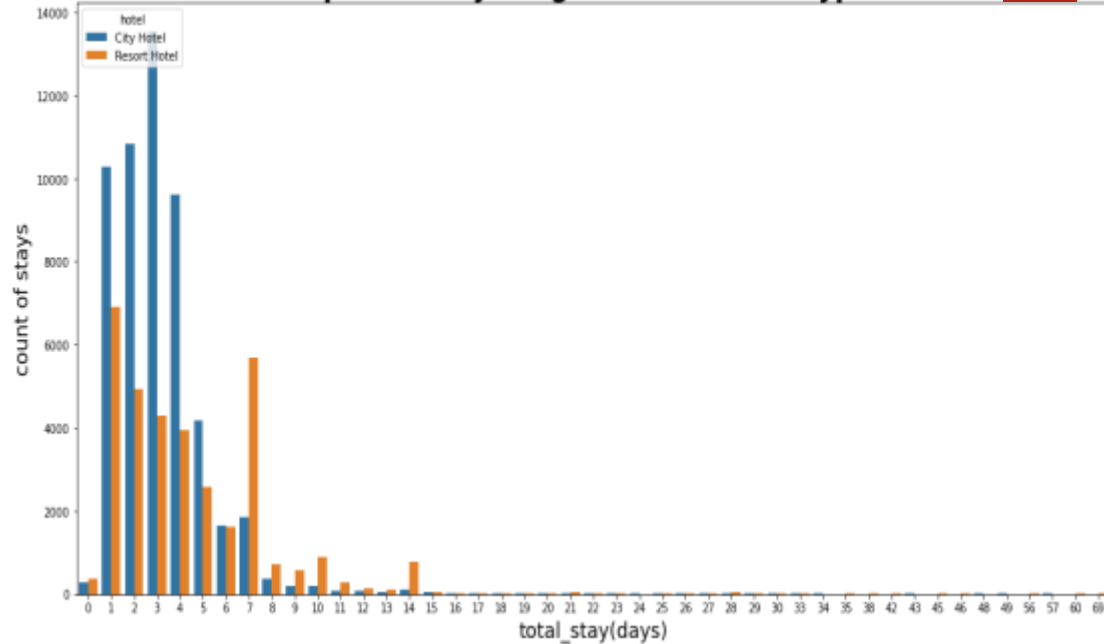
Conclusions:

- 'Online T/A' has the highest cancellation in both type of Hotel
- In order to reduce the booking cancellations hotels need to set the refundable/ no refundable and deposit policies
- In "TA/TO", City hotels has the high cancellation rate compared to resort hotels.
- In "direct" both the hotels has almost same cancellation rate..

% of not getting the same room vs Booking cancellation



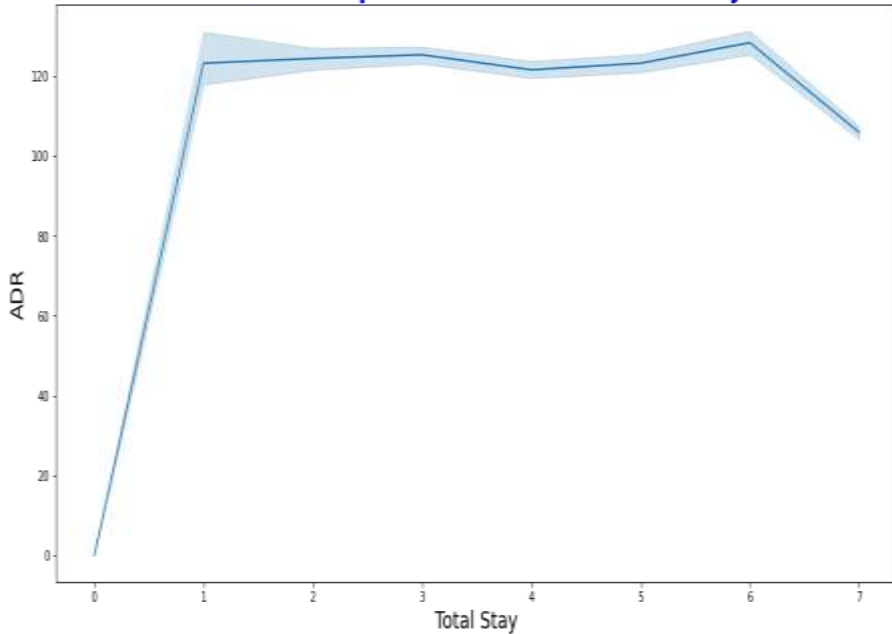
Optimal Stay Length in Both hotel types



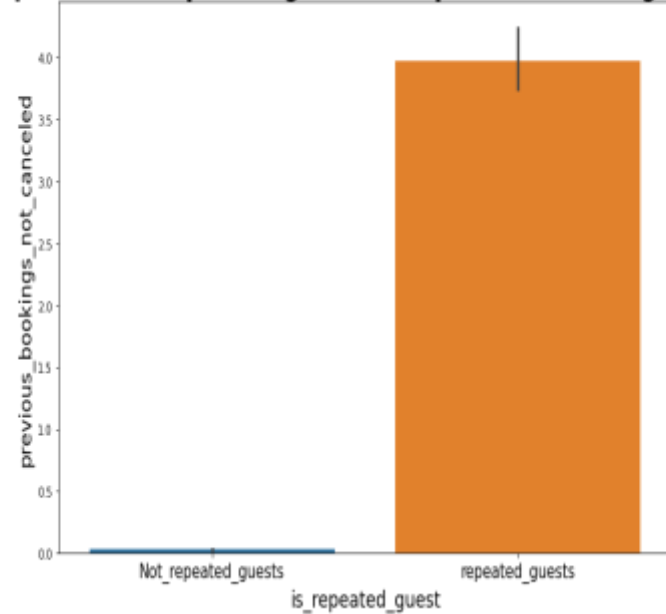
Conclusions:

- It's clear that there is not much (2.5%) effect on cancellation of the bookings even if the guests are not assigned with rooms which they reserved during booking.
- Optimal stay in both the type of hotel is less than 7 days.

Relationship between adr and total stay



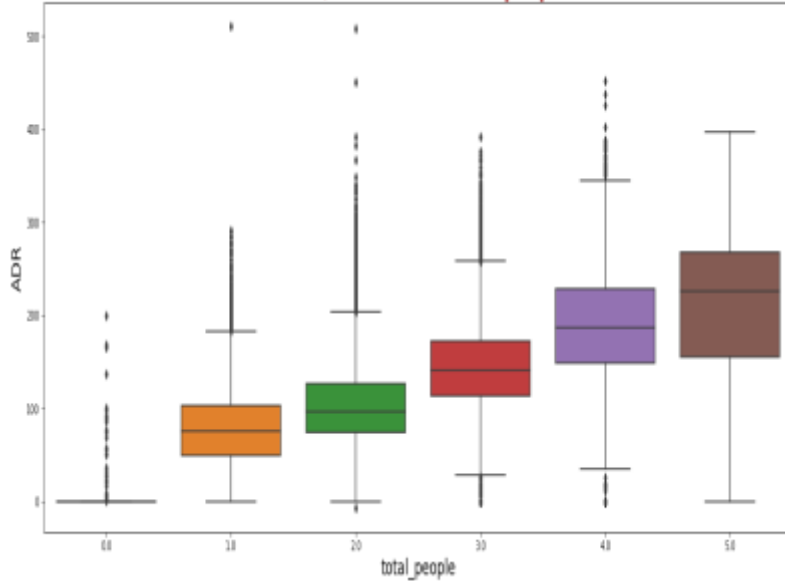
Relationship Between repeated guests and previous bookings nor canceled.



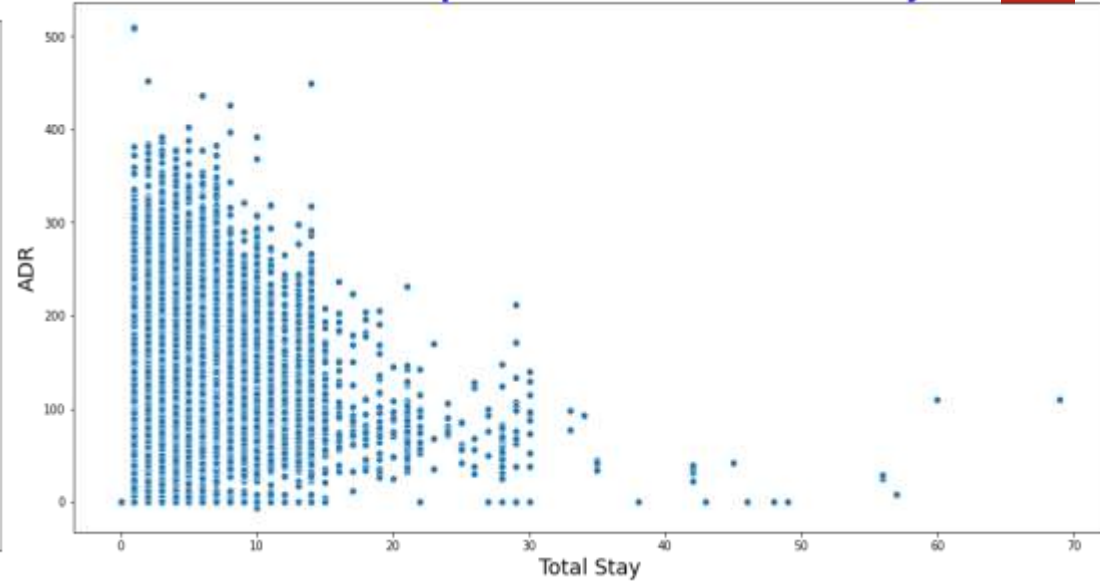
Conclusions:

- As the total stay increases the adr also increases.
- Not Repeated guests are more likely to cancel their bookings.

ADR v/s Total Number of people



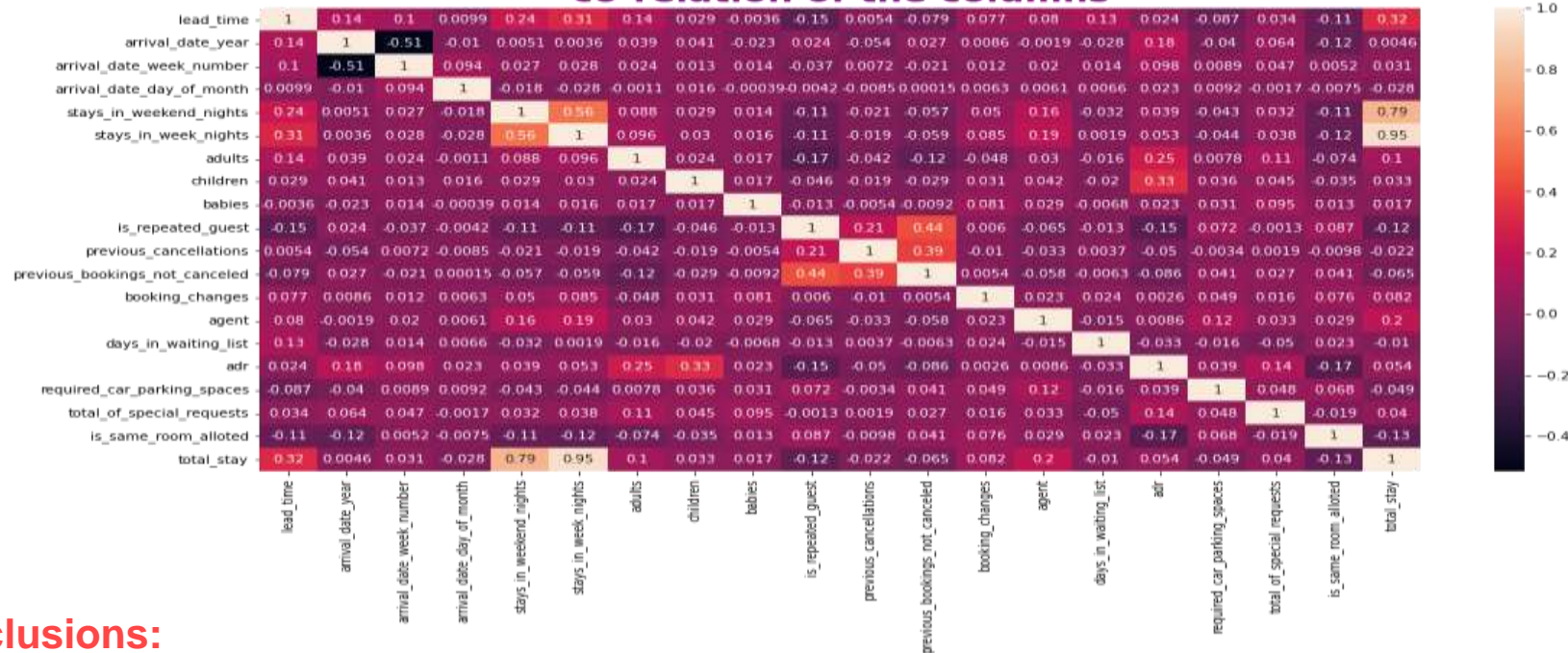
Relationship between adr and total stay



Conclusions:

- As the total number of people increases adr also increases.
- Thus adr and total people are directly proportional to each other.
- From above scatter we can say that as the stay increases adr is decreasing. Thus for longer stays customer can get good adr.

co-relation of the columns



Conclusions:

- is_canceled and same_room_alloted_or_not are negatively correlated. That means customer is unlikely to cancel his bookings if he don't get the same room as per reserved room. We have visualized it above.
- lead_time and total_stay is positively correlated. That means more is the stay of customer more will be the lead time.
- adults, childrens and babies are correlated to each other. That means more the people more will be adr.
- is_repeated_guest and previous_bookings_not_canceled has strong correlation. may be repeated guests are not more likely to cancel their bookings.

THANK YOU