

Hotel Booking Pattern & Analysis

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Analysis of data received from Room Booking Platform

Online platforms such as trivago, goibibo, makemytrip etc. are used for booking hotel rooms, here is a dataset of rooms booked from such platform.

Tasks

- To understand the pattern of bookings, and the general trends followed by users.
- Create a report, as being a part of the platform, for the marketing team.

Let's get started

Importing necessary libraries

```
library(tidyverse)
```

Import data from the csv

```
datas <- read.csv("dataset/hotel_bookings.csv")
```

Get the str of the data

```
str(datas)
```

```
## 'data.frame':   119390 obs. of  32 variables:
## $ hotel          : Factor w/ 2 levels "City Hotel","Resort Hotel": 2 2 2 2 2 2 2 2 2 2 ...
## $ is_canceled    : int   0 0 0 0 0 0 0 0 1 1 ...
## $ lead_time      : int   342 737 7 13 14 14 0 9 85 75 ...
## $ arrival_date_year : int   2015 2015 2015 2015 2015 2015 2015 2015 2015 2015 ...
## $ arrival_date_month : Factor w/ 12 levels "April","August",...: 6 6 6 6 6 6 6 6 6 6 ...
## $ arrival_date_week_number : int   27 27 27 27 27 27 27 27 27 27 ...
## $ arrival_date_day_of_month : int    1 1 1 1 1 1 1 1 1 1 ...
## $ stays_in_weekend_nights : int    0 0 0 0 0 0 0 0 0 0 ...
## $ stays_in_week_nights    : int    0 0 1 1 2 2 2 2 3 3 ...
```

```

## $ adults          : int  2 2 1 1 2 2 2 2 2 2 ...
## $ children        : int  0 0 0 0 0 0 0 0 0 0 ...
## $ babies          : int  0 0 0 0 0 0 0 0 0 0 ...
## $ meal            : Factor w/ 5 levels "BB","FB","HB",...: 1 1 1 1 1 1 1 2 1 3 ...
## $ country         : Factor w/ 178 levels "ABW","AGO","AIA",...: 137 137 60 60 60 60 137 ...
## $ market_segment : Factor w/ 8 levels "Aviation","Complementary",...: 4 4 4 3 7 7 4 4 ...
## $ distribution_channel : Factor w/ 5 levels "Corporate","Direct",...: 2 2 2 1 4 4 2 2 4 4 ...
## $ is_repeated_guest : int  0 0 0 0 0 0 0 0 0 0 ...
## $ previous_cancellations : int  0 0 0 0 0 0 0 0 0 0 ...
## $ previous_bookings_not_canceled: int  0 0 0 0 0 0 0 0 0 0 ...
## $ reserved_room_type : Factor w/ 10 levels "A","B","C","D",...: 3 3 1 1 1 1 3 3 1 4 ...
## $ assigned_room_type : Factor w/ 12 levels "A","B","C","D",...: 3 3 3 1 1 1 3 3 1 4 ...
## $ booking_changes    : int  3 4 0 0 0 0 0 0 0 0 ...
## $ deposit_type       : Factor w/ 3 levels "No Deposit","Non Refund",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ agent              : Factor w/ 334 levels "1","10","103",...: 334 334 334 157 103 103 334 ...
## $ company            : Factor w/ 353 levels "10","100","101",...: 353 353 353 353 353 353 353 ...
## $ days_in_waiting_list : int  0 0 0 0 0 0 0 0 0 0 ...
## $ customer_type      : Factor w/ 4 levels "Contract","Group",...: 3 3 3 3 3 3 3 3 3 3 ...
## $ adr                : num  0 0 75 75 98 ...
## $ required_car_parking_spaces : int  0 0 0 0 0 0 0 0 0 0 ...
## $ total_of_special_requests : int  0 0 0 0 1 1 0 1 1 0 ...
## $ reservation_status : Factor w/ 3 levels "Canceled","Check-Out",...: 2 2 2 2 2 2 2 2 1 1 ...
## $ reservation_status_date : Factor w/ 926 levels "2014-10-17","2014-11-18",...: 122 122 123 123 123 123 123 123 123 123 ...

```

Understanding the columns

The data looks clean enough, with proper column headers, as well

- Hotel has two types:
 - Resort hotel
 - City hotel
- Is_cancelled:
 - “1” if the booking is cancelled
- lead_time
 - No of days between booking and booked date
- Arrival : year, month, week_number, day
- Stay
 - No of weekend nights
 - No of week nights (because of price difference during the weekends)
- No of people: adults, children, babies
- Meal booked has 5 types
 - BB - Bed & Breakfast
 - FB - Full Board (Breakfast, Lunch & Dinner)
 - HB - Half Board (Breakfast + 1 other (dinner or lunch, mostly dinner))
 - SC - No Meal package

- Undefined - No Meal package
- Country (self - explanatory)
- market_segment (group of people who share common characteristic)
- distribution_channel (intermediaries between users and hotel booking eg. websites, travel agents, tour operators)
- is_repeated_guest (has previous booking)
- previous_cancellations (has previously cancelled a booking)
- reserved_room_type (type of room reserved)
- assigned_room_type (type of room assigned, due to high volume this can differ from reserved_room_type)
- booking_changes (no of times changes have been made to the booking)
- deposit_type
 - No Deposit
 - Non Refund - deposit of value equals total cost
 - Refundable - value under the total cost of stay
- agent
 - ID of travel agency that made the booking
- Company
 - ID of the company responsible for booking or payment
- days_in_waiting_list (no of days before the booking was confirmed)
- customer_type
 - Contract
 - Group
 - Transient
 - Transient-party
- adr (average daily rate)
 - $adr = (\text{sum_of_all_expenses})/(\text{total_nights_of_stay})$
- required_car_parking_spaces
- total_of_special_requests
- reservation_status
 - Canceled
 - Check-Out
 - No Show - Customer did not show up
- reservation_status_date
 - Date when the final changes to the entry was made.

Calculating the NA, values

```
colSums(is.na(datas))
```

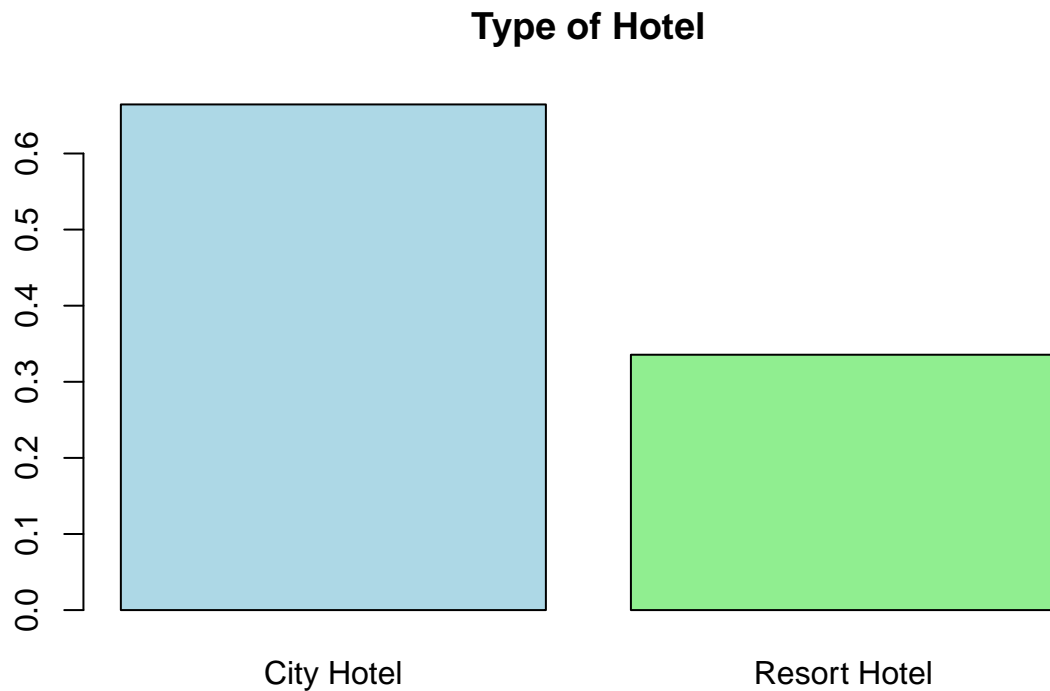
```
##          hotel          is_canceled
##          0          0
##      lead_time      arrival_date_year
##          0          0
##      arrival_date_month      arrival_date_week_number
##          0          0
##      arrival_date_day_of_month      stays_in_weekend_nights
##          0        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

There are no such columns with na values that need to be removed or edited

Hotel

There are just two types of hotel, Resort & City, so a basic barplot would give the idea of the percentage of booking.

```
counts <- prop.table(table(datas$hotel))
barplot(counts, col = c('lightblue', 'lightgreen'), main = "Type of Hotel")
```



Analysis

City hotels are twice as much booked compared to Resort hotels, following reasons can be derived for that.

- City hotels are better options for corporate bookings, and business purposes
- Resort hotels can be a good option for larger parties.

Cancelled bookings

To understand what percentage of bookings are cancelled.

```
cancelled <- prop.table(table(datas$is_canceled))  
barplot(cancelled, main = "Percentage of bookings cancelled", names.arg = c("Not Cancelled", "Cancelled"))
```

Percentage of bookings cancelled

