Hotel Data Visualization With Python

1.INTRODUCTION

PROJECT REPORT

PROJECT TITLE: Hotel Data Visualization With Python

TEAM SIZE:4

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Intro

Have you been curious to find out whether people loved to visit hotels or resorts more? or what sort of meals do people prefer most in hotels? If you are curious just like me, don't worry, these are some of the interesting discoveries and many more we are going to be uncovering in the visualization and analysis of the hotel data.

For someone given to traveling and vacations, visualizing the Hotel booking demand data was an exciting one for me. I was able to discover amazing things about hotels and the behavior patterns of people who visit them and I will love to share those amazing discoveries with you in this article and am sure you gonna love them.

So sit tight as we explore the hotel's data

Data Source

This data was sourced from Kaggle's data repository. Kaggle is a public online repository of data that allows users to access and analyze data from taken from different sources. The hotel data shows quite a number of features that shows different characteristics of different people who visit the hotels. Some of the characteristics include arrival day, lead time, deposit type, and a lot more features. You can find the data here on **Kaggle**.

DATA WRANGLING AND VISUALIZATION

The python programming language is a cool language in the sense that it has a lot of cool libraries that make data visualization, and Data Wrangling and Visualization

The python programming language is a cool language in the sense that it has a lot of cool libraries that make data cleaning, sorting, processing easy. The most popular of them is the panda's library which is built on top of the NumPy library. It manages different tasks in data analysis from loading the data, to cleaning and processing it to even visualizing the data. The popular Matplotlib library is another library that was also built on the panda's library to make visualization very easy through the different functions it provides. The Seaborn library was also developed to create more attractive and illuminating statistical graphics to give insights into the plots

If you need more information about this library and how they work, you can check out their documentation.

Here is a quick overview of what the data looks like

	0	1	2	3	4	5	6	7	8	9	***	119380	119381
hotel	Resort Hotel	***	City Hotel	City Hote									
is_canceled	0	0	0	0	0	0	0	0	1	1	14.	0	0
lead_time	342	737	7	13	14	14	0	9	85	75	***	44	188
arrival_date_year	2015	2015	2015	2015	2015	2015	2015	2015	2015	2015	***	2017	2017
arrival_date_month	July		August	August									
arrival_date_week_number	27	27	27	27	27	27	27	27	27	27	14.	35	35
arrival_date_day_of_month	1	1	1	1	1	1	1	1	1	1		31	31
stays_in_weekend_nights	0	0	0	0	0	0	0	0	0	0	***	1	2
stays_in_week_nights	0	0	1	1	2	2	2	2	3	3	***	3	3
adults	2	2	1	1	2	2	2	2	2	2	ic.	2	2
children	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	***	0.0	0.0
babies	0	0	0	0	0	0	0	0	0	0	***	0	0
meal	ВВ	BB	BB	BB	BB	ВВ	ВВ	FB	BB	нв	***	SC	BE
country	PRT	PRT	GBR	GBR	GBR	GBR	PRT	PRT	PRT	PRT	224	DEU	DEU
market_segment	Direct	Direct	Direct	Corporate	Online TA	Online TA	Direct	Direct	Online TA	Offline TA/TO	***	Online TA	Direc
distribution_channel	Direct	Direct	Direct	Corporate	TA/TO	TA/TO	Direct	Direct	TA/TO	TA/TO		TA/TO	Direc
is_repeated_guest	0	0	0	0	0	0	0	0	0	0	***	0	(
previous_cancellations	0	0	0	0	0	0	0	0	0	0		0	

The codes can be found here on Github.

Let's hop right into a visualization of the different columns

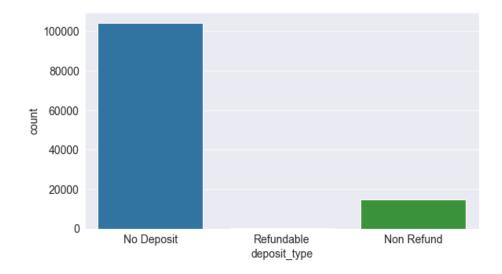
Hotel Type



Observation

We can see that about times two of the total number of people who booked hotels prefer the city hotels to the resort hotels.

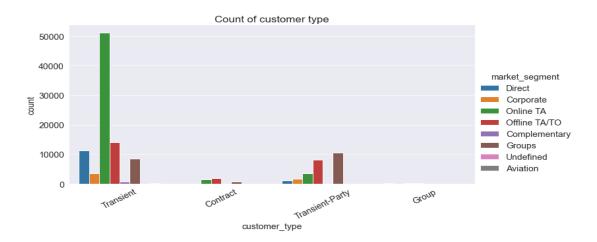
DEPOSITE TYPE



OBSERVATION

This shows that quite a high number of people did not make any deposits

Customer Types in the Market Segments

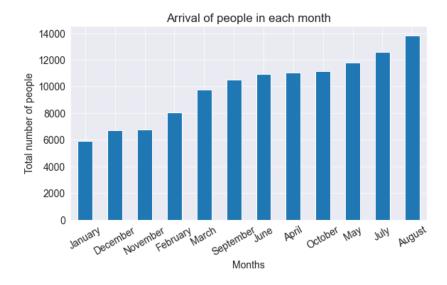


observation

From the chart, we can see:

- The largest number of people who visit the hotel is the transient customers and a huge number of them come from the online TA market segment next to the offline TA/TO
- $\bullet \quad$ A very few people who book at the hotels at contract customers
- A good number of transient-party customers come from the groups market segment next to the offline TA/TO.

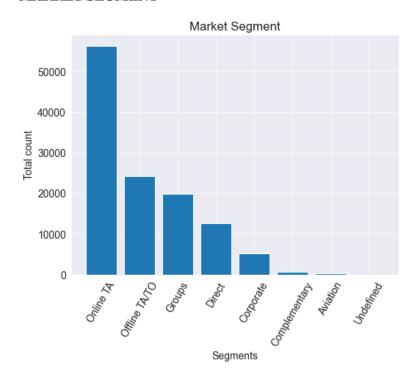
Arrival Months



observation

More people arrived at the hotel in the month of August while the least number of people arrived in the month of January

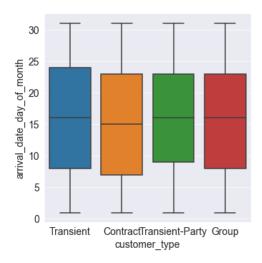
MARKET SEGMENT



Observation

The market segment that patronizes the hotel most omes from the online TA while the least comes from the aviation

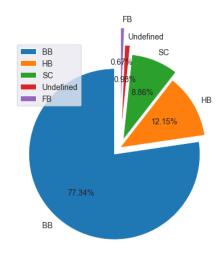
Which day do people arrive most at the hotel



Observation

On average, most customers arrive on the 15th day of the month.

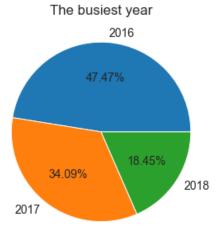
Which meal do cutomer prefer more



Observation

From here, we can see that a lot of customers prefer breakfast in bed(BB) while just a few prefer FB

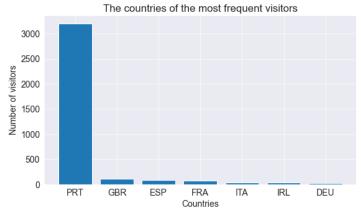
Which year is the busiest year



Observation

From the data, 2016 has been the busiest year

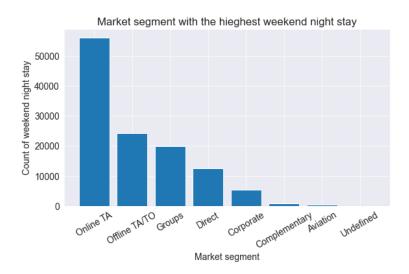
Which country do the most frequent visitors come from



Observation

Most frequent visitors of this hotel are from PRT

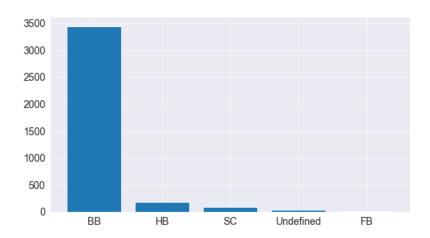
Which market segment has the highest stay in weekend night



Observation

The online TA had more weekend stays next to the offline TA/TO than to the groups and the rest

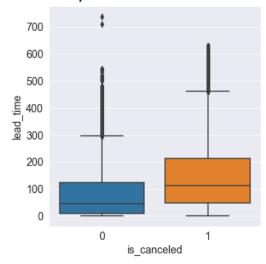
What type of meals do the repeated guests enjoy most?



observation

Most of the returning guests also love breakfast in bed

Relationship between lead time and canceled bookings



People who canceled their booking have a higher lead time on average than those who didn't cancel theirs. A higher variance also exists among those who didn't cancel their booking.

The codes can be found here.

Modeling

After performing a lot of visualization, I had to use *Pandas.getdummies* on all the categorical variables for easy modeling. This is a classification task. So, **Logistic Regression** is the best fit for it (Don't worry, it isn't as technical as it sounds). I imported **confusion_matrix** and **f1_score** to measure the classification accuracy.

Conclusion

- A lot of people loved to visit the city hotels than the resort hotels
- More people arrived at the hotel in the month of August which makes perfect sense because the
 month of August falls in the summer break of the northern hemisphere while the lowest falls in
 the month of January which is quite explainable because January is the beginning of the year and
 book tend to focus more on work
- Majority of people who book at the hotels come online TA market segment since a lot of people spend most of their time on the internet thereby making it easy for them to book rooms at the hotel
- Most customers prefer Breakfast in Bed(BB). After all, who doesn't like enjoyment?