

King Saud University
College of Business Administration
First Semester
2021-2022



# MIS419 Group Project

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Section: 58713

# Introduction:

The dataset is about a hotel booking demand, it comprises booking information for a city hotel and a resort hotel and it includes information such as when the booking was made, the length of the stay, the number of the adults, children and or babies and the number of available parking information. Our objective from this dataset is that we want to give an insight for the hotel reservation system to simplify for the hotel the customers reservations through using classification algorithm, and how can they treat their customers based on their purpose, such as if they have reserved from the booking agency they are coming for a vacation so the services that will be provided for them will be based on their purpose. And if they have reserved from the website then another indicator will be considered to give the customers the appropriate service.

# **Data exploration and Description: Alanoud**

### -The attributes before the preprocessing:

The number of attributes in the dataset before the preprocessing is 32 and 119390 row attributes

is:hotel,is\_canceled,lead\_time,arrival\_date\_year,arrival\_date\_month,arrival\_da te\_week\_number,arrival\_date\_day\_of\_month,stays\_in\_weekend\_nights,stays\_in\_week\_nights,adults,children,babies,meal,country,market\_segment,distributi on\_channel,is\_repeated\_guest,previous\_cancellations,previous\_bookings\_not \_canceled,reserved\_room\_type,assigned\_room\_type,booking\_changes,deposi t\_type,agent,company,days\_in\_waiting\_list,customer\_type,adr,required\_car\_p arking\_spaces,total\_of\_special\_requests,reservation\_status,reservation\_status\_date

#### -Metadata:

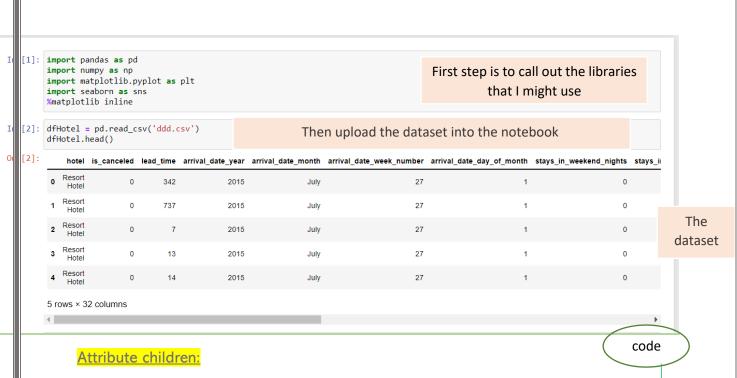
attribute	Туре	Description
Adults	Integer	Number of adults in the reserved room
Children	Integer	
		Number of children in the reserved room
Babies		Number of babies in the reserved room
	Integer	
Agent	Categorical	ID of the online travel intermediary that helped with the reservation
Company	Categorical	ID of the company organization that responsible for handling the expenses the hotel
ArrivalDateDa yOfMonth	Integer	The Day of the month of the arrival to the hotel
ArrivalDateM onth	Categorical	The Month of arrival to the hotel with 12 categories :January to December
ArrivalDateYe ar	Integer	The Year of arrival to the hotel

IsRepeatedGu	Binary	The value indicates if the booking was from a
est		repeated/regular guest (1) is yes and (0) is no
DaysInWaitin	Integer	The Number of days the reservation was in the
gList		waiting list before it was approved to the visitor
Country	Categorical	The Country Categories are represented in the ISO
		3155-3:2013 format
DistributionC		The reservation distribution channel which is an
hannel	Categorical	offline company the categorizes is:
		-TA\TO: the TA means Travel Agents and TO means
		Tour Operators which mean
		Corporate, direct, GDS, undefiend
ReservationSt	Categorical	The last Reservation status which includes of three
atus		categories:
		Canceled – reservation was canceled by the visitor
		Check-Out – the visitor has checked from the hotel roc
		No-Show – the visitor has not showed up to the reserv
StaysInWeeke	Integer	The Number of weekend nights the visitor stayed or
ndNights		reserved to stay at the hotel
StaysInWeekN	Integer	The Number of weeknights the visitor stayed or
ights		reserved to stay at the hotel
IsCanceled	Binary	The Values indicates if the reservation was canceled
		(1) for yes and (0) in no
DepositType	Categorical	Indication on if the visitor made a deposit for the
		reservation the categories are:
		No Deposit – there is no deposit
		Non Refund – a deposit was made as the total cost of
		the reservation
		Refundable – a deposit was made under the total cost
		of the reservation
hotel	binary	This indicates the type of hotel and the values of it:
		City hotel, Resort hotel

This metadata includes the most relevant to the preprocessing or viewed as important attribute for more information about the other attributes click here: <a href="https://www.sciencedirect.com/science/article/pii/S2352340918315191">https://www.sciencedirect.com/science/article/pii/S2352340918315191</a>

#### -Attribute values:

What helped me in finding out the values for the attribute is using Jupyter Notebook which is denote documents that contain both code and rich text elements, such as figures, links, equations



by doing this code "value counts"

this will get the attribute values and the number of the occurrence of that attribute value.

The 0.0 indicates that there are no children in the reservation and the rest of numbers indicates that there is a child in the reservation.

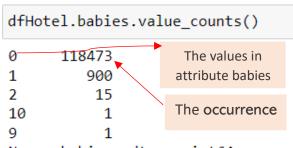
# dfHotel.children.value counts() 0.0 110796 The values in attribute children 1.0 4861 2.0 3652 The occurrence 3.0 76 10.0 1

Name: children, dtype: int64

The outcome of the code

#### Attribute babies:

The O indicates that there are no babies in the reservation and the rest of numbers indicates that there is a baby in the reservation.



Name: babies, dtype: int64

#### Attribute adult

this indicates the number of adults in a reservation for a room the number O indicates that there is no adult which mean no reservation, or the room is for children which we will provide more information after the preprocessing take a place.

221,636 949 12,403

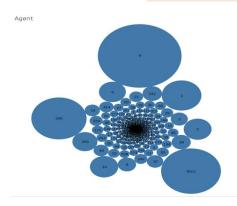
250K 100K 150K 200K 250K 0 200 400 600 800 1000 0K 2K 4K 6K 8K 10K 12K 14K

Adults Babies Children

#### Attribute agent;

the attribute agent has IDs as its values there is some unique IDs like the ones in green box and some repeated IDs like the IDs in a red box

The values in attribute agent



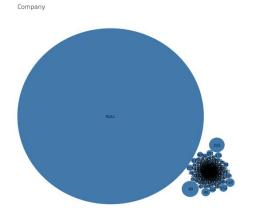
#### dfHotel.agent.value\_counts() 9.0 31961 The occurrence 240.0 13922 1.0 7191 14.0 3640 The black dots 7.0 3539 indicate there is 289.0 more ID that did 1 432.0 1 not come as an 265.0 1 outcome because 93.0 1 the information is 304.0 1 way too big Name: agent, Length:

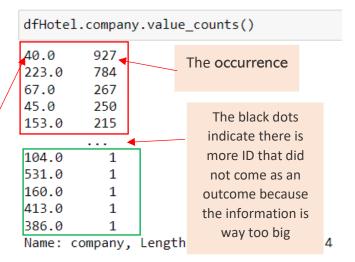
Name: adults, dtype: int64

#### Attribute company

the attribute company has IDs as its values there is some unique IDs like the ones in green box and some repeated IDs like the IDs in a red box

> The values in attribute company





#### Attribute IsRepeatedGuest

The zero indicate that the reservation was not by a repaeated guest and the one indicate that it is a reapeated guest this information could help in sending offers to the regluar

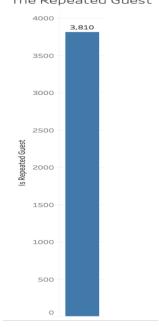
guest and try to attrac and reatain the new gusets.

The percentage of the repatead guest 96.81% and 3.19% for the not repatead guest

isReapeatedGuest dfHotel.is\_repeated\_guest.value\_counts() 115580 The occurrence 1 3810 Name: is\_repeated\_guest, dtype: int64

The values in attribute

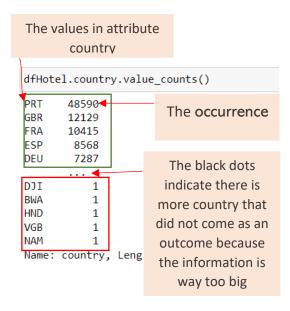
The Repeated Guest



#### Attribute country:

this attribute includes the country name, there is some countries that is repeated frequently like in the read box and some unique countries like in green box

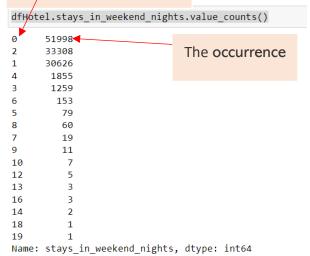




#### Attribute StaysInWeekendNights:

The attribute stays\_in\_weekend\_nights have the number of weekend nights the visitor stayed or reserved to stay at the hotel as its attribute values the percentage of visitors who stayed in weekends are 56.45% ano who did not 43.55%

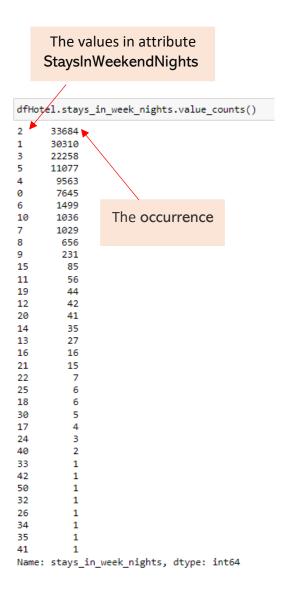
# The values in attribute StaysInWeekendNights

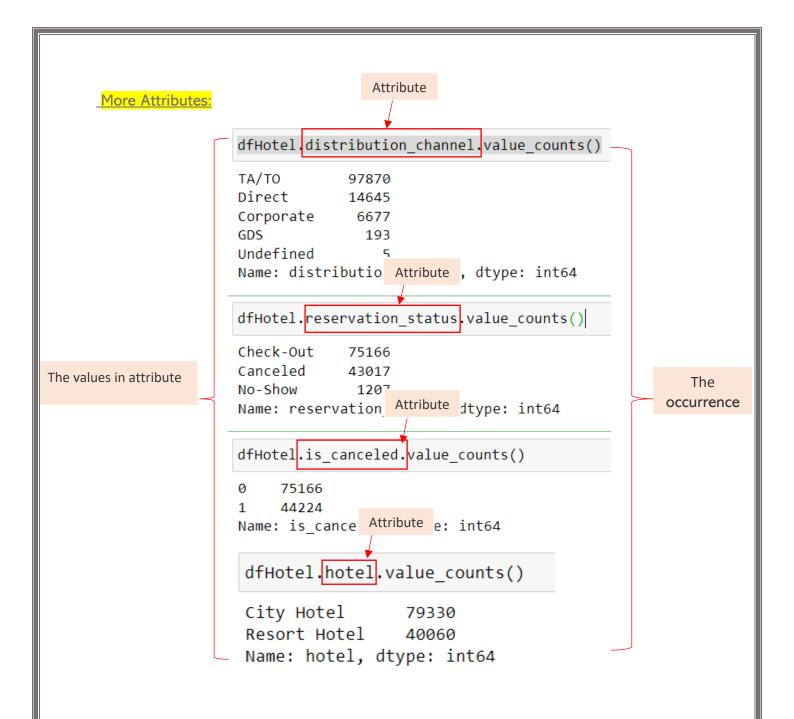


#### Attribute StaysInWeekNights:

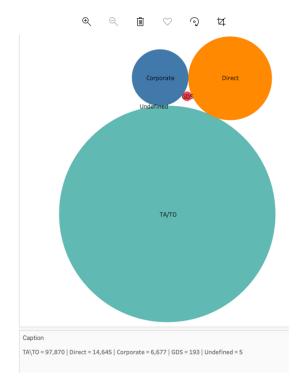
The attribute stays\_in\_week\_nights have the number of weeknights the visitor stayed or reserved to stay at the hotel as its attribute values the percentage of visitors who

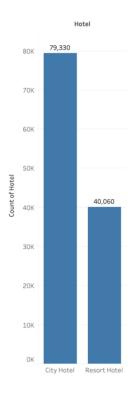
stayed for weeknights are 93.6% and who did not 6.4

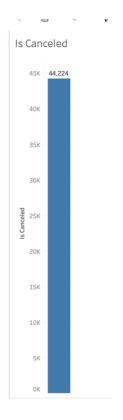




- -for hotel attribute the city hotel is more than the resort hotel by 39270 differences ,the percentage of the city hotel is 66.45% and the resort hotel 33.55%
- for is canceled attribute the visitors seem to not cancel the reservation more than canceling it, the percentage for the guests who canceled the reservation 37.04% and who didn't cancel 62.96%







#### -The attributes after the preprocessing:

Now the dataset includes 28 attribute and 119210 row and the attributes is(hotel,is\_canceled,lead\_time,arrival\_date\_week\_number,stays\_in\_weekend\_nights,stays\_in\_week\_nights,meal,country,market\_segment,distribution\_chan nel,is\_repeated\_guest,previous\_cancellations,previous\_bookings\_not\_cancele d,reserved\_room\_type,assigned\_room\_type,booking\_changes,deposit\_type,a gent,purpous,days\_in\_waiting\_list,customer\_type,adr,required\_car\_parking\_s paces,total\_of\_special\_requests,reservation\_status,reservation\_status\_date,d ate\_of\_arrival)

#### The reasons behind the preprocessing:

The look of the data was scattered because of a lot of dimensionalities and there is a lot of missing data in the attributes, which led to thinking of solutions to make it ready for data mining:

- 1-The column of dates (ArrivalDateDayOfMonth, ArrivalDateYear, ArrivalDateMonth) the best solution for this attributes is to collect them into single attribute which is date\_of\_arrival because the attributes being separate will not imply more information so it's the same (the type of the data preprocessing: Aggregation).
- 2-The country attribute have missing values and the number of the missing values is 407 out 119390 not missing values, which led to put the most repeated country due to the small amount of missing data in this attribute (handling missing values with Mode)
- 3-The Attributes of children ,adults ,babies was separated so it is better to put them together to make the data more useful so we created the attribute status which have adults ,family, groups , couples ,reservation without adult as its attribute value and also there is no reservation value which we got rid of because it's indicates there is no adult or children or babies which it doesn't make sense since the other attribute have data of

the reservation so it might be just a missing value and since it's only 180 out of 119390 we chose to delete it (the type of data prepossessing: Aggregation, Feature creation and handling missing data by eliminate data object).

4-The Agent attribute indicates that it's the ID of the online travel intermediary that helped with the booking So It was IDs and null values as its attribute values and the number of IDs was 103050 and the null values was 16340, so it's a lot of missing values which we need to deal with so the solution is to say to whom there are IDs are the visitors who used Booking website as a way to reserve a room and for the null values are the visitors who used the hotel website as a way to reserve a room (the type of data prepossessing: Feature Creation).

5-The company attribute indicating IDs of the company that are responsible for handling the expenses of the hotel so it's containing IDs and null values as its attribute values, the number of IDs is 6797 and the number of null values is 112593, it's a lot of missing values that we need to deal with, the solution is the visitors who have IDs are on a business trip because the company paid for them for any reason either for checking other branches or a conference in that country. For those who have no IDs then they came for entertainment and vacation (the type of data prepossessing: Feature Creation).

6-is\_canceled and is\_repeated\_guest we changed their attribute values from 0,1 to yes and no to make it more useful (the type of data prepossessing: Feature Creation).

#### The prepressing steps:

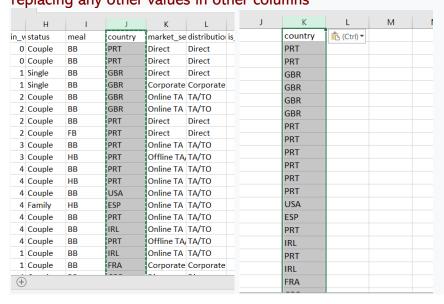
#### country attribute:

1-first I knew that there is missing values because I did this code to every column, and this is the output when I tried it on the column country

```
dfHotel.country.isnull().sum(axis = 0)

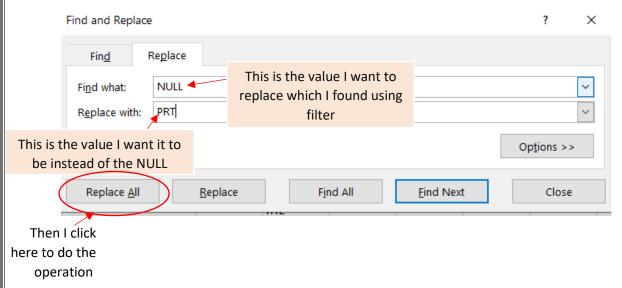
: 488
```

So there is 488 Null value out of 119390 row which is small amount so I can use the most frequent country to solve the missing value, so I copied the column and took it to empty excel sheet, so I find the data and replace more accurately without the danger of replacing any other values in other columns

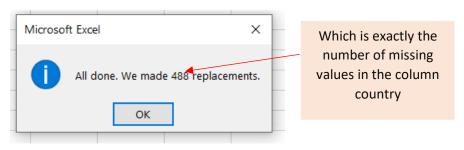


2- then I need to know the most frequent country so I can use find and replace feature in excel to replace the null value with it, so I did the code below and the output is the most frequent country in the country column.

Then since now I know the most frequent value now I will simply use find and replace by clicking Cntrl and F

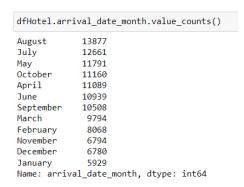


3-then be clicking on replace all ,all the null values are now PRT country



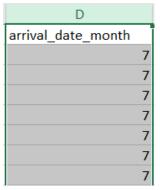
#### Date of arrival attribute:

1-first to fix this all we need is Date formula, but I have a problem which is the ArrivalDateMonth attribute because it's a categorical data type and include the name of month as its attribute value as u can see here

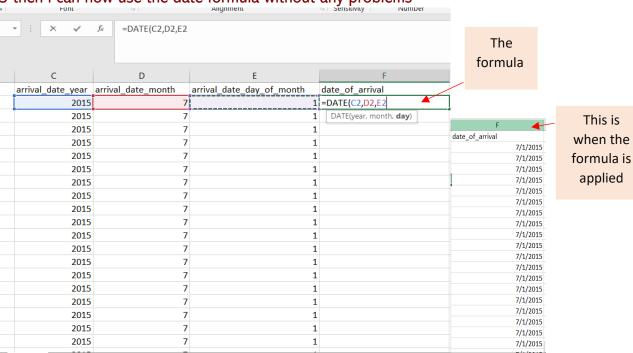


and this doesn't work with the Date formula, so I need to change the data type from categorical to integer like for example if the attribute value is January then it's should be 1

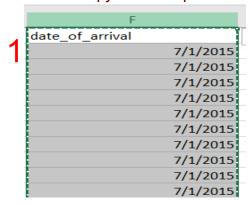
2-so I used the find and replace feature in excel to fix this and this is the outcome the 7 indicate that it's July

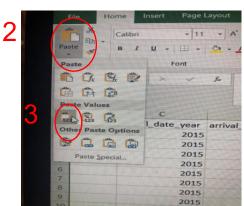


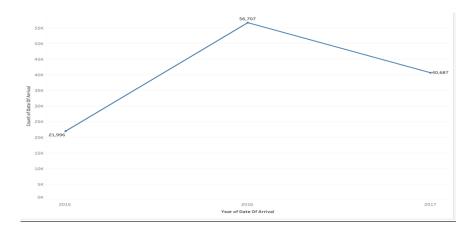
3-then I can now use the date formula without any problems



4-then I deleted the attributes (ArrivalDateDayOfMonth, ArrivalDateYear, ArrivalDateMonth) and just kept the DateOfArrival but before that I need to make the column independent from the other column so the formula still works when I delete them this will be done by selecting the column DateOfArrival and click on copy thin click paste





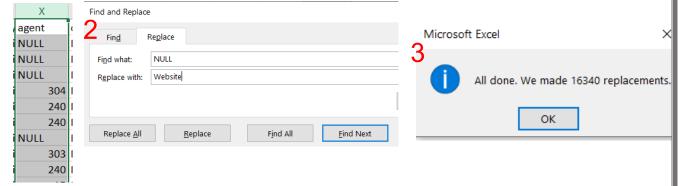


#### Agent attribute:

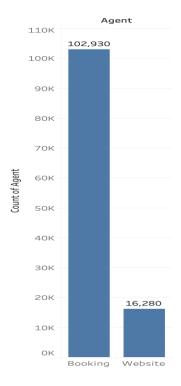
1-snice this attribute only includes IDs and missing values as its attribute value the first thing that I did is to figure out how many missing values in this column which is 16340 out of 119390 row and the percentage 13.69% which is a lot so it's a must to deal with

dfHotel.agent.isnull().sum()
: 16340

Then I went to use find and replace then I replaced the NULL with website



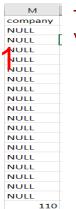
2-then I figure out what the IDs that are in the column using filter and then to replace it with Booking using find and replace feature in excel.



#### Purpose attribute:

1-since the attribute company only includes IDs of the company and missing values as its attribute values the first thing that I did is to figure out how many missing values in this column which is 112593 out of 119390 rows and the percentage 94.33% so it's a lot and we must deal with it.

#### 112593



Then I went to use find and replace then I replaced the NULL with vacation.

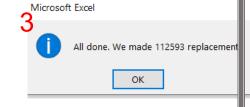
Find and Replace

Find Replace

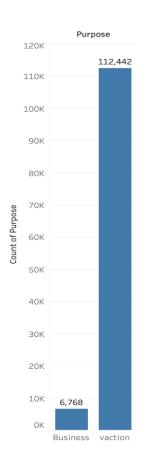
Find What: NULL

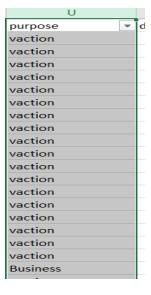
Replace with: vacation

Replace All Replace Find All Find Next



2-then I figure out what the IDs that are in the column using filter and then to replace it with Business using find and replace feature in excel and I change the name of the column from company to purpose to make more sense.





#### Status Attribute:



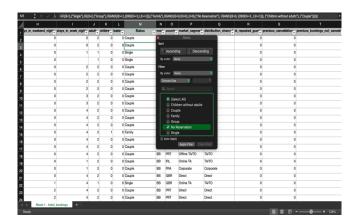
At the beginning of the IF statements, we mentioned that if the column of the adult is equal to 1, the result will be a Single, but if it is greater than 2, it will be a group.

And then we mentioned two conditions in one IF statement, which is if the adult column is greater than or equal to 1 and also if the children or babies column is greater than or equal to 1 then the result will be a family.

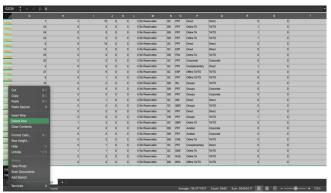
And then in the case of all the columns for adult, children, and babies, it was equal to zero, which means that there is no reservation, but after reviewing the rows in which this option appeared, it was found that there is other information such as the time of arrival and so on, and this is illogical, and their total was

only 180 rows. So, we decide to delete the rows to not affect the dataset with counterproductive result. How did we delete it?

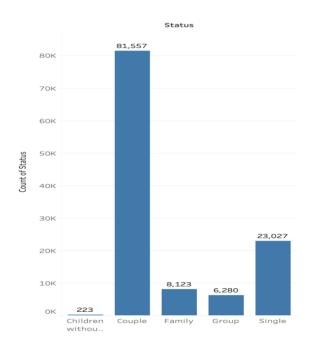
1- First we choose the Sort and filter option then we choose to just show the rows with No reservation.



2- Then we select all the 180 rows and choose the Delete row option.

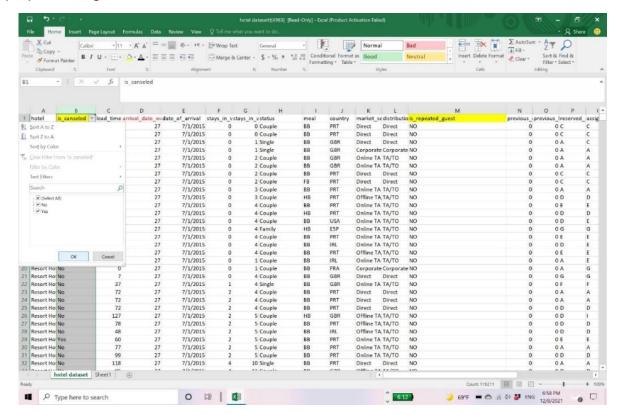


In the last case, we also combined two conditions in one IF statement, which states that if the adult column is equal to 0 and the children column or babies is equal to or greater than 1, the result will be children without adult and if it does not stipulate one of the options mentioned, it will of course be couple.

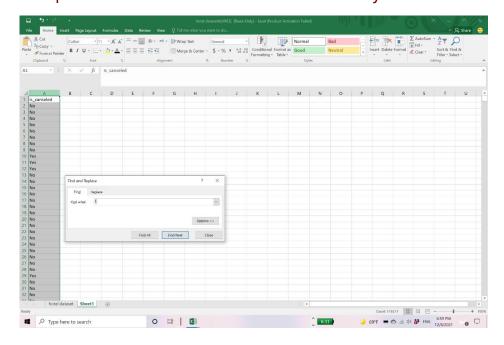


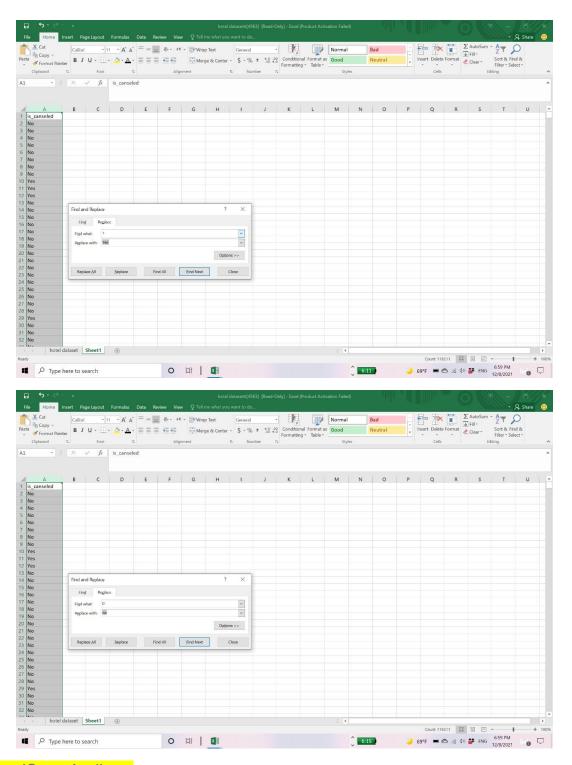
#### ls canceled Attribute:

1-first step is to copy the column and then go to an empty excel to do the preprocessing.



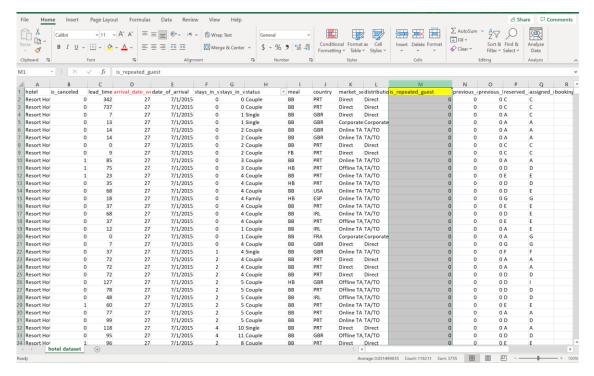
2-we will use find and replace feature to make 0 mean no and 1 mean yes.



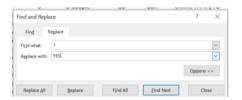


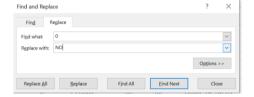
#### IsRepeatedGuest Attribute:

1-first step is to copy the column and then go to an empty excel to do the preprocessing



2-we will use find and replace feature to make 0 mean no and 1 mean yes.





#### The metadata after the preprocessing:

-this metadata includes only the attribute that been through preprocessing

attribute	Туре	Description
purpose	Binary	This attribute indicates
		the purpose of the trip
		And have this two
		attribute value:
		Vacation: indicates that
		the guest is staying for
		vacation
		Business: this indicates
		that the guest is
		staying for work
Status	Categorical	This attribute indicates
		the room was reserved
		to what of this
		categorizes:

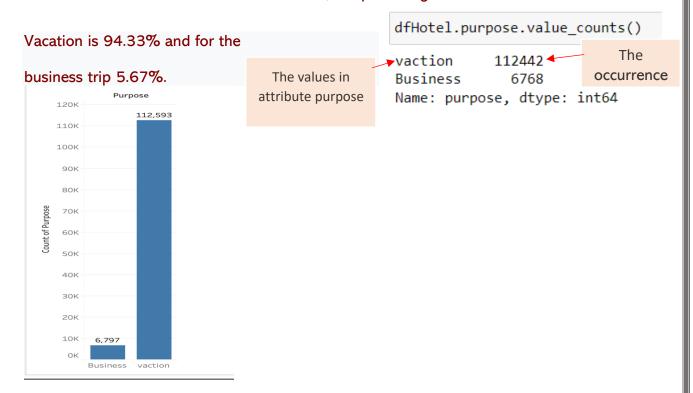
		Single: the room is reserved for only one adult without children or babies Couples: the room is reserved for only two adults without children or babies Groups: the room is reserved for more than two adults without babies or children Family: the room is reserved to one or two or more adults with
		children and babies Children without adults: the room is reserved to children without adults
Agent	Binary	This attribute indicates what are the online intermediary that the guest used to reserve in the hotel And have this two attribute value: Website: indicates that the guest used the hotel website to reserve in the hotel Booking: indicates that the guest used the Booking website to reserve in the hotel
DateOfArrival	Integer	This attribute indicates the day, month, year of arrival
IsCanceled	Binary	The Values indicates if the reservation was canceled the values is yes,no
IsRepeatedGuest	Binary	The value indicates if the booking was from a repeated/regular guest and the values is yes,no

#### -Attribute values:

-this includes only the attribute values that are relevant and need more explanation

#### Purpose Attribute:

-now instead of company and its values it's more useful now like we can see if the guest is coming for a business trip so we can put them on a room away from disturbance and make an offer to them, the percentage for the



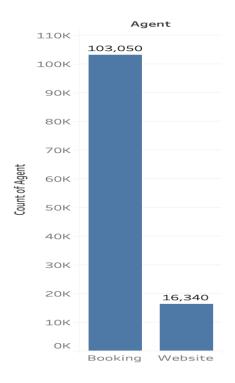
#### Agent Attribute:

-now instead of the old values now it's more useful, like the number of visitors using booking are more than the visitors who use the hotel website which means maybe the website need more work for it to be recognized by the visitors, the percentage for Booking is 86.21% and for the website 13.64%.

The values in attribute Agent

dfHotel.agent.value\_counts()

Booking 102930 The Website 16280 occurrence
Name: agent, dtype: int64



#### Status Attribute:

-now the attribute and its values are more useful, the children without adults could indicates that their parent reserved the room for their kids only ,there was a no reservation but since it just 180 row out of 119390 we deleted it

The percentage for status:

Couple:68.31%

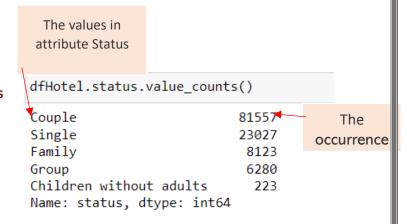
Single: 19.29%

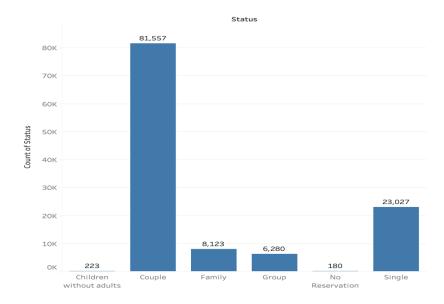
Family:6.8%

Group:5.26%

Children without adults: 1.87%

So the most status of the reservation is reserved for couples.





#### IsRepeatedGuest Attribute:

this attribute includes yes and no as its attribute values the no mean it's not a repeated guest and yes means it's repeated guest the percentage of the repeated guest is 3.15% and 96.85% is not a repeated guest.

#### dfHotel.is\_repeated\_guest.value\_counts()

NO 115455 YES 3755

Name: is\_repeated\_guest, dtype: int64

#### IsCanceled Attribute:

this attribute includes yes and no as its attribute values the no mean it's not a canceled and yes means it's canceled percentage of the cancelation 37.04% and who didn't cancel 62.96%

#### dfHotel.is\_canceled.value\_counts()

No 75011 Yes 44199

Name: is\_canceled, dtype: int64

#### -The dataset before and after the prepossessing:

```
-before the preprocessing
                                                              The number
         In [43]: dfHotel.info()
                                                               of values in
                   <class 'pandas.core.frame.DataFrame'>
 The total
                                                              the attribute
                   RangeIndex: 119390 entries, 0 to 119389
                   Data columns (total 32 columns):
   rows
                                                         Non-Null/Count
                    # Column
                                                       119390 non-null object
                        hote1
                        is_canceled
                                                        119390 non-null
                                                                          int64
                    1
                        lead_time
                                                        119390 non-null
                        arrival_date_year
                                                        119390 non-null
                                                                          int64
                    3
                        arrival_date_month
                                                        119390 non-null
                                                                          object
                        arrival_date_week_number
arrival_date_day_of_month
                    5
                                                        119390 non-null
                                                                          int64
                                                       119390 non-null
                    6
                                                                          int64
                        stays_in_weekend_nights
                    7
                                                        119390 non-null
                                                                          int64
                    8
                                                        119390 non-null
                                                                          int64
                        stays_in_week_nights
                                                        119390 non-null
                    9
                        adults
                                                                          int64
                                                                          float64
                    10 children
                                                        119386 non-null
                    11 babies
                                                        119390 non-null int64
                                                        119390 non-null object
                    12
                        meal
                    13 country
                                                        118902 non-null object
                    14 market_segment
                                                        119390 non-null object
                                                      119390 non-null object
119390 non-null int64
                    15 distribution_channel
                                                                          object
attributes
                    16 is_repeated_guest
                    17
                        previous_cancellations
                                                        119390 non-null
                                                                          int64
                        previous_bookings_not_canceled 119390 non-null int64
                    18
                                                119390 non-null object
119390 non-null object
                    19 reserved_room_type
                    20 assigned_room_type
                                                                                            As you can see
                    21 booking_changes
                                                       119390 non-null int64
                    22 deposit_type
                                                        119390 non-null object
                                                                                               there is a
                    23 agent
                                                        103050 non-null float64
                    24 company
                                                        6797 non-null
                                                                          float64
                                                                                            missing values
                                                      119390 non-null int64
                    25 days_in_waiting_list
                    26
                       customer_type
                                                        119390 non-null object
                    27 adr
                                                        119390 non-null float64
                                                aces 119390 non-null int64
ts 119390 non-null int64
119390 non-null int64
119390 non-null object
                    28 required_car_parking_spaces
                    29 total_of_special_requests
                    30 reservation_status
                    -31 reservation_status_date
                                                        119390 non-null object
                   dtypes: float64(4), int64(16), object(12)
                   memory usage: 29.1+ MB
```

#### -After the preprocessing

The total rows

dfHotel.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 119210 entries, 0 to 119209

Data columns (total 28 columns):

Duca	columns (cocal 20 columns).		
#	Column	Non-Null Count	Dtype
0	hotel	119210 non-null	object
1	is_canceled	119210 non-null	int64
2	lead_time	119210 non-null	int64
3	arrival_date_week_number	119210 non-null	int64
4	date_of_arrival	119210 non-null	object
5	stays_in_weekend_nights	119210 non-null	int64
6	stays_in_week_nights	119210 non-null	int64
7	status	119210 non-null	object
8	meal	119210 non-null	object
9	country	119210 non-null	object
10	market_segment	119210 non-null	object
11	distribution_channel	119210 non-null	object
12	is_repeated_guest	119210 non-null	int64
4.5		11001011	2-4-64

As you can see there is no missing values

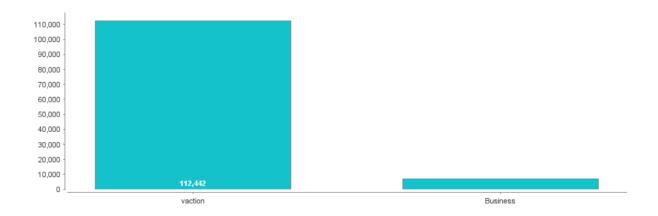
attributes

119210 non-null int64 13 previous\_cancellations 14 previous\_bookings\_not\_canceled 119210 non-null int64 15 reserved\_room\_type 119210 non-null object 16 assigned\_room\_type 119210 non-null object 17 booking\_changes 119210 non-null int64 119210 non-null object 18 deposit\_type 119210 non-null object 119210 non-null object 19 agent 20 purpose 21 days\_in\_waiting\_list 119210 non-null int64 25 adr 119210 non-null object 119210 non-null float64 119210 non-null int64 119210 non-null int64 119210 non-null int64 119210 non-null int64 119210 non-null object 119210 non-null ob 119210 non-null object dtypes: float64( $\overline{1}$ ), int $\overline{64}(12)$ , object(15) memory usage: 25.5+ MB

#### - The limitation:

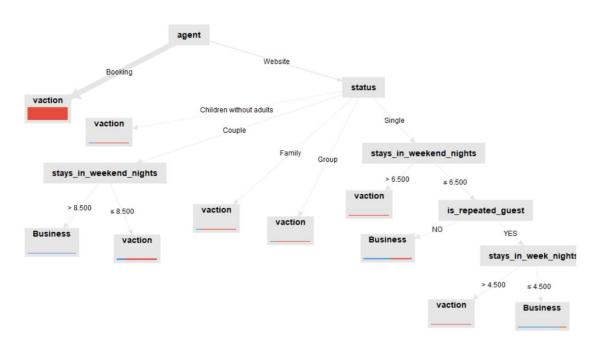
- 1- The first limitation in the data set were the Is Canceled column and the Is Repeated guest column were represented as Integer 0,1 and we converted them to Categorical Yes,No to make it easier to understand, especially in the decision tree.
- 2- The second limitation were in the column of the arrival date year, arrival date month, and the arrival date day of month, each of them was in a separate column, and this is considered illogical to some extent, so we collected them in one column to make it easier to read.
- 3- The third limitation was in the adult, Children and Babies column. It was difficult to read the columns and determine the type of reservation, so we made a new column that combines these three columns so that it is classified if they are single, couple, family, groups or children without adult.
- 4- According to the third limitation, 180 rows were found. Adult, children, and babies all contained zero, with other information in other attributes, and this is also illogical. After discussing the limitation with the project members, it was decided to delete the rows because there is no information that can help us determinate the status, and it constitutes a very small percentage of 100,000 row that we have.
- 5- The fourth limitation were in the Agent and company column it's contained many codes and nulls as well, so they were collected in one column to show whether the booking was through the website or through Booking itself.
- 6- In the end, there were a number of attributes that were not sufficiently explained in the metadata, so we wish we know the owner of the dataset to ask him some questions that will help us.

# Pattern discovery:



We build our model based on this chart comparison, choose "purpose" as class label so we want to demonstrate opportunities that hidden in purpose attribute.

#### The result:



We choose prediction model and decision tree method, our class label is Purpose and the class label indicate tow results from traveling 1(vacation),2 (Business)

#### **Explanation:**

Based on purpose class label we notice the following:

- 1- Who booked through Booking (as travel agent) the purpose from the travel is Vacation
- 2- Who booked through website will be based on another indicator which is (statues)

Statues values are: Group, Family, Single, Couple, Children without adults

- A- If booked through website and statues is Children without adults, purpose is vacation
- B- If booked through website and statues is Family, purpose is vacation
- C- If booked through website and statues is Group, purpose is vacation

If booked through website and statues is Couple, purpose is depending in (Stays in weekend night) Attribute

Stays in weekend nights Attribute indicate: Number of weekend nights (Saturday or Sunday) the guest stayed or booked to stay at the hotel (integer)

If the number of the gust stays in weekend nights was >8.500, purpose will be Business

Else the number of the gust stays in weekend nights was≤ 8.500, purpose will be vacation

D- If booked through website and statues is single

is depending in (Stays in weekend night) Attribute

If the number of the gust stays in weekend nights was >6.500, purpose will be Vacation

Else the number of the gust stays in weekend nights was≤ 6.500, we will look to if the gust is repeated gust through attribute called (is\_repeted\_guest)

So

If guest not repeated guest, purpose will be business

If the guest is repeated guest, we look up for the (stays in week night) attribute

Stays in week night indicate: Number of week nights (Monday to Friday) the guest stayed or booked to stay at the hotel (integer)

If the number of the gust stays in week nights was >4.500, purpose will be Vacation

Else the number of the gust stays in week nights was≤ 4.500, purpose will be Business

#### Our recommendation:

#### 1-In scenario number "1"

Have special collaboration with booking to customize service to people who booked through booking in order to provide seasonal offers, special prices, customize the hotel service ex: massage, hair salon, gym for guest from booking

#### 2-in scenario number "2"

A, B, C, D couple, single, have same purpose (vacation) with different statues segment: Customize hotel website

1-updated offers in real time

2-accesbility

3-easy to navigate

Provide special activates ex: choose specific room, activity for family group

Reservation for table in the restaurant for any kind of occasions Other

4-provide partnership with rental car company to make the customers transportation easier.

5-Provide bus to guide tour across city or island

#### 3- couple and single where their purpose is business

Track period of their business trips and:

- Provide separate section with strong internet connection
- Provide separate elevators used to reach designed (meetings room)
- Provide special cars to pick them form the airport to the hotel or any desired locations

#### Performance:

#### **Confusion Matrix**

	true Business	true vaction	class precision
pred. Business	65	12	84.42%
pred. vaction	1856	32127	94.54%
class recall	3.38%	99.96%	

#### **Explanation:**

Performance: F score

F score for Business: 0.06499 F score for Vacation: 0.9717

#### Based on confusion matrix:

We calculated F1 score (F1 Score is the Harmonic Mean between precision and recall. The range for F1 Score is [0, 1]. It tells you how precise your classifier is (how many instances it classifies correctly), as well as how robust it is (it does not miss a significant number of instances). and the results was indicating increment in vacation as our model predicted

#### **Numbers**

Vacation: 0.9717 close to 1, our recommendation is that hotel offers remain stable for travelers with vacation purpose

Business: 0.0649 which indicate decrement in number of travelers with business purpose, with take in consideration that number of travelers with business purpose are most likely stays in city hotels

We target business with different statues, the hotel should look for numbers of business travelers in city hotels and manage to provide special offers, service for them including: meetings room, quite sections and areas.

#### Furthermore:

#### Based on confusion matrix:

we have high recall value; high recall indicates: there were very few false negatives (test result that is incorrect because the test failed to recognize an existing condition or finding make the condition absent .) in high recall the classifier is more permissive in the criteria for classifying something as positive, in other word our recall return most relevant information about purpose attribute, which indicate that we will focus on travelers with vacation purpose either our class predicted as true or false ,this measure gives our model more flexibility to choose more segment to serve

"32127" travelers have predicted as true vacation

"1856" travelers have predicted as false vacation (false negative)

#### Additionally:

We also have high precision whish mean that an algorithm returns more relevant results than irrelevant ones, with few false positive (test result which wrongly indicates that a particular condition or attribute is present) in our model if we chose to focus on high precision we will focus on travelers with correct business purpose only.

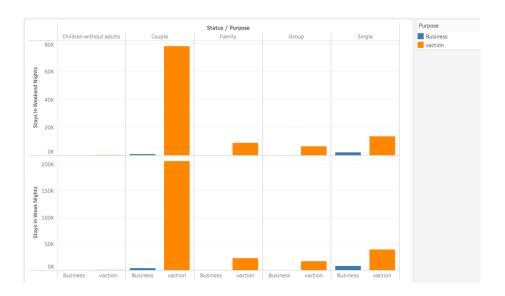
"65" travelers have predicted as true Business

"12" travelers have predicted as false Business (false positive)

#### Finally:

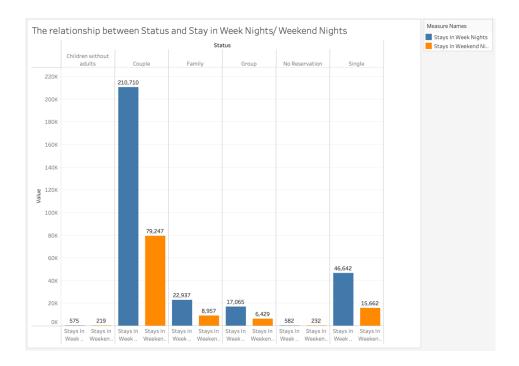
We have an ideal model with high precision and recall with all results predicted correctly.

# Charts and comparisons

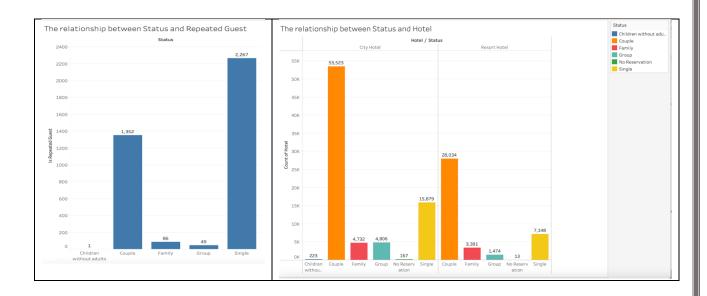


As we see above, when we are compering all Status attribute and the Purpose if it's a cognize eBusiness or Vacation with if they stay in weekend nights or week nights, we r that all of type of Status is stays in Week nights/Weekend nights for vacation is more than business purpose.

So, based on that we should focuses on how to offer a good services for those who come for vacation because they are the majority.



Here we can see the relationship between "Status" and stay in Week nights/Weekend nights. We noticed the guests who stays in Week nights is more than Weekend, so because of that the hotels may focus on how to make special offer in price or Special events and activities in the Weekend to attract people.



We need to know which type of Status is repeating guest and what is the hotel they reserve.

Here is the relationship between Status and Repeated Guest, so we noticed Singles are the most repeated guest and based on the right picture they were reserved City hotel more than resort hotel.

## **Conclusion:**

In conclusion, the result we had after applying classification algorithm in our dataset is the data have been classified to different groups , and as a result we could explain those groups such as giving the appropriate services for the different type of customers and for their different type of purpose. We reached out the final hidden pattern for our model, one of the most helping tool was confusion matrix and F1 score measurement which guide as to discover new segment or target to serve them in sufficient way.

Some of the limitations that we have faced during this project is that using excel for data preprocessing was not as efficient as using python, the data preprocessing require logic thinking and problem solving skills And using Rapidminer as main software there were difficulties during the mining part starting from building the model to the final result, maybe if we could use another software like Jupyter it could be easier and we could find more useful tutorials. And for working in this project we had a zoom meetings during the semester that have reached 15 hours.

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<u>sa=true</u>	
For more charts or if the charts are not	obvious CLICK HERE