

hotel booking project-Copy1

May 25, 2021

Hotel booking project is a project that will predict how likely it is for a customer to cancel their hotel booking

1 data cleaning

```
[54]: #importing libraries

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[55]: #read the data

hotel = pd.read_csv("C:/Users/hp/Downloads/hotel_bookings.csv")
```

```
[56]: #first 5 rows

hotel.head()
```

```
[56]:
```

	hotel	is_canceled	lead_time	arrival_date_year	arrival_date_month	\
0	Resort Hotel	0	342	2015	July	
1	Resort Hotel	0	737	2015	July	
2	Resort Hotel	0	7	2015	July	
3	Resort Hotel	0	13	2015	July	
4	Resort Hotel	0	14	2015	July	

	arrival_date_week_number	arrival_date_day_of_month	\
0	27	1	
1	27	1	
2	27	1	
3	27	1	
4	27	1	

	stays_in_weekend_nights	stays_in_week_nights	adults	children	babies	\
0	0	0	2	0.0	0	
1	0	0	2	0.0	0	

2		0		1	1	0.0	0
3		0		1	1	0.0	0
4		0		2	2	0.0	0

	meal	country	market_segment	distribution_channel	is_repeated_guest	\
0	BB	PRT	Direct	Direct	0	
1	BB	PRT	Direct	Direct	0	
2	BB	GBR	Direct	Direct	0	
3	BB	GBR	Corporate	Corporate	0	
4	BB	GBR	Online TA	TA/TO	0	

	previous_cancellations	previous_bookings_not_canceled	reserved_room_type	\
0	0	0	C	
1	0	0	C	
2	0	0	A	
3	0	0	A	
4	0	0	A	

	assigned_room_type	booking_changes	deposit_type	agent	company	\
0	C	3	No Deposit	NaN	NaN	
1	C	4	No Deposit	NaN	NaN	
2	C	0	No Deposit	NaN	NaN	
3	A	0	No Deposit	304.0	NaN	
4	A	0	No Deposit	240.0	NaN	

	days_in_waiting_list	customer_type	adr	required_car_parking_spaces	\
0	0	Transient	0.0	0	
1	0	Transient	0.0	0	
2	0	Transient	75.0	0	
3	0	Transient	75.0	0	
4	0	Transient	98.0	0	

	total_of_special_requests	reservation_status	reservation_status_date
0	0	Check-Out	7/1/2015
1	0	Check-Out	7/1/2015
2	0	Check-Out	7/2/2015
3	0	Check-Out	7/2/2015
4	1	Check-Out	7/3/2015

```
[57]: # number os rows and number of columns
```

```
hotel.shape
```

```
[57]: (119390, 32)
```

```
[58]: #checking for missing data
```

```
hotel.isnull().sum()
```

```
[58]: hotel          0
      is_canceled    0
      lead_time      0
      arrival_date_year  0
      arrival_date_month  0
      arrival_date_week_number  0
      arrival_date_day_of_month  0
      stays_in_weekend_nights  0
      stays_in_week_nights  0
      adults          0
      children        4
      babies          0
      meal            0
      country         488
      market_segment  0
      distribution_channel  0
      is_repeated_guest  0
      previous_cancellations  0
      previous_bookings_not_canceled  0
      reserved_room_type  0
      assigned_room_type  0
      booking_changes  0
      deposit_type     0
      agent           16340
      company         112593
      days_in_waiting_list  0
      customer_type    0
      adr              0
      required_car_parking_spaces  0
      total_of_special_requests  0
      reservation_status  0
      reservation_status_date  0
      dtype: int64
```

```
[59]: #dealing with missing values
```

```
def data_clean(hotel):
    hotel.fillna(0,inplace=True)
    print(hotel.isnull().sum())
```

```
[60]: #calling the function and now we dont have any null values
```

```
data_clean(hotel)
```

```
hotel          0
is_canceled    0
```

```

lead_time                0
arrival_date_year         0
arrival_date_month        0
arrival_date_week_number  0
arrival_date_day_of_month 0
stays_in_weekend_nights  0
stays_in_week_nights     0
adults                   0
children                 0
babies                   0
meal                     0
country                  0
market_segment           0
distribution_channel      0
is_repeated_guest         0
previous_cancellations    0
previous_bookings_not_canceled 0
reserved_room_type        0
assigned_room_type        0
booking_changes           0
deposit_type              0
agent                     0
company                   0
days_in_waiting_list     0
customer_type             0
adr                       0
required_car_parking_spaces 0
total_of_special_requests 0
reservation_status        0
reservation_status_date   0
dtype: int64

```

```
[61]: #display the columns
```

```
hotel.columns
```

```
[61]: Index(['hotel', 'is_canceled', 'lead_time', 'arrival_date_year',
          'arrival_date_month', 'arrival_date_week_number',
          'arrival_date_day_of_month', 'stays_in_weekend_nights',
          'stays_in_week_nights', 'adults', 'children', 'babies', 'meal',
          'country', 'market_segment', 'distribution_channel',
          'is_repeated_guest', 'previous_cancellations',
          'previous_bookings_not_canceled', 'reserved_room_type',
          'assigned_room_type', 'booking_changes', 'deposit_type', 'agent',
          'company', 'days_in_waiting_list', 'customer_type', 'adr',
          'required_car_parking_spaces', 'total_of_special_requests',
          'reservation_status', 'reservation_status_date'],
          dtype=object)
```

```
dtype='object')
```

```
[62]: #finding the unique instances in the three categories of people
```

```
list=['adults', 'children', 'babies']
for val in list:
    print('{} has uniques values as {}'.format(val,hotel[val].unique()))
```

```
adults has uniques values as [ 2  1  3  4 40 26 50 27 55  0 20  6  5 10]
```

```
children has uniques values as [ 0.  1.  2. 10.  3.]
```

```
babies has uniques values as [ 0  1  2 10  9]
```

```
[63]: #creating a filter for the zeros that exist in all the three categories
```

```
filter=(hotel["adults"]== 0) & (hotel["children"]==0) & (hotel['babies']==0)
hotel[filter]
```

```
[63]:
```

	hotel	is_canceled	lead_time	arrival_date_year	\
--	-------	-------------	-----------	-------------------	---

2224	Resort Hotel	0	1	2015	
2409	Resort Hotel	0	0	2015	
3181	Resort Hotel	0	36	2015	
3684	Resort Hotel	0	165	2015	
3708	Resort Hotel	0	165	2015	
...	
115029	City Hotel	0	107	2017	
115091	City Hotel	0	1	2017	
116251	City Hotel	0	44	2017	
116534	City Hotel	0	2	2017	
117087	City Hotel	0	170	2017	

	arrival_date_month	arrival_date_week_number	\
2224	October	41	
2409	October	42	
3181	November	47	
3684	December	53	
3708	December	53	
...	
115029	June	26	
115091	June	26	
116251	July	28	
116534	July	28	
117087	July	30	

	arrival_date_day_of_month	stays_in_weekend_nights	\
2224	6	0	
2409	12	0	
3181	20	1	
3684	30	1	

3708	30	2
...
115029	27	0
115091	30	0
116251	15	1
116534	15	2
117087	27	0

	stays_in_week_nights	adults	children	babies	meal	country	\
2224	3	0	0.0	0	SC	PRT	
2409	0	0	0.0	0	SC	PRT	
3181	2	0	0.0	0	SC	ESP	
3684	4	0	0.0	0	SC	PRT	
3708	4	0	0.0	0	SC	PRT	
...	
115029	3	0	0.0	0	BB	CHE	
115091	1	0	0.0	0	SC	PRT	
116251	1	0	0.0	0	SC	SWE	
116534	5	0	0.0	0	SC	RUS	
117087	2	0	0.0	0	BB	BRA	

	market_segment	distribution_channel	is_repeated_guest	\
2224	Corporate	Corporate	0	
2409	Corporate	Corporate	0	
3181	Groups	TA/TO	0	
3684	Groups	TA/TO	0	
3708	Groups	TA/TO	0	
...	
115029	Online TA	TA/TO	0	
115091	Complementary	Direct	0	
116251	Online TA	TA/TO	0	
116534	Online TA	TA/TO	0	
117087	Offline TA/TO	TA/TO	0	

	previous_cancellations	previous_bookings_not_canceled	\
2224	0	0	
2409	0	0	
3181	0	0	
3684	0	0	
3708	0	0	
...	
115029	0	0	
115091	0	0	
116251	0	0	
116534	0	0	
117087	0	0	

	reserved_room_type	assigned_room_type	booking_changes	deposit_type	\
2224	A	I	1	No Deposit	
2409	A	I	0	No Deposit	
3181	A	C	0	No Deposit	
3684	A	A	1	No Deposit	
3708	A	C	1	No Deposit	
...	
115029	A	A	1	No Deposit	
115091	E	K	0	No Deposit	
116251	A	K	2	No Deposit	
116534	A	K	1	No Deposit	
117087	A	A	0	No Deposit	

	agent	company	days_in_waiting_list	customer_type	adr	\
2224	0.0	174.0	0	Transient-Party	0.00	
2409	0.0	174.0	0	Transient	0.00	
3181	38.0	0.0	0	Transient-Party	0.00	
3684	308.0	0.0	122	Transient-Party	0.00	
3708	308.0	0.0	122	Transient-Party	0.00	
...	
115029	7.0	0.0	0	Transient	100.80	
115091	0.0	0.0	0	Transient	0.00	
116251	425.0	0.0	0	Transient	73.80	
116534	9.0	0.0	0	Transient-Party	22.86	
117087	52.0	0.0	0	Transient	0.00	

	required_car_parking_spaces	total_of_special_requests	\
2224	0	0	
2409	0	0	
3181	0	0	
3684	0	0	
3708	0	0	
...	
115029	0	0	
115091	1	1	
116251	0	0	
116534	0	1	
117087	0	0	

	reservation_status	reservation_status_date
2224	Check-Out	10/6/2015
2409	Check-Out	10/12/2015
3181	Check-Out	11/23/2015
3684	Check-Out	1/4/2016
3708	Check-Out	1/5/2016
...
115029	Check-Out	6/30/2017

115091	Check-Out	7/1/2017
116251	Check-Out	7/17/2017
116534	Check-Out	7/22/2017
117087	Check-Out	7/29/2017

[180 rows x 32 columns]

```
[64]: #since we lacked some columns in the previous output, now we fix that
      #with set_option in pandas
```

```
pd.set_option('display.max_column',32)
```

```
[65]: filter =(hotel["adults"]== 0) & (hotel["children"]==0) & (hotel['babies']==0)
      hotel[filter]
```

```
[65]:
```

	hotel	is_canceled	lead_time	arrival_date_year	\
2224	Resort Hotel	0	1	2015	
2409	Resort Hotel	0	0	2015	
3181	Resort Hotel	0	36	2015	
3684	Resort Hotel	0	165	2015	
3708	Resort Hotel	0	165	2015	
...	
115029	City Hotel	0	107	2017	
115091	City Hotel	0	1	2017	
116251	City Hotel	0	44	2017	
116534	City Hotel	0	2	2017	
117087	City Hotel	0	170	2017	

	arrival_date_month	arrival_date_week_number	\
2224	October	41	
2409	October	42	
3181	November	47	
3684	December	53	
3708	December	53	
...	
115029	June	26	
115091	June	26	
116251	July	28	
116534	July	28	
117087	July	30	

	arrival_date_day_of_month	stays_in_weekend_nights	\
2224	6	0	
2409	12	0	
3181	20	1	
3684	30	1	
3708	30	2	

...
115029	27	0
115091	30	0
116251	15	1
116534	15	2
117087	27	0

	stays_in_week_nights	adults	children	babies	meal	country	\
2224	3	0	0.0	0	SC	PRT	
2409	0	0	0.0	0	SC	PRT	
3181	2	0	0.0	0	SC	ESP	
3684	4	0	0.0	0	SC	PRT	
3708	4	0	0.0	0	SC	PRT	
...	
115029	3	0	0.0	0	BB	CHE	
115091	1	0	0.0	0	SC	PRT	
116251	1	0	0.0	0	SC	SWE	
116534	5	0	0.0	0	SC	RUS	
117087	2	0	0.0	0	BB	BRA	

	market_segment	distribution_channel	is_repeated_guest	\
2224	Corporate	Corporate	0	
2409	Corporate	Corporate	0	
3181	Groups	TA/TO	0	
3684	Groups	TA/TO	0	
3708	Groups	TA/TO	0	
...	
115029	Online TA	TA/TO	0	
115091	Complementary	Direct	0	
116251	Online TA	TA/TO	0	
116534	Online TA	TA/TO	0	
117087	Offline TA/TO	TA/TO	0	

	previous_cancellations	previous_bookings_not_canceled	\
2224	0	0	
2409	0	0	
3181	0	0	
3684	0	0	
3708	0	0	
...	
115029	0	0	
115091	0	0	
116251	0	0	
116534	0	0	
117087	0	0	

reserved_room_type	assigned_room_type	booking_changes	deposit_type	\
--------------------	--------------------	-----------------	--------------	---

2224	A	I	1	No Deposit
2409	A	I	0	No Deposit
3181	A	C	0	No Deposit
3684	A	A	1	No Deposit
3708	A	C	1	No Deposit
...
115029	A	A	1	No Deposit
115091	E	K	0	No Deposit
116251	A	K	2	No Deposit
116534	A	K	1	No Deposit
117087	A	A	0	No Deposit

	agent	company	days_in_waiting_list	customer_type	adr \
2224	0.0	174.0	0	Transient-Party	0.00
2409	0.0	174.0	0	Transient	0.00
3181	38.0	0.0	0	Transient-Party	0.00
3684	308.0	0.0	122	Transient-Party	0.00
3708	308.0	0.0	122	Transient-Party	0.00
...
115029	7.0	0.0	0	Transient	100.80
115091	0.0	0.0	0	Transient	0.00
116251	425.0	0.0	0	Transient	73.80
116534	9.0	0.0	0	Transient-Party	22.86
117087	52.0	0.0	0	Transient	0.00

	required_car_parking_spaces	total_of_special_requests \
2224	0	0
2409	0	0
3181	0	0
3684	0	0
3708	0	0
...
115029	0	0
115091	1	1
116251	0	0
116534	0	1
117087	0	0

	reservation_status	reservation_status_date
2224	Check-Out	10/6/2015
2409	Check-Out	10/12/2015
3181	Check-Out	11/23/2015
3684	Check-Out	1/4/2016
3708	Check-Out	1/5/2016
...
115029	Check-Out	6/30/2017
115091	Check-Out	7/1/2017

116251	Check-Out	7/17/2017
116534	Check-Out	7/22/2017
117087	Check-Out	7/29/2017

[180 rows x 32 columns]

2 analysing the data

```
[66]: #where the guests come from?
      #spatial analysis

country_analysis = hotel[hotel['is_canceled']==0]['country'].value_counts().
      ↪reset_index()
```

```
[67]: country_analysis
```

```
[67]:   index  country
0     PRT    21071
1     GBR     9676
2     FRA     8481
3     ESP     6391
4     DEU     6069
..    ...      ...
161    BHR         1
162    AIA         1
163    BHS         1
164    TJK         1
165    BDI         1
```

[166 rows x 2 columns]

```
[68]: #rename the columns

country_analysis.columns=['country','no.of guests']
```

```
[69]: country_analysis
```

```
[69]:   country  no.of guests
0     PRT         21071
1     GBR          9676
2     FRA          8481
3     ESP          6391
4     DEU          6069
..    ...           ...
161    BHR              1
162    AIA              1
163    BHS              1
```

```
164     TJK          1
165     BDI          1
```

```
[166 rows x 2 columns]
```

```
[70]: !pip install folium
```

```
Requirement already satisfied: folium in c:\users\hp\anaconda3\lib\site-packages
(0.12.1)
Requirement already satisfied: branca>=0.3.0 in c:\users\hp\anaconda3\lib\site-
packages (from folium) (0.4.2)
Requirement already satisfied: requests in c:\users\hp\anaconda3\lib\site-
packages (from folium) (2.24.0)
Requirement already satisfied: numpy in c:\users\hp\anaconda3\lib\site-packages
(from folium) (1.19.2)
Requirement already satisfied: jinja2>=2.9 in c:\users\hp\anaconda3\lib\site-
packages (from folium) (2.11.2)
Requirement already satisfied: urllib3!=1.25.0,!1.25.1,<1.26,>=1.21.1 in
c:\users\hp\anaconda3\lib\site-packages (from requests->folium) (1.25.11)
Requirement already satisfied: certifi>=2017.4.17 in
c:\users\hp\anaconda3\lib\site-packages (from requests->folium) (2020.6.20)
Requirement already satisfied: chardet<4,>=3.0.2 in
c:\users\hp\anaconda3\lib\site-packages (from requests->folium) (3.0.4)
Requirement already satisfied: idna<3,>=2.5 in c:\users\hp\anaconda3\lib\site-
packages (from requests->folium) (2.10)
Requirement already satisfied: MarkupSafe>=0.23 in
c:\users\hp\anaconda3\lib\site-packages (from jinja2>=2.9->folium) (1.1.1)
```

```
[71]: import folium
      from folium.plugins import HeatMap
```

```
[72]: basemap = folium.Map()
```

```
[73]: basemap
```

```
[73]: <folium.folium.Map at 0x26b6622d670>
```

```
[74]: !pip install plotly
```

```
Requirement already satisfied: plotly in c:\users\hp\anaconda3\lib\site-packages
(4.14.3)
Requirement already satisfied: six in c:\users\hp\anaconda3\lib\site-packages
(from plotly) (1.15.0)
Requirement already satisfied: retrying>=1.3.3 in
c:\users\hp\anaconda3\lib\site-packages (from plotly) (1.3.3)
```

```
[75]: import plotly.express as px
```

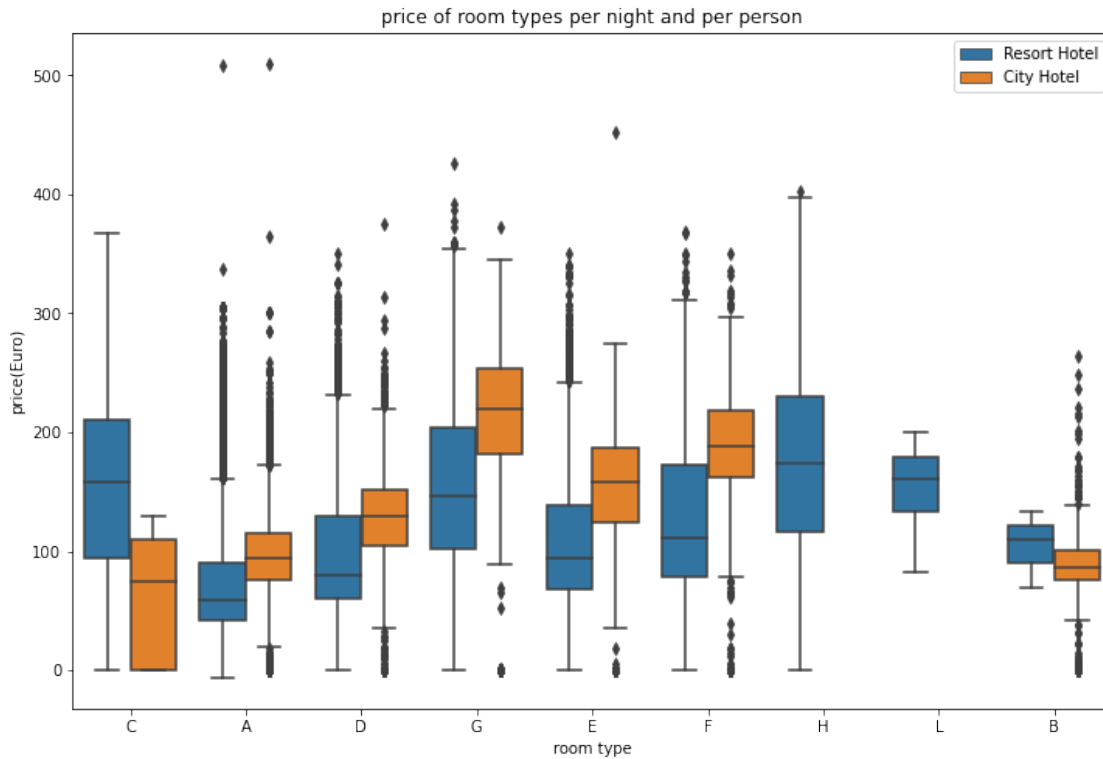
```
[76]: map_guests = px.choropleth(country_analysis,
                                locations = country_analysis['country'],
                                color = country_analysis['no.of guests'],
                                hover_name = country_analysis['country'],
                                title='Home country of guests')
map_guests.show()
```

```
[77]: #how much guests pay for hotel
      #distribution of hotel type

hotel_type = hotel[hotel['is_canceled']== 0]
hotel_type.columns
```

```
[77]: Index(['hotel', 'is_canceled', 'lead_time', 'arrival_date_year',
          'arrival_date_month', 'arrival_date_week_number',
          'arrival_date_day_of_month', 'stays_in_weekend_nights',
          'stays_in_week_nights', 'adults', 'children', 'babies', 'meal',
          'country', 'market_segment', 'distribution_channel',
          'is_repeated_guest', 'previous_cancellations',
          'previous_bookings_not_canceled', 'reserved_room_type',
          'assigned_room_type', 'booking_changes', 'deposit_type', 'agent',
          'company', 'days_in_waiting_list', 'customer_type', 'adr',
          'required_car_parking_spaces', 'total_of_special_requests',
          'reservation_status', 'reservation_status_date'],
          dtype='object')
```

```
[78]: plt.figure(figsize=(12,8))
sns.boxplot(x='reserved_room_type',y='adr',hue='hotel', data=hotel_type)
plt.title("price of room types per night and per person")
plt.xlabel('room type')
plt.ylabel('price(Euro)')
plt.legend()
plt.show()
```



```
[79]: # how does the price per night vary over the year
```

```
[80]: hotel.columns
```

```
[80]: Index(['hotel', 'is_canceled', 'lead_time', 'arrival_date_year',
         'arrival_date_month', 'arrival_date_week_number',
         'arrival_date_day_of_month', 'stays_in_weekend_nights',
         'stays_in_week_nights', 'adults', 'children', 'babies', 'meal',
         'country', 'market_segment', 'distribution_channel',
         'is_repeated_guest', 'previous_cancellations',
         'previous_bookings_not_canceled', 'reserved_room_type',
         'assigned_room_type', 'booking_changes', 'deposit_type', 'agent',
         'company', 'days_in_waiting_list', 'customer_type', 'adr',
         'required_car_parking_spaces', 'total_of_special_requests',
         'reservation_status', 'reservation_status_date'],
        dtype='object')
```

```
[81]: resort_hotel=hotel[(hotel['hotel']=='Resort Hotel') & (hotel['is_canceled']==0)]
      city_hotel=hotel[(hotel['hotel']=='City Hotel') & (hotel['is_canceled']==0)]
```

```
[99]: resort_hotel.head()
```

```

[99]:      hotel  is_canceled  lead_time  arrival_date_year  arrival_date_month  \
0  Resort Hotel           0        342            2015            July
1  Resort Hotel           0        737            2015            July
2  Resort Hotel           0         7            2015            July
3  Resort Hotel           0        13            2015            July
4  Resort Hotel           0        14            2015            July

      arrival_date_week_number  arrival_date_day_of_month  \
0                             27                        1
1                             27                        1
2                             27                        1
3                             27                        1
4                             27                        1

      stays_in_weekend_nights  stays_in_week_nights  adults  children  babies  \
0                             0                     0       2        0.0      0
1                             0                     0       2        0.0      0
2                             0                     1       1        0.0      0
3                             0                     1       1        0.0      0
4                             0                     2       2        0.0      0

      meal  country  market_segment  distribution_channel  is_repeated_guest  \
0  BB      PRT      Direct          Direct              0
1  BB      PRT      Direct          Direct              0
2  BB      GBR      Direct          Direct              0
3  BB      GBR      Corporate        Corporate          0
4  BB      GBR      Online TA        TA/TO              0

      previous_cancellations  previous_bookings_not_canceled  reserved_room_type  \
0                             0                             0                  C
1                             0                             0                  C
2                             0                             0                  A
3                             0                             0                  A
4                             0                             0                  A

      assigned_room_type  booking_changes  deposit_type  agent  company  \
0                      C                3  No Deposit   0.0    0.0
1                      C                4  No Deposit   0.0    0.0
2                      C                0  No Deposit   0.0    0.0
3                      A                0  No Deposit  304.0    0.0
4                      A                0  No Deposit  240.0    0.0

      days_in_waiting_list  customer_type  adr  required_car_parking_spaces  \
0                             0    Transient   0.0                        0
1                             0    Transient   0.0                        0
2                             0    Transient  75.0                        0
3                             0    Transient  75.0                        0

```

4	0	Transient	98.0	0
---	---	-----------	------	---

	total_of_special_requests	reservation_status	reservation_status_date
0	0	Check-Out	7/1/2015
1	0	Check-Out	7/1/2015
2	0	Check-Out	7/2/2015
3	0	Check-Out	7/2/2015
4	1	Check-Out	7/3/2015

```
[100]: #group by arrival date month and price
        #to make it a df, reset index

        resort_hotelfdf = resort_hotel.groupby(['arrival_date_month'])['adr'].mean().
        ↪reset_index()
```

```
[101]: resort_hotelfdf
```

```
[101]:   arrival_date_month      adr
0         April    75.867816
1         August   181.205892
2        December    68.322236
3        February    54.147478
4         January    48.708919
5          July    150.122528
6          June    107.921869
7          March    57.012487
8           May    76.657558
9        November    48.681640
10         October    61.727505
11        September    96.416860
```

```
[102]: city_hotelfdf = city_hotel.groupby(['arrival_date_month'])['adr'].mean().
        ↪reset_index()
```

```
[103]: city_hotelfdf
```

```
[103]:   arrival_date_month      adr
0         April    111.856824
1         August    118.412083
2        December    87.856764
3        February    86.183025
4         January    82.160634
5          July    115.563810
6          June    117.702075
7          March    90.170722
8           May    120.445842
9        November    86.500456
```



```

10         October 101.745956
11         September 112.598452

```

```
[104]: #merge the two dataframes
```

```

final = resort_hotelfdf.merge(city_hotelfdf, on='arrival_date_month')
final.columns = ['months', 'price_for_resort', 'price_for_city']

```

```
[105]: final
```

```

[105]:      months  price_for_resort  price_for_city
0      April          75.867816          111.856824
1      August          181.205892          118.412083
2      December          68.322236           87.856764
3      February          54.147478           86.183025
4      January          48.708919           82.160634
5      July           150.122528          115.563810
6      June           107.921869          117.702075
7      March           57.012487           90.170722
8      May            76.657558          120.445842
9      November          48.681640           86.500456
10     October          61.727505          101.745956
11     September          96.416860          112.598452

```

```
[106]: #sort the months using
```

```
!pip install sorted-months-weekdays
```

Requirement already satisfied: sorted-months-weekdays in
c:\users\hp\anaconda3\lib\site-packages (0.2)

```
[107]: !pip install sort-dataframeby-monthorweek
```

Requirement already satisfied: sort-dataframeby-monthorweek in
c:\users\hp\anaconda3\lib\site-packages (0.4)

```
[108]: import sort_dataframeby_monthorweek as sd
```

```

[109]: def sort_data(df, colname):
        return sd.Sort_Dataframeby_Month(df, colname)

```

```

[110]: final = sort_data(final, 'months')
final

```

```

[110]:      months  price_for_resort  price_for_city
0      January          48.708919           82.160634
1      February          54.147478           86.183025
2      March           57.012487           90.170722

```

3	April	75.867816	111.856824
4	May	76.657558	120.445842
5	June	107.921869	117.702075
6	July	150.122528	115.563810
7	August	181.205892	118.412083
8	September	96.416860	112.598452
9	October	61.727505	101.745956
10	November	48.681640	86.500456
11	December	68.322236	87.856764

```
[111]: #visuals.
#line plot

px.line(final,x='months',y=['price_for_resort', 'price_for_city'],
        title='room price overnight per month')
```

```
[112]: final.columns
```

```
[112]: Index(['months', 'price_for_resort', 'price_for_city'], dtype='object')
```

```
[115]: #analysis demands of hotels
```

```
[116]: rush_resort = resort_hotel['arrival_date_month'].value_counts().reset_index()
rush_resort.columns = ['months','no.of guests']
rush_resort
```

```
[116]:
```

	months	no.of guests
0	August	3257
1	July	3137
2	October	2577
3	March	2573
4	April	2550
5	May	2535
6	February	2308
7	September	2102
8	June	2038
9	December	2017
10	November	1976
11	January	1868

```
[117]: rush_city = city_hotel['arrival_date_month'].value_counts().reset_index()
rush_city.columns = ['months','no.of guests']
rush_city
```

```
[117]:
```

	months	no.of guests
0	August	5381
1	July	4782
2	May	4579

3	June	4366
4	October	4337
5	September	4290
6	March	4072
7	April	4015
8	February	3064
9	November	2696
10	December	2392
11	January	2254

[118]: *#merge dataframes*

```
final_rush = rush_resort.merge(rush_city,on='months')
final_rush.columns = ['months','no of guests in resort','no of guest in city']
final_rush
```

[118]:

	months	no of guests in resort	no of guest in city
0	August	3257	5381
1	July	3137	4782
2	October	2577	4337
3	March	2573	4072
4	April	2550	4015
5	May	2535	4579
6	February	2308	3064
7	September	2102	4290
8	June	2038	4366
9	December	2017	2392
10	November	1976	2696
11	January	1868	2254

[119]: *#hierachy of my months*

```
final_rush = sort_data(final_rush,'months')
final_rush
```

[119]:

	months	no of guests in resort	no of guest in city
0	January	1868	2254
1	February	2308	3064
2	March	2573	4072
3	April	2550	4015
4	May	2535	4579
5	June	2038	4366
6	July	3137	4782
7	August	3257	5381
8	September	2102	4290
9	October	2577	4337
10	November	1976	2696
11	December	2017	2392

```
[120]: #we need trend, so we go for line plot

px.line(final_rush,x='months',y= ['no of guests in resort', 'no of guest in_
↪city'],
        title='total no of guest per months')
```

3 machine learning

```
[121]: hotel.head()
```

```
[121]:
```

	hotel	is_canceled	lead_time	arrival_date_year	arrival_date_month	\
0	Resort Hotel	0	342	2015	July	
1	Resort Hotel	0	737	2015	July	
2	Resort Hotel	0	7	2015	July	
3	Resort Hotel	0	13	2015	July	
4	Resort Hotel	0	14	2015	July	

	arrival_date_week_number	arrival_date_day_of_month	\
0	27	1	
1	27	1	
2	27	1	
3	27	1	
4	27	1	

	stays_in_weekend_nights	stays_in_week_nights	adults	children	babies	\
0	0	0	2	0.0	0	
1	0	0	2	0.0	0	
2	0	1	1	0.0	0	
3	0	1	1	0.0	0	
4	0	2	2	0.0	0	

	meal	country	market_segment	distribution_channel	is_repeated_guest	\
0	BB	PRT	Direct	Direct	0	
1	BB	PRT	Direct	Direct	0	
2	BB	GBR	Direct	Direct	0	
3	BB	GBR	Corporate	Corporate	0	
4	BB	GBR	Online TA	TA/TO	0	

	previous_cancellations	previous_bookings_not_canceled	reserved_room_type	\
0	0	0	C	
1	0	0	C	
2	0	0	A	
3	0	0	A	
4	0	0	A	

	assigned_room_type	booking_changes	deposit_type	agent	company	\
--	--------------------	-----------------	--------------	-------	---------	---

0	C	3	No Deposit	0.0	0.0
1	C	4	No Deposit	0.0	0.0
2	C	0	No Deposit	0.0	0.0
3	A	0	No Deposit	304.0	0.0
4	A	0	No Deposit	240.0	0.0

	days_in_waiting_list	customer_type	adr	required_car_parking_spaces	\
0	0	Transient	0.0		0
1	0	Transient	0.0		0
2	0	Transient	75.0		0
3	0	Transient	75.0		0
4	0	Transient	98.0		0

	total_of_special_requests	reservation_status	reservation_status_date
0	0	Check-Out	7/1/2015
1	0	Check-Out	7/1/2015
2	0	Check-Out	7/2/2015
3	0	Check-Out	7/2/2015
4	1	Check-Out	7/3/2015

```
[122]: #find correlation
hotel.corr()
```

```
[122]:
```

	is_canceled	lead_time	arrival_date_year	\
is_canceled	1.000000	0.293123	0.016660	
lead_time	0.293123	1.000000	0.040142	
arrival_date_year	0.016660	0.040142	1.000000	
arrival_date_week_number	0.008148	0.126871	-0.540561	
arrival_date_day_of_month	-0.006130	0.002268	-0.000221	
stays_in_weekend_nights	-0.001791	0.085671	0.021497	
stays_in_week_nights	0.024765	0.165799	0.030883	
adults	0.060017	0.119519	0.029635	
children	0.005036	-0.037613	0.054636	
babies	-0.032491	-0.020915	-0.013192	
is_repeated_guest	-0.084793	-0.124410	0.010341	
previous_cancellations	0.110133	0.086042	-0.119822	
previous_bookings_not_canceled	-0.057358	-0.073548	0.029218	
booking_changes	-0.144381	0.000149	0.030872	
agent	-0.046529	-0.012640	0.056463	
company	-0.082995	-0.086250	0.033882	
days_in_waiting_list	0.054186	0.170084	-0.056497	
adr	0.047557	-0.063077	0.197580	
required_car_parking_spaces	-0.195498	-0.116451	-0.013684	
total_of_special_requests	-0.234658	-0.095712	0.108531	

	arrival_date_week_number	\
is_canceled	0.008148	

lead_time	0.126871
arrival_date_year	-0.540561
arrival_date_week_number	1.000000
arrival_date_day_of_month	0.066809
stays_in_weekend_nights	0.018208
stays_in_week_nights	0.015558
adults	0.025909
children	0.005515
babies	0.010395
is_repeated_guest	-0.030131
previous_cancellations	0.035501
previous_bookings_not_canceled	-0.020904
booking_changes	0.005508
agent	-0.018244
company	-0.032750
days_in_waiting_list	0.022933
adr	0.075791
required_car_parking_spaces	0.001920
total_of_special_requests	0.026149

	arrival_date_day_of_month \
is_canceled	-0.006130
lead_time	0.002268
arrival_date_year	-0.000221
arrival_date_week_number	0.066809
arrival_date_day_of_month	1.000000
stays_in_weekend_nights	-0.016354
stays_in_week_nights	-0.028174
adults	-0.001566
children	0.014553
babies	-0.000230
is_repeated_guest	-0.006145
previous_cancellations	-0.027011
previous_bookings_not_canceled	-0.000300
booking_changes	0.010613
agent	0.000202
company	0.003724
days_in_waiting_list	0.022728
adr	0.030245
required_car_parking_spaces	0.008683
total_of_special_requests	0.003062

	stays_in_weekend_nights	stays_in_week_nights \
is_canceled	-0.001791	0.024765
lead_time	0.085671	0.165799
arrival_date_year	0.021497	0.030883
arrival_date_week_number	0.018208	0.015558

arrival_date_day_of_month	-0.016354	-0.028174
stays_in_weekend_nights	1.000000	0.498969
stays_in_week_nights	0.498969	1.000000
adults	0.091871	0.092976
children	0.045794	0.044203
babies	0.018483	0.020191
is_repeated_guest	-0.087239	-0.097245
previous_cancellations	-0.012775	-0.013992
previous_bookings_not_canceled	-0.042715	-0.048743
booking_changes	0.063281	0.096209
agent	0.161427	0.195135
company	-0.079977	-0.043641
days_in_waiting_list	-0.054151	-0.002020
adr	0.049342	0.065237
required_car_parking_spaces	-0.018554	-0.024859
total_of_special_requests	0.072671	0.068192

	adults	children	babies	\
is_canceled	0.060017	0.005036	-0.032491	
lead_time	0.119519	-0.037613	-0.020915	
arrival_date_year	0.029635	0.054636	-0.013192	
arrival_date_week_number	0.025909	0.005515	0.010395	
arrival_date_day_of_month	-0.001566	0.014553	-0.000230	
stays_in_weekend_nights	0.091871	0.045794	0.018483	
stays_in_week_nights	0.092976	0.044203	0.020191	
adults	1.000000	0.030440	0.018146	
children	0.030440	1.000000	0.024030	
babies	0.018146	0.024030	1.000000	
is_repeated_guest	-0.146426	-0.032858	-0.008943	
previous_cancellations	-0.006738	-0.024729	-0.007501	
previous_bookings_not_canceled	-0.107983	-0.021072	-0.006550	
booking_changes	-0.051673	0.048952	0.083440	
agent	0.024994	0.050581	0.030266	
company	-0.166778	-0.042622	-0.009459	
days_in_waiting_list	-0.008283	-0.033271	-0.010621	
adr	0.230641	0.324853	0.029186	
required_car_parking_spaces	0.014785	0.056255	0.037383	
total_of_special_requests	0.122884	0.081736	0.097889	

	is_repeated_guest	previous_cancellations	\
is_canceled	-0.084793	0.110133	
lead_time	-0.124410	0.086042	
arrival_date_year	0.010341	-0.119822	
arrival_date_week_number	-0.030131	0.035501	
arrival_date_day_of_month	-0.006145	-0.027011	
stays_in_weekend_nights	-0.087239	-0.012775	
stays_in_week_nights	-0.097245	-0.013992	

adults	-0.146426	-0.006738
children	-0.032858	-0.024729
babies	-0.008943	-0.007501
is_repeated_guest	1.000000	0.082293
previous_cancellations	0.082293	1.000000
previous_bookings_not_canceled	0.418056	0.152728
booking_changes	0.012092	-0.026993
agent	-0.052264	-0.018192
company	0.159723	-0.001190
days_in_waiting_list	-0.022235	0.005929
adr	-0.134314	-0.065646
required_car_parking_spaces	0.077090	-0.018492
total_of_special_requests	0.013050	-0.048384

	previous_bookings_not_canceled \
is_canceled	-0.057358
lead_time	-0.073548
arrival_date_year	0.029218
arrival_date_week_number	-0.020904
arrival_date_day_of_month	-0.000300
stays_in_weekend_nights	-0.042715
stays_in_week_nights	-0.048743
adults	-0.107983
children	-0.021072
babies	-0.006550
is_repeated_guest	0.418056
previous_cancellations	0.152728
previous_bookings_not_canceled	1.000000
booking_changes	0.011608
agent	-0.046296
company	0.110817
days_in_waiting_list	-0.009397
adr	-0.072144
required_car_parking_spaces	0.047653
total_of_special_requests	0.037824

	booking_changes	agent	company \
is_canceled	-0.144381	-0.046529	-0.082995
lead_time	0.000149	-0.012640	-0.086250
arrival_date_year	0.030872	0.056463	0.033882
arrival_date_week_number	0.005508	-0.018244	-0.032750
arrival_date_day_of_month	0.010613	0.000202	0.003724
stays_in_weekend_nights	0.063281	0.161427	-0.079977
stays_in_week_nights	0.096209	0.195135	-0.043641
adults	-0.051673	0.024994	-0.166778
children	0.048952	0.050581	-0.042622
babies	0.083440	0.030266	-0.009459

is_repeated_guest	0.012092	-0.052264	0.159723
previous_cancellations	-0.026993	-0.018192	-0.001190
previous_bookings_not_canceled	0.011608	-0.046296	0.110817
booking_changes	1.000000	0.036478	0.088863
agent	0.036478	1.000000	-0.121536
company	0.088863	-0.121536	1.000000
days_in_waiting_list	-0.011634	-0.040853	-0.022986
adr	0.019618	0.016707	-0.128470
required_car_parking_spaces	0.065620	0.119158	0.038299
total_of_special_requests	0.052833	0.060696	-0.091066

	days_in_waiting_list	adr \
is_canceled	0.054186	0.047557
lead_time	0.170084	-0.063077
arrival_date_year	-0.056497	0.197580
arrival_date_week_number	0.022933	0.075791
arrival_date_day_of_month	0.022728	0.030245
stays_in_weekend_nights	-0.054151	0.049342
stays_in_week_nights	-0.002020	0.065237
adults	-0.008283	0.230641
children	-0.033271	0.324853
babies	-0.010621	0.029186
is_repeated_guest	-0.022235	-0.134314
previous_cancellations	0.005929	-0.065646
previous_bookings_not_canceled	-0.009397	-0.072144
booking_changes	-0.011634	0.019618
agent	-0.040853	0.016707
company	-0.022986	-0.128470
days_in_waiting_list	1.000000	-0.040756
adr	-0.040756	1.000000
required_car_parking_spaces	-0.030600	0.056628
total_of_special_requests	-0.082730	0.172185

	required_car_parking_spaces \
is_canceled	-0.195498
lead_time	-0.116451
arrival_date_year	-0.013684
arrival_date_week_number	0.001920
arrival_date_day_of_month	0.008683
stays_in_weekend_nights	-0.018554
stays_in_week_nights	-0.024859
adults	0.014785
children	0.056255
babies	0.037383
is_repeated_guest	0.077090
previous_cancellations	-0.018492
previous_bookings_not_canceled	0.047653

booking_changes	0.065620
agent	0.119158
company	0.038299
days_in_waiting_list	-0.030600
adr	0.056628
required_car_parking_spaces	1.000000
total_of_special_requests	0.082626

	total_of_special_requests
is_canceled	-0.234658
lead_time	-0.095712
arrival_date_year	0.108531
arrival_date_week_number	0.026149
arrival_date_day_of_month	0.003062
stays_in_weekend_nights	0.072671
stays_in_week_nights	0.068192
adults	0.122884
children	0.081736
babies	0.097889
is_repeated_guest	0.013050
previous_cancellations	-0.048384
previous_bookings_not_canceled	0.037824
booking_changes	0.052833
agent	0.060696
company	-0.091066
days_in_waiting_list	-0.082730
adr	0.172185
required_car_parking_spaces	0.082626
total_of_special_requests	1.000000

[123]: *#correlation with respect to is cancelled*

```
co_relate = hotel.corr()['is_canceled']
co_relate
```

[123]: is_canceled	1.000000
lead_time	0.293123
arrival_date_year	0.016660
arrival_date_week_number	0.008148
arrival_date_day_of_month	-0.006130
stays_in_weekend_nights	-0.001791
stays_in_week_nights	0.024765
adults	0.060017
children	0.005036
babies	-0.032491
is_repeated_guest	-0.084793
previous_cancellations	0.110133

```

previous_bookings_not_canceled    -0.057358
booking_changes                   -0.144381
agent                             -0.046529
company                           -0.082995
days_in_waiting_list             0.054186
adr                               0.047557
required_car_parking_spaces       -0.195498
total_of_special_requests         -0.234658
Name: is_canceled, dtype: float64

```

```
[124]: #finding the most important features
```

```
co_relate.abs().sort_values(ascending=False)
```

```

[124]: is_canceled          1.000000
lead_time                0.293123
total_of_special_requests 0.234658
required_car_parking_spaces 0.195498
booking_changes          0.144381
previous_cancellations    0.110133
is_repeated_guest         0.084793
company                   0.082995
adults                    0.060017
previous_bookings_not_canceled 0.057358
days_in_waiting_list     0.054186
adr                       0.047557
agent                     0.046529
babies                    0.032491
stays_in_week_nights      0.024765
arrival_date_year          0.016660
arrival_date_week_number   0.008148
arrival_date_day_of_month  0.006130
children                   0.005036
stays_in_weekend_nights    0.001791
Name: is_canceled, dtype: float64

```

```
[125]: #
```

```
hotel.groupby('is_canceled')['reservation_status'].value_counts()
```

```

[125]: is_canceled  reservation_status
0          Check-Out          75166
1          Canceled           43017
          No-Show             1207
Name: reservation_status, dtype: int64

```

```
[126]: #exclude unnecessary features
```

```
list_not = ['days_in_waiting_list ', 'arrival_date_year ']
```

```
[127]: #fetch numerical features we have  
#using a list comprehension
```

```
num_features = [col for col in hotel.columns if hotel[col].dtype != 'object']  
↳ and col not in list_not]  
num_features
```

```
[127]: ['is_canceled',  
        'lead_time',  
        'arrival_date_year',  
        'arrival_date_week_number',  
        'arrival_date_day_of_month',  
        'stays_in_weekend_nights',  
        'stays_in_week_nights',  
        'adults',  
        'children',  
        'babies',  
        'is_repeated_guest',  
        'previous_cancellations',  
        'previous_bookings_not_canceled',  
        'booking_changes',  
        'agent',  
        'company',  
        'days_in_waiting_list',  
        'adr',  
        'required_car_parking_spaces',  
        'total_of_special_requests']
```

```
[128]: hotel.columns
```

```
[128]: Index(['hotel', 'is_canceled', 'lead_time', 'arrival_date_year',  
            'arrival_date_month', 'arrival_date_week_number',  
            'arrival_date_day_of_month', 'stays_in_weekend_nights',  
            'stays_in_week_nights', 'adults', 'children', 'babies', 'meal',  
            'country', 'market_segment', 'distribution_channel',  
            'is_repeated_guest', 'previous_cancellations',  
            'previous_bookings_not_canceled', 'reserved_room_type',  
            'assigned_room_type', 'booking_changes', 'deposit_type', 'agent',  
            'company', 'days_in_waiting_list', 'customer_type', 'adr',  
            'required_car_parking_spaces', 'total_of_special_requests',  
            'reservation_status', 'reservation_status_date'],  
            dtype='object')
```

```
[129]:
```

```
cat_not =  
    ↪['arrival_date_year', 'country', 'assigned_room_type', 'booking_changes',  
    ↪'reservation_status', 'days_in_waiting_list']
```

```
[130]: cat_features = [col for col in hotel.columns if hotel[col].dtype == 'object'  
    ↪and col not in cat_not]
```

```
[131]: cat_features
```

```
[131]: ['hotel',  
        'arrival_date_month',  
        'meal',  
        'market_segment',  
        'distribution_channel',  
        'reserved_room_type',  
        'deposit_type',  
        'customer_type',  
        'reservation_status_date']
```

```
[132]: #extracting derived features
```

```
[133]: cat_data = hotel[cat_features]
```

```
[134]: cat_data.dtypes
```

```
[134]: hotel                object  
arrival_date_month        object  
meal                      object  
market_segment            object  
distribution_channel       object  
reserved_room_type        object  
deposit_type              object  
customer_type             object  
reservation_status_date    object  
dtype: object
```

```
[135]: #when you want to block the warning  
import warnings  
from warnings import filterwarnings  
filterwarnings('ignore')
```

```
[136]: cat_data['reservation_status_date'] = pd.  
    ↪to_datetime(cat_data['reservation_status_date'])
```

```
[137]: #creating different columns for month day and year  
  
cat_data['year'] = cat_data['reservation_status_date'].dt.year  
cat_data['month'] = cat_data['reservation_status_date'].dt.month
```

```
cat_data['day'] = cat_data['reservation_status_date'].dt.day
```

```
[138]: #drop the column with the combination of the data
```

```
cat_data.drop('reservation_status_date', axis =1, inplace=True)
```

```
[139]: cat_data.dtypes
```

```
[139]: hotel                object
arrival_date_month      object
meal                   object
market_segment          object
distribution_channel     object
reserved_room_type      object
deposit_type            object
customer_type           object
year                   int64
month                  int64
day                    int64
dtype: object
```

```
[140]: cat_data['cancellation']=hotel['is_canceled']
```

```
[ ]:
```

```
[141]: cat_data.dtypes
```

```
[141]: hotel                object
arrival_date_month      object
meal                   object
market_segment          object
distribution_channel     object
reserved_room_type      object
deposit_type            object
customer_type           object
year                   int64
month                  int64
day                    int64
cancellation            int64
dtype: object
```

```
[142]: #applying feature encoding
cat_data.head()
```

```
[142]:      hotel arrival_date_month meal market_segment distribution_channel \
0  Resort Hotel           July   BB           Direct           Direct
1  Resort Hotel           July   BB           Direct           Direct
2  Resort Hotel           July   BB           Direct           Direct
```

3	Resort Hotel	July	BB	Corporate	Corporate
4	Resort Hotel	July	BB	Online TA	TA/TO

	reserved_room_type	deposit_type	customer_type	year	month	day	\
0	C	No Deposit	Transient	2015	7	1	
1	C	No Deposit	Transient	2015	7	1	
2	A	No Deposit	Transient	2015	7	2	
3	A	No Deposit	Transient	2015	7	2	
4	A	No Deposit	Transient	2015	7	3	

	cancellation
0	0
1	0
2	0
3	0
4	0

```
[143]: ##mean encoding technique
```

```
cat_data['market_segment'].unique()
```

```
[143]: array(['Direct', 'Corporate', 'Online TA', 'Offline TA/TO',
        'Complementary', 'Groups', 'Undefined', 'Aviation'], dtype=object)
```

```
[144]: col_enc = cat_data.columns[0:8]
```

```
[145]: for col in col_enc:
        print(cat_data.groupby([col])['cancellation'].mean().to_dict())
        print('\n')
```

```
{'City Hotel': 0.41726963317786464, 'Resort Hotel': 0.27763354967548676}
```

```
{'April': 0.4079718640093787, 'August': 0.3775311666786769, 'December':
0.3497050147492625, 'February': 0.3341596430342092, 'January':
0.3047731489289931, 'July': 0.37453597662112, 'June': 0.4145717158789652,
'March': 0.3215233816622422, 'May': 0.39665846832329743, 'November':
0.3123344127171033, 'October': 0.3804659498207885, 'September':
0.3917015607156452}
```

```
{'BB': 0.3738489871086556, 'FB': 0.5989974937343359, 'HB': 0.3446034709258107,
'SC': 0.3723943661971831, 'Undefined': 0.2446535500427716}
```

```
{'Aviation': 0.21940928270042195, 'Complementary': 0.13055181695827725,
'Corporate': 0.1873465533522191, 'Direct': 0.15341900682214818, 'Groups':
0.6106203624249155, 'Offline TA/TO': 0.34316032866757507, 'Online TA':
```

```
0.3672114312020823, 'Undefined': 1.0}
```

```
{'Corporate': 0.22075782537067545, 'Direct': 0.17459883919426425, 'GDS':  
0.19170984455958548, 'TA/TO': 0.41025850618166954, 'Undefined': 0.8}
```

```
{'A': 0.3910737958462218, 'B': 0.3291592128801431, 'C': 0.33047210300429186,  
'D': 0.3177959481277017, 'E': 0.29288446824789593, 'F': 0.30376251294442524,  
'G': 0.3643744030563515, 'H': 0.40765391014975044, 'L': 0.3333333333333333, 'P':  
1.0}
```

```
{'No Deposit': 0.28377022390841067, 'Non Refund': 0.9936244601357374,  
'Refundable': 0.2222222222222222}
```

```
{'Contract': 0.3096172718351325, 'Group': 0.1022530329289428, 'Transient':  
0.4074632028835102, 'Transient-Party': 0.2542986785543703}
```

```
[146]: for col in col_enc:  
        dict = cat_data.groupby([col])['cancellation'].mean().to_dict()  
        cat_data[col] = cat_data[col].map(dict)
```

```
[147]: cat_data.head()
```

```
[147]:      hotel  arrival_date_month      meal  market_segment  \  
0  0.277634      0.374536  0.373849      0.153419  
1  0.277634      0.374536  0.373849      0.153419  
2  0.277634      0.374536  0.373849      0.153419  
3  0.277634      0.374536  0.373849      0.187347  
4  0.277634      0.374536  0.373849      0.367211  
  
      distribution_channel  reserved_room_type  deposit_type  customer_type  \  
0      0.174599      0.330472      0.28377      0.407463  
1      0.174599      0.330472      0.28377      0.407463  
2      0.174599      0.391074      0.28377      0.407463  
3      0.220758      0.391074      0.28377      0.407463  
4      0.410259      0.391074      0.28377      0.407463  
  
      year  month  day  cancellation  
0  2015      7      1              0  
1  2015      7      1              0  
2  2015      7      2              0  
3  2015      7      2              0
```


4 2015 7 3 0

```
[148]: num_features
```

```
[148]: ['is_canceled',
        'lead_time',
        'arrival_date_year',
        'arrival_date_week_number',
        'arrival_date_day_of_month',
        'stays_in_weekend_nights',
        'stays_in_week_nights',
        'adults',
        'children',
        'babies',
        'is_repeated_guest',
        'previous_cancellations',
        'previous_bookings_not_canceled',
        'booking_changes',
        'agent',
        'company',
        'days_in_waiting_list',
        'adr',
        'required_car_parking_spaces',
        'total_of_special_requests']
```

```
[149]: entire_df = pd.concat([cat_data,hotel[num_features]], axis=1)
```

```
[150]: entire_df.head()
```

```
[150]:
```

	hotel	arrival_date_month	meal	market_segment	\
0	0.277634	0.374536	0.373849	0.153419	
1	0.277634	0.374536	0.373849	0.153419	
2	0.277634	0.374536	0.373849	0.153419	
3	0.277634	0.374536	0.373849	0.187347	
4	0.277634	0.374536	0.373849	0.367211	

	distribution_channel	reserved_room_type	deposit_type	customer_type	\
0	0.174599	0.330472	0.28377	0.407463	
1	0.174599	0.330472	0.28377	0.407463	
2	0.174599	0.391074	0.28377	0.407463	
3	0.220758	0.391074	0.28377	0.407463	
4	0.410259	0.391074	0.28377	0.407463	

	year	month	day	cancellation	is_canceled	lead_time	arrival_date_year	\
0	2015	7	1	0	0	342	2015	
1	2015	7	1	0	0	737	2015	
2	2015	7	2	0	0	7	2015	

3	2015	7	2	0	0	13	2015
4	2015	7	3	0	0	14	2015

	arrival_date_week_number	arrival_date_day_of_month	\
0	27	1	
1	27	1	
2	27	1	
3	27	1	
4	27	1	

	stays_in_weekend_nights	stays_in_week_nights	adults	children	babies	\
0	0	0	2	0.0	0	
1	0	0	2	0.0	0	
2	0	1	1	0.0	0	
3	0	1	1	0.0	0	
4	0	2	2	0.0	0	

	is_repeated_guest	previous_cancellations	previous_bookings_not_canceled	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	

	booking_changes	agent	company	days_in_waiting_list	adr	\
0	3	0.0	0.0	0	0.0	
1	4	0.0	0.0	0	0.0	
2	0	0.0	0.0	0	75.0	
3	0	304.0	0.0	0	75.0	
4	0	240.0	0.0	0	98.0	

	required_car_parking_spaces	total_of_special_requests
0	0	0
1	0	0
2	0	0
3	0	0
4	0	1

```
[151]: #
entire_df.drop('cancellation', axis=1, inplace=True)
```

```
[152]: entire_df.shape
```

```
[152]: (119390, 31)
```

```
[153]: #handling outliers
```

```
entire_df.head()
```

```
[153]:
```

	hotel	arrival_date_month	meal	market_segment	\
0	0.277634	0.374536	0.373849	0.153419	
1	0.277634	0.374536	0.373849	0.153419	
2	0.277634	0.374536	0.373849	0.153419	
3	0.277634	0.374536	0.373849	0.187347	
4	0.277634	0.374536	0.373849	0.367211	

	distribution_channel	reserved_room_type	deposit_type	customer_type	\
0	0.174599	0.330472	0.28377	0.407463	
1	0.174599	0.330472	0.28377	0.407463	
2	0.174599	0.391074	0.28377	0.407463	
3	0.220758	0.391074	0.28377	0.407463	
4	0.410259	0.391074	0.28377	0.407463	

	year	month	day	is_canceled	lead_time	arrival_date_year	\
0	2015	7	1	0	342	2015	
1	2015	7	1	0	737	2015	
2	2015	7	2	0	7	2015	
3	2015	7	2	0	13	2015	
4	2015	7	3	0	14	2015	

	arrival_date_week_number	arrival_date_day_of_month	\
0	27	1	
1	27	1	
2	27	1	
3	27	1	
4	27	1	

	stays_in_weekend_nights	stays_in_week_nights	adults	children	babies	\
0	0	0	2	0.0	0	
1	0	0	2	0.0	0	
2	0	1	1	0.0	0	
3	0	1	1	0.0	0	
4	0	2	2	0.0	0	

	is_repeated_guest	previous_cancellations	previous_bookings_not_canceled	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	

	booking_changes	agent	company	days_in_waiting_list	adr	\
0	3	0.0	0.0	0	0.0	
1	4	0.0	0.0	0	0.0	

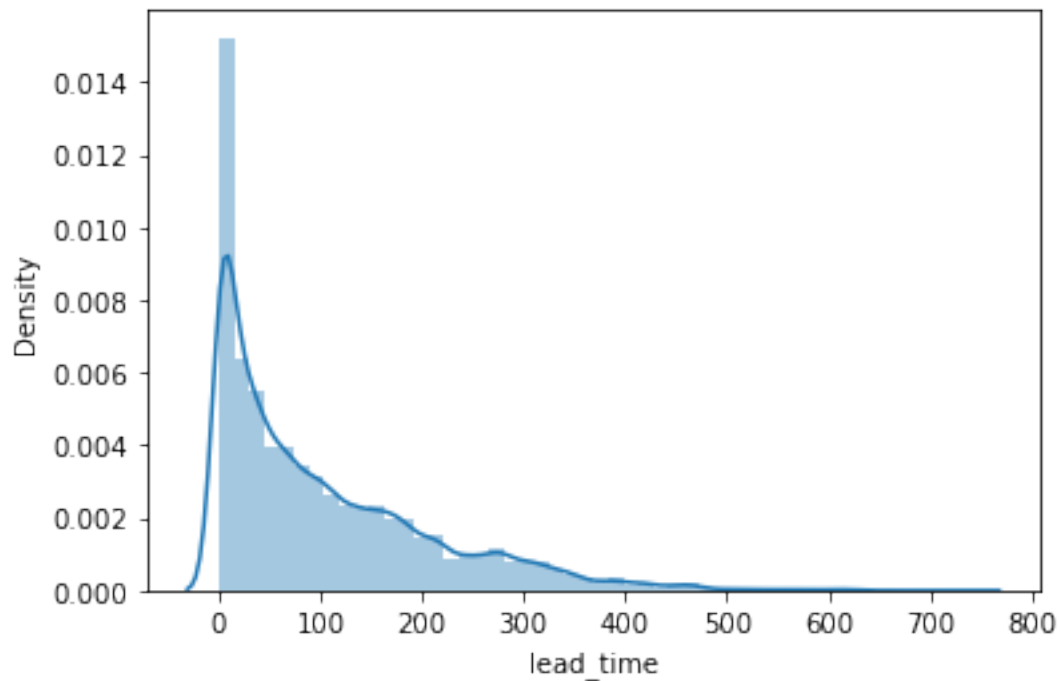
2	0	0.0	0.0	0	75.0
3	0	304.0	0.0	0	75.0
4	0	240.0	0.0	0	98.0

	required_car_parking_spaces	total_of_special_requests
0	0	0
1	0	0
2	0	0
3	0	0
4	0	1

```
[154]: #distribution of lead time
```

```
sns.distplot(entire_df['lead_time'])
```

```
[154]: <AxesSubplot:xlabel='lead_time', ylabel='Density'>
```



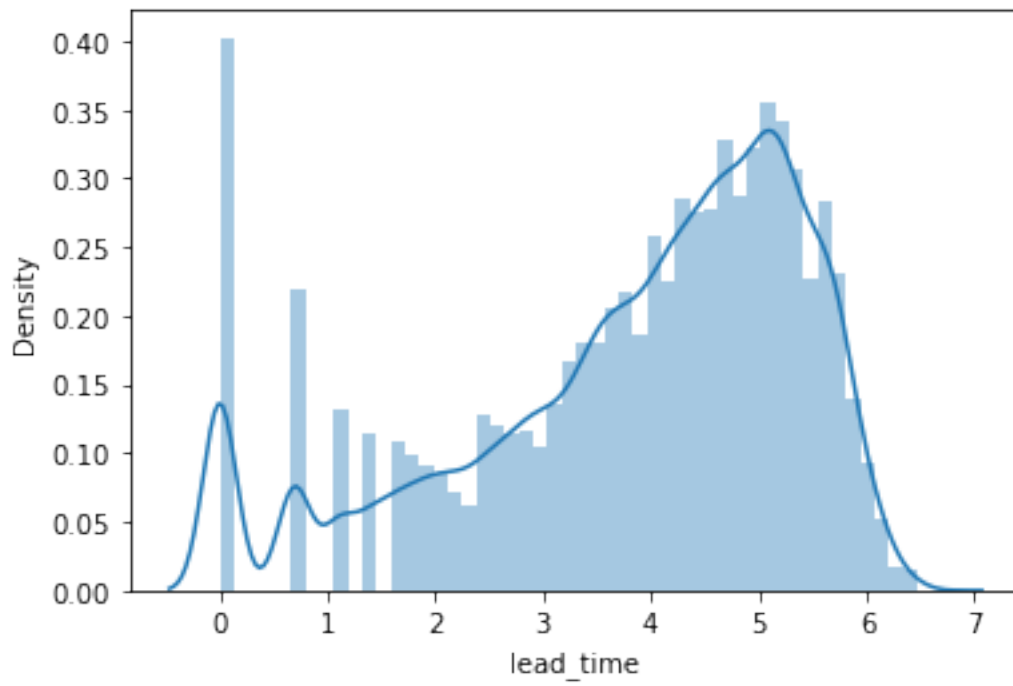
```
[155]: #find the log of these
```

```
def handle_outlier(col):
    entire_df[col] = np.log1p(entire_df[col])
```

```
[156]: handle_outlier('lead_time')
```

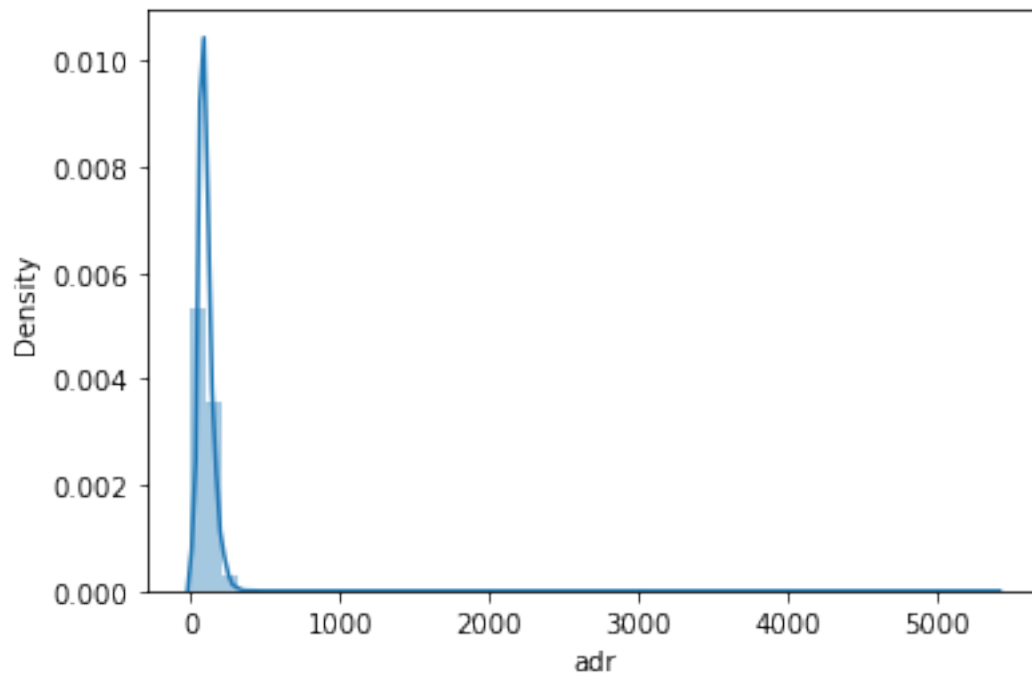
```
[157]: sns.distplot(entire_df['lead_time'])
```

```
[157]: <AxesSubplot:xlabel='lead_time', ylabel='Density'>
```



```
[158]: sns.distplot(entire_df['adr'])
```

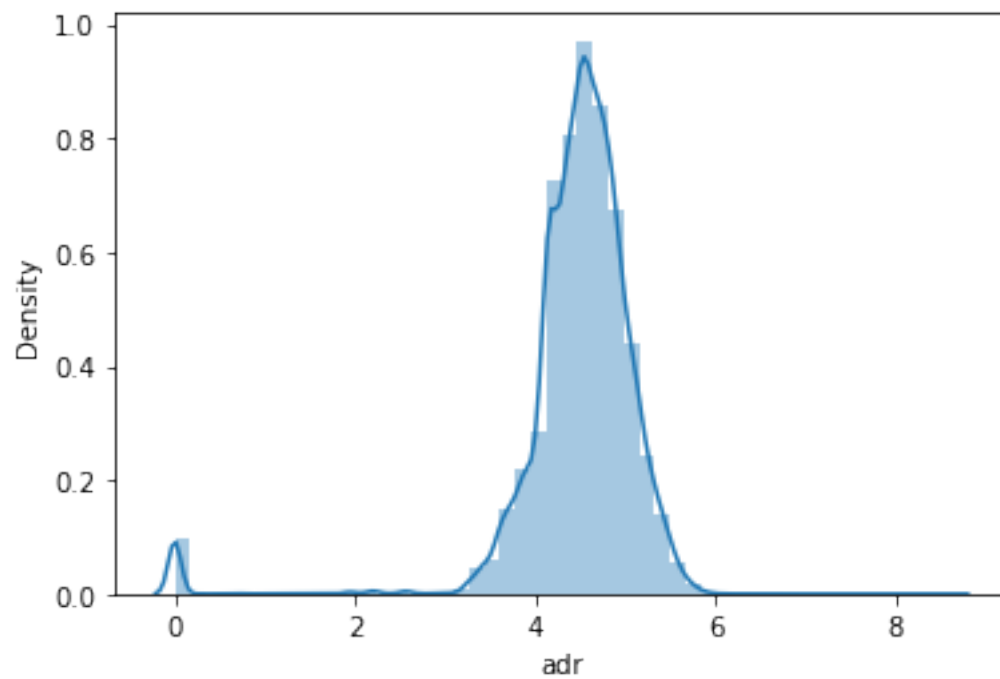
```
[158]: <AxesSubplot:xlabel='adr', ylabel='Density'>
```



```
[159]: handle_outlier('adr')
```

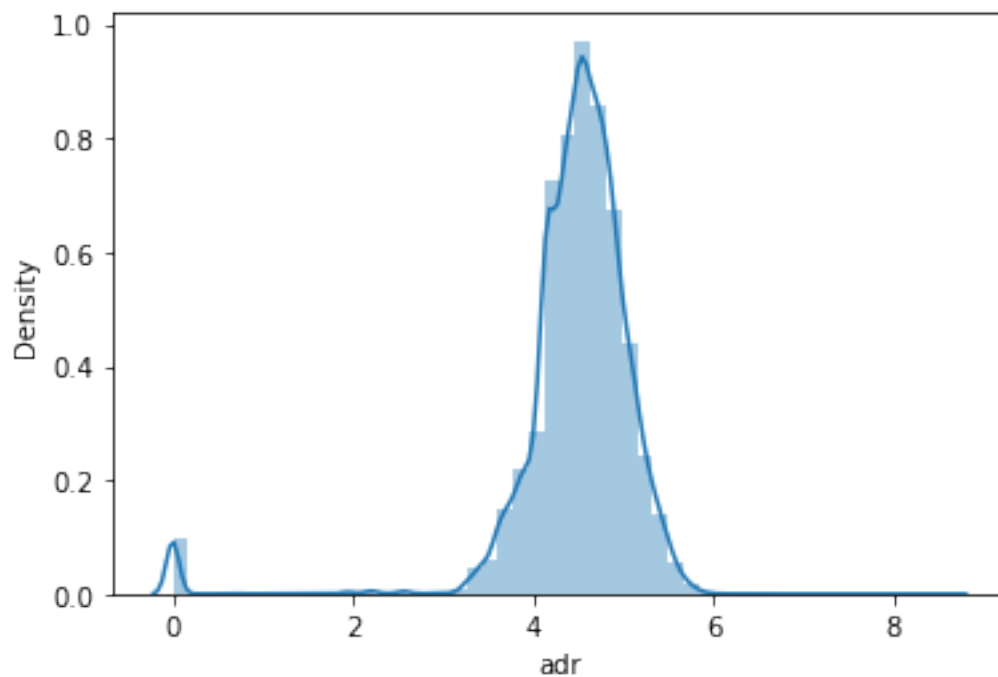
```
[160]: sns.distplot(entire_df['adr'].dropna())
```

```
[160]: <AxesSubplot:xlabel='adr', ylabel='Density'>
```



```
[161]: sns.distplot(entire_df['adr'])
```

```
[161]: <AxesSubplot:xlabel='adr', ylabel='Density'>
```



```
[162]: #applying feature importance  
#most important features  
  
entire_df.isnull().sum()
```

```
[162]: hotel                0  
arrival_date_month        0  
meal                      0  
market_segment            0  
distribution_channel       0  
reserved_room_type        0  
deposit_type              0  
customer_type             0  
year                      0  
month                     0  
day                       0  
is_canceled               0  
lead_time                 0
```

```

arrival_date_year          0
arrival_date_week_number  0
arrival_date_day_of_month  0
stays_in_weekend_nights   0
stays_in_week_nights      0
adults                     0
children                   0
babies                     0
is_repeated_guest          0
previous_cancellations     0
previous_bookings_not_canceled 0
booking_changes            0
agent                      0
company                    0
days_in_waiting_list      0
adr                        1
required_car_parking_spaces 0
total_of_special_requests  0
dtype: int64

```

```
[163]: entire_df.dropna(inplace=True)
```

```
[164]: #dependent and independent feature

#dependent feature
y = entire_df['is_canceled']

#independent features
x=entire_df.drop('is_canceled',axis=1)

```

```
[165]: from sklearn.linear_model import Lasso
from sklearn.feature_selection import SelectFromModel
```

```
[166]: feature_selmodel = SelectFromModel(Lasso(alpha=0.005, random_state=0))
```

```
[167]: feature_selmodel.fit(x,y)
```

```
[167]: SelectFromModel(estimator=Lasso(alpha=0.005, random_state=0))
```

```
[168]: feature_selmodel.get_support()
```

```
[168]: array([False, False, False, False, False, False,  True, False,  True,
         True,  True,  True,  True,  True,  True,  True,  True, False,
         True, False, False,  True,  True,  True, False,  True,  True,
         True,  True,  True])
```

```
[169]: cols = x.columns
```



```
[170]: selected_feat = cols[feature_selmodel.get_support()]

[171]: print('total features {}'.format(x.shape[1]))
       print('selected features {}'.format(len(selected_feat)))

total features 30
selected features 19

[172]: x = x[selected_feat]
```

4 logistic regression

```
[173]: #applying machine learning
       #cross validation of data

[174]: from sklearn.model_selection import train_test_split
       from sklearn.linear_model import LogisticRegression

[188]: X_train, X_test, y_train, y_test = train_test_split(x,y,test_size=0.25,
       ↪random_state=42)

[189]: logreg = LogisticRegression()

[190]: logreg.fit(X_train,y_train)

[190]: LogisticRegression()

[191]: y_pred = logreg.predict(X_test)

[192]: y_pred

[192]: array([0, 0, 0, ..., 0, 1, 0], dtype=int64)

[184]: from sklearn.metrics import confusion_matrix

[185]: confusion_matrix(y_test, y_pred)

[185]: array([[15450,  3292],
       [ 5181,  5925]], dtype=int64)

[193]: from sklearn.metrics import accuracy_score

[194]: accuracy_score(y_test,y_pred)

[194]: 0.7235995711605467

[195]: from sklearn.model_selection import cross_val_score
```

```
[196]: score = cross_val_score(logreg, x, y, cv=10)
```

```
[197]: score.mean()
```

```
[197]: 0.6832624512461718
```

5 applying various algorithmn on this data.

```
[199]: #importing the models from sklearn
```

```
from sklearn.naive_bayes import GaussianNB
from sklearn.linear_model import LogisticRegression
from sklearn.neighbors import KNeighborsClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.tree import DecisionTreeClassifier
```

```
[200]: #initializing the model
```

```
models = []

models.append(('LogisticRegression', LogisticRegression()))
models.append(('Naive bayes', GaussianNB()))
models.append(('RandomForest', RandomForestClassifier()))
models.append(('Decision tree ', DecisionTreeClassifier()))
models.append(('KNN', KNeighborsClassifier()))
```

```
[201]: #fit the models
```

```
for name,model in models:
    print(name)
    model.fit(X_train, y_train)

    predictions = model.predict(X_test)

    from sklearn.metrics import confusion_matrix
    print(confusion_matrix(predictions, y_test))
    print('\n')

    print(accuracy_score(predictions, y_test))
    print('\n')
```

```
LogisticRegression
[[16661  6204]
 [ 2046 4937]]
```

```
0.7235995711605467
```

```
Naive bayes
[[ 7091   767]
 [11616 10374]]
```

0.5851313320825516

```
RandomForest
[[18647   997]
 [   60 10144]]
```

0.9645872420262664

```
Decision tree
[[18188   591]
 [   519 10550]]
```

0.9628115786652373

```
KNN
[[18640   901]
 [   67 10240]]
```

0.9675690163495042

[]: