

AtliQ Grands

Your Healing Place



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About Us

A hospitality leader for two decades, AtliQ Grands boasts a collection of opulent Indian retreats. Scattered across Delhi, Mumbai, Bengaluru, and Hyderabad, their properties offer a haven of luxurious amenities and diverse accommodations, ranging from grand hotels to stunning resorts. AtliQ Grands has built a legacy of exceptional service and unforgettable culinary experiences.



Problem Statement

The Indian luxury hotel chain AtliQ Grands, a hospitality veteran of 20 years, is struggling. They're losing ground due to both aggressive competitors and internal blunders. To fight back, they're turning to data analysis, but lack the in-house expertise. The revenue team is seeking a third-party to analyze guest data and provide valuable insights



AtliQ Hotels Data Analysis Project

```
import pandas as pd
```

==> 1. Data Import and Data Exploration

Datasets

We have 5 csv file

- dim_date.csv
- dim_hotels.csv
- dim_rooms.csv
- fact_aggregated_bookings
- fact_bookings.csv

Read bookings data in a dataframe

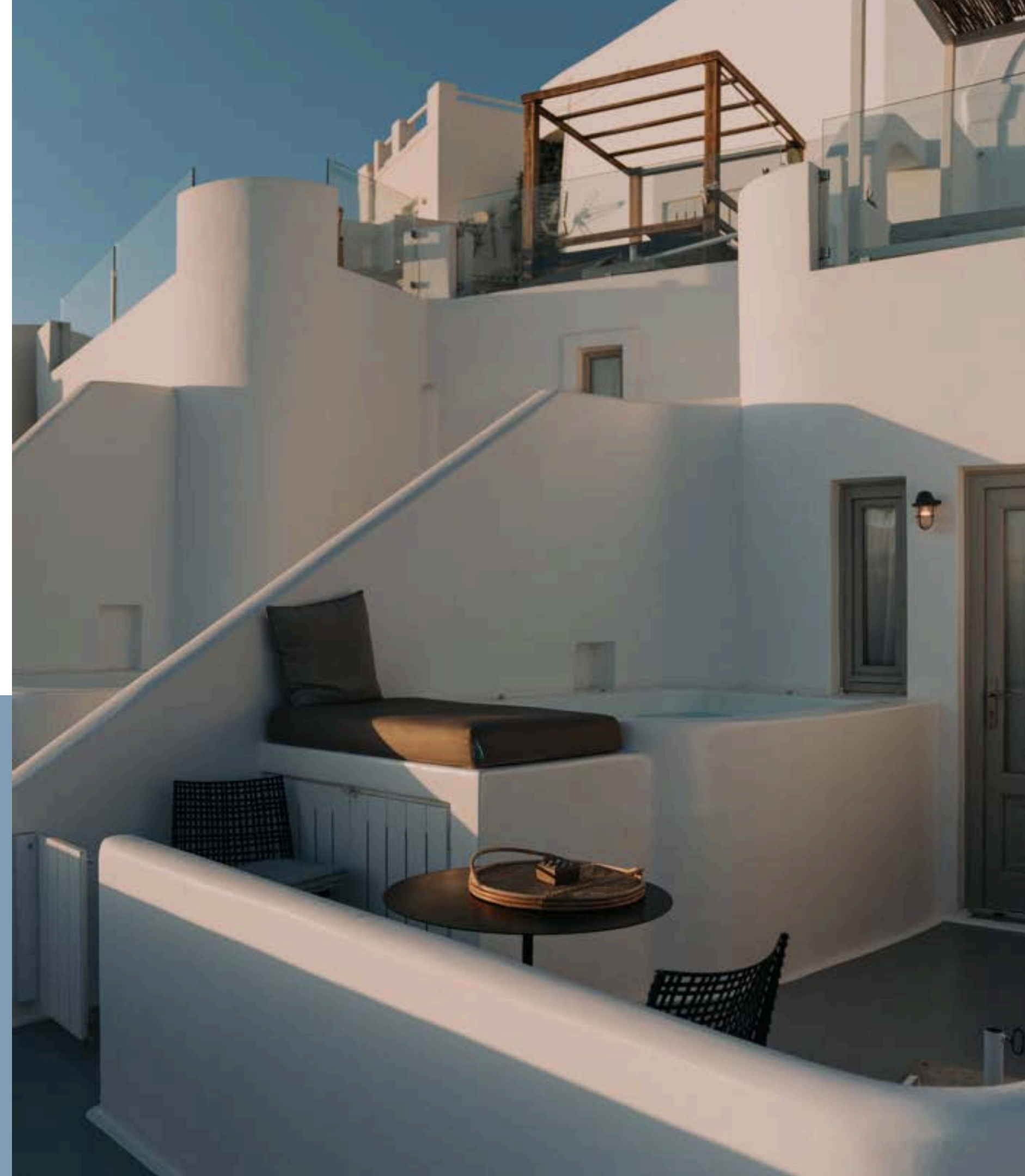
```
df_bookings = pd.read_csv('datasets/fact_bookings.csv')
```

Explore bookings data

```
df_bookings.head()
```

	booking_id	property_id	booking_date	check_in_date	checkout_date
0	May012216558RT11	16558	27-04-22	1/5/2022	2/5/2022
1	May012216558RT12	16558	30-04-22	1/5/2022	2/5/2022
2	May012216558RT13	16558	28-04-22	1/5/2022	4/5/2022
3	May012216558RT14	16558	28-04-22	1/5/2022	2/5/2022
4	May012216558RT15	16558	27-04-22	1/5/2022	2/5/2022

	no_guests	room_category	booking_platform	ratings_given	booking_status
0	-3.0	RT1	direct online	1.0	Checked Out
1	2.0	RT1	others	NaN	Cancelled
2	2.0	RT1	logtrip	5.0	Checked





```
Out
3      -2.0      RT1      others      NaN
Cancelled
4      4.0      RT1      direct online      5.0      Checked
Out
      revenue_generated      revenue_realized
0              10010              10010
1               9100               3640
2            9100000               9100
3               9100               3640
4             10920             10920

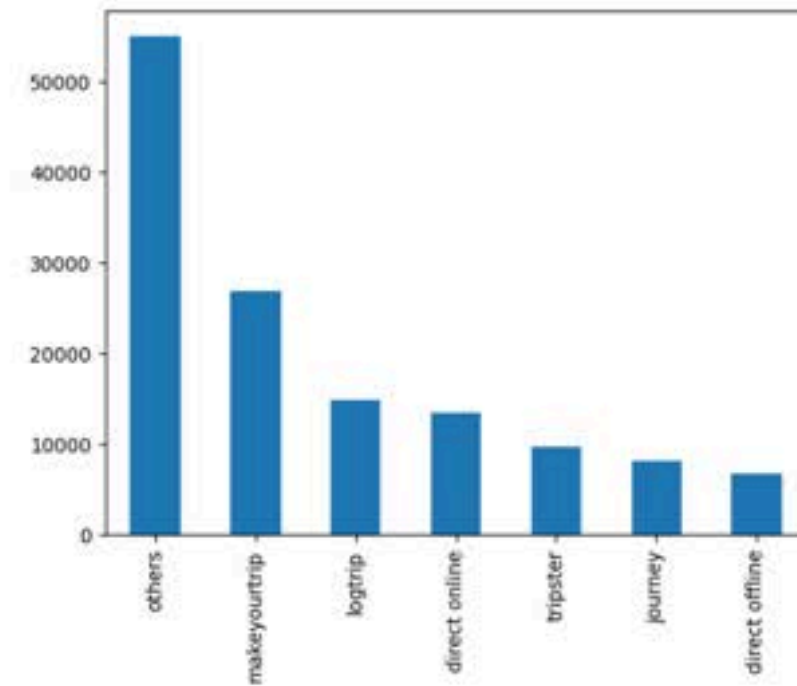
df_bookings.shape
(134590, 12)

df_bookings.room_category.unique()
array(['RT1', 'RT2', 'RT3', 'RT4'], dtype=object)

df_bookings.booking_platform.unique()
array(['direct online', 'others', 'logtrip', 'tripster',
      'makeyourtrip',
      'journey', 'direct offline'], dtype=object)

df_bookings.booking_platform.value_counts()
others          55066
makeyourtrip    26898
logtrip         14756
direct online   13379
tripster        9630
journey         8106
direct offline   6755
Name: booking_platform, dtype: int64

df_bookings.booking_platform.value_counts().plot(kind="bar")
<AxesSubplot: >
```



```
df_bookings.describe()
      property_id      no_guests      ratings_given      revenue_generated
\count  134590.000000  134587.000000  56683.000000  1.345900e+05
mean    18061.113493     2.036170     3.619004  1.537805e+04
std     1093.055847     1.034885     1.235009  9.303604e+04
min     16558.000000    -17.000000     1.000000  6.500000e+03
25%     17558.000000     1.000000     3.000000  9.900000e+03
50%     17564.000000     2.000000     4.000000  1.350000e+04
75%     18563.000000     2.000000     5.000000  1.800000e+04
max     19563.000000     6.000000     5.000000  2.856000e+07

      revenue_realized
```




```
count    134590.000000
mean      12696.123256
std        6928.108124
min        2600.000000
25%        7600.000000
50%       11700.000000
75%       15300.000000
max       45220.000000
```

Read rest of the files

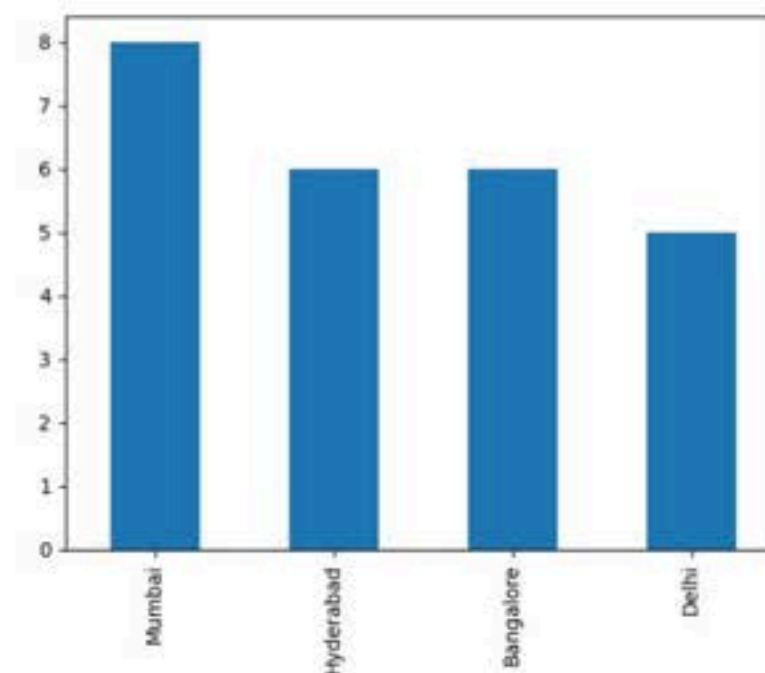
```
df_date = pd.read_csv('datasets/dim_date.csv')
df_hotels = pd.read_csv('datasets/dim_hotels.csv')
df_rooms = pd.read_csv('datasets/dim_rooms.csv')
df_agg_bookings = pd.read_csv('datasets/fact_aggregated_bookings.csv')

df_hotels.shape
(25, 4)

df_hotels.head(3)
  property_id property_name category city
0      16558   Atliq Grands   Luxury  Delhi
1      16559   Atliq Exotica   Luxury  Mumbai
2      16560   Atliq City    Business  Delhi

df_hotels.category.value_counts()
Luxury    16
Business    9
Name: category, dtype: int64

df_hotels.city.value_counts().plot(kind="bar")
<AxesSubplot: >
```

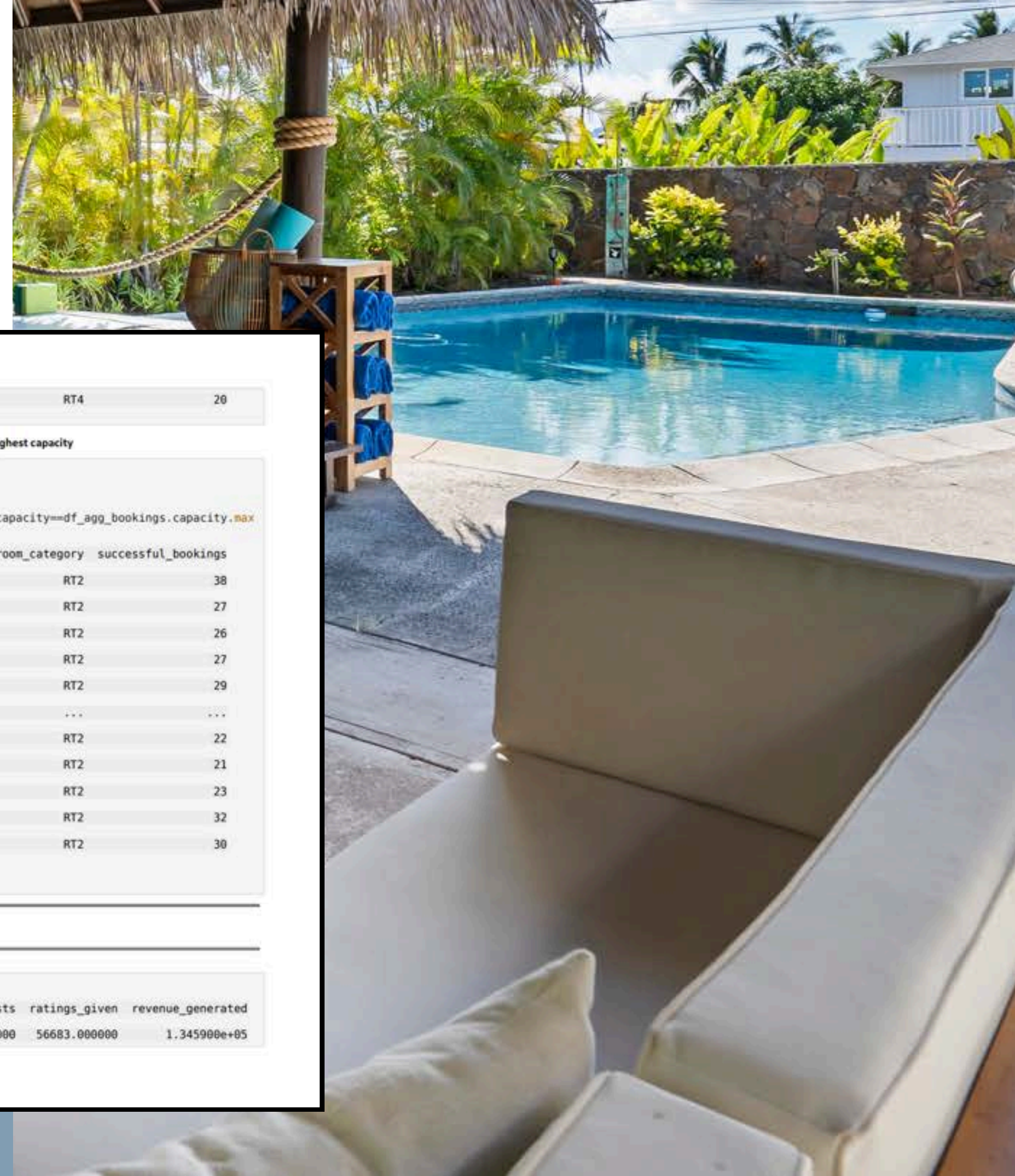


Exercise: Explore aggregate bookings ***

```
df_agg_bookings.head(3)
  property_id check_in_date room_category successful_bookings
capacity
0      16559    1-May-22         RT1             25
1      19562    1-May-22         RT1             28
2      19563    1-May-22         RT1             23
capacity
```

Exercise-1. Find out unique property ids in aggregate bookings dataset

```
df_agg_bookings.property_id.unique()
array([16559, 19562, 19563, 17558, 16558, 17560, 19558, 19560, 17561,
       16560, 16561, 16562, 16563, 17559, 17562, 17563, 18558, 18559,
       18561, 18562, 18563, 19559, 19561, 17564, 18560], dtype=int64)
```

Exercise-2. Find out total bookings per property_id

```
df_agg_bookings.groupby("property_id")["successful_bookings"].sum()
property_id
16558    3153
16559    7338
16560    4693
16561    4418
16562    4820
16563    7211
17558    5053
17559    6142
17560    6013
17561    5183
17562    3424
17563    6337
17564    3982
18558    4475
18559    5256
18560    6638
18561    6458
18562    7333
18563    4737
19558    4400
19559    4729
19560    6079
19561    5736
19562    5812
19563    5413
Name: successful_bookings, dtype: int64
```

Exercise-3. Find out days on which bookings are greater than capacity

```
df_agg_bookings[df_agg_bookings.successful_bookings>df_agg_bookings.capacity]
property_id check_in_date room_category successful_bookings capacity
3          17558      1-May-22         RT1              30      19.0
12          16563      1-May-22         RT1             100      41.0
4136        19558     11-Jun-22         RT2              50      39.0
6209        19560      2-Jul-22         RT1             123      26.0
8522        19559     25-Jul-22         RT1              35      24.0
```

```
9194    18563    31-Jul-22    RT4              20
18.0
```

Exercise-4. Find out properties that have highest capacity

```
df_agg_bookings.capacity.max()
50.0
df_agg_bookings[df_agg_bookings.capacity==df_agg_bookings.capacity.max()]
property_id check_in_date room_category successful_bookings capacity
27          17558      1-May-22         RT2              38      50.0
128          17558      2-May-22         RT2              27      50.0
229          17558      3-May-22         RT2              26      50.0
328          17558      4-May-22         RT2              27      50.0
428          17558      5-May-22         RT2              29      50.0
...          ...          ...          ...          ...
8728         17558     27-Jul-22         RT2              22      50.0
8828         17558     28-Jul-22         RT2              21      50.0
8928         17558     29-Jul-22         RT2              23      50.0
9028         17558     30-Jul-22         RT2              32      50.0
9128         17558     31-Jul-22         RT2              30      50.0
[92 rows x 5 columns]
```

==> 2. Data Cleaning

```
df_bookings.describe()
property_id no_guests ratings_given revenue_generated
count 134590.000000 134587.000000 56683.000000 1.345900e+05
```




mean	18061.113493	2.036170	3.619004	1.537805e+04
std	1093.055847	1.034885	1.235009	9.303604e+04
min	16558.000000	-17.000000	1.000000	6.500000e+03
25%	17558.000000	1.000000	3.000000	9.900000e+03
50%	17564.000000	2.000000	4.000000	1.350000e+04
75%	18563.000000	2.000000	5.000000	1.800000e+04
max	19563.000000	6.000000	5.000000	2.856000e+07
revenue_realized				
count	134590.000000			
mean	12696.123256			
std	6928.108124			
min	2600.000000			
25%	7600.000000			
50%	11700.000000			
75%	15300.000000			
max	45220.000000			

(1) Clean invalid guests

```
df_bookings[df_bookings.no_guests<=0]
```

	booking_id	property_id	booking_date	check_in_date	\
0	May012216558RT11	16558	27-04-22	1/5/2022	
3	May012216558RT14	16558	28-04-22	1/5/2022	
17924	May122218559RT44	18559	12/5/2022	12/5/2022	
18020	May122218561RT22	18561	8/5/2022	12/5/2022	
18119	May122218562RT311	18562	5/5/2022	12/5/2022	
18121	May122218562RT313	18562	10/5/2022	12/5/2022	
56715	Jun082218562RT12	18562	5/6/2022	8/6/2022	
119765	Jul202219560RT220	19560	19-07-22	20-07-22	
134586	Jul312217564RT47	17564	30-07-22	31-07-22	

	checkout_date	no_guests	room_category	booking_platform	\
ratings_given					
0	2/5/2022	-3.0	RT1	direct online	
1.0					
3	2/5/2022	-2.0	RT1	others	
NaN					
17924	14-05-22	-10.0	RT4	direct online	
NaN					
18020	14-05-22	-12.0	RT2	makeyourtrip	

NaN				
18119	17-05-22	-6.0	RT3	direct offline
5.0				
18121	17-05-22	-4.0	RT3	direct online
NaN				
56715	13-06-22	-17.0	RT1	others
NaN				
119765	22-07-22	-1.0	RT2	others
NaN				
134586	1/8/2022	-4.0	RT4	logtrip
2.0				

	booking_status	revenue_generated	revenue_realized
0	Checked Out	10010	10010
3	Cancelled	9100	3640
17924	No Show	20900	20900
18020	Cancelled	9000	3600
18119	Checked Out	16800	16800
18121	Cancelled	14400	5760
56715	Checked Out	6500	6500
119765	Checked Out	13500	13500
134586	Checked Out	38760	38760

As you can see above, number of guests having less than zero value represents data error. We can ignore these records.

```
df_bookings = df_bookings[df_bookings.no_guests>0]
```

```
df_bookings.shape
```

```
(134578, 12)
```

(2) Outlier removal in revenue generated

```
df_bookings.revenue_generated.min(),  
df_bookings.revenue_generated.max()
```

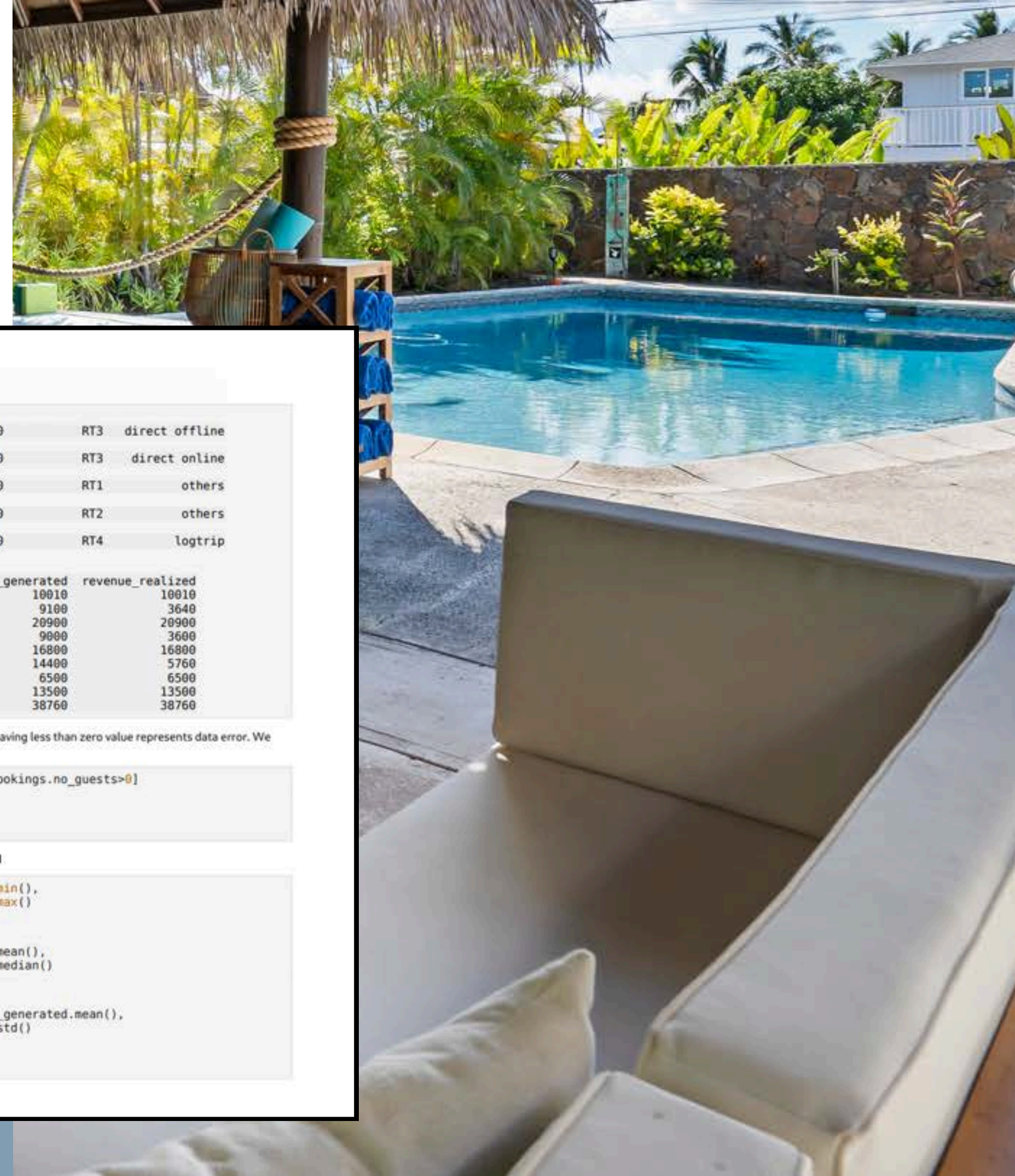
```
(6500, 28560000)
```

```
df_bookings.revenue_generated.mean(),  
df_bookings.revenue_generated.median()
```

```
(15378.036937686695, 13500.0)
```

```
avg, std = df_bookings.revenue_generated.mean(),  
df_bookings.revenue_generated.std()
```

```
higher_limit = avg + 3*std  
higher_limit
```





```
294498.50173207896
lower_limit = avg - 3*std
lower_limit
-263742.4278567056

df_bookings[df_bookings.revenue_generated<=0]

Empty DataFrame
Columns: [booking_id, property_id, booking_date, check_in_date,
checkout_date, no_guests, room_category, booking_platform,
ratings_given, booking_status, revenue_generated, revenue_realized]
Index: []

df_bookings[df_bookings.revenue_generated>higher_limit]

   booking_id  property_id booking_date check_in_date \
2    May012216558RT13      16558   28-04-22    1/5/2022
111  May012216559RT32      16559   29-04-22    1/5/2022
315  May012216562RT22      16562   28-04-22    1/5/2022
562  May012217559RT118      17559   26-04-22    1/5/2022
129176 Jul282216562RT26      16562   21-07-22   28-07-22

   checkout_date  no_guests room_category booking_platform
ratings_given \
2      4/5/2022        2.0          RT1          logtrip
5.0
111     2/5/2022        6.0          RT3    direct online
NaN
315     4/5/2022        2.0          RT2    direct offline
3.0
562     2/5/2022        2.0          RT1           others
NaN
129176   29-07-22        2.0          RT2    direct online
3.0

   booking_status  revenue_generated  revenue_realized
2      Checked Out           9100000             9100
111    Checked Out          28560000            28560
315    Checked Out          12600000            12600
562      Cancelled           2000000             4420
129176  Checked Out          10000000            12600

df_bookings = df_bookings[df_bookings.revenue_generated<=higher_limit]
df_bookings.shape
(134573, 12)

df_bookings.revenue_realized.describe()
```

```
count      134573.000000
mean       12695.983585
std        6927.791692
min         2600.000000
25%        7600.000000
50%       11700.000000
75%       15300.000000
max       45220.000000
Name: revenue_realized, dtype: float64

higher_limit = df_bookings.revenue_realized.mean() +
3*df_bookings.revenue_realized.std()
higher_limit
33479.358661845814

df_bookings[df_bookings.revenue_realized>higher_limit]

   booking_id  property_id booking_date check_in_date \
137  May012216559RT41      16559   27-04-22    1/5/2022
139  May012216559RT43      16559   1/5/2022    1/5/2022
143  May012216559RT47      16559   28-04-22    1/5/2022
149  May012216559RT413      16559   24-04-22    1/5/2022
222  May012216560RT45      16560   30-04-22    1/5/2022
...
134328 Jul312219560RT49      19560   31-07-22   31-07-22
134331 Jul312219560RT412      19560   31-07-22   31-07-22
134467 Jul312219562RT45      19562   28-07-22   31-07-22
134474 Jul312219562RT412      19562   25-07-22   31-07-22
134581 Jul312217564RT42      17564   31-07-22   31-07-22

   checkout_date  no_guests room_category booking_platform
ratings_given \
137     7/5/2022        4.0          RT4           others
NaN
139     2/5/2022        6.0          RT4          tripster
3.0
143     3/5/2022        3.0          RT4           others
5.0
149     7/5/2022        5.0          RT4          logtrip
NaN
222     3/5/2022        5.0          RT4           others
3.0
...
134328   2/8/2022        6.0          RT4    direct online
5.0
134331   1/8/2022        6.0          RT4           others
2.0
134467   1/8/2022        6.0          RT4    makeyourtrip
```





```
4.0
134474    6/8/2022    5.0    RT4    direct offline
5.0
134581    1/8/2022    4.0    RT4    makeyourtrip
4.0

   booking_status  revenue_generated  revenue_realized
137    Checked Out             38760             38760
139    Checked Out             45220             45220
143    Checked Out             35530             35530
149    Checked Out             41990             41990
222    Checked Out             34580             34580
...          ...
134328    Checked Out             39900             39900
134331    Checked Out             39900             39900
134467    Checked Out             39900             39900
134474    Checked Out             37050             37050
134581    Checked Out             38760             38760
```

[1299 rows x 12 columns]

One observation we can have in above dataframe is that all rooms are RT4 which means presidential suit. Now since RT4 is a luxurious room it is likely their rent will be higher. To make a fair analysis, we need to do data analysis only on RT4 room types

```
df_bookings[df_bookings.room_category=="RT4"].revenue_realized.describe()
```

```
count    16071.000000
mean     23439.308444
std       9048.599076
min        7600.000000
25%      19000.000000
50%      26600.000000
75%      32300.000000
max       45220.000000
Name: revenue_realized, dtype: float64
```

```
# mean + 3*standard deviation
23439+3*9048
```

50583

Here higher limit comes to be 50583 and in our dataframe above we can see that max value for revenue realized is 45220. Hence we can conclude that there is no outlier and we don't need to do any data cleaning on this particular column

```
df_bookings[df_bookings.booking_id=="May012216558RT213"]
```

```
Empty DataFrame
Columns: [booking_id, property_id, booking_date, check_in_date,
checkout_date, no_guests, room_category, booking_platform,
ratings_given, booking_status, revenue_generated, revenue_realized]
Index: []
```

```
df_bookings.isnull().sum()
```

```
booking_id      0
property_id     0
booking_date    0
check_in_date   0
checkout_date   0
no_guests       0
room_category   0
booking_platform 0
ratings_given   77897
booking_status  0
revenue_generated 0
revenue_realized 0
dtype: int64
```

Total values in our dataframe is 134576. Out of that 77899 rows has null rating. Since there are many rows with null rating, we should not filter these values. Also we should not replace this rating with a median or mean rating etc

Exercise-1. In aggregate bookings find columns that have null values. Fill these null values with whatever you think is the appropriate substitute (possible ways is to use mean or median)

```
df_agg_bookings.isnull().sum()
```

```
property_id      0
check_in_date    0
room_category    0
successful_bookings 0
capacity         2
dtype: int64
```

```
df_agg_bookings[df_agg_bookings.capacity.isna()]
```

	property_id	check_in_date	room_category	successful_bookings
capacity				
8	17561	1-May-22	RT1	22
NaN				
14	17562	1-May-22	RT1	12
NaN				

```
df_agg_bookings.capacity.median()
```




```
25.0
df_agg_bookings.capacity.fillna(df_agg_bookings.capacity.median(),
inplace=True)
df_agg_bookings.loc[[8,15]]
```

	property_id	check_in_date	room_category	successful_bookings
capacity				
8	17561	1-May-22	RT1	22
25.0				
15	17563	1-May-22	RT1	21
25.0				

Exercise-2. In aggregate bookings find out records that have successful_bookings value greater than capacity. Filter those records

```
df_agg_bookings[df_agg_bookings.successful_bookings>df_agg_bookings.ca
capacity]
```

	property_id	check_in_date	room_category	successful_bookings
capacity				
3	17558	1-May-22	RT1	30
19.0				
12	16563	1-May-22	RT1	100
41.0				
4136	19558	11-Jun-22	RT2	50
39.0				
6209	19560	2-Jul-22	RT1	123
26.0				
8522	19559	25-Jul-22	RT1	35
24.0				
9194	18563	31-Jul-22	RT4	20
18.0				

```
df_agg_bookings.shape
(9200, 5)

df_agg_bookings =
df_agg_bookings[df_agg_bookings.successful_bookings<=df_agg_bookings.c
apacity]
df_agg_bookings.shape
(9194, 5)
```

==> 3. Data Transformation

Create occupancy percentage column

```
df_agg_bookings.head(3)
```

	property_id	check_in_date	room_category	successful_bookings
capacity				
0	16559	1-May-22	RT1	25
30.0				
1	19562	1-May-22	RT1	28
30.0				
2	19563	1-May-22	RT1	23
30.0				

```
df_agg_bookings['occ_pct'] = df_agg_bookings.apply(lambda row:
row['successful_bookings']/row['capacity'], axis=1)
```

You can use following approach to get rid of SettingWithCopyWarning

```
new_col = df_agg_bookings.apply(lambda row:
row['successful_bookings']/row['capacity'], axis=1)
df_agg_bookings = df_agg_bookings.assign(occ_pct=new_col.values)
df_agg_bookings.head(3)
```

	property_id	check_in_date	room_category	successful_bookings
capacity \				
0	16559	1-May-22	RT1	25
30.0				
1	19562	1-May-22	RT1	28
30.0				
2	19563	1-May-22	RT1	23
30.0				

```
occ_pct
0 0.833333
1 0.933333
2 0.766667
```

Convert it to a percentage value

```
df_agg_bookings['occ_pct'] = df_agg_bookings['occ_pct'].apply(lambda
x: round(x*100, 2))
df_agg_bookings.head(3)
```

	property_id	check_in_date	room_category	successful_bookings
capacity \				
0	16559	1-May-22	RT1	25
30.0				



```
1      19562      1-May-22      RT1      28
30.0
2      19563      1-May-22      RT1      23
30.0

occ_pct
0      83.33
1      93.33
2      76.67

df_bookings.head()

   booking_id  property_id booking_date check_in_date
checkout_date \
1  May012216558RT12      16558      30-04-22      1/5/2022
2/5/2022
4  May012216558RT15      16558      27-04-22      1/5/2022
2/5/2022
5  May012216558RT16      16558      1/5/2022      1/5/2022
3/5/2022
6  May012216558RT17      16558      28-04-22      1/5/2022
6/5/2022
7  May012216558RT18      16558      26-04-22      1/5/2022
3/5/2022

no_guests room_category booking_platform ratings_given
booking_status \
1      2.0      RT1      others      NaN
Cancelled
4      4.0      RT1      direct online      5.0      Checked
Out
5      2.0      RT1      others      4.0      Checked
Out
6      2.0      RT1      others      NaN
Cancelled
7      2.0      RT1      logtrip      NaN      No
Show

revenue_generated revenue_realized
1      9100      3640
4      10920      10920
5      9100      9100
6      9100      3640
7      9100      9100

df_agg_bookings.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 9194 entries, 0 to 9199
Data columns (total 6 columns):
```

#	Column	Non-Null Count	Dtype
0	property_id	9194 non-null	int64
1	check_in_date	9194 non-null	object
2	room_category	9194 non-null	object
3	successful_bookings	9194 non-null	int64
4	capacity	9194 non-null	float64
5	occ_pct	9194 non-null	float64

dtypes: float64(2), int64(2), object(2)
memory usage: 502.8+ KB

There are various types of data transformations that you may have to perform based on the need. Few examples of data transformations are,

1. Creating new columns
2. Normalization
3. Merging data
4. Aggregation

==> 4. Insights Generation

1. What is an average occupancy rate in each of the room categories?

```
df_agg_bookings.head(3)

   property_id check_in_date room_category  successful_bookings
capacity \
0      16559      1-May-22      RT1      25
30.0
1      19562      1-May-22      RT1      28
30.0
2      19563      1-May-22      RT1      23
30.0

occ_pct
0      83.33
1      93.33
2      76.67

df_agg_bookings.groupby("room_category")["occ_pct"].mean()

room_category
RT1      57.889643
RT2      58.009756
RT3      58.020213
RT4      59.277925
Name: occ_pct, dtype: float64
```




I don't understand RT1, RT2 etc. Print room categories such as Standard, Premium, Elite etc along with average occupancy percentage

```
df = pd.merge(df_agg_bookings, df_rooms, left_on="room_category",
              right_on="room_id")
df.head(4)
```

	property_id	check_in_date	room_category	successful_bookings
capacity \	16559	1-May-22	RT1	25
0				
30.0	19562	1-May-22	RT1	28
1				
30.0	19563	1-May-22	RT1	23
2				
30.0	16558	1-May-22	RT1	18
3				
19.0				

	occ_pct	room_id	room_class
0	83.33	RT1	Standard
1	93.33	RT1	Standard
2	76.67	RT1	Standard
3	94.74	RT1	Standard

```
df.drop("room_id",axis=1, inplace=True)
df.head(4)
```

	property_id	check_in_date	room_category	successful_bookings
capacity \	16559	1-May-22	RT1	25
0				
30.0	19562	1-May-22	RT1	28
1				
30.0	19563	1-May-22	RT1	23
2				
30.0	16558	1-May-22	RT1	18
3				
19.0				

	occ_pct	room_class
0	83.33	Standard
1	93.33	Standard
2	76.67	Standard
3	94.74	Standard

```
df.groupby("room_class")["occ_pct"].mean()
```

room_class	
Elite	58.009756
Premium	58.028213
Presidential	59.277925

```
Standard      57.889643
Name: occ_pct, dtype: float64

df[df.room_class=="Standard"].occ_pct.mean()
57.88964285714285
```

2. Print average occupancy rate per city

```
df_hotels.head(3)
```

	property_id	property_name	category	city
0	16558	Atliq Grands	Luxury	Delhi
1	16559	Atliq Exotica	Luxury	Mumbai
2	16560	Atliq City	Business	Delhi

```
df = pd.merge(df, df_hotels, on="property_id")
df.head(3)
```

	property_id	check_in_date	room_category	successful_bookings
capacity \	16559	1-May-22	RT1	25
0				
30.0	16559	2-May-22	RT1	20
1				
30.0	16559	3-May-22	RT1	17
2				
30.0				

	occ_pct	room_class	property_name	category	city
0	83.33	Standard	Atliq Exotica	Luxury	Mumbai
1	66.67	Standard	Atliq Exotica	Luxury	Mumbai
2	56.67	Standard	Atliq Exotica	Luxury	Mumbai

```
df.groupby("city")["occ_pct"].mean()
```

city	
Bangalore	56.332376
Delhi	61.507341
Hyderabad	58.120652
Mumbai	57.909181

Name: occ_pct, dtype: float64

3. When was the occupancy better? Weekday or Weekend?

```
df_date.head(3)
```

	date	mm	yy	week no	day_type
0	01-May-22	May	22	W 19	weekend
1	02-May-22	May	22	W 19	weekeday
2	03-May-22	May	22	W 19	weekeday



```
df = pd.merge(df, df_date, left_on="check_in_date", right_on="date")
df.head(3)
```

	property_id	check_in_date	room_category	successful_bookings
capacity \				
0	16559	10-May-22	RT1	18
30.0				
1	16559	10-May-22	RT2	25
41.0				
2	16559	10-May-22	RT3	20
32.0				

	occ_pct	room_class	property_name	category	city	date	mm
yy \							
0	60.00	Standard	Atliq Exotica	Luxury	Mumbai	10-May-22	May
22							
1	60.98	Elite	Atliq Exotica	Luxury	Mumbai	10-May-22	May
22							
2	62.50	Premium	Atliq Exotica	Luxury	Mumbai	10-May-22	May
22							

	week no	day_type
0	W 20	weekeday
1	W 20	weekeday
2	W 20	weekeday

```
df.groupby("day_type")["occ_pct"].mean().round(2)
```

```
day_type
weekeday    50.88
weekend     72.34
Name: occ_pct, dtype: float64
```

4: In the month of June, what is the occupancy for different cities

```
df_june_22 = df[df["mm yy"]=="Jun 22"]
df_june_22.head(4)
```

	property_id	check_in_date	room_category	successful_bookings
capacity \				
2200	16559	10-Jun-22	RT1	20
30.0				
2201	16559	10-Jun-22	RT2	26
41.0				
2202	16559	10-Jun-22	RT3	20
32.0				
2203	16559	10-Jun-22	RT4	11
18.0				

	occ_pct	room_class	property_name	category	city	date
--	---------	------------	---------------	----------	------	------

\							
2200	66.67	Standard	Atliq Exotica	Luxury	Mumbai	10-Jun-22	
2201	63.41	Elite	Atliq Exotica	Luxury	Mumbai	10-Jun-22	
2202	62.50	Premium	Atliq Exotica	Luxury	Mumbai	10-Jun-22	
2203	61.11	Presidential	Atliq Exotica	Luxury	Mumbai	10-Jun-22	

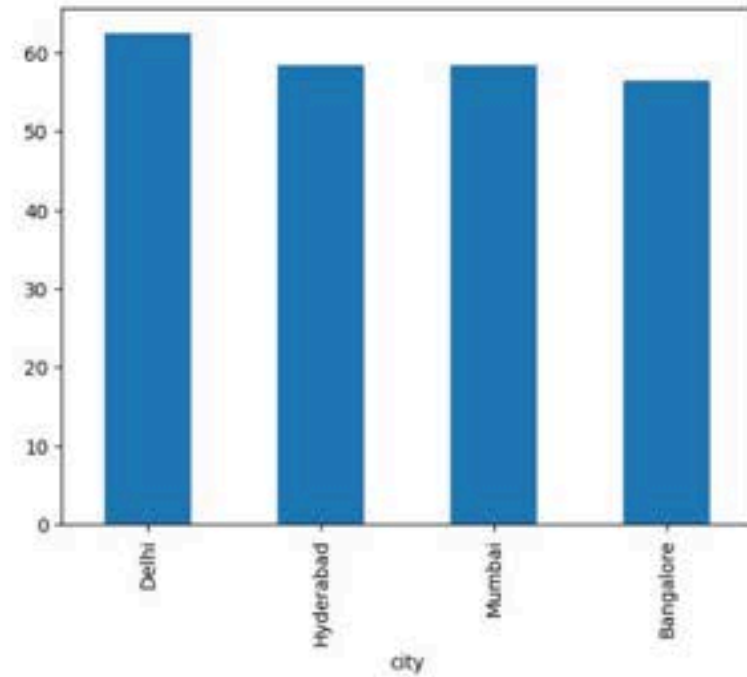
	mm	yy	week no	day_type
2200	Jun	22	W 24	weekeday
2201	Jun	22	W 24	weekeday
2202	Jun	22	W 24	weekeday
2203	Jun	22	W 24	weekeday

```
df_june_22.groupby('city')
['occ_pct'].mean().round(2).sort_values(ascending=False)
```

```
city
Delhi      62.47
Hyderabad  58.46
Mumbai     58.38
Bangalore  56.44
Name: occ_pct, dtype: float64
```

```
df_june_22.groupby('city')
['occ_pct'].mean().round(2).sort_values(ascending=False).plot(kind="bar")
```

```
<AxesSubplot: xlabel='city'>
```

5: We got new data for the month of august. Append that to existing data

```
df_august = pd.read_csv("datasets/new_data_august.csv")
df_august.head(3)
```

	property_id	property_name	category	city	room_category
room_class \					
0	16559	Atliq Exotica	Luxury	Mumbai	RT1
Standard					
1	19562	Atliq Bay	Luxury	Bangalore	RT1
Standard					
2	19563	Atliq Palace	Business	Bangalore	RT1
Standard					

	check_in_date	mmm yy	week no	day_type	successful_bookings
capacity \					
0	01-Aug-22	Aug-22	W 32	weekeday	30
30					
1	01-Aug-22	Aug-22	W 32	weekeday	21
30					
2	01-Aug-22	Aug-22	W 32	weekeday	23

```
30
    occ%
0  100.00
1   70.00
2   76.67

df_august.columns
Index(['property_id', 'property_name', 'category', 'city',
      'room_category',
      'room_class', 'check_in_date', 'mmm yy', 'week no', 'day_type',
      'successful_bookings', 'capacity', 'occ%'],
      dtype='object')

df.columns
Index(['property_id', 'check_in_date', 'room_category',
      'successful_bookings',
      'capacity', 'occ_pct', 'room_class', 'property_name',
      'category',
      'city', 'date', 'mmm yy', 'week no', 'day_type'],
      dtype='object')

df_august.shape
(7, 13)

df.shape
(6497, 14)

latest_df = pd.concat([df, df_august], ignore_index = True, axis = 0)
latest_df.tail(10)
```

	property_id	check_in_date	room_category	successful_bookings
capacity \				
6494	16563	31-Jul-22	RT2	32
38.0				
6495	16563	31-Jul-22	RT3	14
20.0				
6496	16563	31-Jul-22	RT4	13
18.0				
6497	16559	01-Aug-22	RT1	30
30.0				
6498	19562	01-Aug-22	RT1	21
30.0				
6499	19563	01-Aug-22	RT1	23
30.0				
6500	19558	01-Aug-22	RT1	30
40.0				
6501	19560	01-Aug-22	RT1	20



```
26.0
6502      17561    01-Aug-22      RT1         18
26.0
6503      17564    01-Aug-22      RT1         10
16.0
```

	occ_pct	room_class	property_name	category	city
date \					
6494 Jul-22	84.21	Elite	Atliq Palace	Business	Delhi
6495 Jul-22	70.00	Premium	Atliq Palace	Business	Delhi
6496 Jul-22	72.22	Presidential	Atliq Palace	Business	Delhi
6497 Jul-22	NaN	Standard	Atliq Exotica	Luxury	Mumbai
6498 Jul-22	NaN	Standard	Atliq Bay	Luxury	Bangalore
6499 Jul-22	NaN	Standard	Atliq Palace	Business	Bangalore
6500 Jul-22	NaN	Standard	Atliq Grands	Luxury	Bangalore
6501 Jul-22	NaN	Standard	Atliq City	Business	Bangalore
6502 Jul-22	NaN	Standard	Atliq Blu	Luxury	Mumbai
6503 Jul-22	NaN	Standard	Atliq Seasons	Business	Mumbai

	mm	yy	week	no	day_type	occ%
6494 Jul-22	Jul	22	W	32	weekend	NaN
6495 Jul-22	Jul	22	W	32	weekend	NaN
6496 Jul-22	Jul	22	W	32	weekend	NaN
6497 Aug-22	Aug	22	W	32	weekeday	100.00
6498 Aug-22	Aug	22	W	32	weekeday	70.00
6499 Aug-22	Aug	22	W	32	weekeday	76.67
6500 Aug-22	Aug	22	W	32	weekeday	75.00
6501 Aug-22	Aug	22	W	32	weekeday	76.92
6502 Aug-22	Aug	22	W	32	weekeday	69.23
6503 Aug-22	Aug	22	W	32	weekeday	62.50

```
latest_df.shape
(6504, 15)
```

6. Print revenue realized per city

```
df_bookings.head()
```

	booking_id	property_id	booking_date	check_in_date
checkout_date \				
1 May012216558RT12	16558	30-04-22	1/5/2022	
2/5/2022				
4 May012216558RT15	16558	27-04-22	1/5/2022	
2/5/2022				
5 May012216558RT16	16558	1/5/2022	1/5/2022	
3/5/2022				
6 May012216558RT17	16558	28-04-22	1/5/2022	
6/5/2022				
7 May012216558RT18	16558	26-04-22	1/5/2022	
3/5/2022				

	no_guests	room_category	booking_platform	ratings_given
booking_status \				
1 Cancelled	2.0	RT1	others	NaN
4 Out	4.0	RT1	direct online	5.0
5 Out	2.0	RT1	others	4.0
6 Cancelled	2.0	RT1	others	NaN
7 Show	2.0	RT1	logtrip	NaN

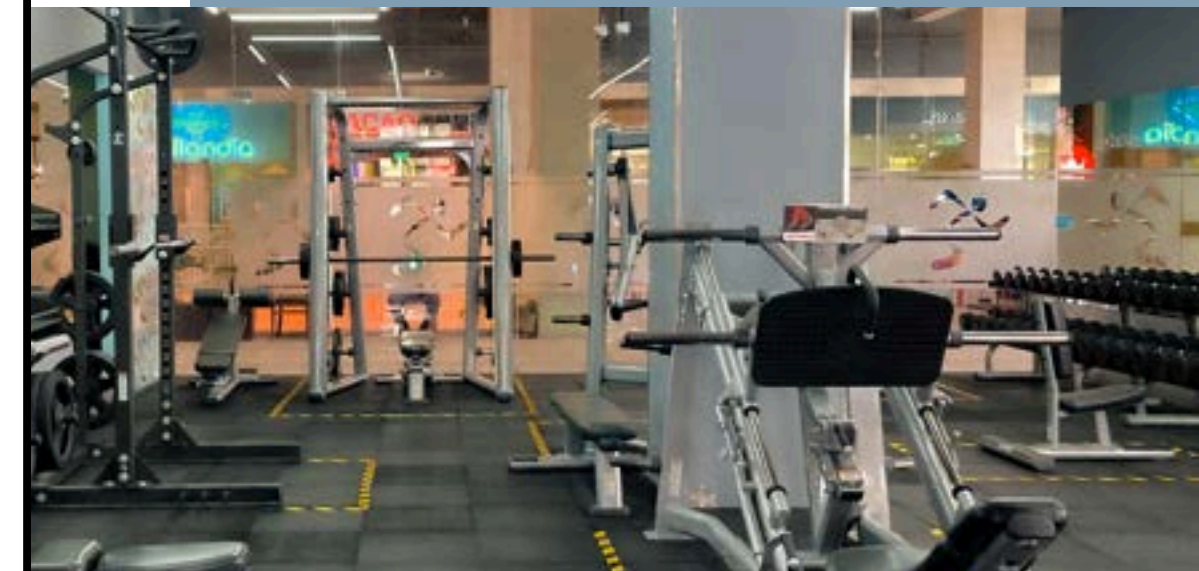
	revenue_generated	revenue_realized
1	9100	3640
4	10920	10920
5	9100	9100
6	9100	3640
7	9100	9100

```
df_hotels.head(3)
```

	property_id	property_name	category	city
0	16558	Atliq Grands	Luxury	Delhi
1	16559	Atliq Exotica	Luxury	Mumbai
2	16560	Atliq City	Business	Delhi

```
df_bookings_all = pd.merge(df_bookings, df_hotels, on="property_id")
df_bookings_all.head(3)
```

	booking_id	property_id	booking_date	check_in_date
checkout_date \				
0 May012216558RT12	16558	30-04-22	1/5/2022	
2/5/2022				





```
1 May012216558RT15 16558 27-04-22 1/5/2022
2/5/2022
2 May012216558RT16 16558 1/5/2022 1/5/2022
3/5/2022
```

no_guests	room_category	booking_platform	ratings_given
2.0	RT1	others	NaN
Cancelled			
1	4.0	RT1 direct online	5.0
Out			
2	2.0	RT1 others	4.0
Out			

revenue_generated	revenue_realized	property_name	category	city
0	9100	3640 Atliq Grands	Luxury	Delhi
1	10920	10920 Atliq Grands	Luxury	Delhi
2	9100	9100 Atliq Grands	Luxury	Delhi

```
df_bookings_all.groupby("city")["revenue_realized"].sum()

city
Bangalore 420383550
Delhi      294404488
Hyderabad 325179310
Mumbai     668569251
Name: revenue_realized, dtype: int64
```

7. Print month by month revenue

```
df_date.head(3)
```

	date	mmm	yy	week no	day_type
0	01-May-22	May	22	W 19	weekend
1	02-May-22	May	22	W 19	weekeday
2	03-May-22	May	22	W 19	weekeday

```
df_date["mmm yy"].unique()

array(['May 22', 'Jun 22', 'Jul 22'], dtype=object)
```

```
df_bookings_all.head(3)
```

booking_id	property_id	booking_date	check_in_date
0	May012216558RT12	16558	30-04-22 1/5/2022
1	May012216558RT15	16558	27-04-22 1/5/2022
2	May012216558RT16	16558	1/5/2022 1/5/2022

```
no_guests room_category booking_platform ratings_given
booking_status \
0 2.0 RT1 others NaN
Cancelled
1 4.0 RT1 direct online 5.0 Checked
Out
2 2.0 RT1 others 4.0 Checked
Out
```

revenue_generated	revenue_realized	property_name	category	city
0	9100	3640 Atliq Grands	Luxury	Delhi
1	10920	10920 Atliq Grands	Luxury	Delhi
2	9100	9100 Atliq Grands	Luxury	Delhi

```
df_date.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 92 entries, 0 to 91
Data columns (total 4 columns):
# Column Non-Null Count Dtype
---
0 date 92 non-null object
1 mmm yy 92 non-null object
2 week no 92 non-null object
3 day_type 92 non-null object
dtypes: object(4)
memory usage: 3.0+ KB

df_date["date"] = pd.to_datetime(df_date["date"])
df_date.head(3)
```

	date	mmm	yy	week no	day_type
0	2022-05-01	May	22	W 19	weekend
1	2022-05-02	May	22	W 19	weekeday
2	2022-05-03	May	22	W 19	weekeday

```
df_bookings_all.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 134573 entries, 0 to 134572
Data columns (total 15 columns):
# Column Non-Null Count Dtype
---
0 booking_id 134573 non-null object
1 property_id 134573 non-null int64
2 booking_date 134573 non-null object
3 check_in_date 134573 non-null object
4 checkout_date 134573 non-null object
5 no_guests 134573 non-null float64
6 room_category 134573 non-null object
```





```
7 booking_platform 134573 non-null object
8 ratings_given 56676 non-null float64
9 booking_status 134573 non-null object
10 revenue_generated 134573 non-null int64
11 revenue_realized 134573 non-null int64
12 property_name 134573 non-null object
13 category 134573 non-null object
14 city 134573 non-null object
dtypes: float64(2), int64(3), object(10)
memory usage: 16.4+ MB

df_bookings_all["check_in_date"] =
pd.to_datetime(df_bookings_all["check_in_date"])
df_bookings_all.head(4)
```

	booking_id	property_id	booking_date	check_in_date	checkout_date
0	May012216558RT12	16558	30-04-22	2022-01-05	2/5/2022
1	May012216558RT15	16558	27-04-22	2022-01-05	2/5/2022
2	May012216558RT16	16558	1/5/2022	2022-01-05	3/5/2022
3	May012216558RT17	16558	28-04-22	2022-01-05	6/5/2022

	no_guests	room_category	booking_platform	ratings_given
0	2.0	RT1	others	NaN
Cancelled				
1	4.0	RT1	direct online	5.0
Out				
2	2.0	RT1	others	4.0
Out				
3	2.0	RT1	others	NaN
Cancelled				

	revenue_generated	revenue_realized	property_name	category	city
0	9100	3640	Atliq Grands	Luxury	Delhi
1	10920	10920	Atliq Grands	Luxury	Delhi
2	9100	9100	Atliq Grands	Luxury	Delhi
3	9100	3640	Atliq Grands	Luxury	Delhi

```
df_bookings_all = pd.merge(df_bookings_all, df_date,
left_on="check_in_date", right_on="date")
df_bookings_all.head(3)
```

	booking_id	property_id	booking_date	check_in_date	checkout_date
0	May052216558RT11	16558	15-04-22	2022-05-05	

```
7/5/2022
1 May052216558RT12 16558 30-04-22 2022-05-05
7/5/2022
2 May052216558RT13 16558 1/5/2022 2022-05-05
6/5/2022
```

	no_guests	room_category	booking_platform	ratings_given
0	3.0	RT1	tripster	5.0
Out				
1	2.0	RT1	others	NaN
Cancelled				
2	3.0	RT1	direct offline	5.0
Out				

	revenue_generated	revenue_realized	property_name	category	city
0	10010	10010	Atliq Grands	Luxury	Delhi
1	9100	3640	Atliq Grands	Luxury	Delhi
2	10010	10010	Atliq Grands	Luxury	Delhi

	date	mmm	yy	week no	day_type
0	2022-05-05	May	22	W 19	weekeday
1	2022-05-05	May	22	W 19	weekeday
2	2022-05-05	May	22	W 19	weekeday

```
df_bookings_all.groupby("mmm yy")["revenue_realized"].sum()
```

mmm	yy	revenue_realized
Jul	22	389940912
Jun	22	377191229
May	22	408375641

Name: revenue_realized, dtype: int64

Exercise-1. Print revenue realized per hotel type

```
df_bookings_all.property_name.unique()
array(['Atliq Grands', 'Atliq Exotica', 'Atliq City', 'Atliq Blu',
      'Atliq Bay', 'Atliq Palace', 'Atliq Seasons'], dtype=object)

df_bookings_all.groupby("property_name")
["revenue_realized"].sum().round(2).sort_values()
```

property_name	revenue_realized
Atliq Seasons	66086735
Atliq Grands	211462134
Atliq Bay	259996918





```
Atliq Blu      260851922
Atliq City    285798439
Atliq Palace  304081863
Atliq Exotica 320258588
Name: revenue_realized, dtype: int64
```

Exercise-2 Print average rating per city

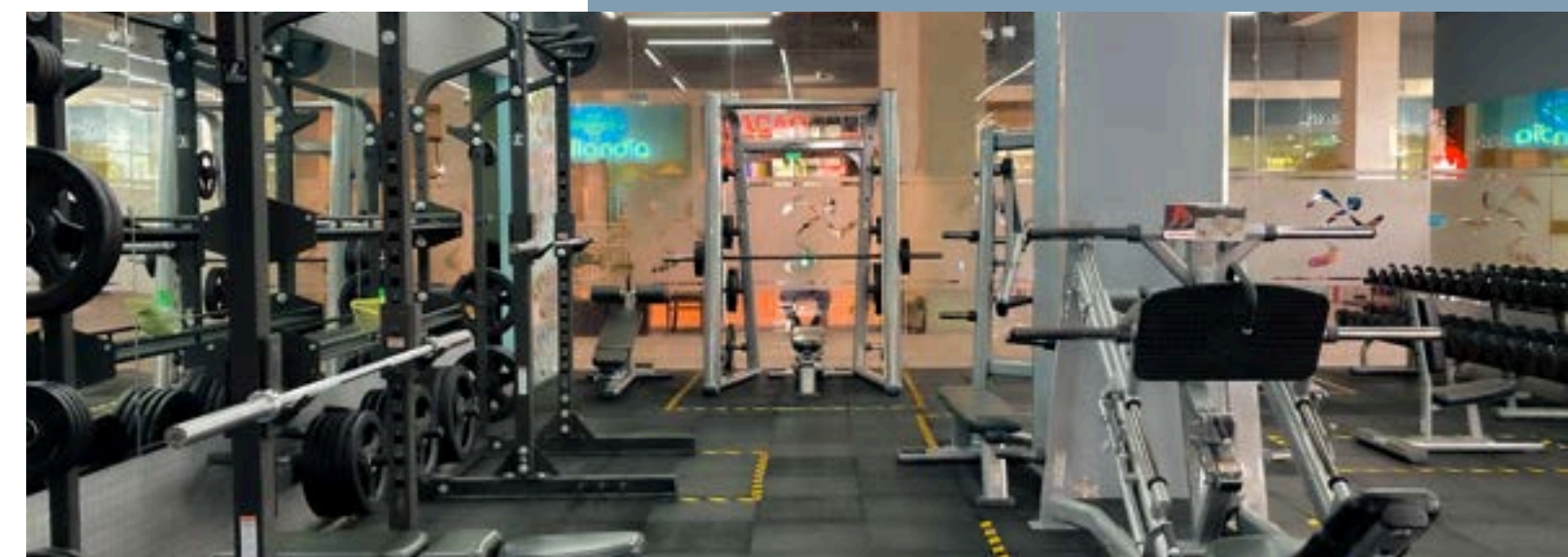
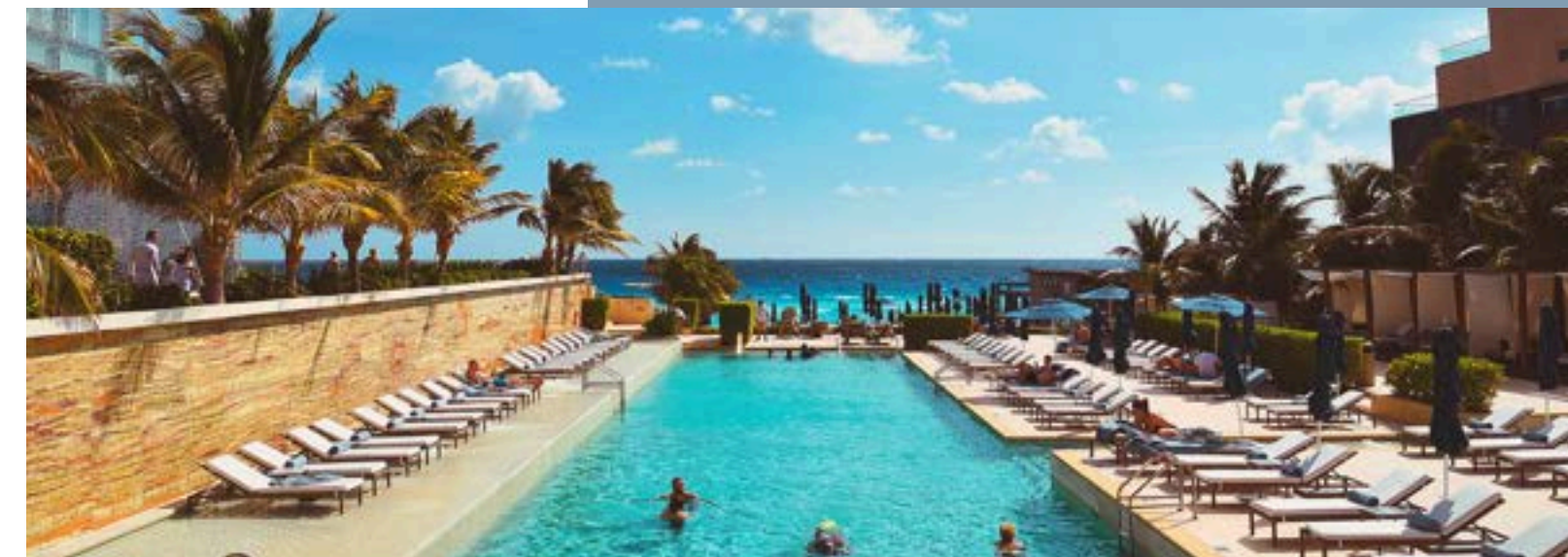
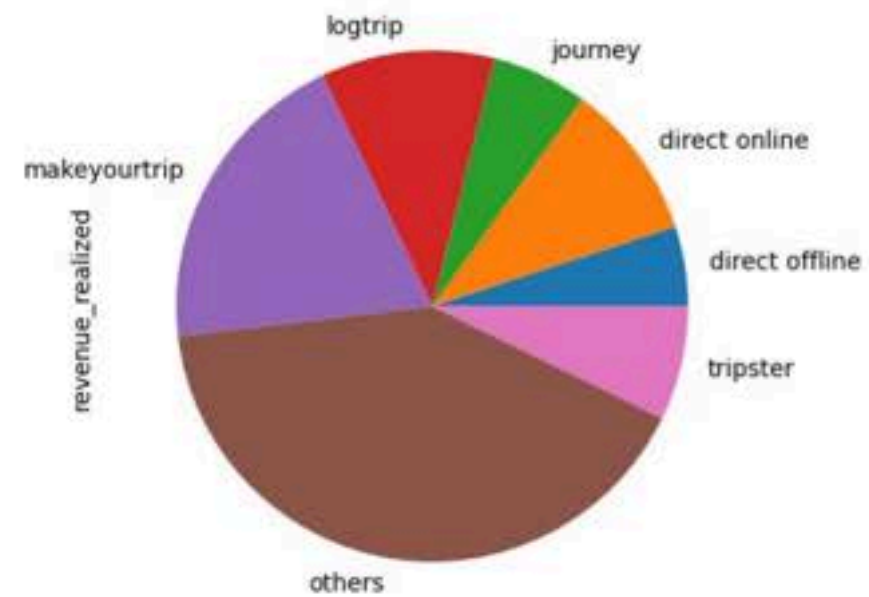
```
df_bookings_all.groupby("city")["ratings_given"].mean().round(2)

city
Bangalore    3.41
Delhi        3.78
Hyderabad    3.66
Mumbai       3.65
Name: ratings_given, dtype: float64
```

Exercise-3 Print a pie chart of revenue realized per booking platform

```
df_bookings_all.groupby("booking_platform")
["revenue_realized"].sum().plot(kind="pie")

<AxesSubplot: ylabel='revenue_realized'>
```





AtliQ Grands

Thank you

