

+ Code + Text

✓ RAW
Disk

◆ Gemini



```
[5] import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
16 #Importing Files in a file
```

```
import pandas as pd

file_paths = ["/content/dim_date.csv", "/content/dim_hotels.csv", "/content/dim_rooms.csv", "/content/fact_bookings.csv", "/content/fact_aggregat

dataframes = []
for file_path in file_paths:
    if file_path.endswith('.csv'):
        df = pd.read_csv(file_path)
    elif file_path.endswith('.xlsx'):
        df = pd.read_excel(file_path)
    # Add other file format handling as needed
    else:
        print(f"Unsupported file format: {file_path}")
        continue
    dataframes.append(df)

# Now you have a list of dataframes, you can process them as needed
for df in dataframes:
    print(df.head())
```



	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
3	04-May-22	May	22	W	19	weekeday
4	05-May-22	May	22	W	19	weekeday

+ Code + Text

✓

RAM
Disk

↑

↓

↻

💬

⚙️

📄

🗑️

⋮

Gemini

^

```
0 05 May 22 05 May 22 W 15 weekday
property_id property_name category city
0 16558 Atliq Grands Luxury Delhi
1 16559 Atliq Exotica Luxury Mumbai
2 16560 Atliq City Business Delhi
3 16561 Atliq Blu Luxury Delhi
4 16562 Atliq Bay Luxury Delhi

room_id room_class
0 RT1 Standard
1 RT2 Elite
2 RT3 Premium
3 RT4 Presidential

booking_id property_id booking_date check_in_date checkout_date \
0 May012216558RT11 16558 2022-04-27 2022-05-01 2022-05-02
1 May012216558RT12 16558 2022-04-30 2022-05-01 2022-05-02
2 May012216558RT13 16558 2022-04-28 2022-05-01 2022-05-04
3 May012216558RT14 16558 2022-04-28 2022-05-01 2022-05-02
4 May012216558RT15 16558 2022-04-27 2022-05-01 2022-05-02

no_guests room_category booking_platform ratings_given booking_status \
0 3 RT1 direct online 1.0 Checked Out
1 2 RT1 others NaN Cancelled
2 2 RT1 logtrip 5.0 Checked Out
3 2 RT1 others NaN Cancelled
4 4 RT1 direct online 5.0 Checked Out

revenue_generated revenue_realized
0 10010 10010
1 9100 3640
2 9100 9100
3 9100 3640
4 10920 10920

property_id check_in_date room_category successful_bookings capacity
0 16559 01-May-22 RT1 25 30
1 19562 01-May-22 RT1 28 30
2 19563 01-May-22 RT1 23 30
3 17558 01-May-22 RT1 13 19
4 16558 01-May-22 RT1 18 19
```

+ Code + Text

✓ RAM
Disk

+

Gemini

^

```
0s ▶ filepath1='/content/dim_date.csv'  
filepath2='/content/dim_hotels.csv'  
filepath3='/content/dim_rooms.csv'  
filepath4='/content/fact_aggregated_bookings.csv'  
filepath5='/content/fact_bookings.csv'
```

```
0s ▶ dim_date=pd.read_csv(filepath1)  
dim_hotels=pd.read_csv(filepath2)  
dim_rooms=pd.read_csv(filepath3)  
fact_aggregate=pd.read_csv(filepath4)  
fact_bookings=pd.read_csv(filepath5)
```

```
0s [12] dim_date.head()
```



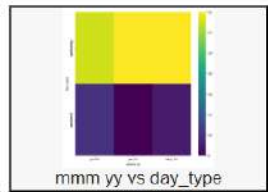
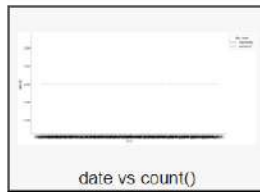
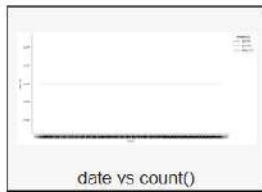
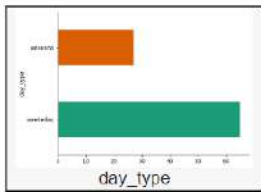
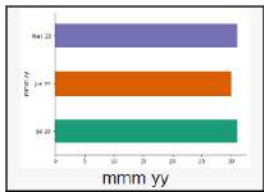
1 to 5 of 5 entries [Filter](#)

index	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
3	04-May-22	May 22	W 19	weekeday
4	05-May-22	May 22	W 19	weekeday

+ Code + Text

✓ RAM
Disk

◆ Gemini



✓ [13] dim_date=dim_date[dim_date['week no']!= 'W 32']
dim_date



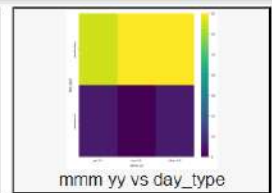
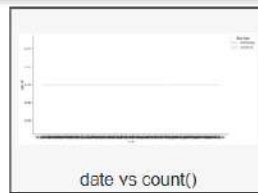
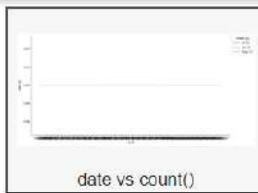
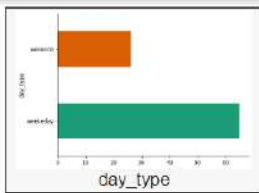
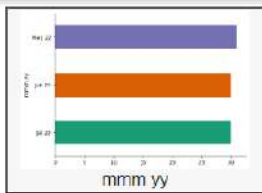
1 to 25 of 91 entries Filter ?

index	date	mmm yy	week no	day_type
0	01-May-22	May 22	W 19	weekend
1	02-May-22	May 22	W 19	weekday
2	03-May-22	May 22	W 19	weekday
3	04-May-22	May 22	W 19	weekday
4	05-May-22	May 22	W 19	weekday
5	06-May-22	May 22	W 19	weekday
6	07-May-22	May 22	W 19	weekend
7	08-May-22	May 22	W 20	weekend
8	09-May-22	May 22	W 20	weekday
9	10-May-22	May 22	W 20	weekday
10	11-May-22	May 22	W 20	weekday
11	12-May-22	May 22	W 20	weekday
12	13-May-22	May 22	W 20	weekday
13	14-May-22	May 22	W 20	weekend
14	15-May-22	May 22	W 21	weekend
15	16-May-22	May 22	W 21	weekday
16	17-May-22	May 22	W 21	weekday
17	18-May-22	May 22	W 21	weekday

+ Code + Text

✓ RAM
Disk

◆ Gemini



✓ [14] dim_date.info()



```
<class 'pandas.core.frame.DataFrame'>
Index: 91 entries, 0 to 90
Data columns (total 4 columns):
#   Column    Non-Null Count  Dtype
---  -
0    date      91 non-null     object
1    mmm yy    91 non-null     object
2    week no   91 non-null     object
3    day_type  91 non-null     object
dtypes: object(4)
memory usage: 5.6+ KB
```

✓ [15] dim_hotels.head()

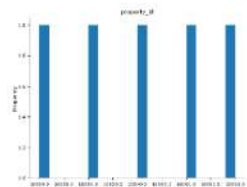


	property_id	property_name	category	city
0	16558	Atliq Grands	Luxury	Delhi
1	16559	Atliq Exotica	Luxury	Mumbai
2	16560	Atliq City	Business	Delhi
3	16561	Atliq Blu	Luxury	Delhi
4	16562	Atliq Bay	Luxury	Delhi

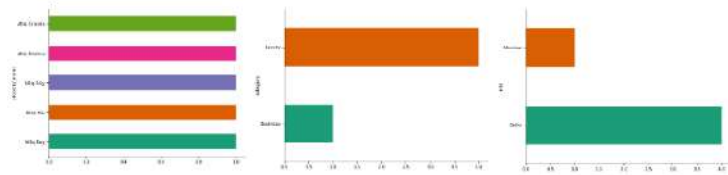
✓ [15]



Distributions



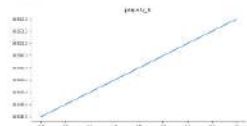
Categorical distributions



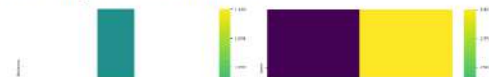
Time series



Values



2-d categorical distributions



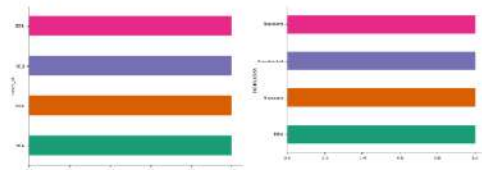
✓ [16] dim_rooms.head()



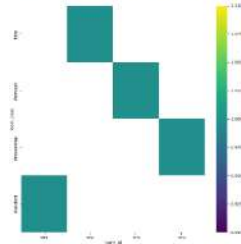
	room_id	room_class
0	RT1	Standard
1	RT2	Elite
2	RT3	Premium
3	RT4	Presidential



Categorical distributions



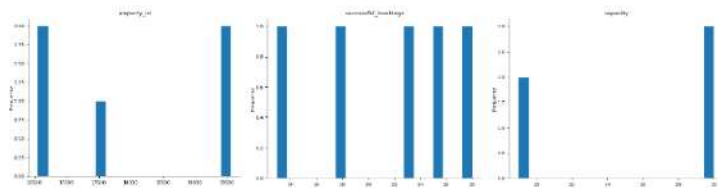
2-d categorical distributions



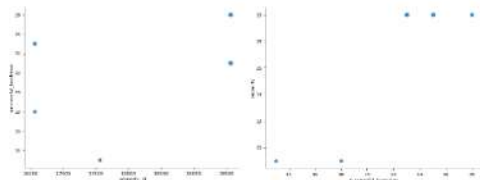
[17] fact_aggregate.head()

	property_id	check_in_date	room_category	successful_bookings	capacity
0	16559	01-May-22	RT1	25	30
1	19562	01-May-22	RT1	28	30
2	19563	01-May-22	RT1	23	30
3	17558	01-May-22	RT1	13	19
4	16558	01-May-22	RT1	18	19

Distributions



2-d distributions



Time series



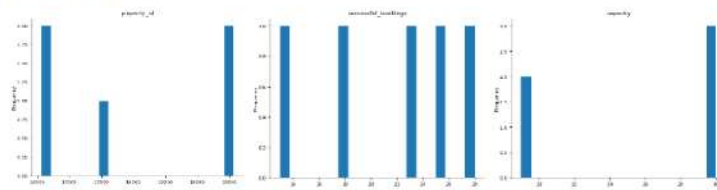
+ Code + Text

✓ RAM
Disk

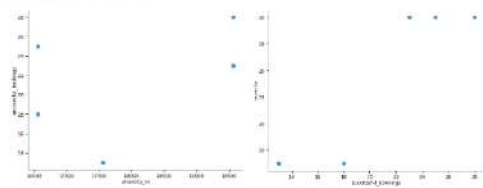
◆ Gemini

✓ 0s [17]	4	16558	01-May-22	RT1	18	19
-----------	---	-------	-----------	-----	----	----

Distributions



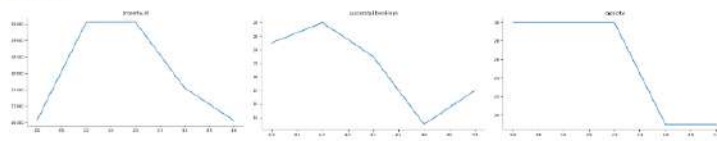
2-d distributions



Time series





Values




+ Code + Text

✓ RAM ☐ Disk ☐ Gemini ^

✓ 0s  fact_aggregate.info()

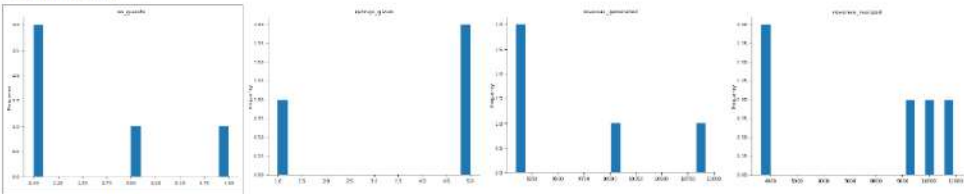
 <class 'pandas.core.frame.DataFrame'>
RangeIndex: 9200 entries, 0 to 9199
Data columns (total 5 columns):
column Non-Null Count Dtype
--- ---
0 property_id 9200 non-null int64
1 check_in_date 9200 non-null object
2 room_category 9200 non-null object
3 successful_bookings 9200 non-null int64
4 capacity 9200 non-null int64
dtypes: int64(3), object(2)
memory usage: 359.5+ KB

✓ 0s [19] fact_bookings.head()

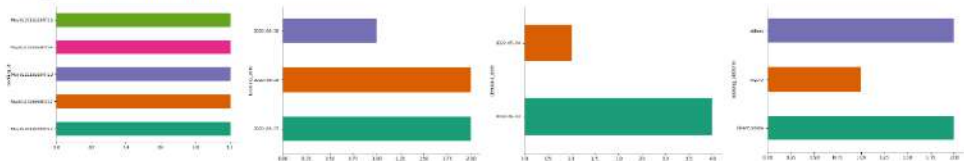


	booking_id	property_id	booking_date	check_in_date	checkout_date	no_guests	room_category	booking_platform	ratings_given	booking
0	May012216558RT11	16558	2022-04-27	2022-05-01	2022-05-02	3	RT1	direct online	1.0	Chec
1	May012216558RT12	16558	2022-04-30	2022-05-01	2022-05-02	2	RT1	others	NaN	C
2	May012216558RT13	16558	2022-04-28	2022-05-01	2022-05-04	2	RT1	logtrip	5.0	Chec
3	May012216558RT14	16558	2022-04-28	2022-05-01	2022-05-02	2	RT1	others	NaN	C
4	May012216558RT15	16558	2022-04-27	2022-05-01	2022-05-02	4	RT1	direct online	5.0	Chec

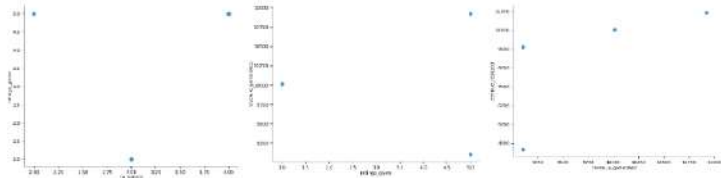
Distributions



Categorical distributions



2-d distributions



Time series



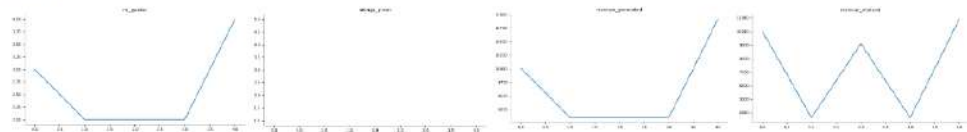
Values



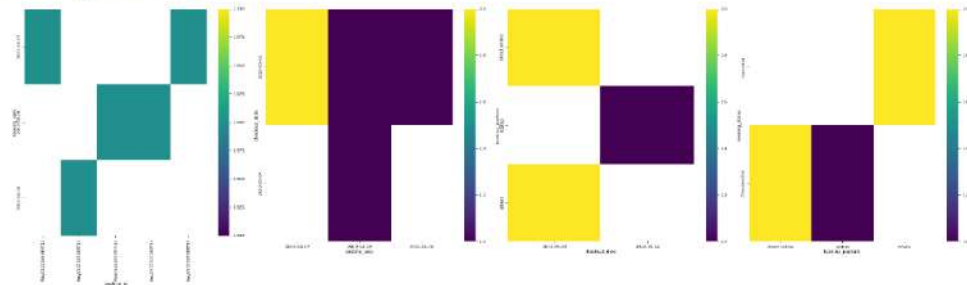
Time series



Values



2-d categorical distributions



0s fact_bookings.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 134590 entries, 0 to 134589
Data columns (total 12 columns):
Column Non-Null Count Dtype

0 booking_id 134590 non-null object
1 property_id 134590 non-null int64
2 booking_date 134590 non-null object
3 check_in_date 134590 non-null object
4 checkout_date 134590 non-null object
5 no_guests 134590 non-null int64
6 room_category 134590 non-null object
7 booking_platform 134590 non-null object
8 ratings_given 56683 non-null float64
9 booking_status 134590 non-null object
10 revenue_generated 134590 non-null int64
11 revenue_realized 134590 non-null int64
dtypes: float64(1), int64(4), object(7)
memory usage: 12.3+ MB

0s [21] fact_bookings.describe()

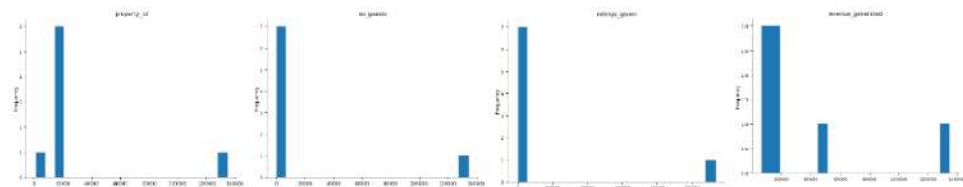
	property_id	no_guests	ratings_given	revenue_generated	revenue_realized
count	134590.000000	134590.000000	56683.000000	134590.000000	134590.000000
mean	18061.113493	2.036808	3.619004	14916.013188	12696.123256
std	1093.055847	1.031766	1.235009	6452.868072	6928.108124
min	16558.000000	1.000000	1.000000	6500.000000	2600.000000
25%	17558.000000	1.000000	3.000000	9900.000000	7600.000000

+ Code + Text

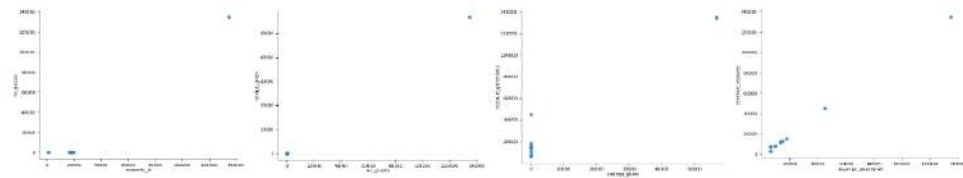
✓ RAM
Disk Gemini ^

50%	17564.000000	2.000000	4.000000	13500.000000	11700.000000
75%	18563.000000	2.000000	5.000000	18000.000000	15300.000000
max	19563.000000	6.000000	5.000000	45220.000000	45220.000000

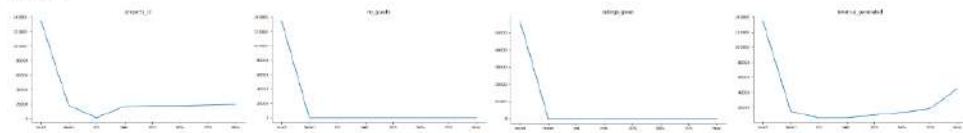
Distributions



2-d distributions



Values



✓ [22] fact_bookings.isnull().sum()

```
08 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    77907
    booking_status    0
    revenue_generated 0
    revenue_realized 0
    dtype: int64
```

```
08 [26] revenue_df=pd.merge(dim_hotels,fact_bookings,how='left',on='property_id')
    revenue_df.head()
```

check_in_date	checkout_date	no_guests	room_category	booking_platform	ratings_given	booking_status	revenue_generated	revenue_realized
2022-05-01	2022-05-02	3	RT1	direct online	1.0	Checked Out	10010	10010
2022-05-01	2022-05-02	2	RT1	others	NaN	Cancelled	9100	3640
2022-05-01	2022-05-04	2	RT1	logtrip	5.0	Checked Out	9100	9100
2022-05-01	2022-05-02	2	RT1	others	NaN	Cancelled	9100	3640
2022-05-01	2022-05-02	4	RT1	direct online	5.0	Checked Out	10920	10920

+ Code + Text

✓ RAM Disk Gemini ^

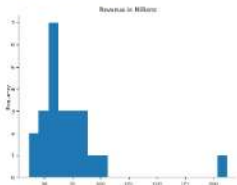
```
✓ [27] # Individual Hotel Revenue
0s hotel_revenue=revenue_df.groupby(['property_name','city']).agg({'revenue_realized':'sum'}).rename(columns={'revenue_realized':'Revenue in Millions'})
hotel_revenue.reset_index(inplace=True)
hotel_revenue['Revenue in Millions']=hotel_revenue['Revenue in Millions']/1000000
hotel_revenue['Revenue in Millions']=hotel_revenue['Revenue in Millions'].round(2)
hotel_revenue
```



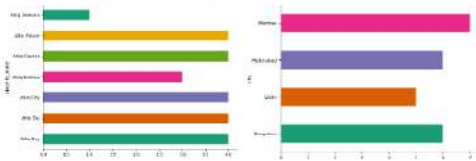
1 to 24 of 24 entries Filter ?

Index	property_name	city	Revenue in Millions
0	Atliq Bay	Bangalore	82.44
1	Atliq Bay	Delhi	56.44
2	Atliq Bay	Hyderabad	69.26
3	Atliq Bay	Mumbai	51.91
4	Atliq Blu	Bangalore	72.96
5	Atliq Blu	Delhi	57.93
6	Atliq Blu	Hyderabad	56.04
7	Atliq Blu	Mumbai	73.92
8	Atliq City	Bangalore	81.88
9	Atliq City	Delhi	54.93
10	Atliq City	Hyderabad	61.01
11	Atliq City	Mumbai	88.0
12	Atliq Exotica	Bangalore	60.02
13	Atliq Exotica	Hyderabad	47.84
14	Atliq Exotica	Mumbai	212.44
15	Atliq Grands	Bangalore	54.49
16	Atliq Grands	Delhi	36.06
17	Atliq Grands	Hyderabad	46.25
18	Atliq Grands	Mumbai	74.73
19	Atliq Palace	Bangalore	68.6
20	Atliq Palace	Delhi	89.14
21	Atliq Palace	Hyderabad	44.84

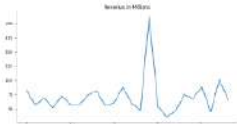
✓ [27] Distributions



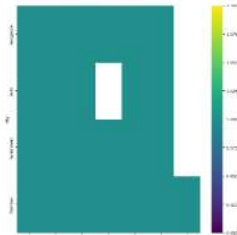
Categorical distributions



Values

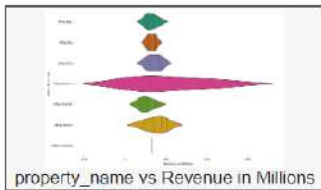
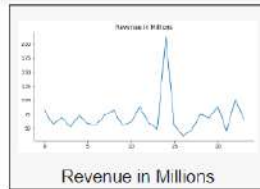
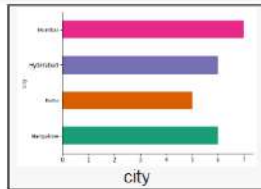
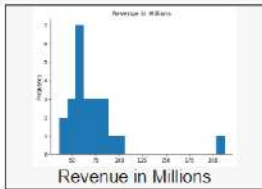


2-d categorical distributions



+ Code + Text

✓ RAM ☐ Disk ☐ Gemini ^



✓ 0s [28] revenue_pivot=hotel_revenue.pivot(index='city',columns='property_name',values='Revenue in Millions')
revenue_pivot



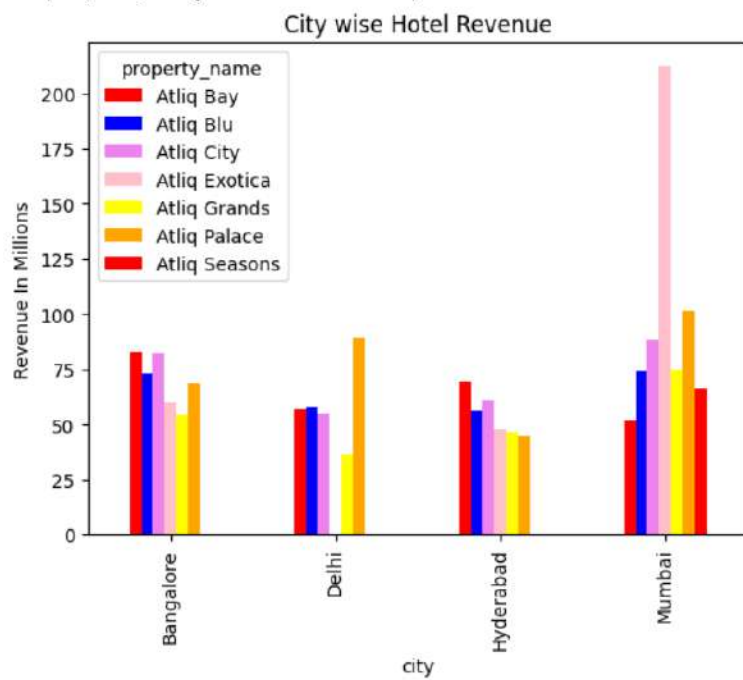
1 to 4 of 4 entries Filter ?

city	Atliq Bay	Atliq Blu	Atliq City	Atliq Exotica	Atliq Grands	Atliq Palace	Atliq Seasons
Bangalore	82.44	72.96	81.88	60.02	54.49	68.6	NaN
Delhi	56.44	57.93	54.93	NaN	36.06	89.14	NaN
Hyderabad	69.26	56.04	61.01	47.84	46.25	44.84	NaN
Mumbai	51.91	73.92	88.0	212.44	74.73	101.51	66.13

Show 25 per page

```
[35] revenue_pivot.plot(kind='bar',color=['red','blue','violet','pink','yellow','orange'])
plt.ylabel('Revenue In Millions')
plt.title('City wise Hotel Revenue')
```

Text(0.5, 1.0, 'City wise Hotel Revenue')



+ Code + Text

✓ RAM Disk Gemini ^

```
0s # Citywise Revenue In Millions
city_revenue=revenue_df.groupby(['city']).agg({'revenue_realized':'sum'}).rename(columns={'revenue_realized':'Revenue in Millions'})
city_revenue['Revenue in Millions']=city_revenue['Revenue in Millions']/1000000
city_revenue['Revenue in Millions']=city_revenue['Revenue in Millions'].round(2)
city_revenue.sort_values(by='Revenue in Millions',ascending=False)
```



1 to 4 of 4 entries

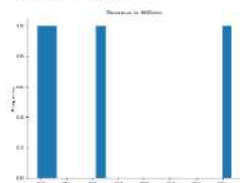
city	Revenue in Millions
Mumbai	668.64
Bangalore	420.4
Hyderabad	325.23
Delhi	294.5

Show per page

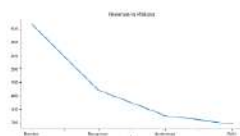


Like what you see? Visit the [data table notebook](#) to learn more about interactive tables.

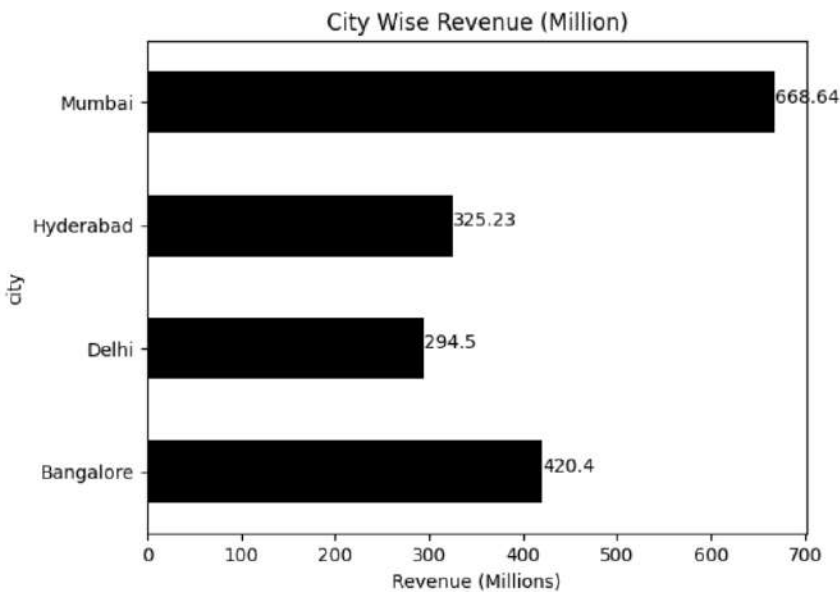
Distributions



Values



```
ax=city_revenue.plot(kind='barh',legend=False,color='black')
plt.xlabel('Revenue (Millions)')
plt.title('City Wise Revenue (Million)')
for index, value in enumerate(city_revenue['Revenue in Millions']):
    ax.text(value, index, str(value))
plt.show()
```



+ Code + Text

✓ RAM
Disk

✦ Gemini

```
revenue_tr=pd.merge(dim_date,revenue_df,how='left',left_on='date',right_on='check_in_date')
revenue_trend=revenue_tr.groupby(['week no','property_name']).agg({'revenue_realized':'sum'}).rename(columns={'revenue_realized':'Revenue in Mi
revenue_trend['Revenue in Millions']=revenue_trend['Revenue in Millions']/1000000
revenue_trend['Revenue in Millions']=revenue_trend['Revenue in Millions'].round(2)
revenue_trend.reset_index(inplace=True)
revenue_trend
```



1 to 25 of 91 entries Filter ?

Index	week no	property_name	Revenue in Millions
0	W 19	Atliq Bay	20.87
1	W 19	Atliq Blu	20.98
2	W 19	Atliq City	23.32
3	W 19	Atliq Exotica	25.74
4	W 19	Atliq Grands	17.55
5	W 19	Atliq Palace	24.32
6	W 19	Atliq Seasons	5.41
7	W 20	Atliq Bay	21.37
8	W 20	Atliq Blu	21.47
9	W 20	Atliq City	23.38
10	W 20	Atliq Exotica	26.15
11	W 20	Atliq Grands	17.17
12	W 20	Atliq Palace	24.7
13	W 20	Atliq Seasons	5.2
14	W 21	Atliq Bay	17.58
15	W 21	Atliq Blu	17.76
16	W 21	Atliq City	19.56
17	W 21	Atliq Exotica	20.68
18	W 21	Atliq Grands	14.28
19	W 21	Atliq Palace	20.52
20	W 21	Atliq Seasons	4.55
21	W 22	Atliq Bay	20.83

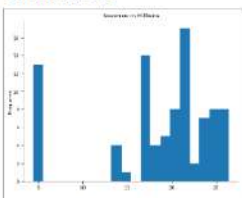
+ Code + Text

✓ RAM
Disk

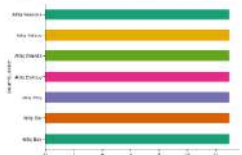
◆ Gemini



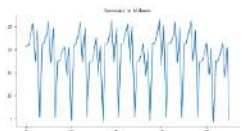
Distributions



Categorical distributions



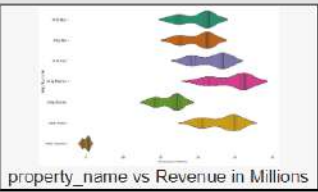
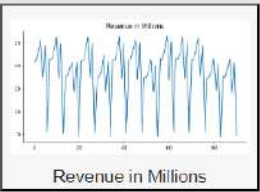
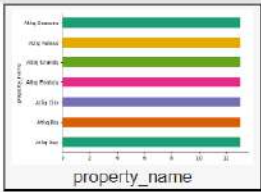
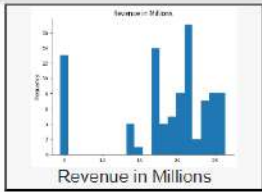
Values



+ Code + Text

✓ RAM
Disk

◆ Gemini




```
[92] # Weekly Trend of Hotel's Revenues
pivot_data=revenue_trend.pivot(index='week no',columns='property_name',values='Revenue in Millions')
pivot_data.plot(kind='line',marker='o', color=['red','blue','violet','pink','yellow','orange'])
plt.ylabel('Revenue (Millions)')
plt.title('Weekly Revenue Trend')
```

Text(0.5, 1.0, 'Weekly Revenue Trend')



+ Code + Text

✓ RAM
Disk Gemini ^

0s  # Week over Week Revenue Trend
atliq_revenue_trend=revenue_tr.groupby(['week no']).agg({'revenue_realized':'sum'}).rename(columns={'revenue_realized':'Revenue in Millions'})
atliq_revenue_trend['Revenue in Millions']=atliq_revenue_trend['Revenue in Millions']/1000000
atliq_revenue_trend['Revenue in Millions']=atliq_revenue_trend['Revenue in Millions'].round(2)
atliq_revenue_trend['Prev week Revenue']=atliq_revenue_trend['Revenue in Millions'].shift(1)
atliq_revenue_trend['Change Percentage']=((atliq_revenue_trend['Revenue in Millions']/atliq_revenue_trend['Prev week Revenue'])-1)*100
atliq_revenue_trend

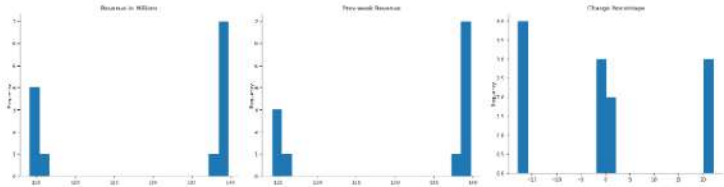
 Revenue in Millions Prev week Revenue Change Percentage 

week no

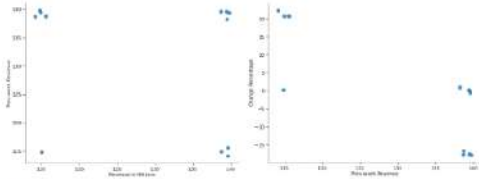
W 19	138.18	NaN	NaN
W 20	139.44	138.18	0.911854
W 21	114.92	139.44	-17.584624
W 22	138.72	114.92	20.710059
W 23	115.57	138.72	-16.688293
W 24	139.58	115.57	20.775288
W 25	138.67	139.58	-0.651956
W 26	114.15	138.67	-17.682267
W 27	139.56	114.15	22.260184
W 28	139.38	139.56	-0.128977
W 29	139.73	139.38	0.251112
W 30	114.81	139.73	-17.834395
W 31	115.04	114.81	0.200331

[38]

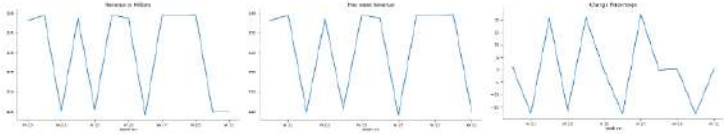
Distributions



2-d distributions



Values

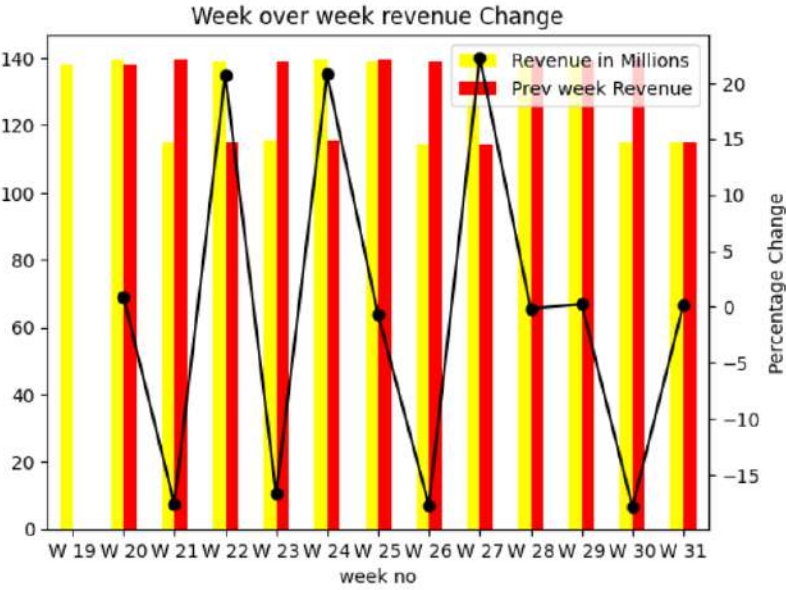


Next steps: [Generate code with atliq_revenue_trend](#) [View recommended plots](#) [New interactive sheet](#)

```
[40] atliq_revenue_trend[['Revenue in Millions','Prev week Revenue']].plot(kind='bar',color=['yellow', 'red'])
atliq_revenue_trend['Change Percentage'].plot(secondary_y=True,color='black',marker='o')
plt.ylabel('Percentage Change')
plt.title('Week over week revenue change')
```

```
[40] atliq_revenue_trend[['Revenue in Millions','Prev week Revenue']].plot(kind='bar',color=['yellow', 'red'])
atliq_revenue_trend['Change Percentage'].plot(secondary_y=True,color='black',marker='o')
plt.ylabel('Percentage Change')
plt.title('Week over week revenue Change')
```

Text(0.5, 1.0, 'Week over week revenue Change')



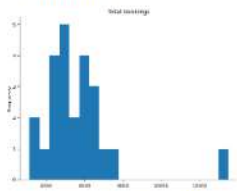
```
[41] # Hotelwise Bookings
hotel_bookings=revenue_df.groupby(['property_name','city']).agg({'booking_id':'nunique'}).rename(columns={'booking_id':'Total bookings'})
hotel_bookings.reset_index(inplace=True)
hotel_bookings
```

1 to 24 of 24 entriesFilter?

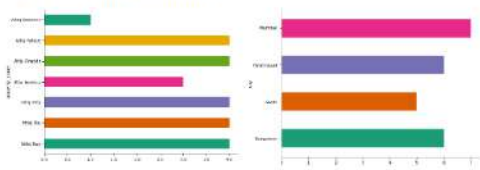
index	property_name	city	Total bookings
0	Atliq Bay	Bangalore	5812
1	Atliq Bay	Delhi	4820
2	Atliq Bay	Hyderabad	7333
3	Atliq Bay	Mumbai	3424
4	Atliq Blu	Bangalore	5736
5	Atliq Blu	Delhi	4418
6	Atliq Blu	Hyderabad	6458
7	Atliq Blu	Mumbai	5183
8	Atliq City	Bangalore	5979
9	Atliq City	Delhi	4693
10	Atliq City	Hyderabad	6638
11	Atliq City	Mumbai	6013
12	Atliq Exotica	Bangalore	4705
13	Atliq Exotica	Hyderabad	5256
14	Atliq Exotica	Mumbai	13480
15	Atliq Grands	Bangalore	4371
16	Atliq Grands	Delhi	3153
17	Atliq Grands	Hyderabad	4475
18	Atliq Grands	Mumbai	5036
19	Atliq Palace	Bangalore	5413
20	Atliq Palace	Delhi	7147
21	Atliq Palace	Hyderabad	4728
22	Atliq Palace	Mumbai	6337
23	Atliq Seaside	Mumbai	5689

✓ [41]
Cs

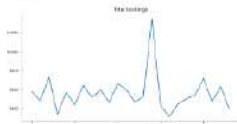
Distributions



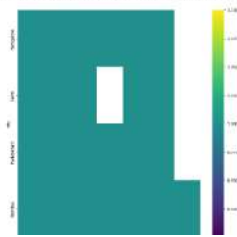
Categorical distributions



Values



2-d categorical distributions



```
[44] booking_pivot=hotel_bookings.pivot(index='city',columns='property_name',values='Total bookings')
      booking_pivot.plot(kind='bar', color=['red','blue','violet','pink','yellow','orange'])
      plt.ylabel('Total Bookings')
      plt.title('Hotel & City wise Bookings')
```

Text(0.5, 1.0, 'Hotel & City wise Bookings')



+ Code + Text

✓ RAM
Disk

◆ Gemini



✓
00

```
[45] # Hotelwise cancellation percentage, No Show Percentage, Check out percentage
df1=revenue_df.groupby(['property_name','booking_status']).agg({'booking_id':'nunique'}).reset_index().pivot(index='property_name',columns='booking_status',values='booking_id')
df2=revenue_df.groupby(['property_name']).agg({'booking_id':'nunique'}).rename(columns={'booking_id':'Toatal Bookings'})
df3=pd.merge(df1,df2,how='inner',on='property_name')
df3
```



Cancelled Checked Out No Show Toatal Bookings



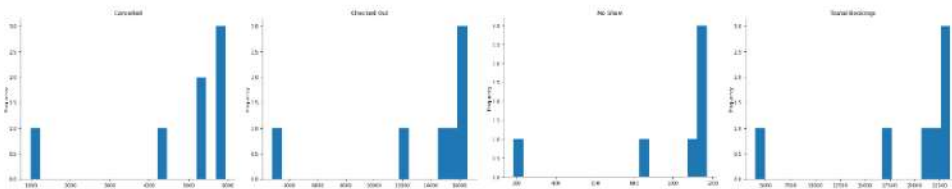
property_name



Atliq Bay	5314	14965	1110	21389
Atliq Blu	5373	15267	1155	21795
Atliq City	5811	16365	1147	23323
Atliq Exotica	5713	16557	1171	23441
Atliq Grands	4273	11914	848	17035
Atliq Palace	5949	16532	1144	23625
Atliq Seasons	987	2811	184	3982



Distributions



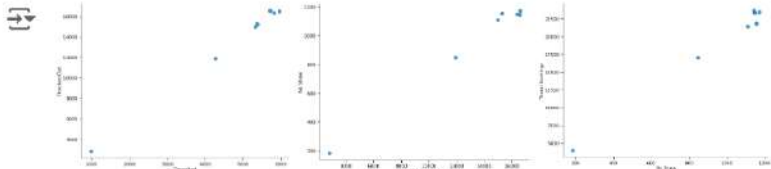
2-d distributions



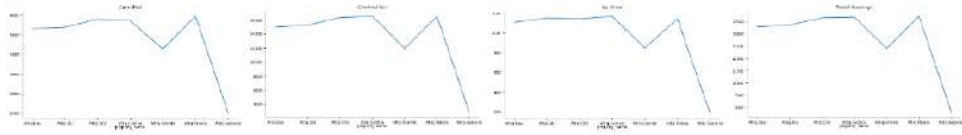
+ Code + Text

✓ RAM
Disk Gemini ^

2-d distributions



Values



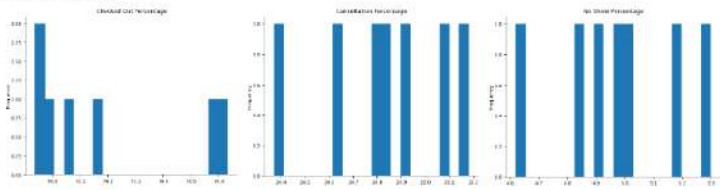
Next steps: [Generate code with df3](#) [View recommended plots](#) [New interactive sheet](#)

```
[46] df3['Cancellation Percentage']=df3['Cancelled']*100/df3['Toatal Bookings']
df3['Checked Out Percentage']=df3['Checked Out']*100/df3['Toatal Bookings']
df3['No Show Percentage']=df3['No Show']*100/df3['Toatal Bookings']
df4=df3[['Checked Out Percentage','Cancellation Percentage','No Show Percentage']]
df4
```

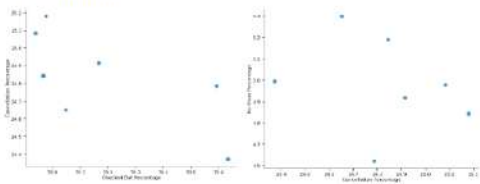
	Checked Out Percentage	Cancellation Percentage	No Show Percentage
property_name			
Atliq Bay	69.965870	24.844546	5.189583
Atliq Blu	70.048176	24.652443	5.299381
Atliq City	70.166788	24.915320	4.917892

🎮	Atliq City	70.166788	24.915320	4.917892
🔗	Atliq Exotica	70.632652	24.371827	4.995521
	Atliq Grands	69.938362	25.083651	4.977986
	Atliq Palace	69.976720	25.180952	4.842328
	Atliq Seasons	70.592667	24.786539	4.620794

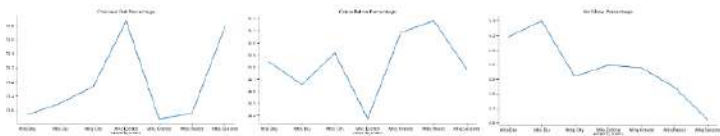
Distributions



2-d distributions

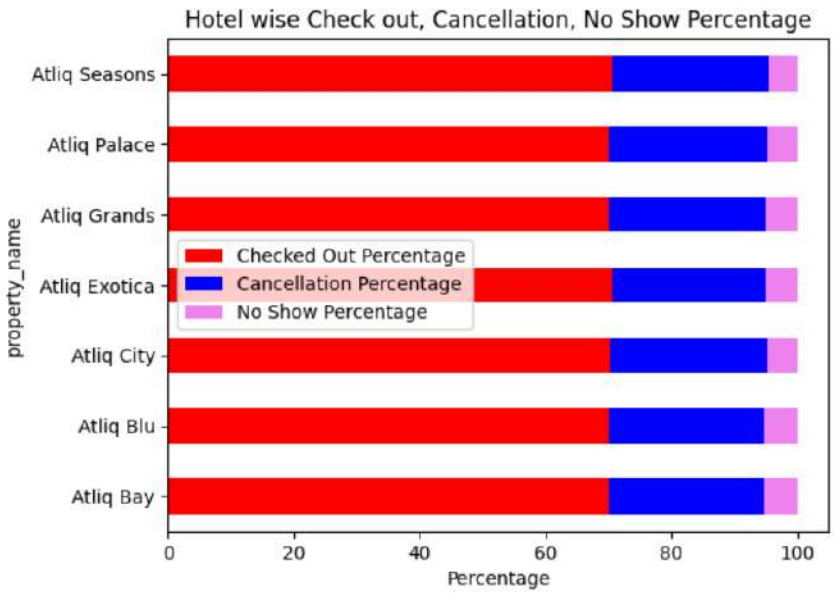


Values



```
df4.plot(kind='barh',stacked=True,color=['red','blue','violet'])
plt.xlabel('Percentage')
plt.title('Hotel wise Check out, Cancellation, No Show Percentage')
```

Text(0.5, 1.0, 'Hotel wise Check out, Cancellation, No Show Percentage')



```
[49] # booking platform wise
platform_df=revenue_df.groupby(['booking_platform']).agg({'booking_id':'nunique'}).rename(columns={'booking_id':'Total Bookings'})
platform_df.sort_values(by='Total Bookings',ascending=False)
```

✓ Cs

▶

↕

Total Bookings

booking_platform

others

55066

makeyourtrip

26898

logtrip

14756

direct online

13379

tripster

9630

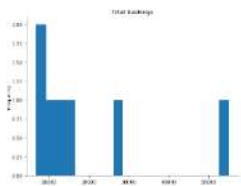
journey

8106

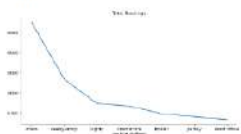
direct offline

6755

Distributions

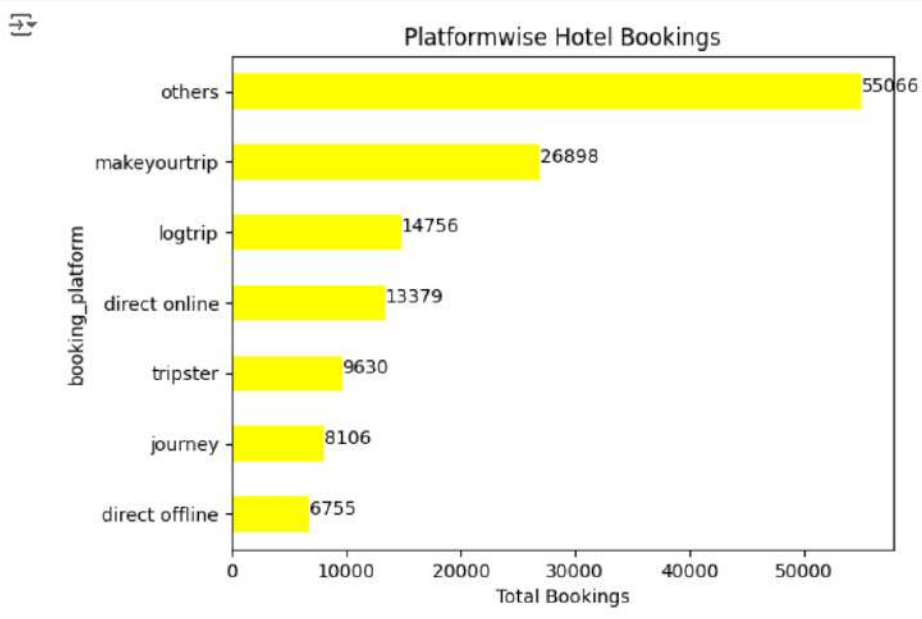


Values



```
[50] platform_df=platform_df.sort_values(by='Total Bookings')
```

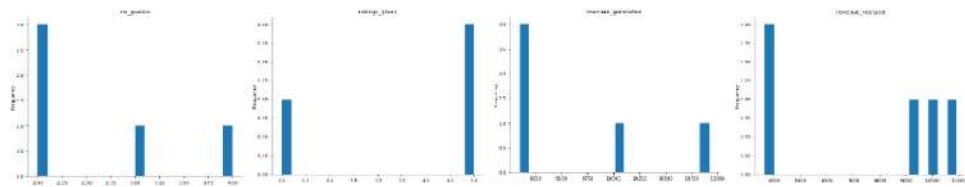
```
ax=platform_df.plot(kind='barh',legend=False, color='yellow')
plt.xlabel('Total Bookings')
plt.title('Platformwise Hotel Bookings')
for index, value in enumerate(platform_df['Total Bookings']):
    ax.text(value, index, str(value))
plt.show()
```



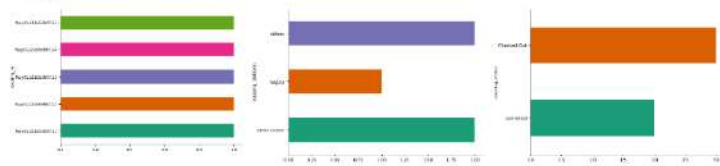
```
rooms_df=pd.merge(dim_rooms,fact_bookings,how='left',left_on='room_id',right_on='room_category')
rooms_df.head()
```

	room_id	room_class	booking_id	property_id	booking_date	check_in_date	checkout_date	no_guests	room_category	booking_platform	rate
0	RT1	Standard	May012216558RT11	16558	2022-04-27	2022-05-01	2022-05-02	3	RT1	direct online	100
1	RT1	Standard	May012216558RT12	16558	2022-04-30	2022-05-01	2022-05-02	2	RT1	others	100
2	RT1	Standard	May012216558RT13	16558	2022-04-28	2022-05-01	2022-05-04	2	RT1	logtrip	100
3	RT1	Standard	May012216558RT14	16558	2022-04-28	2022-05-01	2022-05-02	2	RT1	others	100
4	RT1	Standard	May012216558RT15	16558	2022-04-27	2022-05-01	2022-05-02	4	RT1	direct online	100

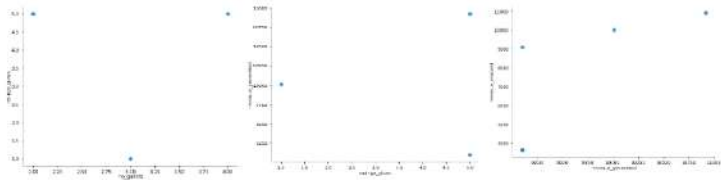
Distributions



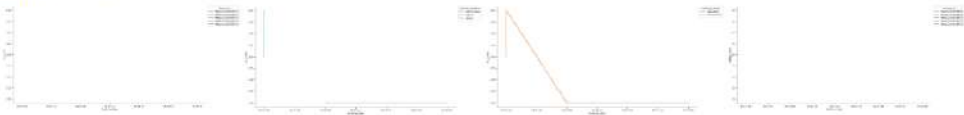
Categorical distributions



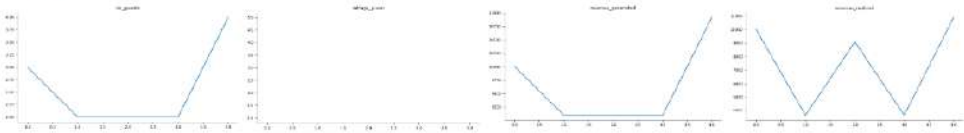
2-d distributions



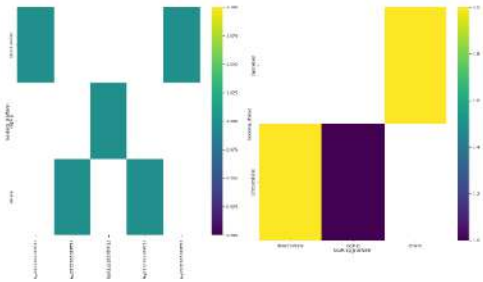
Time series



Values



2-d categorical distributions



+ Code + Text

✓ RAM
Disk

◆ Gemini



```
[53] # Room Categorywise Bookings
room_bookings=rooms_df.groupby(['room_class']).agg({'booking_id':'nunique'})
room_bookings['Percentage']=room_bookings['booking_id']*100/room_bookings['booking_id'].sum()
room_bookings['Percentage']=room_bookings['Percentage'].round(2)
room_bookings
```



booking_id Percentage



room_class



Elite 49505 36.78

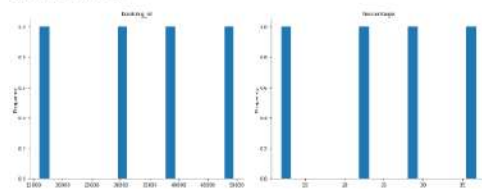


Premium 30566 22.71

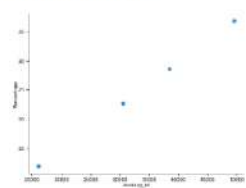
Presidential 16073 11.94

Standard 38446 28.57

Distributions



2-d distributions



✓

55

Booking Trends weekly
weekly_bookings=revenue_tr.groupby(['week no','property_name']).agg({'booking_id':'nunique'}).rename(columns={'booking_id':'Total Bookings'})
weekly_bookings.reset_index(inplace=True)
weekly_bookings_pivot=weekly_bookings.pivot(index='week no',columns='property_name',values='Total Bookings')
weekly_bookings_pivot



week no	Atliq Bay	Atliq Blu	Atliq City	Atliq Exotica	Atliq Grands	Atliq Palace	Atliq Seasons
W 19	1745	1771	1897	1907	1411	1913	321
W 20	1746	1789	1906	1906	1384	1908	319
W 21	1441	1465	1586	1546	1134	1600	270
W 22	1726	1754	1896	1913	1389	1933	323
W 23	1421	1475	1559	1624	1149	1591	270
W 24	1756	1784	1916	1937	1394	1929	325
W 25	1734	1779	1897	1895	1381	1930	327
W 26	1424	1459	1568	1574	1141	1581	270
W 27	1758	1777	1883	1911	1379	1933	321
W 28	1753	1781	1867	1902	1380	1924	322
W 29	1750	1784	1922	1921	1408	1909	324
W 30	1427	1457	1569	1569	1145	1585	268
W 31	1451	1450	1566	1557	1130	1593	274


```
[62] weekly_bookings_pivot.plot(kind='barh',legend='false')
plt.ylabel('Total Bookings')
plt.title('Weekly Hotel Bookings')
```

Text(0.5, 1.0, 'Weekly Hotel Bookings')



```
[63] x=revenue_tr.groupby(['week no','booking_status']).agg({'booking_id':'nunique'})
x.reset_index(inplace=True)
x
```

✓ [63]

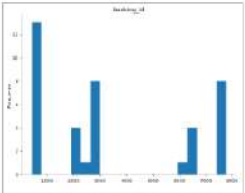
🔍

	week no	booking_status	booking_id
0	W 19	Cancelled	2779
1	W 19	Checked Out	7628
2	W 19	No Show	558
3	W 20	Cancelled	2711
4	W 20	Checked Out	7699
5	W 20	No Show	548
6	W 21	Cancelled	2226
7	W 21	Checked Out	6331
8	W 21	No Show	485
9	W 22	Cancelled	2772
10	W 22	Checked Out	7585
11	W 22	No Show	577
12	W 23	Cancelled	2222
13	W 23	Checked Out	6399
14	W 23	No Show	468
15	W 24	Cancelled	2797
16	W 24	Checked Out	7688
17	W 24	No Show	556
18	W 25	Cancelled	2748
19	W 25	Checked Out	7657

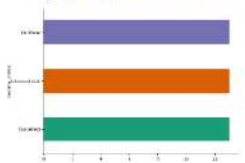
+ Code		+ Text			✓		RAM	Disk		▼	◆ Gemini	^
✓	0%	[63]	↕	20	W 25	No Show	538					
				21	W 26	Cancelled	2299					
				22	W 26	Checked Out	6292					
				23	W 26	No Show	426					
				24	W 27	Cancelled	2691					
				25	W 27	Checked Out	7734					
				26	W 27	No Show	537					
				27	W 28	Cancelled	2643					
				28	W 28	Checked Out	7757					
				29	W 28	No Show	529					
				30	W 29	Cancelled	2733					
				31	W 29	Checked Out	7777					
				32	W 29	No Show	508					
				33	W 30	Cancelled	2210					
				34	W 30	Checked Out	6348					
				35	W 30	No Show	462					
				36	W 31	Cancelled	2193					
				37	W 31	Checked Out	6347					
				38	W 31	No Show	481					



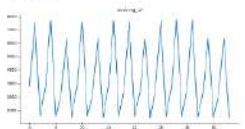
Distributions






Categorical distributions



Values



```
0% y=revenue_tr.groupby(['week no']).agg({'booking_id':'nunique'}).rename(columns={'booking_id':'Total Bookings'})  
y.reset_index(inplace=True)  
y
```



	week no	Total Bookings
0	W 19	10965
1	W 20	10958
2	W 21	9042
3	W 22	10934
4	W 23	9089
5	W 24	11041
6	W 25	10943
7	W 26	9017
8	W 27	10962
9	W 28	10929
10	W 29	11018
11	W 30	9020
12	W 31	9021

+ Code + Text

✓ RAM
Disk

+ Gemini



```
x1=x.pivot(index='week no',columns='booking_status',values='booking_id')
hotel_booking=pd.merge(x1,y,on='week no')
hotel_booking['Cancellation Percentage']=hotel_booking['Cancelled']*100/hotel_booking['Total Bookings']
hotel_booking['Checked Out Percentage']=hotel_booking['Checked Out']*100/hotel_booking['Total Bookings']
hotel_booking['No Show Percentage']=hotel_booking['No Show']*100/hotel_booking['Total Bookings']
booking_status_trend=hotel_booking[['week no','Checked Out Percentage','Cancellation Percentage','No Show Percentage']]
booking_status_trend
```



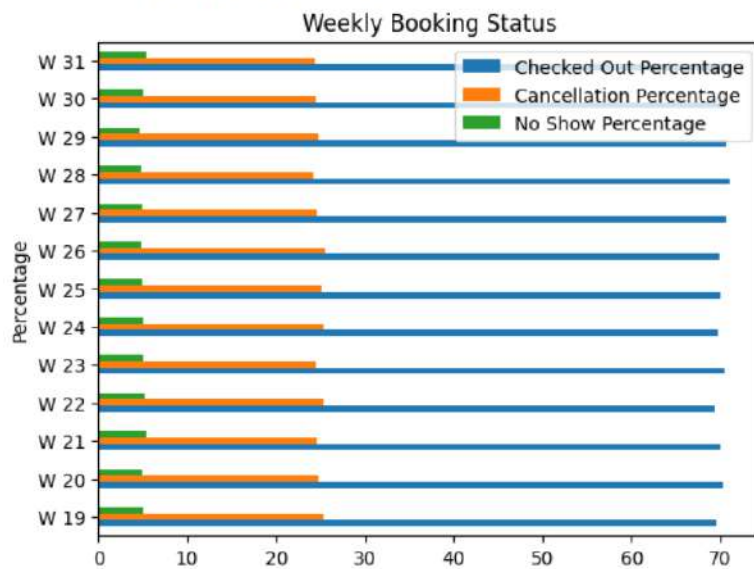
	week no	Checked Out Percentage	Cancellation Percentage	No Show Percentage
0	W 19	69.566803	25.344277	5.088919
1	W 20	70.259171	24.739916	5.000913
2	W 21	70.017695	24.618447	5.363858
3	W 22	69.370770	25.352113	5.277117
4	W 23	70.403785	24.447134	5.149081
5	W 24	69.631374	25.332850	5.035776
6	W 25	69.971671	25.111944	4.916385
7	W 26	69.779306	25.496285	4.724409
8	W 27	70.552819	24.548440	4.898741
9	W 28	70.976302	24.183365	4.840333
10	W 29	70.584498	24.804865	4.610637
11	W 30	70.376940	24.501109	5.121951
12	W 31	70.358053	24.309943	5.332003





```
18 x1=booking_status_trend.set_index('week no')  
x1.plot(kind='barh',legend='false')  
plt.ylabel('Percentage')  
plt.title('Weekly Booking Status')
```

```
Text(0.5, 1.0, 'Weekly Booking Status')
```



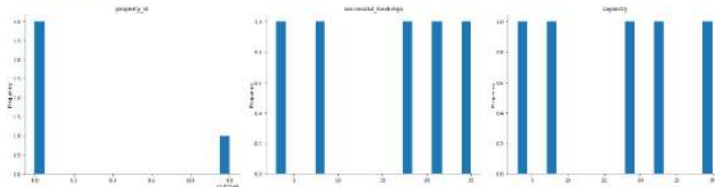
```
08 [68] oc=pd.merge(dim_hotels,fact_aggregate,how='left',on='property_id')  
occ_df=pd.merge(dim_date,oc,how='left',left_on='date',right_on='check_in_date')  
occ_df.head()
```


+ Code + Text

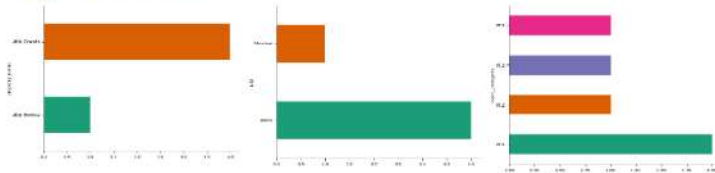
✓ RAM Disk Gemini ^

	date	mm	yy	week no	day_type	property_id	property_name	category	city	check_in_date	room_category	successful_bookings	capacity
0	2022-05-01	May	22	W 19	weekend	16558	Atliq Grands	Luxury	Delhi	2022-05-01	RT1	18	19
1	2022-05-01	May	22	W 19	weekend	16558	Atliq Grands	Luxury	Delhi	2022-05-01	RT2	21	22
2	2022-05-01	May	22	W 19	weekend	16558	Atliq Grands	Luxury	Delhi	2022-05-01	RT3	8	8
3	2022-05-01	May	22	W 19	weekend	16558	Atliq Grands	Luxury	Delhi	2022-05-01	RT4	3	3
4	2022-05-01	May	22	W 19	weekend	16559	Atliq Exotica	Luxury	Mumbai	2022-05-01	RT1	25	30

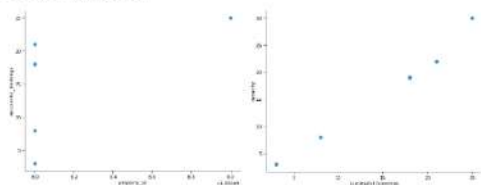
Distributions



Categorical distributions



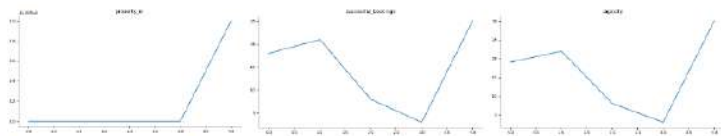
2-d distributions



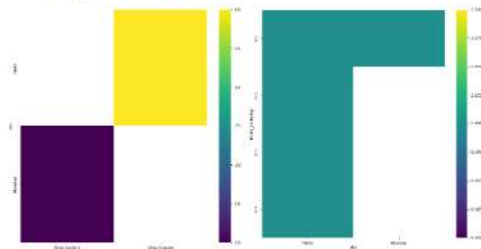
Time series



Values



2-d categorical distributions



+ Code + Text

✓ RAM
Disk

◆ Gemini

^

```
✓ [69] # Hotelwise Occupancy  
hotel_occ=occ_df.groupby(['city','property_name']).agg({'successful_bookings':'sum','capacity':'sum'})  
hotel_occ['Occupancy']=hotel_occ['successful_bookings']*100/hotel_occ['capacity']  
hotel_occ.reset_index(inplace=True)  
hotel_occ
```



	city	property_name	successful_bookings	capacity	Occupancy
0	Bangalore	Atliq Bay	5736	8736	65.659341
1	Bangalore	Atliq Blu	5669	10647	53.245046
2	Bangalore	Atliq City	5904	9009	65.534466
3	Bangalore	Atliq Exotica	4645	8645	53.730480
4	Bangalore	Atliq Grands	4316	9737	44.325768
5	Bangalore	Atliq Palace	5347	10010	53.416583
6	Delhi	Atliq Bay	4762	8918	53.397623
7	Delhi	Atliq Blu	4362	6643	65.663104
8	Delhi	Atliq City	4635	8645	53.614806
9	Delhi	Atliq Grands	3114	4732	65.807270
10	Delhi	Atliq Palace	7054	10647	66.253405
11	Hyderabad	Atliq Bay	7246	11011	65.806920
12	Hyderabad	Atliq Blu	6374	9737	65.461641
13	Hyderabad	Atliq City	6553	9919	66.065128
14	Hyderabad	Atliq Exotica	5192	11648	44.574176
15	Hyderabad	Atliq Grands	4420	8281	53.375196

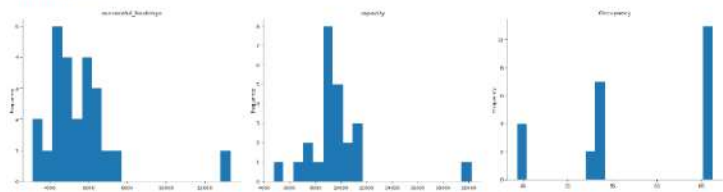
+ Code + Text

✓ RAM
Disk Gemini ^

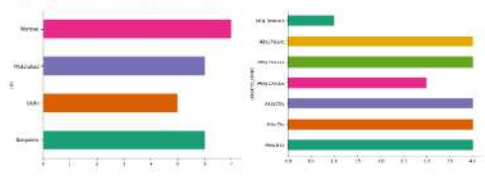
✓ 0% [69]

15	Hyderabad	Atliq Grands	4420	8281	53.375196
16	Hyderabad	Atliq Palace	4669	8827	52.894528
17	Mumbai	Atliq Bay	3388	7553	44.856348
18	Mumbai	Atliq Blu	5120	7735	66.192631
19	Mumbai	Atliq City	5940	11193	53.068882
20	Mumbai	Atliq Exotica	13325	20202	65.958816
21	Mumbai	Atliq Grands	4975	9282	53.598362
22	Mumbai	Atliq Palace	6259	9464	66.134827
23	Mumbai	Atliq Seasons	3934	8827	44.567803

Distributions



Categorical distributions



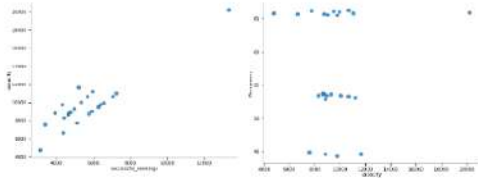
+ Code + Text

✓ RAM
Disk

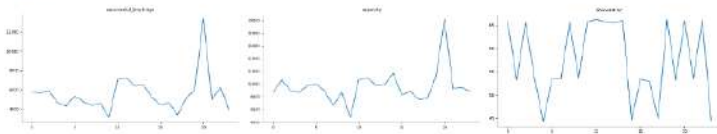
◆ Gemini



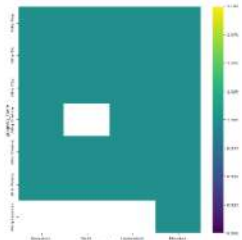
[69] 2-d distributions



Values



2-d categorical distributions



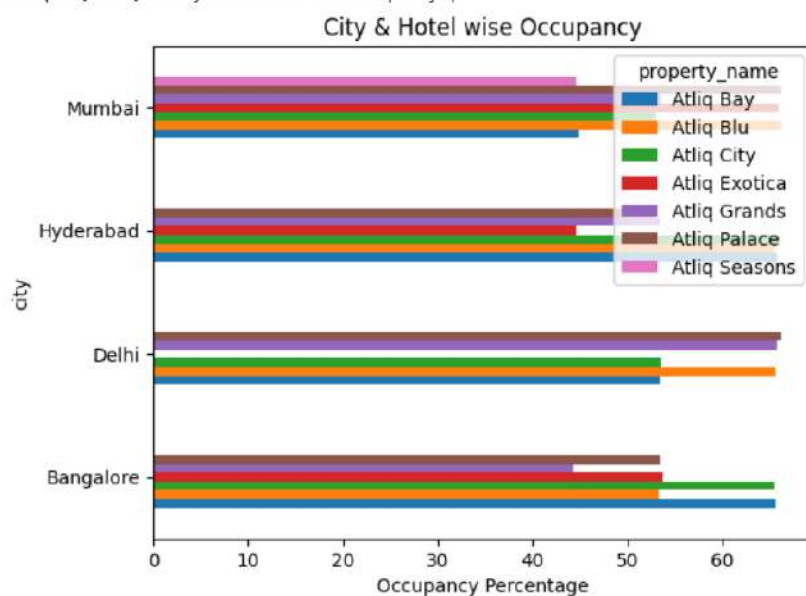
+ Code + Text

✓ RAM
Disk

◆ Gemini

```
[70] hotel_occ_pivot=hotel_occ.pivot(index='city',columns='property_name',values='Occupancy')  
hotel_occ_pivot.plot(kind='barh')  
plt.xlabel('Occupancy Percentage')  
plt.title('City & Hotel wise Occupancy')
```

Text(0.5, 1.0, 'City & Hotel wise Occupancy')



```
[71] # Room Class wise Occupancy  
room_occ=pd.merge(dim_rooms,occ_df,how='left',left_on='room_id',right_on='room_category')  
room_occ_df=room_occ.groupby(['room class']).agg({'successful bookings':'sum','capacity':'sum'})
```

```
[71] # Room Class wise Occupancy
room_occ=pd.merge(dim_rooms,occ_df,how='left',left_on='room_id',right_on='room_category')
room_occ_df=room_occ.groupby(['room_class']).agg({'successful_bookings':'sum','capacity':'sum'})
room_occ_df['Occupancy']=room_occ_df['successful_bookings']*100/room_occ_df['capacity']
room_occ_df['Occupancy']=room_occ_df['Occupancy'].round(2)
room_occ_df
```

↗

successful_bookings capacity Occupancy

room_class

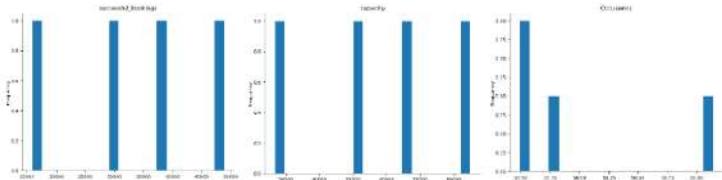
Elite	48916	84994	57.55
Premium	30186	52507	57.49
Presidential	15876	26845	59.14
Standard	37961	65702	57.78

📊

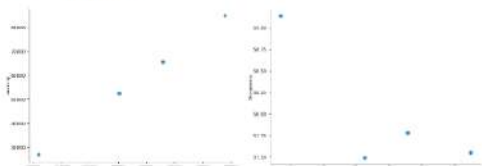
📄

✎

Distributions



2-d distributions



```
[72] room_df=room_occ_df[['occupancy']]
room_df=room_df.sort_values(by='occupancy')
room_df
```

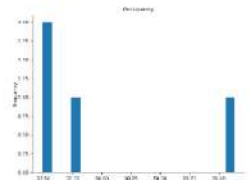
↕

Occupancy	
room_class	
Premium	57.49
Elite	57.55
Standard	57.78
Presidential	59.14

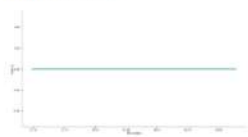
📊

✎

Distributions



Time series

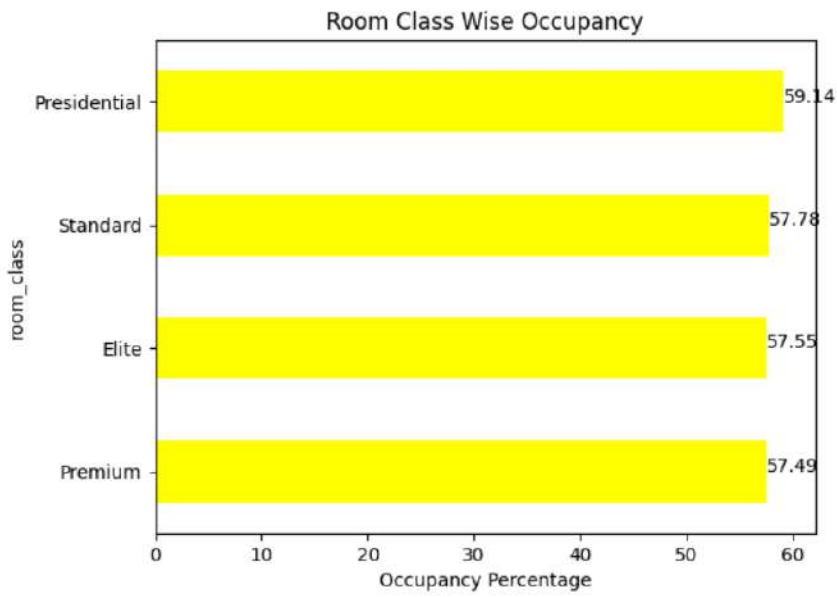


+ Code + Text

✓ RAM  Disk  Gemini ^

0s

```
ax=room_df.plot(kind='barh',legend=False, color='yellow')
plt.xlabel('Occupancy Percentage')
plt.title('Room Class Wise Occupancy')
for index, value in enumerate(room_df['Occupancy']):
    ax.text(value, index, str(value))
plt.show()
```



+ Code + Text

✓ RAM
Disk

◆ Gemini



✓
0a



Daywise Occupancy

```
occ_df['Day Name']=occ_df['date'].dt.day_name()
occ_df['Day No']=occ_df['date'].dt.dayofweek
day_occ=occ_df.groupby(['Day Name','Day No','property_name']).agg({'successful_bookings':'sum','capacity':'sum'})
day_occ['Occupancy']=day_occ['successful_bookings']*100/day_occ['capacity']
day_occx=day_occ[['Occupancy']]
z=day_occx.reset_index()
z.sort_values(by='Day No')
```



	Day Name	Day No	property_name	Occupancy
13	Monday	0	Atliq Seasons	38.620143
7	Monday	0	Atliq Bay	51.816776
8	Monday	0	Atliq Blu	55.618204
9	Monday	0	Atliq City	52.690502
10	Monday	0	Atliq Exotica	51.063094
11	Monday	0	Atliq Grands	46.590909
12	Monday	0	Atliq Palace	53.324946
35	Tuesday	1	Atliq Bay	52.010050
36	Tuesday	1	Atliq Blu	54.409988
37	Tuesday	1	Atliq City	52.130733
38	Tuesday	1	Atliq Exotica	50.613656
41	Tuesday	1	Atliq Seasons	38.937351
39	Tuesday	1	Atliq Grands	46.416084
40	Tuesday	1	Atliq Palace	53.702372

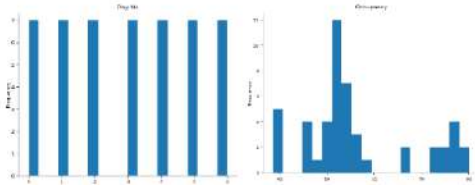


✓	0s	⏮	47	Wednesday	2	Atliq Palace	53.163192
		⏮	48	Wednesday	2	Atliq Seasons	40.444092
			42	Wednesday	2	Atliq Bay	51.894086
			43	Wednesday	2	Atliq Blu	54.389851
			44	Wednesday	2	Atliq City	52.871073
			45	Wednesday	2	Atliq Exotica	50.959378
			46	Wednesday	2	Atliq Grands	46.896853
			34	Thursday	3	Atliq Seasons	39.651071
			33	Thursday	3	Atliq Palace	53.001438
			32	Thursday	3	Atliq Grands	46.569056
			31	Thursday	3	Atliq Exotica	50.821089
			30	Thursday	3	Atliq City	52.907187
			29	Thursday	3	Atliq Blu	55.296013
			28	Thursday	3	Atliq Bay	51.662157
			0	Friday	4	Atliq Bay	52.164670
			1	Friday	4	Atliq Blu	55.074507
			2	Friday	4	Atliq City	52.690502
			3	Friday	4	Atliq Exotica	50.734659
			4	Friday	4	Atliq Grands	46.918706
			5	Friday	4	Atliq Palace	53.342919
			6	Friday	4	Atliq Seasons	40.126883

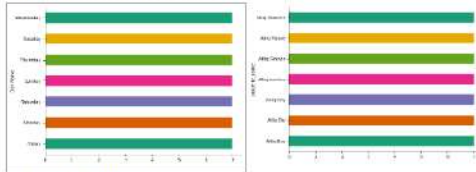
0s

16	Saturday	5	Atliq City	76.182737
20	Saturday	5	Atliq Seasons	57.414750
19	Saturday	5	Atliq Palace	76.078361
18	Saturday	5	Atliq Grands	66.936189
17	Saturday	5	Atliq Exotica	72.687986
15	Saturday	5	Atliq Blu	80.084575
14	Saturday	5	Atliq Bay	74.313877
26	Sunday	6	Atliq Palace	76.671459
21	Sunday	6	Atliq Bay	74.565133
22	Sunday	6	Atliq Blu	78.574305
23	Sunday	6	Atliq City	76.417479
25	Sunday	6	Atliq Grands	67.351399
27	Sunday	6	Atliq Seasons	56.780333
24	Sunday	6	Atliq Exotica	73.500432

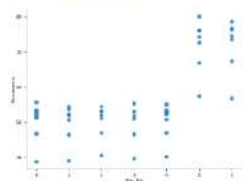
Distributions



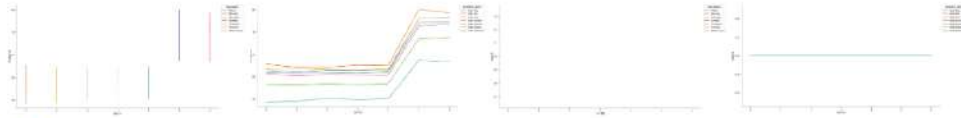
Categorical distributions



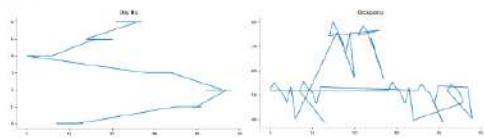
2-d distributions



Time series



Values

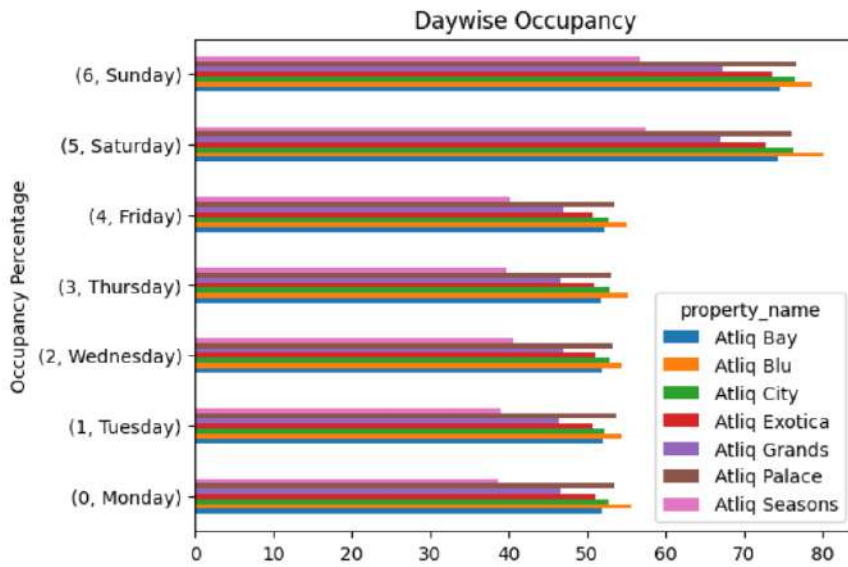


+ Code + Text

✓ RAM
Disk Gemini ^

```
✓ 18 z_pivot=z.pivot(index=['Day No','Day Name'],columns='property_name',values='Occupancy')  
z_pivot.plot(kind='barh',legend='false')  
plt.ylabel('Occupancy Percentage')  
plt.title('Daywise Occupancy')
```

Text(0.5, 1.0, 'Daywise Occupancy')



```
✓ 08 [78] weekly_occ=occ_df.groupby(['week no','property_name']).agg(['successful_bookings':'sum','capacity':'sum'])  
weekly_occx=weekly_occ.reset_index()  
weekly_occx['Occupancy']=weekly_occx['successful_bookings']*100/weekly_occx['capacity']  
weekly_occx
```

+ Code + Text

✓ RAM
Disk

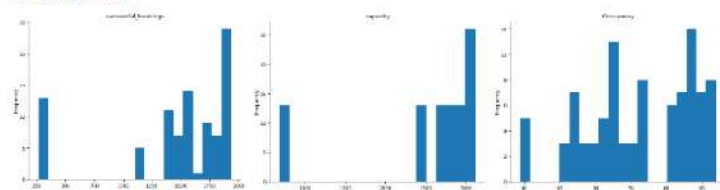
◆ Gemini



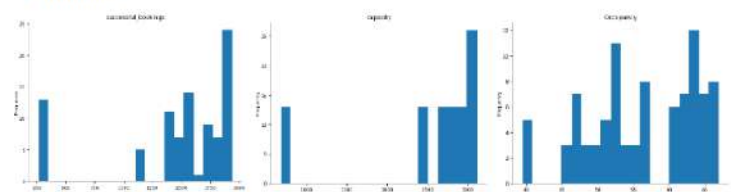
	week no	property_name	successful_bookings	capacity	Occupancy
0	W 19	Atliq Bay	1745	2786	62.634602
1	W 19	Atliq Blu	1771	2674	66.230366
2	W 19	Atliq City	1897	2982	63.615023
3	W 19	Atliq Exotica	1907	3115	61.219904
4	W 19	Atliq Grands	1411	2464	57.264610
...
86	W 31	Atliq City	1566	2982	52.515091
87	W 31	Atliq Exotica	1557	3115	49.983949
88	W 31	Atliq Grands	1130	2464	45.860390
89	W 31	Atliq Palace	1593	2996	53.170895
90	W 31	Atliq Seasons	274	679	40.353461

91 rows × 5 columns

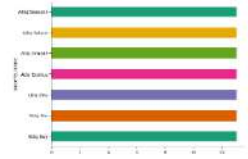
Distributions



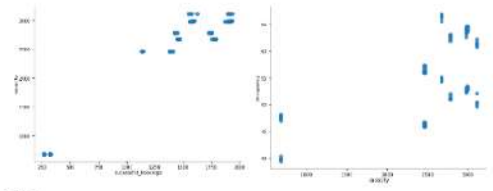
✓ [78] Distributions



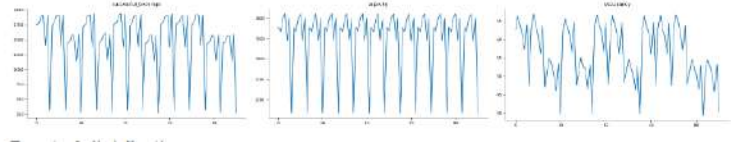
Categorical distributions



2-d distributions



Values

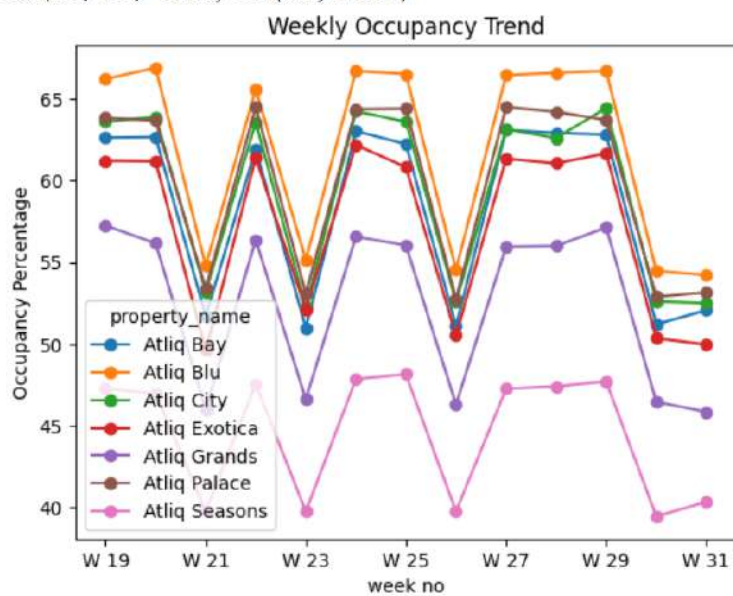


+ Code + Text

✓ RAM
Disk Gemini ^

```
[79] weekly_occx_pivot=weekly_occx.pivot(index='week no',columns='property_name',values='Occupancy')  
weekly_occx_pivot.plot(kind='line',marker='o')  
plt.ylabel('Occupancy Percentage')  
plt.title('Weekly Occupancy Trend')
```

Text(0.5, 1.0, 'Weekly Occupancy Trend')



```
[80] adr=revenue_tr.groupby(['property_name']).agg({'booking_id':'nunique','revenue_realized':'sum'})
```

+ Code + Text

✓ RAM
Disk Gemini ^

```
✓ [81] adr['ADR']=adr['revenue_realized']/adr['booking_id']  
0% adr['ADR']=adr['ADR'].round(0)  
adr_x=adr[['ADR']]  
adry=adr_x.sort_values(by='ADR')  
adr_x
```



ADR



property_name



Atliq Bay 12160.0

Atliq Blu 11972.0

Atliq City 12250.0

Atliq Exotica 13663.0

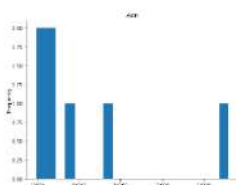
Atliq Grands 12416.0

Atliq Palace 12872.0

Atliq Seasons 16597.0



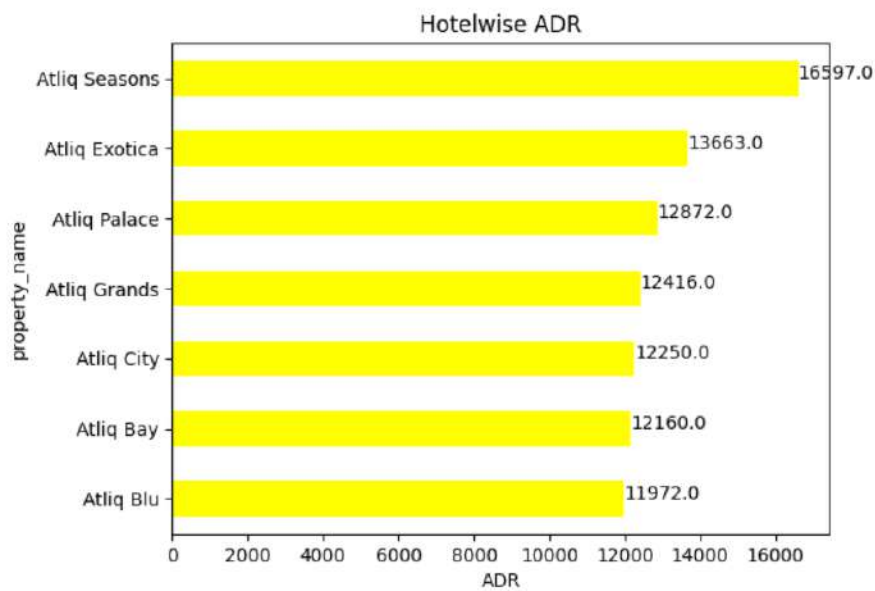
Distributions



+ Code + Text

✓ RAM
Disk Gemini ^

```
✓ [82] ax=adry.plot(kind='barh',legend=False,color='yellow')  
      plt.xlabel('ADR')  
      plt.title('Hotelwise ADR')  
      for index, value in enumerate(adry['ADR']):  
          ax.text(value, index, str(value))  
      plt.show()
```



```
✓ [83] min_date=dim_date['date'].min()  
      max_date=dim_date['date'].max()
```

+ Code + Text



RAM
Disk



◆ Gemini

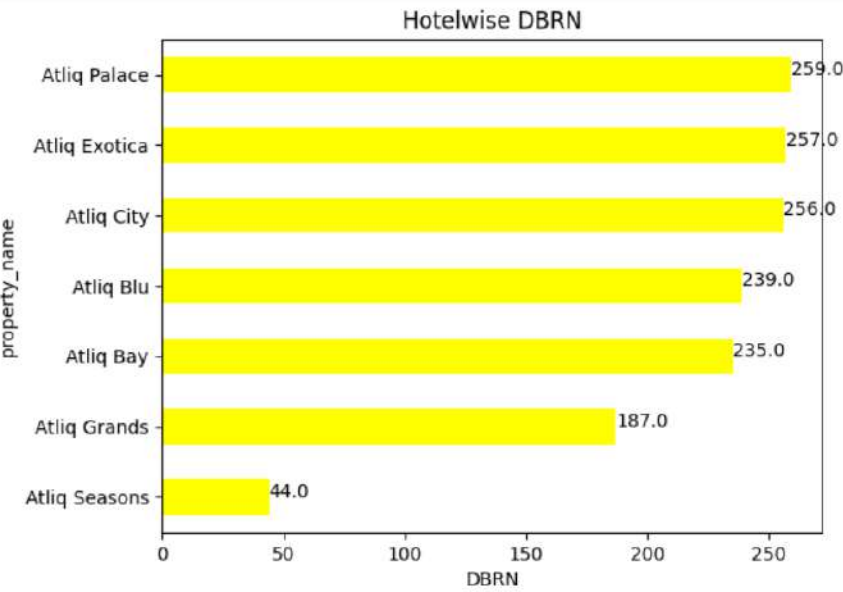


✓ [83] no_of_days=(max_date-min_date)/np.timedelta64(1,'D')
no_of_days



⇄ 90.0

✓ [84] #DBRN
DBRN=adr[['booking_id']]/no_of_days
DBRN['booking_id']=DBRN['booking_id'].round(0)
DBRNx=DBRN.sort_values(by='booking_id')




✓ [86] ax=DBRNx.plot(kind='barh',legend=False,color='yellow')
plt.xlabel('DBRN')
plt.title('Hotelwise DBRN')
for index, value in enumerate(DBRNx['booking_id']):
ax.text(value, index, str(value))
plt.show()



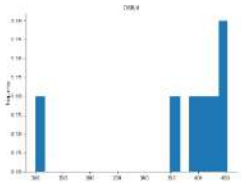
```
[87] #DSRN
hotel_capacity=pd.merge(dim_hotels,fact_aggregate,how='left',on='property_id').groupby(['property_name']).agg({'capacity':'sum'})
hotel_capacity['DSRN']=hotel_capacity/no_of_days
hotel_capacity['DSRN']=hotel_capacity['DSRN'].round(0)
hotel_capacity
hotel_capacityx=hotel_capacity.sort_values(by='DSRN')
hotel_capacityy=hotel_capacityx[['DSRN']]
hotel_capacityy
```

Da  

property_name	DSRN
Atliq Seasons	99.0
Atliq Grands	360.0
Atliq Blu	390.0
Atliq Bay	407.0
Atliq City	435.0
Atliq Palace	438.0
Atliq Exotica	455.0

Distributions



Time series



0s



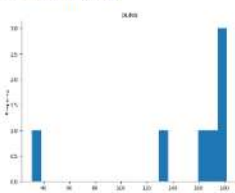
```
# DURN
m=revenue_tr[revenue_tr['booking_status']=='checked out']
n=m.groupby(['property_name']).agg({'booking_id':'nunique'})
n['DURN']=n['booking_id']/no_of_days
n['DURN']=n['DURN'].round(0)
n1=n.sort_values(by='DURN')
n2=n1[['DURN']]
n2
```



	DURN
property_name	
Atliq Seasons	31.0
Atliq Grands	131.0
Atliq Bay	164.0
Atliq Blu	168.0
Atliq City	179.0
Atliq Palace	181.0
Atliq Exotica	182.0



Distributions



```
ax=n2.plot(kind='barh',legend=False, color='yellow')
plt.xlabel('DURN')
plt.title('Hotelwise DURN')
for index, value in enumerate(n2['DURN']):
    ax.text(value, index, str(value))
plt.show()
```

Hotelwise DURN

