

```

1 import pandas as pd
2 import matplotlib.pyplot as plt
3 import seaborn as sns
4 import mysql.connector
5
6
7 db = mysql.connector.connect(host = "localhost",
8                             username = "username",
9                             password = "pass",
10                             database = "db_ec")
11
12 cur = db.cursor()
13 cur = db.cursor(buffered=True)
14

```

```

1 #List all unique cities where customers are located.
2 query = """ select distinct customer_city
3 from customers """
4
5 cur.execute(query)
6
7 data = cur.fetchall()
8
9 df = pd.DataFrame(data)
10 df.head()

```

	0
0	franca
1	sao bernardo do campo
2	sao paulo
3	mogi das cruces
4	campinas

```

1 #Count the number of orders placed in 2017.
2 query = """
3 SELECT COUNT(DISTINCT order_id)
4 FROM orders
5 WHERE YEAR(order_purchase_timestamp) = 2017
6 """
7
8 cur.execute(query)
9 data = cur.fetchone()
10 print("Total unique orders placed in 2017 are", data[0])
11

```

Total unique orders placed in 2017 are 45101

```

1 #Find the total sales per category.
2 query = """ select upper(products.product_category) category,
3 round(sum(payments.payment_value),2) sales
4 from products join order_items
5 on products.product_id = order_items.product_id
6 join payments
7 on payments.order_id = order_items.order_id
8 group by category
9 """
10
11 cur.execute(query)
12
13 data = cur.fetchall()
14
15 df = pd.DataFrame(data, columns = ["Category", "Sales"])
16 df

```

	Category	Sales
0	PERFUMERY	506738.66
1	FURNITURE DECORATION	1430176.39
2	TELEPHONY	486882.05
3	BED TABLE BATH	1712553.67
4	AUTOMOTIVE	852294.33
...
69	CDS MUSIC DVDS	1199.43
70	LA CUISINE	2913.53
71	FASHION CHILDREN'S CLOTHING	785.67
72	PC GAMER	2174.43
73	INSURANCE AND SERVICES	324.51

74 rows × 2 columns

```

1 #Calculate the percentage of orders that were paid in installments.
2 query = """ select ((sum(case when payment_installments >= 1 then 1
3 else 0 end))/count(*))*100 from payments
4 """
5
6 cur.execute(query)
7
8 data = cur.fetchall()
9
10 "the percentage of orders that were paid in installments is", data[0][0]

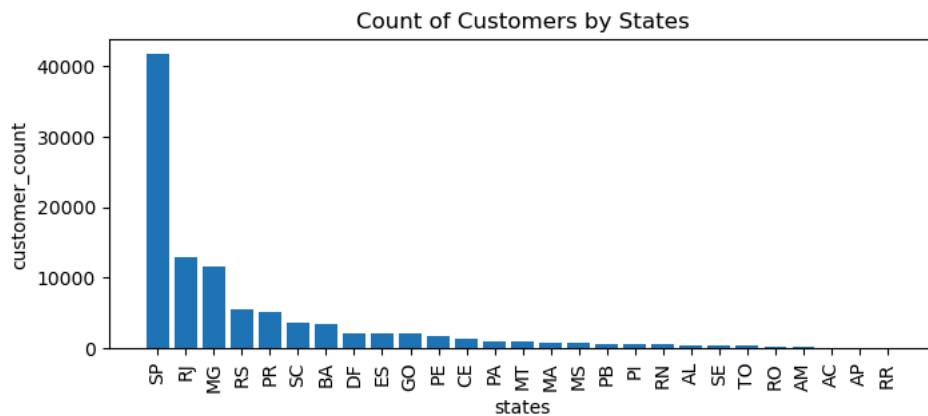
```

('the percentage of orders that were paid in installments is',
Decimal('99.9981'))

```

1 #Count the number of customers from each state.
2 query = """ select customer_state ,count(customer_id)
3 from customers group by customer_state
4 """
5
6 cur.execute(query)
7
8 data = cur.fetchall()
9 df = pd.DataFrame(data, columns = ["state", "customer_count" ])
10 df = df.sort_values(by = "customer_count", ascending= False)
11
12 plt.figure(figsize = (8,3))
13 plt.bar(df["state"], df["customer_count"])
14 plt.xticks(rotation = 90)
15 plt.xlabel("states")
16 plt.ylabel("customer_count")
17 plt.title("Count of Customers by States")
18 plt.show()

```



```

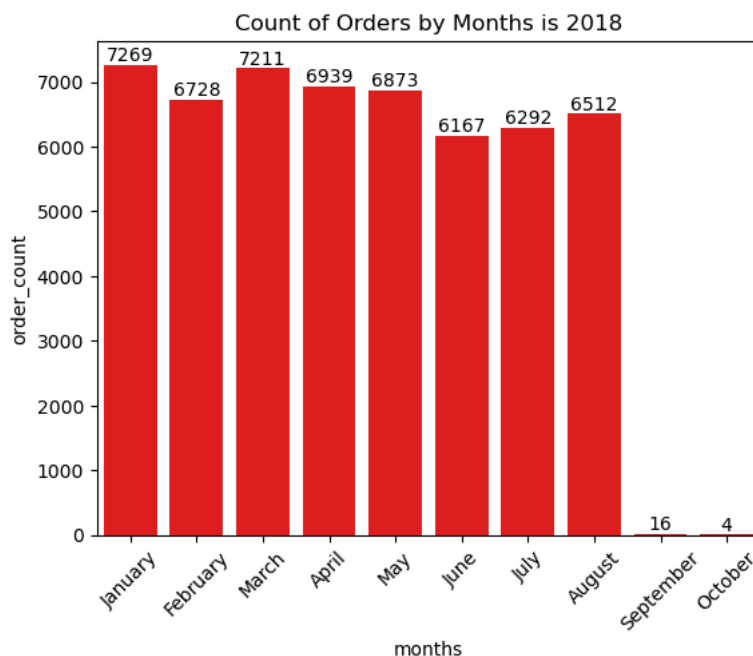
1 #Calculate the number of orders per month in 2018.
2 query = """ select monthname(order_purchase_timestamp) months, count(order_id) order_count
3 from orders where year(order_purchase_timestamp) = 2018
4 group by months
5 """
6
7 cur.execute(query)

```

```

8
9 data = cur.fetchall()
10 df = pd.DataFrame(data, columns = ["months", "order_count"])
11 o = ["January", "February", "March", "April", "May", "June", "July", "August", "September", "October"]
12
13 ax = sns.barplot(x = df["months"], y = df["order_count"], data = df, order = o, color = "red")
14 plt.xticks(rotation = 45)
15 ax.bar_label(ax.containers[0])
16 plt.title("Count of Orders by Months is 2018")
17
18 plt.show()

```



```

1 #Find the average number of products per order, grouped by customer city.
2 query = """with count_per_order as
3 (select orders.order_id, orders.customer_id, count(order_items.order_id) as oc
4 from orders join order_items
5 on orders.order_id = order_items.order_id
6 group by orders.order_id, orders.customer_id)
7
8 select customers.customer_city, round(avg(count_per_order.oc),2) average_orders
9 from customers join count_per_order
10 on customers.customer_id = count_per_order.customer_id
11 group by customers.customer_city order by average_orders desc
12 """
13
14 cur.execute(query)
15
16 data = cur.fetchall()
17 df = pd.DataFrame(data, columns = ["customer city", "average products/order"])
18 df.head(10)

```

	customer city	average products/order
0	padre carvalho	7.00
1	celso ramos	6.50
2	datas	6.00
3	candido godoi	6.00
4	matias olimpico	5.00
5	cidelandia	4.00
6	picarra	4.00
7	morro de sao paulo	4.00
8	teixeira soares	4.00
9	curralinho	4.00

```

1 #Calculate the percentage of total revenue contributed by each product category.
2 query = """select upper(products.product_category) category,
3 round((sum(payments.payment_value)/(select sum(payment_value) from payments))*100,2) sales_percentage

```

```

4 from products join order_items
5 on products.product_id = order_items.product_id
6 join payments
7 on payments.order_id = order_items.order_id
8 group by category order by sales_percentage desc"""
9
10
11 cur.execute(query)
12 data = cur.fetchall()
13 df = pd.DataFrame(data, columns = ["Category", "percentage distribution"])
14 df.head()

```

	Category	percentage distribution
0	BED TABLE BATH	10.70
1	HEALTH BEAUTY	10.35
2	COMPUTER ACCESSORIES	9.90
3	FURNITURE DECORATION	8.93
4	WATCHES PRESENT	8.93

```

1 #Identify the correlation between product price and the number of times a product has been purchased.
2 import numpy as np
3 cur = db.cursor()
4 query = """select products.product_category,
5 count(order_items.product_id),
6 round(avg(order_items.price),2)
7 from products join order_items
8 on products.product_id = order_items.product_id
9 group by products.product_category"""
10
11 cur.execute(query)
12 data = cur.fetchall()
13 df = pd.DataFrame(data, columns = ["Category", "order_count", "price"])
14
15 arr1 = df["order_count"]
16 arr2 = df["price"]
17
18 a = np.corrcoef([arr1, arr2])
19 print("the correlation is", a[0][-1])

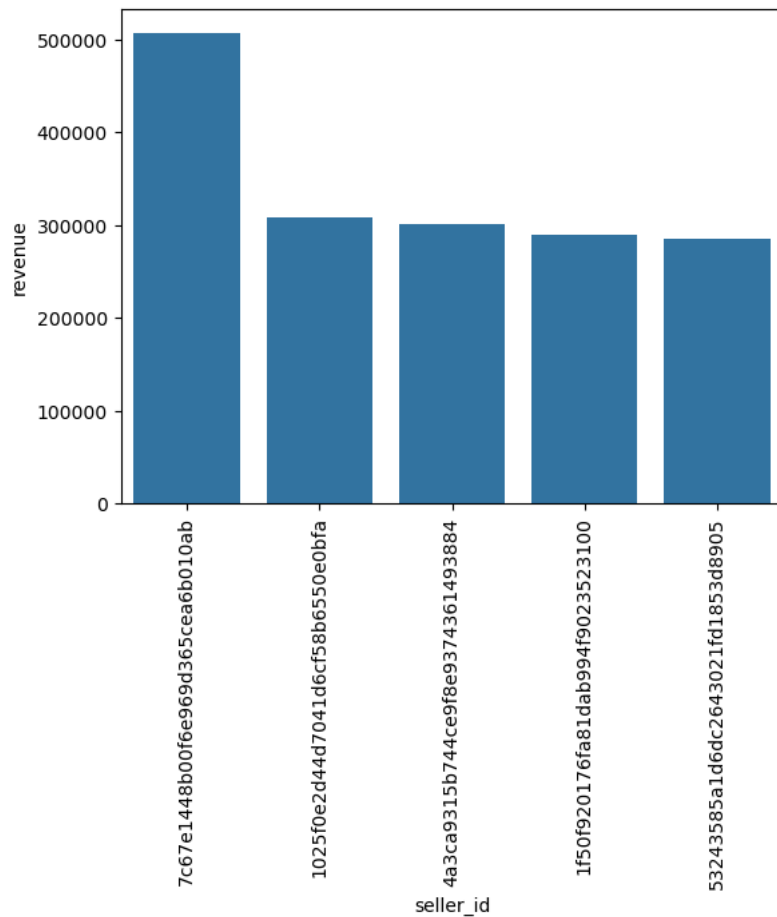
```

the correlation is -0.10631514167157562

```

1 #Calculate the total revenue generated by each seller, and rank them by revenue.
2 query = """ select *, dense_rank() over(order by revenue desc) as rn from
3 (select order_items.seller_id, sum(payments.payment_value)
4 revenue from order_items join payments
5 on order_items.order_id = payments.order_id
6 group by order_items.seller_id) as a """
7
8 cur.execute(query)
9 data = cur.fetchall()
10 df = pd.DataFrame(data, columns = ["seller_id", "revenue", "rank"])
11 df = df.head()
12 sns.barplot(x = "seller_id", y = "revenue", data = df)
13 plt.xticks(rotation = 90)
14 plt.show()

```



```

1 #Calculate the moving average of order values for each customer over their order history.
2 query = """select customer_id, order_purchase_timestamp, payment,
3 avg(payment) over(partition by customer_id order by order_purchase_timestamp
4 rows between 2 preceding and current row) as mov_avg
5 from
6 (select orders.customer_id, orders.order_purchase_timestamp,
7 payments.payment_value as payment
8 from payments join orders
9 on payments.order_id = orders.order_id) as a"""
10 cur.execute(query)
11 data = cur.fetchall()
12 df = pd.DataFrame(data)
13 df

```

	0	1	2	3
0	00012a2ce6f8dcda20d059ce98491703	2017-11-14 16:08:26	114.74	114.739998
1	000161a058600d5901f007fab4c27140	2017-07-16 09:40:32	67.41	67.410004
2	0001fd6190edaaf884bcdf3d49edf079	2017-02-28 11:06:43	195.42	195.419998
3	0002414f95344307404f0ace7a26f1d5	2017-08-16 13:09:20	179.35	179.350006
4	000379cdec625522490c315e70c7a9fb	2018-04-02 13:42:17	107.01	107.010002
...
103881	ffec9f79fd8c764f843e9951b11341	2018-03-29 16:59:26	71.23	27.120001
103882	ffeda5b6d849fbd39689bb92087f431	2018-05-22 13:36:02	63.13	63.130001
103883	ffff42319e9b2d713724ae527742af25	2018-06-13 16:57:05	214.13	214.130005
103884	fffa3172527f765de70084a7e53aae8	2017-09-02 11:53:32	45.50	45.500000
103885	fffe8b65bbe3087b653a978c870db99	2017-09-29 14:07:03	18.37	18.370001

103886 rows × 4 columns

```

1 #Calculate the cumulative sales per month for each year.
2 query = """select years, months , payment, sum(payment)
3 over(order by years, months) cumulative_sales from
4 (select year(orders.order_purchase_timestamp) as years,
5 month(orders.order_purchase_timestamp) as months,
6 round(sum(payments.payment_value),2) as payment from orders join payments

```

```

7 on orders.order_id = payments.order_id
8 group by years, months order by years, months) as a
9 """
10 cur.execute(query)
11 data = cur.fetchall()
12 df = pd.DataFrame(data)
13 df

```

	0	1	2	3
0	2016	9	252.24	252.24
1	2016	10	59090.48	59342.72
2	2016	12	19.62	59362.34
3	2017	1	138488.04	197850.38
4	2017	2	291908.01	489758.39
5	2017	3	449863.60	939621.99
6	2017	4	417788.03	1357410.02
7	2017	5	592918.82	1950328.84
8	2017	6	511276.38	2461605.22
9	2017	7	592382.92	3053988.14
10	2017	8	674396.32	3728384.46
11	2017	9	727762.45	4456146.91
12	2017	10	779677.88	5235824.79
13	2017	11	1194882.80	6430707.59
14	2017	12	878401.48	7309109.07
15	2018	1	1115004.18	8424113.25
16	2018	2	992463.34	9416576.59
17	2018	3	1159652.12	10576228.71
18	2018	4	1160785.48	11737014.19
19	2018	5	1153982.15	12890996.34
20	2018	6	1023880.50	13914876.84
21	2018	7	1066540.75	14981417.59
22	2018	8	1022425.32	16003842.91
23	2018	9	4439.54	16008282.45
24	2018	10	589.67	16008872.12

```

1 #Calculate the year-over-year growth rate of total sales.
2 query = """with a as(select year(orders.order_purchase_timestamp) as years,
3 round(sum(payments.payment_value),2) as payment from orders join payments
4 on orders.order_id = payments.order_id
5 group by years order by years)
6
7 select years, ((payment - lag(payment, 1) over(order by years))/
8 lag(payment, 1) over(order by years)) * 100 from a"""
9
10 cur.execute(query)
11 data = cur.fetchall()
12 df = pd.DataFrame(data, columns = ["years", "yoy % growth"])
13 df

```

	years	yoy % growth
0	2016	NaN
1	2017	12112.703761
2	2018	20.000924

```

1 #Calculate the retention rate of customers, defined as the percentage of customers who make another purchase within 6
2 query = """with a as (select customers.customer_id,
3 min(orders.order_purchase_timestamp) first_order
4 from customers join orders
5 on customers.customer_id = orders.customer_id
6 group by customers.customer_id),
7

```

```

8 b as (select a.customer_id, count(distinct orders.order_purchase_timestamp) next_order
9 from a join orders
10 on orders.customer_id = a.customer_id
11 and orders.order_purchase_timestamp > first_order
12 and orders.order_purchase_timestamp <
13 date_add(first_order, interval 6 month)
14 group by a.customer_id)
15
16 select 100 * (count( distinct a.customer_id)/ count(distinct b.customer_id))
17 from a left join b
18 on a.customer_id = b.customer_id ;""
19
20 cur.execute(query)
21 data = cur.fetchall()
22

```

[(None,)]

```

1 #Identify the top 3 customers who spent the most money in each year.
2 query = """select years, customer_id, payment, d_rank
3 from
4 (select year(orders.order_purchase_timestamp) years,
5 orders.customer_id,
6 sum(payments.payment_value) payment,
7 dense_rank() over(partition by year(orders.order_purchase_timestamp)
8 order by sum(payments.payment_value) desc) d_rank
9 from orders join payments
10 on payments.order_id = orders.order_id
11 group by year(orders.order_purchase_timestamp),
12 orders.customer_id) as a
13 where d_rank <= 3 ;""
14
15 cur.execute(query)
16 data = cur.fetchall()
17 df = pd.DataFrame(data, columns = ["years","id","payment","rank"])
18 sns.barplot(x = "id", y = "payment", data = df, hue = "years")
19 plt.xticks(rotation = 90)
20 plt.show()

```

