1. List all unique cities where customers are located.

```
query = """ select distinct customer_city from customers """

cur.execute(query)

data = cur.fetchall()

data
# df = pd.DataFrame(data)
# df.head()
```

```
[2]: [('franca',),
       ('sao bernardo do campo',),
       ('sao paulo',),
       ('mogi das cruzes',),
      ('campinas',),
       ('jaragua do sul',),
       ('timoteo',),
       ('curitiba',),
       ('belo horizonte',),
       ('montes claros',),
       ('rio de janeiro',),
       ('lencois paulista',),
       ('caxias do sul',),
       ('piracicaba',),
      ('guarulhos',),
       ('pacaja',),
       ('florianopolis',),
       ('aparecida de goiania',),
       ('santo andre',),
```

2. Count the number of orders placed in 2017.

```
[12]: query = """ select count( order_id) from orders where year(order_purchase_timestamp) = 2017 """

cur.execute(query)

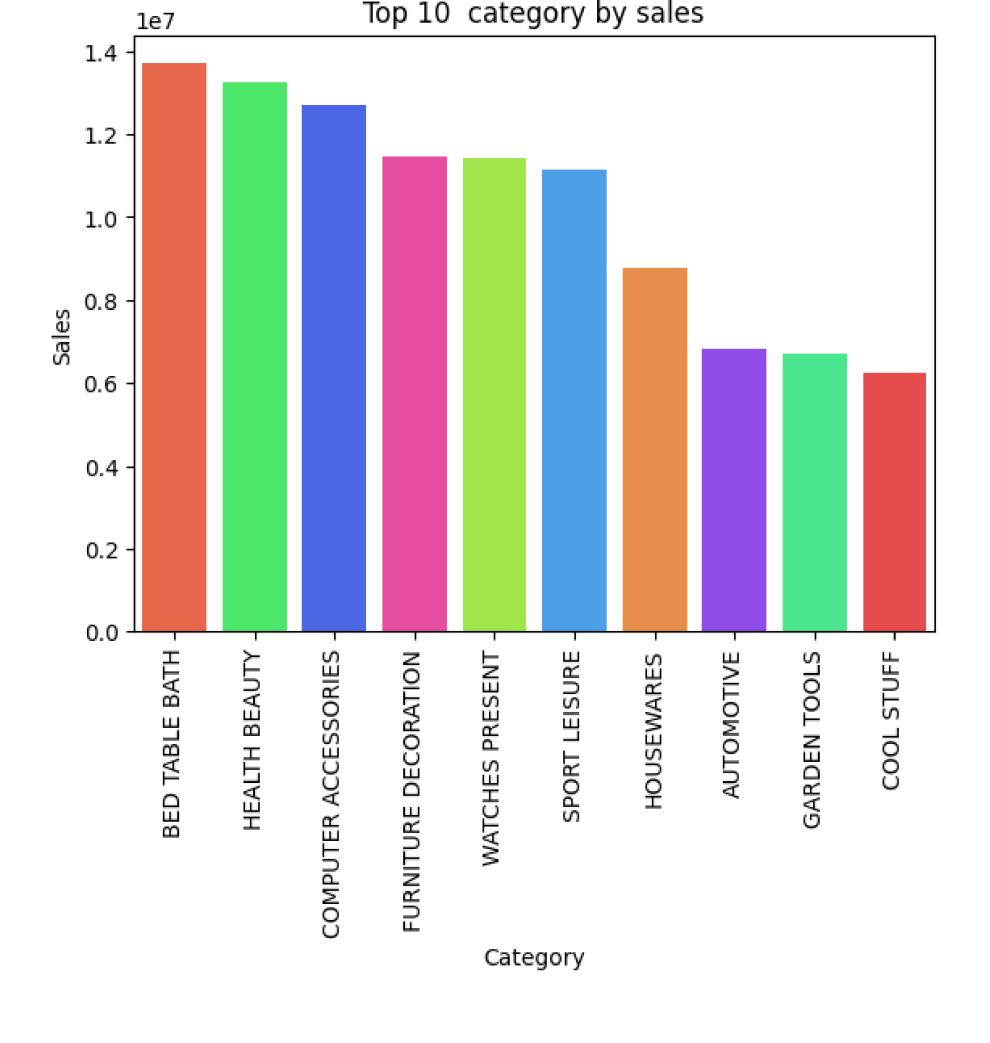
data = cur.fetchall()

data[0][0]
```

[12]: 90202

3. Find the top 10 category by sales.

```
query = """
select upper(products.product_category) category , round(sum(payments.payment_value),2) as paymentsa
from payments join order_items on payments.order_id = order_items.order_id
join products on products.product_id = order_items.product_id
group by category order by paymentsa desc limit 10
10.00.00
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns =["Category", "Sales"])
c = ['#FF5733', '#33FF57', '#3357FF', '#FF33A1', '#A1FF33', '#33A1FF', '#FF8C33', '#8C33FF', '#33FF8C', '#FF3333']
bar_plot = sns.barplot(x="Category", y="Sales", data=df , palette=c)
plt.xticks(rotation=90)
plt.title("Top 10 category by sales")
plt.show()
```



4. Calculate the percentage of orders that were paid in installments.

```
•[5]: query = """

select (sum(case when payment_installments then 1 else 0 end)/count(*))*100 from payments

"""

cur.execute(query)

data = cur.fetchall()

data

[5]: [(Decimal('99.9981'),)]
```

5. Count the number of customers from each state.

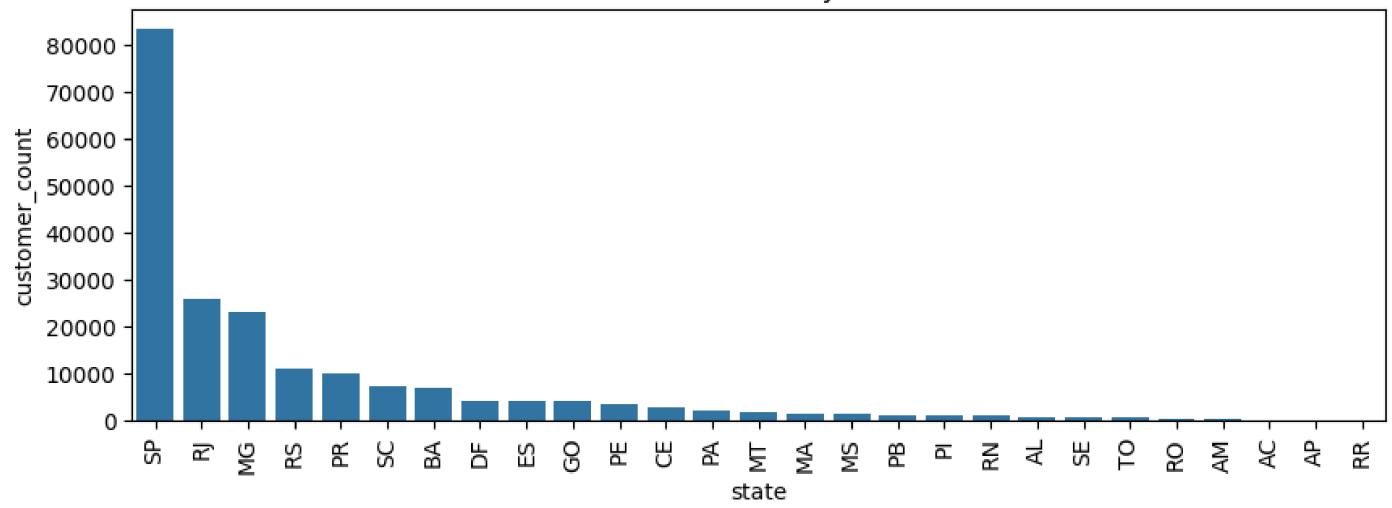
```
[71]: query = """
SELECT customer_state , count(customer_id) as customer FROM customers group by customer_state order by customer desc;
"""

cur.execute(query)

data = cur.fetchall()
    df = pd.DataFrame(data , columns =["state","customer_count"])
    df = df.sort_values(by = "customer_count" , ascending = False)

plt.figure(figsize = (10,3.3))
    sns.barplot(x = "state" ,y = "customer_count" ,data = df)
    plt.xticks(rotation = 90)
    plt.title("Count of order by state")
    plt.show()
    df
```

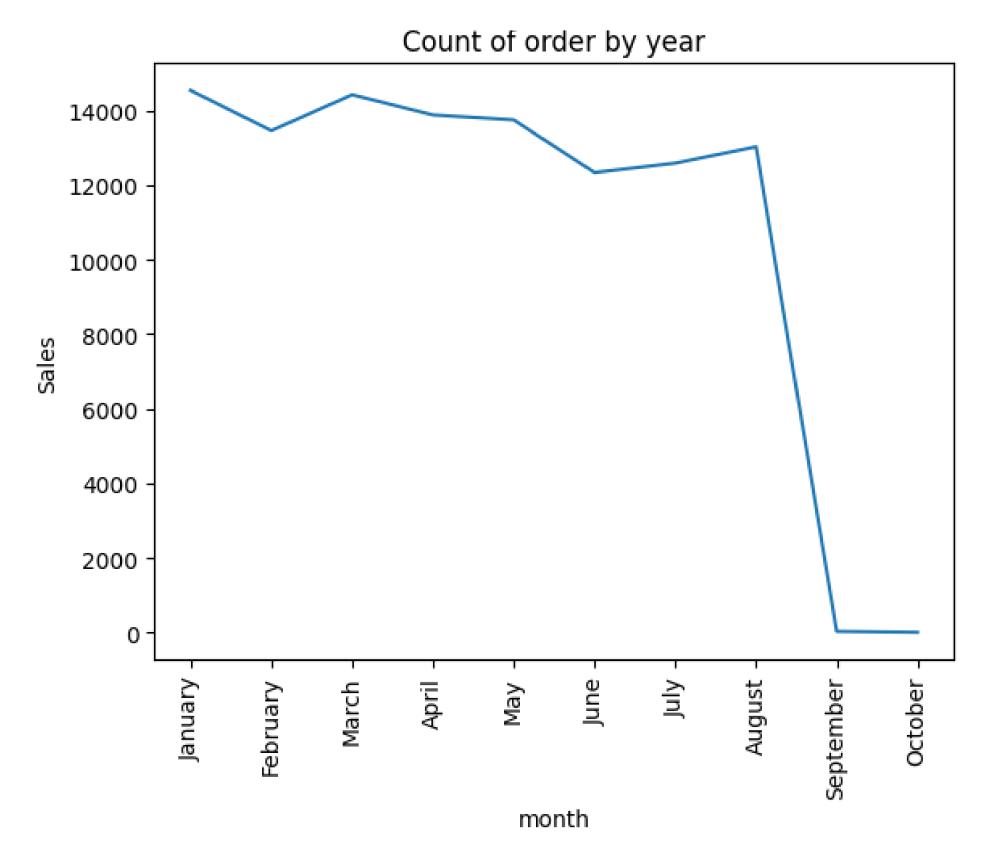
Count of order by state



[71]:		state	customer_count
	0	SP	83492
	1	RJ	25704
	2	MG	23270
	3	RS	10932
	4	PR	10090
	5	SC	7274
	6	ВА	6760

6 Calculate the number of orders per month in 2018.

```
[73]: query = """
      SELECT
          months,
          order count
      FROM (
          SELECT
              MONTHNAME(order_purchase_timestamp) AS months,
              COUNT(order_id) AS order_count,
              MONTH(order_purchase_timestamp) AS monthnum
          FROM
              orders
          WHERE
              YEAR(order_purchase_timestamp) = 2018
          GROUP BY
              months, monthnum
      ) AS monthly_orders
      ORDER BY
          monthnum;
      cur.execute(query)
      data = cur.fetchall()
      df = pd.DataFrame(data , columns =["month", "Sales"])
      sns.lineplot(data= df, x = "month", y = "Sales" )
      plt.xticks(rotation=90)
      plt.title("Count of order by year")
      plt.show()
      df
```



[73]:		month	Sales
	0	January	14538
	1	February	13456
	2	March	14422

```
    April 13878
    May 13746
    June 12334
    July 12584
    August 13024
    September 32
    October 8
```

[7]:

7 Find the average number of products per order, grouped by customer city.

```
query = """

with a as(select orders.order_id , orders.customer_id, count(order_items.order_item_id) as nikhil from order_items join orders on orders.order_id = orders.order_id , orders.customer_id)

select customers.customer_city ,round(avg(a.nikhil),2) from a join customers on a.customer_id = customers.customer_id group by customers.customer_city

"""

cur.execute(query)

data = cur.fetchall()

df = pd.DataFrame(data)

df
```

	0	1
0	sao jose dos campos	4.55
1	indaial	4.46
2	treze tilias	5.09
3	sao paulo	4.62
4	rio de janeiro	4.59
4105	aurora do para	4.00
4106	nova america	4.00
4107	tibau do sul	4.00
4108	lagoa da canoa	4.00
4109	buriti	12.00

[7]:

4110 rows × 2 columns

8 Calculate the percentage of total revenue contributed by each product category.

```
query = """
     select upper(products.product_category) , round((round(sum(payments.payment_value),2) / (select sum(payment_value) from payments))*100,2) from products
     payments.order_id = order_items.order_id group by products.product_category
     cur.execute(query)
     data = cur.fetchall()
     df = pd.DataFrame(data)
     df
[8]:
                                      0
                                            1
                             PERFUMERY 12.66
      0
                 FURNITURE DECORATION 35.73
      1
      2
                             TELEPHONY 12.17
      3 FASHION BAGS AND ACCESSORIES
```

BED TABLE BATH 42.79

0.03

0.07

0.02

0.05

0.01

CDS MUSIC DVDS

FASHION CHILDREN'S CLOTHING

INSURANCE AND SERVICES

LA CUISINE

PC GAMER

4

69

70

71

72

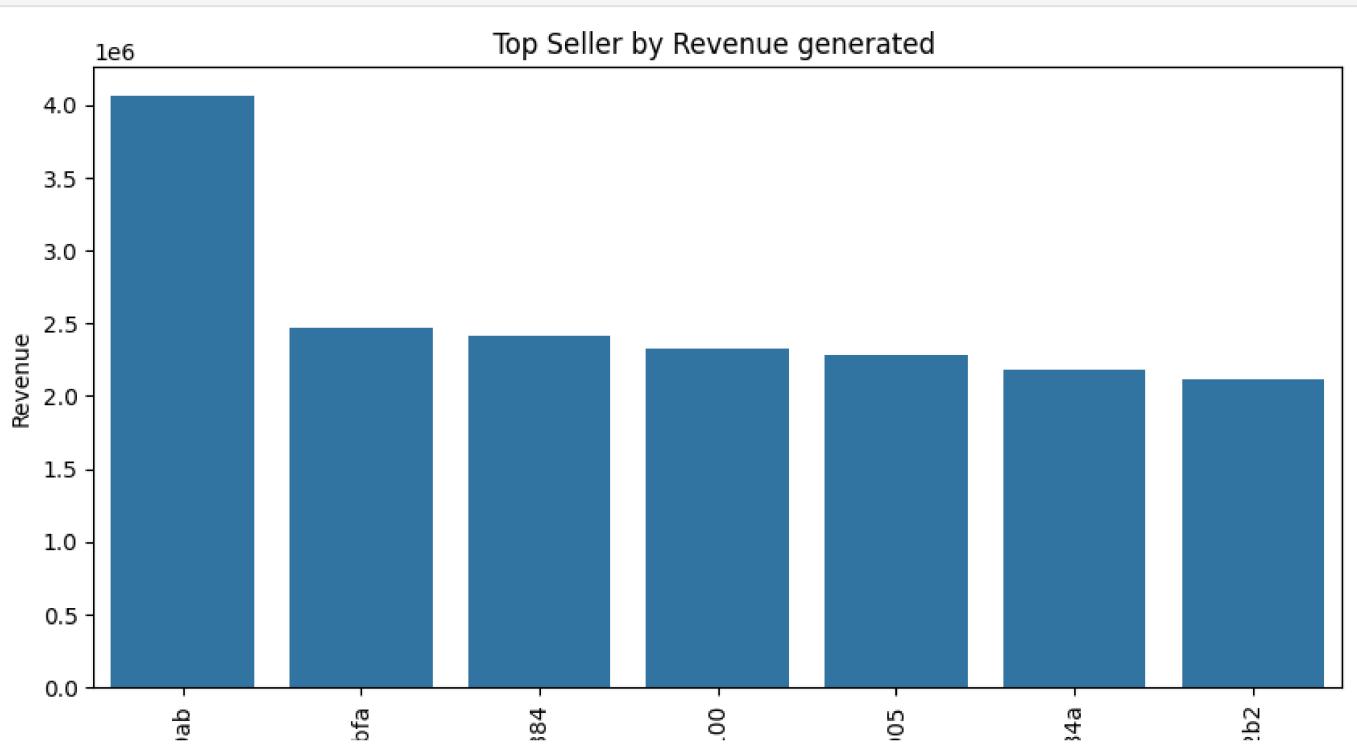
73

9 Identify the correlation between product price and the number of times a product has been purchased.

10 Calculate the total revenue generated by each seller, and rank them by revenue.

```
[81]: query = """
select * , rank() over(order by income desc) as ranks from
   (
    select sellers.seller_id , round(sum(payments.payment_value),2) as income from sellers join order_items on order_items.seller_id = sellers.seller_id
    join payments on order_items.order_id = payments.order_id group by sellers.seller_id
   ) as a
    """
    cur.execute(query)
```

```
data = cur.fetchall()
df = pd.DataFrame(data, columns = ["Seller" , "Revenue" , "Renk"])
df = df.head(7)
plt.figure(figsize = (10,5))
sns.barplot(x = "Seller", y = "Revenue" , data = df)
plt.title(" Top Seller by Revenue generated ")
plt.xticks(rotation = 90)
plt.show()
```



11 Calculate the moving average of order values for each customer over their order history.

```
[61]: query = """
select customer_id , order_purchase_timestamp , avg(payment_value) over(partition by customer_id order by order_purchase_timestamp rows between 2 preced
    (select orders.customer_id , orders.order_purchase_timestamp , payments.payment_value
    from orders join payments on orders.order_id = payments.order_id) as a
    """

cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns = ["customer" ,"time" ,"moving_avg"])
df
```

[61]:		customer	time	moving_avg
	0	00012a2ce6f8dcda20d059ce98491703	2017-11-14 16:08:26	114.739998
	1	00012a2ce6f8dcda20d059ce98491703	2017-11-14 16:08:26	114.739998
	2	00012a2ce6f8dcda20d059ce98491703	2017-11-14 16:08:26	114.739998
	3	00012a2ce6f8dcda20d059ce98491703	2017-11-14 16:08:26	114.739998
	4	000161a058600d5901f007fab4c27140	2017-07-16 09:40:32	67.410004
	415539	ffffa3172527f765de70084a7e53aae8	2017-09-02 11:53:32	45.500000
	415540	ffffe8b65bbe3087b653a978c870db99	2017-09-29 14:07:03	18.370001
	415541	ffffe8b65bbe3087b653a978c870db99	2017-09-29 14:07:03	18.370001
	415542	ffffe8b65bbe3087b653a978c870db99	2017-09-29 14:07:03	18.370001

12 Calculate the cumulative sales per month for each year.

```
query = """
select years, months, payment, sum(payment) over(order by years, months) as comunative_sales from
  (select year(orders.order_purchase_timestamp) as years, month(orders.order_purchase_timestamp) as months, round(sum(payments.payment_value),2) as payment join payments on orders.order_id = payments.order_id group by years, months order by years, months) as a
    """

cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns = ["year", "month", "avg", "moving_avg"])
df
```

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	year	month	avg	moving_avg
0	2016	9	1008.96	1008.96
1	2016	10	236361.92	237370.88
2	2016	12	78. 4 8	237449.36
3	2017	1	553952.16	791401.52
4	2017	2	1167632.04	1959033.56
5	2017	3	1799454.40	3758487.96
6	2017	4	1671152.12	5429640.08
7	2017	5	2371675.28	7801315.36
8	2017	6	2045105.52	9846420.88
9	2017	7	2369531.68	12215952.56

13 Calculate the year-over-year growth rate of total sales.

```
[12]: query = """
  with a as(
    select year(orders.order_purchase_timestamp) as years ,
    round(sum(payments.payment_value),2) revenue
  from orders join payments
    on payments.order_id = orders.order_id group by years order by years)
    select years , round((revenue-back)/back,6) * 100 from
    (select years ,revenue , lag(revenue ,1) over(order by years) back from a) as b
    """

    cur.execute(query)
    data = cur.fetchall()
    df = pd.DataFrame(data, columns = ["year", "yoy"])
    df
```

[12]:		year	yoy
	0	2016	NaN
	1	2017	12112.7038
	2	2018	20.0009

14 Identify the top 3 customers who spent the most money in each year.

```
[83]: query ="""
```

```
query ="""
select customer_id , years , money , ranks from
(select customer_id , money , dense_rank() over(partition by years order by money desc) as ranks, years from
(select orders.customer_id , year(orders.order_purchase_timestamp) as years , round(sum(payments.payment_value),2) as money
  from orders join payments on orders.order_id = payments.order_id group by orders.customer_id , year(orders.order_purchase_timestamp) order by money de
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data ,columns = ["customer_id","year","payment","rank"])
palette = ['#FF5733', '#33FF57', '#3357FF', '#FF33A1', '#A1FF33']
sns.barplot(x= "customer_id", y = "payment",data = df , hue= "year",palette=palette)
plt.xticks(rotation = 90)
plt.show()
```

