All unique cities where customers are located.

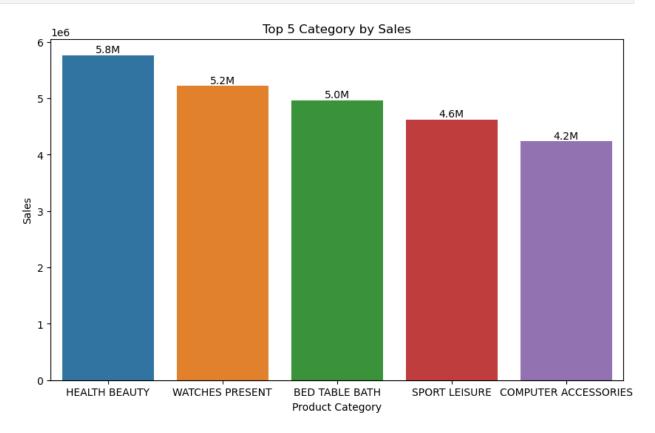
```
query = """select distinct customer_city from customers """
cur.execute(query)
dt = cur.fetchall()
df = pd.DataFrame(dt, columns = ["city"])
df
                       city
0
                     franca
1
      sao bernardo do campo
2
                  sao paulo
3
            mogi das cruzes
4
                   campinas
4114
                     siriji
4115
        natividade da serra
4116
               monte bonito
4117
                 sao rafael
4118
       eugenio de castro
[4119 rows x 1 columns]
```

Number of orders placed in 2017.

Total sales per category.

```
query = """select upper(p.product category), round(sum(oi.price +
oi.freight value),2) as total sales
           from order items as oi
           join products as p
           on oi.product id = p.product id
           group by p.product category
           order by total sales desc"""
cur.execute(query)
dt = cur.fetchall()
df = pd.DataFrame(dt, columns = ["Product Category" , "Sales"])
df
               Product Category
                                       Sales
0
                  HEALTH BEAUTY
                                 5764992.28
1
                WATCHES PRESENT
                                 5222166.43
2
                 BED TABLE BATH 4966726.88
3
                  SPORT LEISURE
                                 4626625.92
4
           COMPUTER ACCESSORIES 4237089.59
                                    6395.64
                        FLOWERS
69
70
                HOUSE COMFORT 2
                                    4682.32
71
                 CDS MUSIC DVDS
                                    3819.96
72
    FASHION CHILDREN'S CLOTHING
                                    2661.44
         INSURANCE AND SERVICES
                                    1298.04
73
[74 rows x 2 columns]
# Visualization of top 5 product category
fd = df.head(5)
plt.figure(figsize = (10,6))
xa = sns.barplot(x = 'Product Category', y = 'Sales', data = fd )
plt.title("Top 5 Category by Sales")
def formated num(number):
    if number >= 1e6:
        return f'{number / 1e6:.1f}M'
    if number >= 1e3:
        return f'{number / 1e3:.1f}k'
    else :
        return str(number)
```

```
for bars in xa.containers:
    xa.bar_label(bars, labels = [formated_num(label) for label in
bars.datavalues])
```



Percentage of orders that were paid in installments.

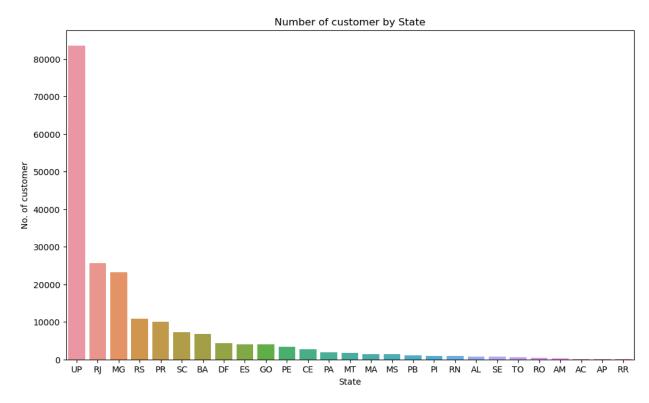
Count the number of customers from each state

```
query = """select customer_state, count(customer_id)
from customers
group by customer_state"""
cur.execute(query)
```

```
dt = cur.fetchall()
df = pd.DataFrame(dt, columns = ["State", "No. of customer"])
df = df.sort_values(by = "No. of customer", ascending = False )

plt.figure(figsize = (12,7))
sns.barplot(x = "State", y = "No. of customer", data = df )
plt.title("Number of customer by State")

Text(0.5, 1.0, 'Number of customer by State')
```



Number of orders per month in 2018.

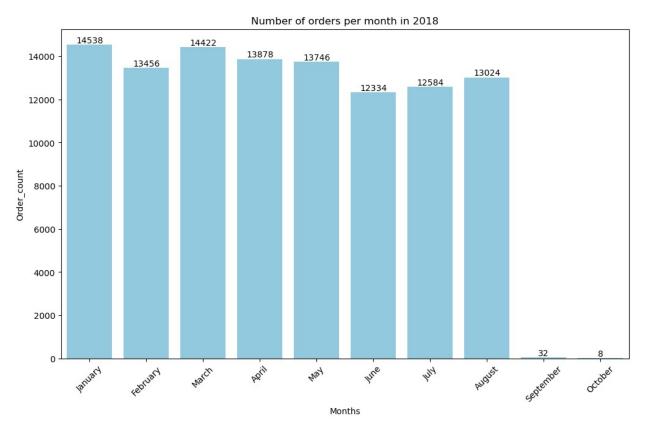
```
query = """select monthname(order_purchase_timestamp) as Months ,
count(order_id) as Number_of_order
from orders where year(order_purchase_timestamp) = 2018
group by Months
order by Months"""

cur.execute(query)

dt = cur.fetchall()

df = pd.DataFrame(dt, columns = ["Months","Order_count"])
df
```

```
Order_count
      Months
0
       April
                     13878
1
      August
                     13024
2
    February
                     13456
3
     January
                     14538
4
        July
                     12584
5
        June
                     12334
6
       March
                     14422
7
                     13746
         May
8
     October 0
                         8
9
                        32
   September
od = ["January", "February", "March", "April", "May", "June", "July",
"August", "September", "October"]
plt.figure(figsize = (12,7))
ax = sns.barplot(x = "Months", y = "Order_count", data = df, order = df)
od, color = "skyblue")
plt.xticks(rotation = 45)
ax.bar_label(ax.containers[0])
plt.title("Number of orders per month in 2018")
plt.show()
```



Average number of products per order, grouped by customer city

```
query = """SELECT
    customers.customer city,
    ROUND(AVG(order_count.oc), 2) AS average_od
FROM
    customers
        JOIN
    (SELECT
        orders.order id,
            orders.customer id,
            COUNT(order items.order item id) AS oc
    FROM
        orders
    JOIN order items ON orders.order id = order items.order id
    GROUP BY orders.order_id , orders.customer_id) AS order_count ON
customers.customer_id = order_count.customer i\overline{d}
GROUP BY customers.customer city
order by average_od desc"""
cur.execute(query)
dt = cur.fetchall()
df = pd.DataFrame(dt, columns = ["City", "Avg order"])
df
                City Avg order
      padre carvalho
                         28.00
1
         celso ramos
                         26.00
2
               datas
                         24.00
3
       candido godoi
                         24.00
4
      matias olimpio
                         20.00
4105
              tuiuti
                          4.00
                          4.00
4106 aurora do para
4107
                          4.00
        nova america
4108
        tibau do sul
                          4.00
4109 lagoa da canoa
                          4.00
[4110 rows x 2 columns]
```

Percentage of total revenue contributed by each product category

```
query = """select p.product_category,
round((sum(oi.price + oi.freight_value) / (select sum(price) +
sum(freight_value) from order_items)) *100,2) as
percentage_of_total_revenue
from order_items as oi
join products as p
on oi.product_id = p.product_id
group by p.product_category
```

```
order by percentage of total revenue desc"""
cur.execute(query)
dt = cur.fetchall()
df = pd.DataFrame(dt, columns =
["Category", "Percentage contribution"])
df
                        Category
                                  Percentage contribution
0
                  HEALTH BEAUTY
                                                     18.19
1
                Watches present
                                                     16.48
2
                                                     15.67
                 bed table bath
3
                  sport leisure
                                                     14.60
4
           computer accessories
                                                     13.37
69
                         flowers
                                                      0.02
70
    Fashion Children's Clothing
                                                      0.01
71
                House Comfort 2
                                                      0.01
72
                 cds music dvds
                                                      0.01
73
         insurance and services
                                                      0.00
[74 rows x 2 columns]
```

Correlation between product price and the number of times a product has been purchased.

```
query = """select pd.product category, count(od.product id) as
order count,
round(avg(od.price),2) as avg price
from products as pd join order items as od
on pd.product id = od.product id
group by pd.product_category"""
cur.execute(query)
dt = cur.fetchall()
df = pd.DataFrame(dt, columns =
["Category", "order count", "avg price"])
arr1 = df["order count"]
arr2 = df["avg price"]
a = np.corrcoef([arr1,arr2])
print("Correlation between product price and the number of times a
product has been purchased is :",a)
Correlation between product price and the number of times a product
has been purchased is : [[ 1. -0.10631514]
 [-0.10631514 1.
```

Total revenue generated by each seller, and rank them by revenue.

```
guery = """select *, dense rank() over(order by revenue by seller
desc) as Rankk from
(select oi.seller_id , sum(p.payment_value) as revenue_by_seller
from order items as oi join payments as p
on oi.order id = p.order id
group by oi.seller id) as sp;"""
cur.execute(query)
dt = cur.fetchall()
df = pd.DataFrame(dt, columns = ["seller id", "Revenue", "Rank"])
print(df)
fd = df.head(5)
print(fd)
                             seller id
                                             Revenue Rank
      7c67e1448b00f6e969d365cea6b010ab 2.028668e+06
0
                                                         1
1
      1025f0e2d44d7041d6cf58b6550e0bfa
                                       1.232888e+06
                                                         2
2
      4a3ca9315b744ce9f8e9374361493884 1.204981e+06
                                                         3
3
      1f50f920176fa81dab994f9023523100
                                        1.161014e+06
                                                         4
4
      53243585a1d6dc2643021fd1853d8905 1.139612e+06
                                                         5
. . .
3090 ad14615bdd492b01b0d97922e87cb87f 7.684000e+01 3082
3091
     702835e4b785b67a084280efca355756 7.424000e+01 3083
3092 4965a7002cca77301c82d3f91b82e1a9 6.544000e+01 3084
3093 77128dec4bec4878c37ab7d6169d6f26 6.088000e+01 3085
3094 cf6f6bc4df3999b9c6440f124fb2f687 4.888000e+01 3086
[3095 \text{ rows } x \text{ 3 columns}]
                          seller id
                                          Revenue Rank
  7c67e1448b00f6e969d365cea6b010ab 2.028668e+06
                                                      1
                                                      2
  1025f0e2d44d7041d6cf58b6550e0bfa 1.232888e+06
                                                      3
2 4a3ca9315b744ce9f8e9374361493884 1.204981e+06
  1f50f920176fa81dab994f9023523100 1.161014e+06
                                                      4
4 53243585a1d6dc2643021fd1853d8905 1.139612e+06
                                                      5
```

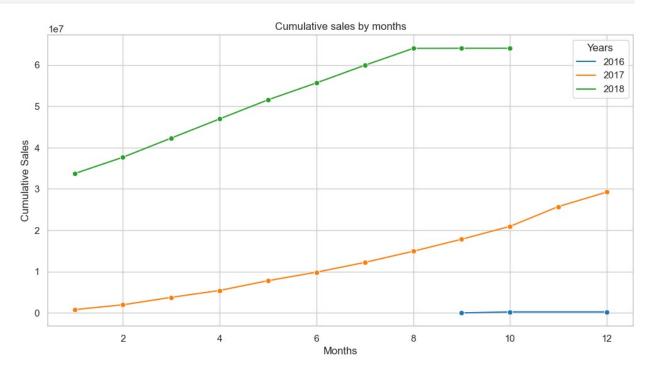
Moving average of order values for each customer over their order history.

```
dt = cur.fetchall()
df = pd.DataFrame(dt, columns =
["Customer id", "order purchase timestamp", "payments", "Moving
Average"1)
df
                              Customer id order purchase timestamp
payments \
        00012a2ce6f8dcda20d059ce98491703
                                               2017-11-14 16:08:26
114.74
        00012a2ce6f8dcda20d059ce98491703
                                               2017-11-14 16:08:26
114.74
        00012a2ce6f8dcda20d059ce98491703
                                               2017-11-14 16:08:26
2
114.74
        00012a2ce6f8dcda20d059ce98491703
                                               2017-11-14 16:08:26
114.74
        000161a058600d5901f007fab4c27140
                                               2017-07-16 09:40:32
4
67.41
. . .
415539 ffffa3172527f765de70084a7e53aae8
                                               2017-09-02 11:53:32
45.50
                                               2017-09-29 14:07:03
415540
        ffffe8b65bbe3087b653a978c870db99
18.37
415541
        ffffe8b65bbe3087b653a978c870db99
                                               2017-09-29 14:07:03
18.37
415542
        ffffe8b65bbe3087b653a978c870db99
                                               2017-09-29 14:07:03
18.37
415543
        ffffe8b65bbe3087b653a978c870db99
                                               2017-09-29 14:07:03
18.37
        Moving Average
0
                114.74
1
                114.74
2
                114.74
3
                114.74
4
                 67.41
                 45.50
415539
                 18.37
415540
415541
                 18.37
415542
                 18.37
                 18.37
415543
[415544 rows x 4 columns]
```

Cumulative sales per month for each year.

```
guery = """select Years, Months, Sales,
round(sum(Sales) over(order by Years, Months),2) as Cumulative sales
from
(select year(od.order purchase timestamp) as Years,
month(od.order purchase timestamp) as Months,
round(sum(p.payment_value),2) as Sales
from orders as od join payments as p
on od.order id = p.order id
group by Years, Months
order by Years, Months) as cu; """
cur.execute(query)
dt = cur.fetchall()
df = pd.DataFrame(dt, columns = ["Years", "Months", "Sales", "Cumulative
Sales"1)
df
                                Cumulative Sales
    Years
           Months
                         Sales
0
     2016
                9
                       1008.96
                                          1008.96
1
     2016
               10
                     236361.92
                                        237370.88
2
     2016
               12
                         78.48
                                        237449.36
3
     2017
                1
                     553952.16
                                        791401.52
4
     2017
                2
                    1167632.04
                                       1959033.56
5
                3
     2017
                    1799454.40
                                      3758487.96
6
     2017
                4
                   1671152.12
                                       5429640.08
7
                5
     2017
                    2371675.28
                                      7801315.36
8
     2017
                6
                   2045105.52
                                      9846420.88
9
                7
     2017
                    2369531.68
                                      12215952.56
10
     2017
                8
                    2697585.28
                                      14913537.84
11
     2017
                9
                   2911049.80
                                      17824587.64
12
     2017
               10
                    3118711.52
                                      20943299.16
13
     2017
               11
                   4779531.20
                                      25722830.36
14
               12
                    3513605.92
     2017
                                      29236436.28
15
     2018
                1
                    4460016.72
                                      33696453.00
16
     2018
                2
                   3969853.36
                                      37666306.36
17
                3
     2018
                   4638608.48
                                      42304914.84
18
     2018
                4
                   4643141.92
                                      46948056.76
19
                5
     2018
                   4615928.60
                                      51563985.36
20
     2018
                6
                   4095522.00
                                      55659507.36
21
                7
     2018
                   4266163.00
                                      59925670.36
22
     2018
                8
                   4089701.29
                                      64015371.65
23
                9
     2018
                      17758.16
                                      64033129.81
24
     2018
               10
                       2358.68
                                      64035488.49
plt.figure(figsize = (12,6))
sns.set(style = 'whitegrid')
sns.lineplot(x = 'Months', y = 'Cumulative Sales', hue = 'Years', data
= df, marker = "o", palette = "tab10", legend = "full")
```

```
plt.title("Cumulative sales by months")
plt.show()
```



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

```
query = """with year sales as (select
year(od.order_purchase_timestamp) as Years,
round(sum(p.payment value),2) as Sales
from orders as od join payments as p
on od.order id = p.order id
group by Years
order by Years)
select years, Sales, lag(Sales, 1) over(order by Years) as
previous year sales,
round((Sales - lag(Sales, 1) over(order by Years))/lag(Sales, 1)
over(order by Years) * 100,2) as yoy growth Percent
from year sales"""
cur.execute(query)
dt = cur.fetchall()
df = pd.DataFrame(dt, columns = ["Year", "Sales", "Previous year
sale","yoy Growth %"])
df
               Sales Previous year sale yoy Growth %
   Year
0 2016
           237449.36
                                     NaN
                                                   NaN
```

```
      1
      2017
      28998986.91
      237449.36
      12112.7

      2
      2018
      34799052.21
      28998986.91
      20.0
```

Calculate the retention rate of customers, defined as the percentage of customers who make another purchase within 6 months of their first purchase.

```
query = """with fo as (select customer id,
min(order purchase timestamp) as first order
from orders
group by customer id),
co as (select fo.customer id,
count(distinct(od.order purchase timestamp)) as count of order
from fo join orders as od
on fo.customer id = od.customer id
and od.order purchase timestamp > first order
and od.order purchase timestamp < date add(first order, interval 6
month)
group by fo.customer id)
select (count(co.customer id) / count(fo.customer id))*100 as
Retentaion rate
from fo join co
on fo.customer id = co.customer id"""
cur.execute(query)
dt = cur.fetchall()
df = pd.DataFrame(dt, columns = ["Retention rate"])
df
# Since none of our customers buys again with us in 6 months so we
have no Retention rate
 Retention rate
0
            None
```

Top 3 customers who spent the most money in each year.

```
query = """select * from
  (select year(od.order_purchase_timestamp) as Years, od.customer_id,
  sum(p.payment_value) as payment,
  dense_rank() over(partition by year(od.order_purchase_timestamp) order
  by sum(p.payment_value) desc) as Rankk
  from orders as od join payments as p
  on od.order_id = p.order_id
  group by year(od.order_purchase_timestamp), od.customer_id) as rn
  where Rankk <= 3"""

cur.execute(query)
  dt = cur.fetchall()</pre>
```

```
df = pd.DataFrame(dt, columns = ["Year", "Customer
ID", "Payments", "Rank"])
plt.figure(figsize = (10,6))
sns.barplot(x = "Customer ID", y = "Payments", hue = "Year", data = df)
plt.xticks(rotation = 90)
plt.show()
```

