List all unique cities where customers are located.

Count the number of orders placed in 2017.

```
query = """ select count(order_id) from orders where
year(order_purchase_timestamp) = 2017 """

cur.execute(query)

data = cur.fetchall()

"total order placed in 2017 are", data [0] [0]

('total order placed in 2017 are', 135303)
```

Find the total sales per category.

```
query = """ select upper(products.product category) category,
round(sum(payment value),2) sales
from products join order items
on products.product id = order items.product id
join payments
on payments.order id = order items.order id
GROUP BY products.product category
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns = ["Category", "Sales"] )
print (df)
                                        Sales
                       Category
                                 13681943.82
0
                      PERFUMERY
1
           FURNITURE DECORATION 38614762.58
2
                      TELEPHONY 13145815.37
3
                 BED TABLE BATH 46238949.12
                     AUTOMOTIVE 23011946.94
4
                 CDS MUSIC DVDS
69
                                    32384.61
                     LA CUISINE
70
                                    78665.31
71
    FASHION CHILDREN'S CLOTHING
                                    21213.09
                       PC GAMER
72
                                    58709.61
73
         INSURANCE AND SERVICES
                                     8761.77
[74 rows x 2 columns]
```

Calculate the percentage of orders that were paid in installments.

```
query = """ select (sum(case when payment_installments >= 1 then 1
else 0 end))/count(*)*100 from payments

cur.execute(query)

data = cur.fetchall()
"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'))
```

Count the number of customers from each state.

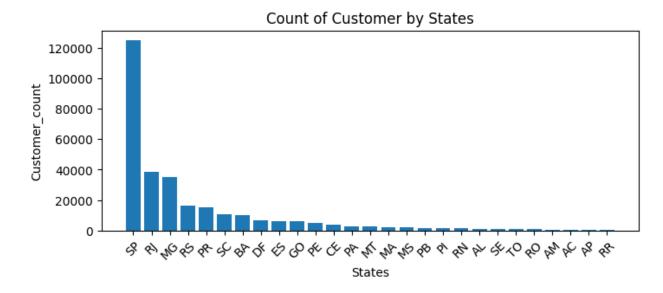
```
query = """ select customer_state ,count(customer_id)
from customers group by customer_state"""

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 = (8,3))
plt.bar(df["state"],df["customer_count"])
plt.xticks(rotation = 45)
plt.xlabel("States")
plt.ylabel("Customer_count")
plt.title("Count of Customer by States")
plt.show

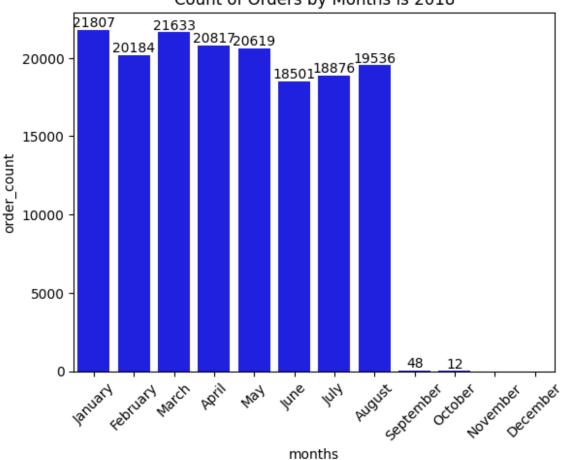
<function matplotlib.pyplot.show(close=None, block=None)>
```



Calculate the number of orders per month in 2018

```
query = """ select monthname(order_purchase_timestamp) months,
count(order_id) order_count
from orders where year(order_purchase_timestamp) = 2018
```

Count of Orders by Months is 2018



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

```
query = """ WITH count per order AS
(select orders.order id, orders.customer id,
count(order items.order id) as oc
from orders join order items
ON orders.order id = order items.order id
group by orders.order id, orders.customer id)
select customers.customer city, ROUND(AVG(count per order.oc), 2) AS
average orders
from customers join count per order
on customers.customer_id = count_per_order.customer_id
group by customers.customer city
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns = ["Customer City", "Average
Products/Orders"])
df.head(10)
         Customer City Average Products/Orders
             sao paulo
                                          10.41
1
   sao jose dos campos
                                          10.25
                                          10.57
          porto alegre
3
               indaial
                                          10.04
4
          treze tilias
                                          11.45
5
        rio de janeiro
                                          10.32
6
                                          12.00
          mario campos
7
                                           9.00
               guariba
8
                cuiaba
                                          10.78
9
                franca
                                          11.26
```

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

```
query = """ select upper(products.product_category) category,
ROUND((SUM(payment_value) / (SELECT SUM(payment_value) FROM payments))
* 100, 2) sales_percentage
from products join order_items
on products.product_id = order_items.product_id
join payments
on payments.order_id = order_items.order_id
GROUP BY products.product_category ORDER BY sales_percentage DESC;
```

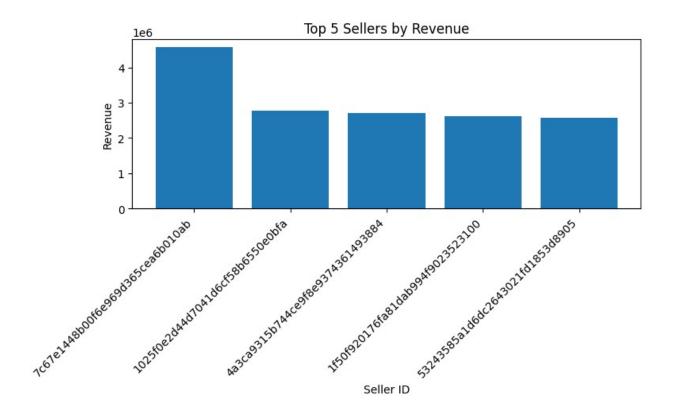
```
0.00
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns = ["Category", "Percentage
Distribution"])
df.head(10)
                          Percentage Distribution
               Category
0
         BED TABLE BATH
                                             96.28
          HEALTH BEAUTY
1
                                             93.18
  COMPUTER ACCESSORIES
                                             89.13
3
   FURNITURE DECORATION
                                             80.40
        WATCHES PRESENT
                                             80.35
5
          SPORT LEISURE
                                             78.26
6
                                             61.55
             HOUSEWARES
7
                                             47.91
             AUTOMOTIVE
8
           GARDEN TOOLS
                                             47.13
             COOL STUFF
                                             43.83
```

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

```
query = """SELECT
products.product category, COUNT(order items.product id) AS
total orders,
ROUND(AVG(order_items.price), 2) AS avg_price
FROM products JOIN order items ON products.product id =
order items.product id
GROUP BY products.product category; """
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns = ["Category", "order count",
"avg_price"])
arr1 = df["order count"].values
arr2 = df["avg_price"].values
corr mat = np.corrcoef(arr1, arr2)
print("The Correlation is", corr_mat[0, 1])
The Correlation is -0.10631514167157556
```

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

```
query = """
SELECT
    DENSE RANK() OVER (ORDER BY revenue DESC) AS rn
FROM (
    SELECT
        oi.seller id,
        SUM(p.payment value) AS revenue
    FROM order items AS oi
    JOIN payments AS p
      ON oi.order id = p.order id
    GROUP BY
        oi.seller id
) AS sales per seller;
# Execute and fetch
cur.execute(query)
data = cur.fetchall()
# assume `data` is already fetched:
df = pd.DataFrame(data, columns=["seller id", "revenue", "rank"])
df top5 = df.head()
fig, ax = plt.subplots(figsize=(8, 5))
df top5 = df.nsmallest(5, "rank")
# Define the tick positions
x pos = range(len(df top5))
ax.bar(x pos, df top5["revenue"])
# First tell Matplotlib where the ticks go...
ax.set xticks(x pos)
# ...then what their labels are
ax.set xticklabels(df top5["seller id"], rotation=45, ha="right")
ax.set xlabel("Seller ID")
ax.set_ylabel("Revenue")
ax.set title("Top 5 Sellers by Revenue")
plt.subplots adjust(bottom=0.25)
plt.tight layout()
plt.show()
```



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

```
# 1) Define a properly balanced, multi-line SQL string:
query = """
SELECT
    customer id,
    order purchase timestamp,
    payment,
    AVG(payment) OVER (
        PARTITION BY customer id
        ORDER BY order_purchase_timestamp
        ROWS BETWEEN 2 PRECEDING AND CURRENT ROW
    ) AS mov_avg
FROM (
    SELECT
        o.customer id,
        o.order purchase timestamp,
        p.payment_value AS payment
    FROM payments AS p
    JOIN orders AS o
      ON p.order id = o.order id
) AS a
ORDER BY customer id, order purchase timestamp;
```

```
0.00
# 2) Execute and fetch:
cur.execute(query)
data = cur.fetchall()
# 3) Load into a DataFrame (with column names matching the SELECT
list):
df = pd.DataFrame(
    data,
    columns=[
        "customer id",
        "order_purchase_timestamp",
        "payment", "mov avg" ]
)
# 4) Inspect the first few rows:
print(df.head())
                        customer id order purchase timestamp
                                                               payment
  00012a2ce6f8dcda20d059ce98491703
                                         2017-11-14 16:08:26
                                                                114.74
1 00012a2ce6f8dcda20d059ce98491703
                                         2017-11-14 16:08:26
                                                                114.74
2 00012a2ce6f8dcda20d059ce98491703
                                         2017-11-14 16:08:26
                                                                114.74
                                         2017-11-14 16:08:26
3 00012a2ce6f8dcda20d059ce98491703
                                                                114.74
4 00012a2ce6f8dcda20d059ce98491703
                                         2017-11-14 16:08:26
                                                                114.74
      mov avg
  114.739998
  114.739998
  114.739998
3 114.739998
4 114.739998
```

Calculate the cumulative sales per month for each year.

```
query = """ SELECT
    years,
    months,
    payment,
    SUM(payment) OVER (
        ORDER BY years, months
```

```
) AS cumulative sales
FROM (
    SELECT
        YEAR(o.order purchase timestamp)
                                            AS years,
        MONTH(o.order purchase timestamp) AS months,
        ROUND(SUM(p.payment value), 2)
                                            AS payment
    FROM orders AS o
    JOIN payments AS p
      ON o.order id = p.order id
    GROUP BY
        YEAR(o.order purchase timestamp),
        MONTH(o.order purchase timestamp)
) AS monthly_totals
ORDER BY
    years,
    months;
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(
    data,
    columns=[
        "years",
        "months"
        "payment", "cumulative sales" ]
print(df)
           months
                                  cumulative sales
    years
                        payment
0
     2016
                 9
                        2270.16
                                      2.270160e+03
1
     2016
                10
                      531814.32
                                      5.340845e+05
2
     2016
                12
                         176.58
                                      5.342611e+05
3
     2017
                 1
                     1246392.36
                                      1.780653e+06
4
                 2
     2017
                     2627172.09
                                      4.407826e+06
5
                 3
     2017
                     4048772.40
                                      8.456598e+06
6
     2017
                 4
                     3760092.27
                                      1.221669e+07
7
                 5
     2017
                     5336269.38
                                      1.755296e+07
8
     2017
                 6
                     4601487.42
                                      2.215445e+07
9
                 7
     2017
                                      2.748589e+07
                     5331446.28
10
                 8
     2017
                     6069566.88
                                      3.355546e+07
11
     2017
                 9
                     6549862.05
                                      4.010532e+07
12
                10
                                      4.712242e+07
     2017
                     7017100.92
13
     2017
                11
                    10753945.20
                                      5.787637e+07
14
     2017
                12
                                      6.578198e+07
                     7905613.32
15
     2018
                 1
                    10035037.61
                                      7.581702e+07
                 2
16
     2018
                     8932170.06
                                      8.474919e+07
                 3
17
     2018
                    10436869.08
                                      9.518606e+07
18
     2018
                 4
                                      1.056331e+08
                    10447069.33
                 5
19
     2018
                                      1.160190e+08
                    10385839.36
```

```
20
    2018
                   9214924.51
                                    1.252339e+08
    2018
               7
21
                   9598866.74
                                    1.348328e+08
22
    2018
               8
                   9201827.89
                                    1.440346e+08
23
    2018
               9
                      39955.86
                                    1.440745e+08
24
    2018
               10
                       5307.03
                                    1.440798e+08
```

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

```
query = """
WITH a AS (
    SELECT
        YEAR(o.order_purchase_timestamp) AS years,
        ROUND(SUM(p.payment value), 2)
                                           AS payment
    FROM orders AS o
    JOIN payments AS p
      ON o.order id = p.order id
    GROUP BY YEAR(o.order purchase timestamp)
SELECT
    years,
    -- compute (this year - last year) / last year * 100
    ((payment
       - LAG(payment, 1) OVER (ORDER BY years)
    / LAG(payment, 1) OVER (ORDER BY years)
    ) * 100
                                         AS yoy pct growth
FROM a
ORDER BY years;
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(
    columns=["years", "yoy_pct_growth"]
)
print(df)
   years yoy_pct_growth
    2016
                     NaN
1
    2017
            12112.703758
    2018
               20.000924
```

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 a AS (
    SELECT
        c.customer id,
        MIN(o.order_purchase_timestamp) AS first_order
    FROM customers AS c
    JOIN orders AS o
      ON c.customer id = o.customer id
    GROUP BY c.customer id
),
b AS (
    SELECT
        a.customer id
    FROM a
    JOIN orders AS o
      ON o.customer id = a.customer id
     AND o.order purchase timestamp > a.first order
     AND o.order_purchase_timestamp < DATE ADD(a.first order, INTERVAL
    GROUP BY a.customer id
SELECT
    100.0
    * COUNT(DISTINCT b.customer id)
    / COUNT(DISTINCT a.customer_id) AS retention rate
FROM a
LEFT JOIN b
 ON a.customer id = b.customer id;
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns=["retention rate"])
print(df)
# Since None of our Customers is Repated That's why our Retention Rate
is 0
  retention rate
         0.00000
```

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

```
query = """
SELECT
    years,
    customer id,
    payment,
    d_rank
FROM (
    SELECT
        YEAR(o.order purchase timestamp) AS years,
        o.customer_id,
        SUM(p.payment value) AS payment,
        DENSE RANK() OVER (
            PARTITION BY YEAR(o.order purchase timestamp)
            ORDER BY SUM(p.payment value) DESC
        ) AS d rank
    FROM orders AS o
    JOIN payments AS p
      ON o.order id = p.order id
    GROUP BY
        YEAR(o.order_purchase_timestamp),
        o.customer id
) AS ranked payments
WHERE d rank <= 3
ORDER BY years, d rank;
cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns=["year", "customer_id", "payment",
"d rank"])
plt.figure(figsize=(10, 6))
sns.barplot(
    data=df,
    x="year",
    y="payment",
    hue="customer id",
    dodge=True
plt.title("Top 3 Customers by Total Payment per Year")
plt.xlabel("Year")
plt.ylabel("Total Payment")
plt.legend(title="Customer ID", bbox to anchor=(1.05, 1), loc="upper
left")
```

plt.tight_layout()
plt.show()

