List all unique cities where customers are located.

```
In [11]:
    query = """ select distinct customer_city from customers"""
    cur.execute(query)
    data = cur.fetchall()
    data

Out[11]: [('franca',),
        ('sao bernardo do campo',),
        ('sao paulo',),
        ('mogi das cruzes',),
        ('campinas',),
        ('jaragua do sul',),
        ('timoteo')
```

2. Count the number of orders placed in 2017.

```
In [14]:
    query = """ select count(order_id) from orders where year(order_purchase_timestamp) = 2017 """
    cur.execute(query)
    data = cur.fetchall()
    "total orders placed in 2017 are", data[0][0]
```

Out[14]: ('total orders placed in 2017 are', 225505)

Find the total sales per category.

```
In [2]:
    query = """ select upper(products.product_category) category,
    round(sum(payments.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 category
    """

    cur.execute(query)
    data = cur.fetchall()

    df =pd.DataFrame(data, columns = ["Category", "Sales"])
    df
```

Out[2]:	Category		Sales
	0	PERFUMERY	3.243127e+07
	1	ELIPNITURE DECORATION	0.1521200+07

Calculate the percentage of orders that were paid in installments.

```
In [30]:
    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",data[0][0]
```

Out[30]: ('the percentage of orders that were paid in installments', Decimal('99.9981'))

5. Count the number of customers from each state.

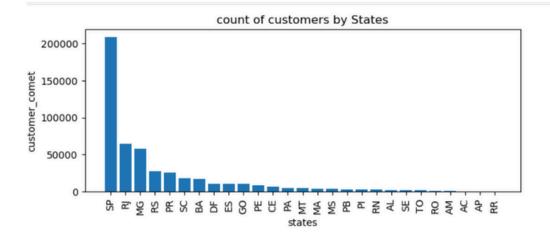
```
In [17]:
    query = """ select customer_state ,count(customer_id)
    from customers group by customer_state
    """

    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 = 90)
    plt.xlabel("states")
    plt.ylabel("customer_comet")
    plt.ylabel("customer_comet")
    plt.title("count of customers by States")

plt.show()
```



1. Calculate the number of orders per month in 2018.

1. Calculate the number of orders per month in 2018.

```
In [5]:
             import pandas as pd
             import matplotlib.pyplot as plt
             import seaborn as sns
             import mysql.connector
             db = mysql.connector.connect(host = "localhost",
                                                  username = "root",
                                                  password = "maharshi@1234",
                                                  database = "ecommerce")
             cur = db.cursor()
In [19]:
             query = """ select monthname(order_purchase_timestamp) months, count(order_id) order_count
             from orders where year(order_purchase_timestamp) = 2018
             group by months
             cur.execute(query)
             data = cur.fetchall()
            data = Cdf. Tetchal()

df = pd.DataFrame(data, columns = ["months","order_count"])

o = ["january","feburary","March","April","May","June","July","August","September","October"]

ax = sns.barplot(x = df["months"],y = df["order_count"],data = df, order = o, color = "red")
             plt.xticks(rotation = 45)
             ax.bar_label(ax.containers[0])
             plt.title("Count of Orders by Months 2018")
             plt.show()
```

Count of Orders by Months 2018 36055 34695 34365 35000 32560 30835 31460 30000 10000 5000 September 0 August Reburary March MIN POIN May months

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

```
In []:
    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) 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/order"])
    df.head(10)
```

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

```
In [26]:
In [30]:
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
          import mysql.connector
          db = mysql.connector.connect(host = "localhost",
                                       username = "root",
                                       password = "maharshi@1234",
                                       database = "ecommerce")
          cur = db.cursor()
          query = """
          select upper(products.product_category) category,
          round((sum(payments.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 category order by sales_percentage desc
          cur.execute(query)
          data = cur.fetchall()
          df = pd.DataFrame(data,columns = ["Category", "percentage distribution"])
          df.head()
```

Out[30]:

Category percentage distribution

0	BED TABLE BATH	171.16
1	HEALTH BEAUTY	165.65
2	COMPUTER ACCESSORIES	158.45

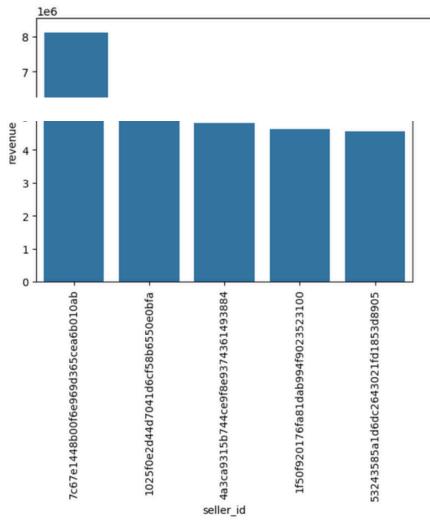
Identify the correlation between product price and the number of times a product has been purchased.

```
In [40]:
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
          import mysql.connector
          import numpy as np
          db = mysql.connector.connect(host = "localhost",
                                       username = "root",
                                       password = "maharshi@1234",
                                       database = "ecommerce")
          cur = db.cursor()
          query = """select products.product_category,
          count(order_items.product_id),
          round(avg(order_items.price),2)
          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","price"])
          arr1 = df["order_count"]
          arr2 = df["price"]
          a= np.corrcoef([arr1,arr2])
          print("the correleation between price and number of times a product has been purchased is", a[0][1])
```

the correleation between price and number of times a product has been purchased is -0.10631514167157563

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

```
In [8]:
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         import mysql.connector
         import numpy as np
         db = mysql.connector.connect(host = "localhost",
                                      username = "root",
                                      password = "maharshi@1234",
                                      database = "ecommerce")
         cur = db.cursor()
         query = """ select *, dense_rank() over(order by revenue desc) as rn from
         (select order_items.seller_id, sum(payments.payment_value)
         revenue from order_items join payments
         on order_items.order_id = payments.order_id
         group by order_items.seller_id) as a"""
         cur.execute(query)
         data = cur.fetchall()
         df = pd.DataFrame(data, columns = ["seller_id", "revenue", "rank"])
         sns.barplot(x = "seller_id", y = "revenue", data = df)
         plt.xticks(rotation = 90)
         plt.show()
         data
```



```
Out[8]: [('7c67e1448b00f6e969d365cea6b010ab', 8114670.516834259, 1), ('1025f0e2d44d7041d6cf58b6550e0bfa', 4931552.6374435425, 2), ('4a3ca9315b744ce9f8e9374361493884', 4819924.316244602, 3), ('1f50f920176fa81dab994f9023523100', 4644054.722041845, 4), ('53243585a1d6dc2643021fd1853d8905', 4558449.287963867, 5), ('da862b14ab17ac2831f4ac5b9dab84a', 4355500, 109034538, 6)
```

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

Advanced_question

```
In [6]:
        import pandas as pd
        import matplotlib.pyplot as plt
         import seaborn as sns
        import mysql.connector
         import numpy as np
        db = mysql.connector.connect(host = "localhost",
                                    username = "root",
                                    password = "maharshi@1234",
                                    database = "ecommerce")
        cur = db.cursor()
        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
         (select orders.customer_id, orders.order_purchase_timestamp,
        payments.payment_value as payment
         from payments join orders
        on payments.order_id = orders.order_id) as a"""
        cur.execute(query)
        data = cur.fetchall()
        df = pd.DataFrame(data)
        df
Out[6]:
              0 00012a2ce6f8dcda20d059ce98491703 2017-11-14 16:08:26 114.74 114.739998
              1 00012a2ce6f8dcda20d059ce98491703 2017-11-14 16:08:26 114.74 114.739998
                                                                              18.57
   20///15
              TITIE8DD5DDE3U87D5538978C87UQD99 ZU17-U9-Z9 14:U7:U5
                                                                                       18.370001
   2077716
              ffffe8b65bbe3087b653a978c870db99 2017-09-29 14:07:03
                                                                              18.37
                                                                                       18.370001
   2077717
              ffffe8b65bbe3087b653a978c870db99 2017-09-29 14:07:03
                                                                                       18.370001
                                                                              18.37
              ffffe8b65bbe3087b653a978c870db99 2017-09-29 14:07:03
   2077718
                                                                              18.37
                                                                                       18.370001
              ffffe8b65bbe3087b653a978c870db99 2017-09-29 14:07:03
                                                                              18.37
                                                                                       18.370001
  2077719
```

2077720 rows × 4 columns

[7]:

2016

0

9

calculate the cummulative sales per month for each year

2

5044.80 5.044800e+03

```
query = """select years, months , payment, sum(payment)
over(order by years, months) cumulativesales from
(select year(orders.order_purchase_timestamp) as years,
month(orders.order_purchase_timestamp) as months,
round(sum(payments.payment_value),2) as payment from orders 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)
df
```

3

calculate the year over year growth rate of total sales.

```
In [2]:
         query = """with a as(select year(orders.order_purchase_timestamp) as years,
         round(sum(payments.payment_value),2) as payment from orders join payments
         on orders.order_id = payments.order_id
         group by years order by years)
         select years, ((payment - lag(payment, 1) over(order by years))/
         lag(payment,1) over(order by years)) * 100 from a"""
         cur.execute(query)
         data = cur.fetchall()
         df = pd.DataFrame(data, columns = ["years", "yoy % growth" ])
Out[2]:
           years yoy % growth
         0 2016
                          NaN
         1 2017
                  12112.703758
```

calculate the retention rate of customers, defined as the percentage of customers who make another purchase

```
query = """with a as (select customers.customer_id,
min(orders.order_purchase_timestamp) first_order
from customers join orders
on customers.customer_id = orders.customer_id
group by customers.customer_id),
b as (select a.customer_id, count(distinct orders.order_purchase_timestamp) next_order
from a join orders
on orders.customer_id = a.customer_id
and orders.order_purchase_timestamp > first_order
and orders.order_purchase_timestamp <
date_add(first_order, interval 6 month)
group by a.customer_id)
select 100 * (count( distinct a.customer_id)/ count(distinct b.customer_id))
from a left join b
on a.customer_id = b.customer_id ;"""
cur.execute(query)
data = cur.fetchall()
```

: [(None,)]

2 2018

20.000924

identify the top3 customers who spent the most money in each year.

```
9]:
     query = """select years, customer_id, payment, d_rank
     from
     (select year(orders.order_purchase_timestamp) years,
     orders.customer_id,
     sum(payments.payment_value) payment,
     dense_rank() over(partition by year(orders.order_purchase_timestamp)
     order by sum(payments.payment_value) desc) d_rank
     from orders join payments
     on payments.order_id = orders.order_id
     group by year(orders.order_purchase_timestamp),
     orders.customer_id) as a
     where d_rank <= 3 ;"""
     cur.execute(query)
     data = cur.fetchall()
     df = pd.DataFrame(data, columns = ["years", "id", "payment", "rank"])
     sns.barplot(x = "id", y = "payment", data = df, hue = "years")
     plt.xticks(rotation = 90)
     plt.show()
```

Out[19]: <function matplotlib.pyplot.show(close=None, block=None)>

9

