```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import calendar
from datetime import datetime
import numpy as np
import mysql.connector

db = mysql.connector.connect(host = "localhost",
                             user = "root",
                             password = "Nivetha-arunjp",
                             database = "ecommerce")

cur = db.cursor()
```

1. List unique cities where customers are located.

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

cur.execute(query)

data = cur.fetchall()
df = pd.DataFrame(data, columns = ['Customer_City'])
df.head()
```

```
           Customer_City
0                 franca
1  sao bernardo do campo
2              sao paulo
3        mogi das cruzes
4               campinas
```

2. 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 orders placed in 2017 are", data[0][0]
```

```
('total orders placed in 2017 are', 45101)
```

3. Find the total sales per category.

```
query = """ WITH distinct_order_items AS (
    SELECT DISTINCT order_id, product_id, price
    FROM order_items
),
distinct_payments AS (
    SELECT DISTINCT order_id, payment_value
    FROM payments
),
item_payment_combined AS (
    SELECT doi.product_id, dp.payment_value
    FROM distinct_order_items doi
    JOIN distinct_payments dp ON doi.order_id = dp.order_id
)

SELECT
    p.product_category AS category,
    ROUND(SUM(ipc.payment_value), 2) AS total_sales
FROM item_payment_combined ipc
JOIN products p ON ipc.product_id = p.product_id
GROUP BY p.product_category """

cur.execute(query)

data = cur.fetchall()
df = pd.DataFrame(data, columns = ['Category', 'Sales'])
df
```

	Category	Sales
0	telephony	406549.62
1	bed table bath	1453936.81
2	computer accessories	1146724.92
3	Furniture Decoration	1021687.60
4	perfumery	469457.76
..	...	...
69	cds music dvds	954.99
70	La Cuisine	2579.41
71	Fashion Children's Clothing	785.67
72	PC Gamer	1717.05
73	insurance and services	324.51

[74 rows x 2 columns]

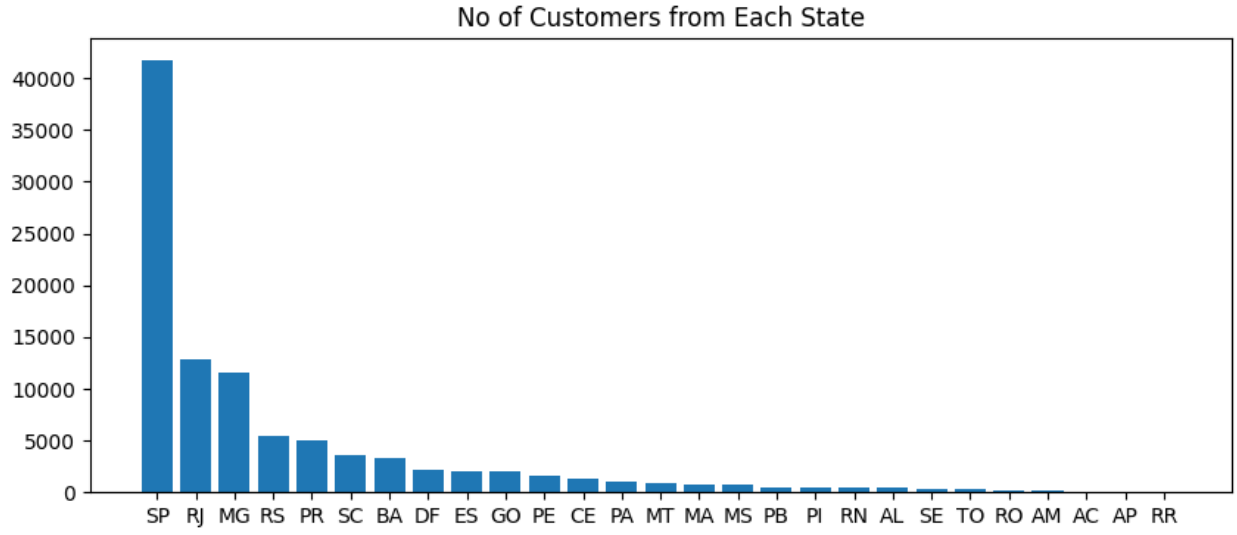
4. Calculate the percentage of orders that were paid in installments.

```
query = """ select (sum(payment_installments >=1))/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'))
```

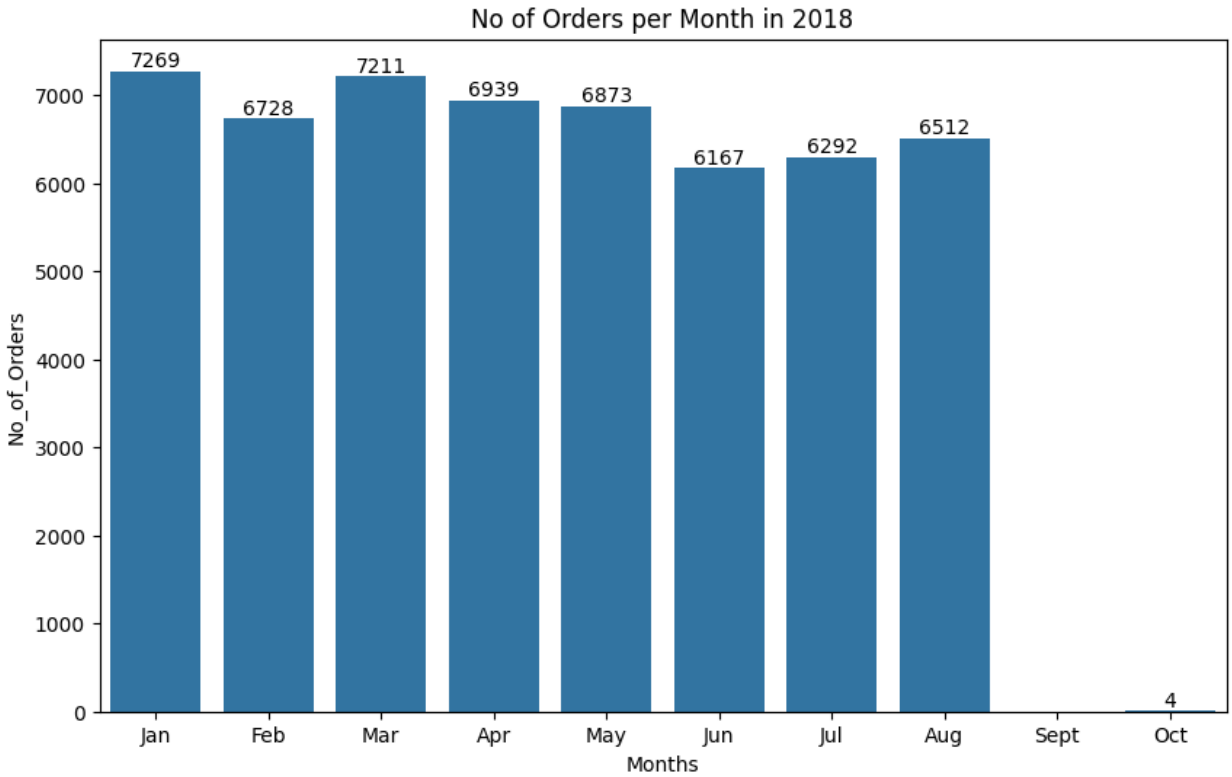
5. Count the number of customers from each state.

```
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=(10,4))
plt.bar(df['State'], df['Customer_Count'])
plt.title("No of Customers from Each State")
plt.show()
```



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

```
query = """ select monthname(order_purchase_timestamp),
count(order_id)
from orders where order_purchase_timestamp = 2018
group by monthname(order_purchase_timestamp)"""

cur.execute(query)

data = cur.fetchall()
df = pd.DataFrame(data, columns = ['Months','No_of_Orders'])
df['Months'] = df['Months'].apply(lambda x: datetime.strptime(x,
'%B').strftime('%b'))

month_order= ['Jan', 'Feb', 'Mar', 'Apr', 'May', 'Jun', 'Jul', 'Aug',
'Sept', 'Oct']

plt.figure(figsize=(10,6))
ax = sns.barplot(x = df["Months"], y = df["No_of_Orders"], data = df,
order = month_order)
ax.bar_label(ax.containers[0])
plt.title("No of Orders per Month in 2018")
plt.show()
```



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

```
query = """WITH distinct_order_items AS (
    SELECT DISTINCT order_id, product_id
    FROM order_items
),
product_count_per_order AS (
    SELECT order_id, COUNT(product_id) AS product_count
    FROM distinct_order_items
    GROUP BY order_id
),
order_customer_city AS (
    SELECT o.order_id, c.customer_city
    FROM orders o
    JOIN customers c ON o.customer_id = c.customer_id
)

SELECT
    occ.customer_city,
    ROUND(AVG(pco.product_count), 2) AS avg_products_per_order
FROM product_count_per_order pco
JOIN order_customer_city occ ON pco.order_id = occ.order_id
GROUP BY occ.customer_city
ORDER BY avg_products_per_order DESC """

cur.execute(query)

data = cur.fetchall()
df = pd.DataFrame(data, columns = ['Customer_City', 'Average_Orders'])
df.head(10)
```

	Customer_City	Average_Orders
0	padre carvalho	7.00
1	picarra	4.00
2	curralinho	4.00
3	teixeira soares	3.00
4	ipua	2.75
5	celso ramos	2.50
6	pauliceia	2.00
7	nova hartz	2.00
8	doutor ulysses	2.00
9	senador cortes	2.00

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

```
query = """ WITH distinct_order_items AS (
    SELECT DISTINCT order_id, product_id
    FROM order_items
),
distinct_payments AS (
    SELECT DISTINCT order_id, payment_value
    FROM payments
),
joined_data AS (
    SELECT doi.product_id, dp.payment_value
    FROM distinct_order_items doi
    JOIN distinct_payments dp ON doi.order_id = dp.order_id
),
category_sales AS (
    SELECT
        p.product_category AS category,
        SUM(jd.payment_value) AS category_revenue
    FROM joined_data jd
    JOIN products p ON jd.product_id = p.product_id
    GROUP BY p.product_category
),
total_revenue AS (
    SELECT SUM(category_revenue) AS total FROM category_sales
)

SELECT
    cs.category,
    ROUND((cs.category_revenue * 100.0 / tr.total), 2) AS
revenue_percentage
FROM category_sales cs, total_revenue tr
ORDER BY revenue_percentage DESC;
 """



cur.execute(query)

data = cur.fetchall()
df = pd.DataFrame(data, columns = ['Category', 'Percentage
Distribution'])
df.head(10)
```

```
               Category  Percentage Distribution
0         HEALTH BEAUTY                     8.88
1        bed table bath                     8.65
2       Watches present                     8.03
```

```
3         sport leisure                      7.15
4  computer accessories                      6.82
5  Furniture Decoration                      6.08
6             housewares                     4.89
7             Cool Stuff                     4.37
8             automotive                     4.26
9           Garden tools                     3.69
```

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

```
query = """ select p.product_category, count(distinct
oi.product_id,oi.order_id), round(avg(oi.price),2)
from products p
join order_items oi on
p.product_id = oi.product_id
group by p.product_category order by p.product_category desc """

cur.execute(query)

data = cur.fetchall()
df = pd.DataFrame(data, columns = ['Category', 'Order Count',
'Price'])
df.head()

arr1 = df['Order Count']
arr2 = df['Price']

a = np.corrcoef([arr1,arr2])
print(" the correlation between the product price and the number of
times a product purchased is", a[0][1])
```

```
 the correlation between the product price and the number of times a
product purchased is -0.10448921163702476
```

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

```
query = """ WITH distinct_order_items AS (
    SELECT DISTINCT order_id, seller_id, product_id
    FROM order_items
),
```

```
distinct_payments AS (
    SELECT DISTINCT order_id, payment_value
    FROM payments
),
seller_revenue_data AS (
    SELECT doi.seller_id, dp.payment_value
    FROM distinct_order_items doi
    JOIN distinct_payments dp ON doi.order_id = dp.order_id
),
seller_revenue AS (
    SELECT
        s.seller_id,
        SUM(srd.payment_value) AS total_revenue
    FROM seller_revenue_data srd
    JOIN sellers s ON srd.seller_id = s.seller_id
    GROUP BY s.seller_id
)

SELECT
    seller_id,
    ROUND(total_revenue, 2) AS total_revenue,
    RANK() OVER (ORDER BY total_revenue DESC) AS revenue_rank
FROM seller_revenue
ORDER BY revenue_rank """

cur.execute(query)
data = cur.fetchall()
df = pd. DataFrame(data, columns = ['Seller_ID','Revenue','Rank'])
df.head()
```

	Seller_ID	Revenue	Rank
0	4a3ca9315b744ce9f8e9374361493884	271189.74	1
1	7c67e1448b00f6e969d365cea6b010ab	257388.49	2
2	4869f7a5dfa277a7dca6462dcf3b52b2	257263.34	3
3	53243585a1d6dc2643021fd1853d8905	234594.81	4
4	da8622b14eb17ae2831f4ac5b9dab84a	231693.69	5

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

```
query = """ WITH distinct_orders AS (
    SELECT DISTINCT
        o.customer_id,
        o.order_id,
        p.payment_value,
        o.order_purchase_timestamp
    FROM orders o
```

```
    JOIN payments p ON o.order_id = p.order_id
),
ordered_orders AS (
    SELECT
        customer_id,
        order_id,
        payment_value,
        order_purchase_timestamp,
        ROW_NUMBER() OVER (PARTITION BY customer_id ORDER BY
order_purchase_timestamp) AS order_num
    FROM distinct_orders
),
moving_avg AS (
    SELECT
        customer_id,
        order_purchase_timestamp,
        payment_value,
        ROUND(AVG(payment_value) OVER (
            PARTITION BY customer_id
            ORDER BY order_num
            ROWS BETWEEN 2 PRECEDING AND CURRENT ROW
        ), 2) AS moving_avg_order_value
    FROM ordered_orders
)

SELECT * FROM moving_avg
ORDER BY customer_id """

cur.execute(query)
data = cur.fetchall()
df = pd. DataFrame(data, columns =
['Customer_ID','Order_Timestamp','Price','Moving Average'])
df
```

```
                              Customer_ID      Order_Timestamp   Price
\
0       00012a2ce6f8dcda20d059ce98491703  2017-11-14 16:08:26  114.74

1       000161a058600d5901f007fab4c27140  2017-07-16 09:40:32   67.41

2       0001fd6190edaaf884bcaf3d49edf079  2017-02-28 11:06:43  195.42

3       0002414f95344307404f0ace7a26f1d5  2017-08-16 13:09:20  179.35

4       000379cdec625522490c315e70c7a9fb  2018-04-02 13:42:17  107.01

...                                  ...                  ...     ...

103266  fffecc9f79fd8c764f843e9951b11341  2018-03-29 16:59:26    0.64
```

```
103267  fffeda5b6d849fbd39689bb92087f431  2018-05-22 13:36:02   63.13

103268  ffff42319e9b2d713724ae527742af25  2018-06-13 16:57:05  214.13

103269  ffffa3172527f765de70084a7e53aae8  2018-09-02 11:53:32   45.50

103270  ffffe8b65bbe3087b653a978c870db99  2017-09-29 14:07:03   18.37


        Moving Average
0               114.74
1                67.41
2               195.42
3               179.35
4               107.01
...                ...
103266           27.12
103267           63.13
103268          214.13
103269           45.50
103270           18.37

[103271 rows x 4 columns]
```

12. Calculate the cumulative sales per month for each year.

```
query = """ WITH monthly_sales AS (
    SELECT
        YEAR(o.order_purchase_timestamp) AS order_year,
        MONTH(o.order_purchase_timestamp) AS order_month,
        SUM(DISTINCT p.payment_value) AS monthly_sales
    FROM orders o
    JOIN payments p ON o.order_id = p.order_id
    GROUP BY YEAR(o.order_purchase_timestamp),
MONTH(o.order_purchase_timestamp)
),
cumulative_sales AS (
    SELECT
        order_year,
        order_month,
        monthly_sales,
        SUM(monthly_sales) OVER (
            PARTITION BY order_year
            ORDER BY order_month
            ROWS BETWEEN UNBOUNDED PRECEDING AND CURRENT ROW
        ) AS cumulative_sales
```

```
    FROM monthly_sales
)

SELECT
    order_year,
    order_month,
    monthly_sales,
    ROUND(cumulative_sales, 2) AS cumulative_sales
FROM cumulative_sales
ORDER BY order_year, order_month """

cur.execute(query)
data = cur.fetchall()
df = pd. DataFrame(data, columns = ['Year','Month','Sales','Cumulative
Sales'])
df
```

	Year	Month	Sales	Cumulative Sales
0	2016	9	252.239994	252.24
1	2016	10	55379.960037	55632.20
2	2016	12	19.620001	55651.82
3	2017	1	122628.520095	122628.52
4	2017	2	252840.189707	375468.71
5	2017	3	373576.479756	749045.19
6	2017	4	352872.829841	1101918.02
7	2017	5	484773.930227	1586691.95
8	2017	6	395519.030382	1982210.98
9	2017	7	434337.579863	2416548.56
10	2017	8	510292.640042	2926841.20
11	2017	9	566847.230360	3493688.43
12	2017	10	600322.420186	4094010.85
13	2017	11	868663.619959	4962674.47
14	2017	12	627272.669885	5589947.14
15	2018	1	784639.359469	784639.36
16	2018	2	698751.360006	1483390.72
17	2018	3	884408.719452	2367799.44
18	2018	4	882028.050555	3249827.49
19	2018	5	897287.410337	4147114.90
20	2018	6	848065.400578	4995180.30
21	2018	7	846513.729134	5841694.03
22	2018	8	782948.190654	6624642.22
23	2018	9	4260.330132	6628902.55
24	2018	10	589.669998	6629492.22

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

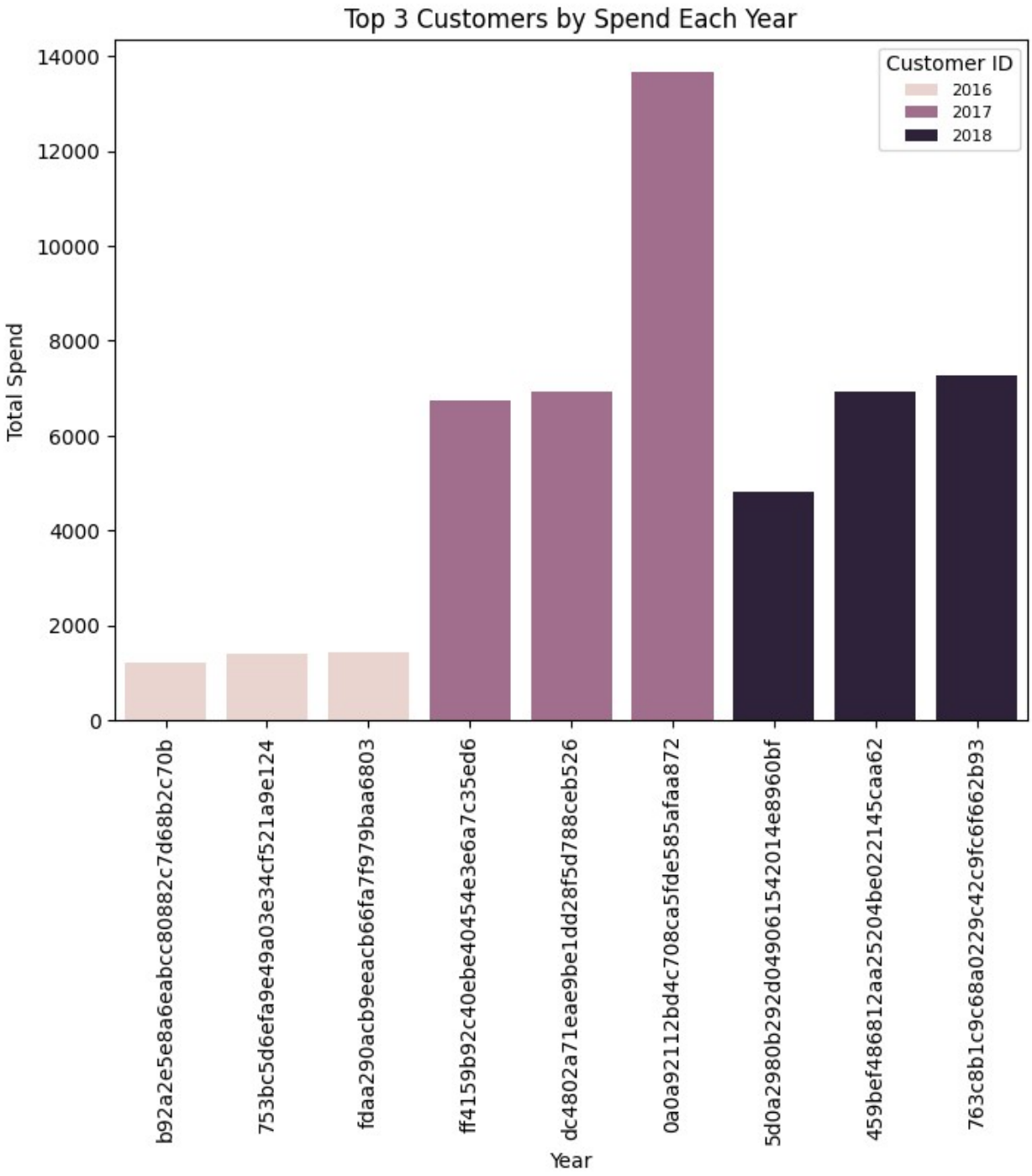
```
query = """ with a as(select year(orders.order_purchase_timestamp) as
years,
round(sum(payments.payment_value),2) as payment from orders join
payments
on orders.order_id = payments.order_id
group by years order by years)

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"])
df
```

	Years	YoY % Growth
0	2016	NaN
1	2017	12112.703761
2	2018	20.000924

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

```
query = """ WITH customer_yearly_spend AS (
    SELECT
        c.customer_id,
        c.customer_unique_id,
        YEAR(o.order_purchase_timestamp) AS order_year,
        ROUND(SUM(DISTINCT p.payment_value), 2) AS total_spent
    FROM customers c
    JOIN orders o ON c.customer_id = o.customer_id
    JOIN payments p ON o.order_id = p.order_id
    GROUP BY c.customer_id, c.customer_unique_id,
YEAR(o.order_purchase_timestamp)
),
ranked_customers AS (
    SELECT *,
        RANK() OVER (PARTITION BY order_year ORDER BY total_spent
DESC) AS ranked
    FROM customer_yearly_spend
)
SELECT
    order_year,
```

```
    customer_unique_id,
    total_spent,
    ranked
FROM ranked_customers
WHERE ranked <= 3
ORDER BY order_year, ranked desc"""

cur.execute(query)
data = cur.fetchall()
df = pd.DataFrame(data, columns =
['Year','Customer_ID','Total_Spend','Rank'])

plt.figure(figsize=(8, 6))
barplot = sns.barplot(data=df, x='Customer_ID', y='Total_Spend',
hue='Year')

plt.legend(title='Customer ID', prop={'size': 8})
plt.title("Top 3 Customers by Spend Each Year")
plt.ylabel("Total Spend")
plt.xlabel("Year")
plt.xticks(rotation=90)
plt.show()
```


Top 3 Customers by Spend Each Year
Customer ID
2016
2017
2018
Total Spend
14000
12000
10000
8000
6000
4000
2000
0
b92a2e5e8a6eabcc80882c7d68b2c70b
753bc5d6efa9e49a03e34cf521a9e124
fdaa290acb9eeacb66fa7f979baa6803
ff4159b92c40ebe40454e3e6a7c35ed6
dc4802a71eae9be1dd28f5d788ceb526
0a0a92112bd4c708ca5fde585afaa872
5d0a2980b292d049061542014e8960bf
459bef486812aa25204be022145caa62
763c8b1c9c68a0229c42c9fc6f662b93
Year

15. 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 first_orders AS (
  SELECT customer_id,
         MIN(order_purchase_timestamp) AS first_order_date
  FROM orders
  GROUP BY customer_id
),
repeat_orders AS (
  SELECT DISTINCT o.customer_id, o.order_id,
o.order_purchase_timestamp
  FROM orders o
  JOIN first_orders f ON o.customer_id = f.customer_id
  WHERE o.order_purchase_timestamp > f.first_order_date
    AND o.order_purchase_timestamp <= DATE_ADD(f.first_order_date,
INTERVAL 6 MONTH)
),
retained_customers AS (
  SELECT DISTINCT customer_id
  FROM repeat_orders
)
SELECT
  ROUND(COUNT(DISTINCT r.customer_id) * 100.0 / COUNT(DISTINCT
f.customer_id), 2) AS retention_rate_percentage
FROM first_orders f
LEFT JOIN retained_customers r ON f.customer_id = r.customer_id """

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

```
[(Decimal('0.00'),)]
```

From the above result we conlcude that none of the customers made a repeat purchase within 6 months of their first purchase.