

E-COMMERCE ANALYSIS PROJECT

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ABOUT PROJECT

This project involves loading e-commerce transaction data from a CSV file into a MySQL database using Python. After establishing the connection, SQL queries are executed through Python to analyze the data and generate actionable insights. The analysis includes solving 15 business queries such as sales trends, customer segmentation, and product performance. The project demonstrates the integration of Python, MySQL, and data visualization tools for comprehensive data analysis and decision-making.

QUERY1

- ✓ List all unique cities where customers are located.

```
▶ query = """ select distinct customer_city from customers """  
  
cur.execute(query)  
  
data = cur.fetchall()  
  
df = pd.DataFrame(data)  
df.head()
```



0

0

franca

1

sao bernardo do campo

2

sao paulo

3

mogi das cruces

4

campinas



QUERY2

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)
```

QUERY3

✓ Find the total sales per category.

```
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  
"""
```

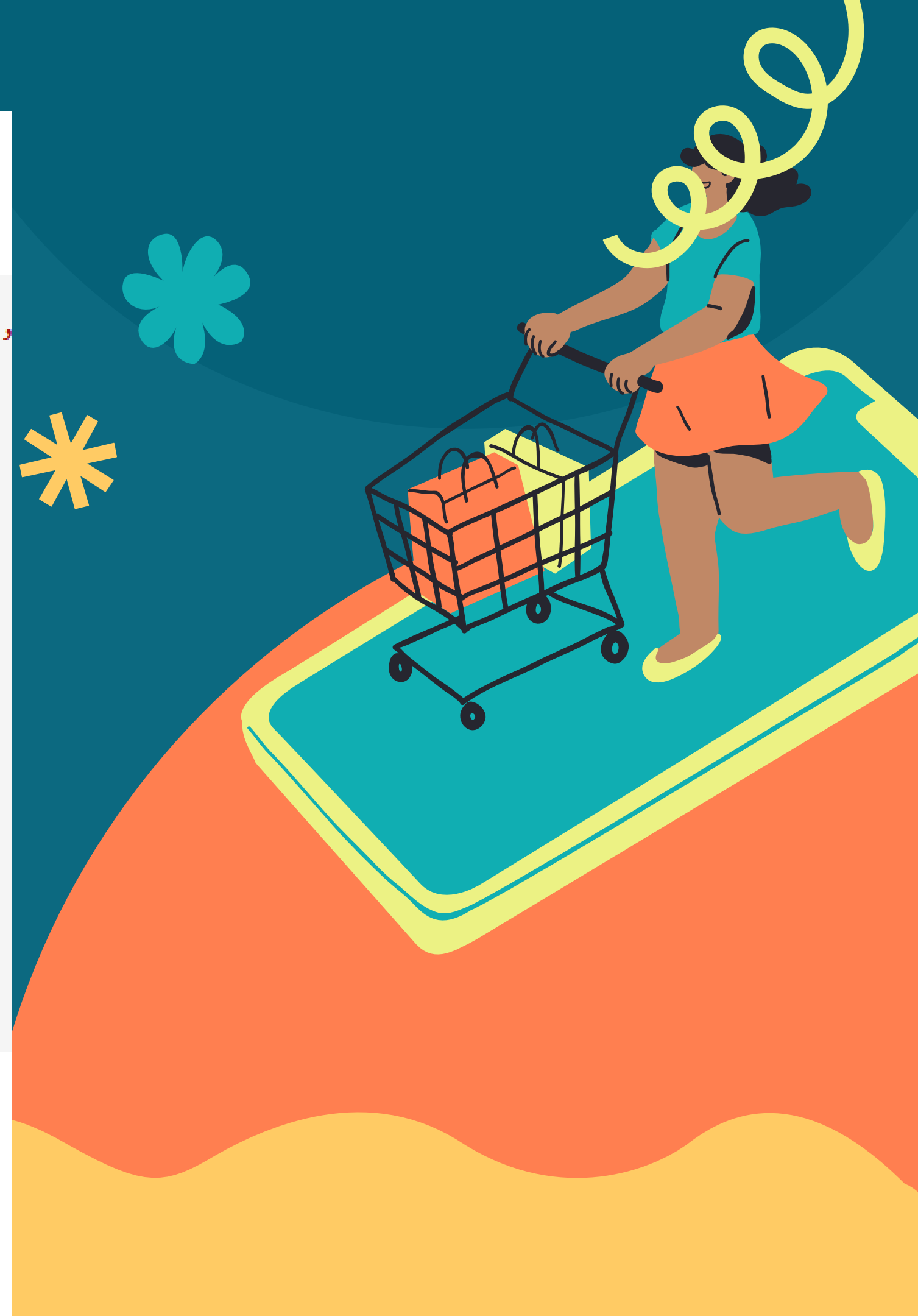
```
cur.execute(query)
```

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

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



	Category	Sales
0	PERFUMERY	506738.66
1	FURNITURE DECORATION	1430176.39



QUERY4

✓ 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'))
```

QUERY5

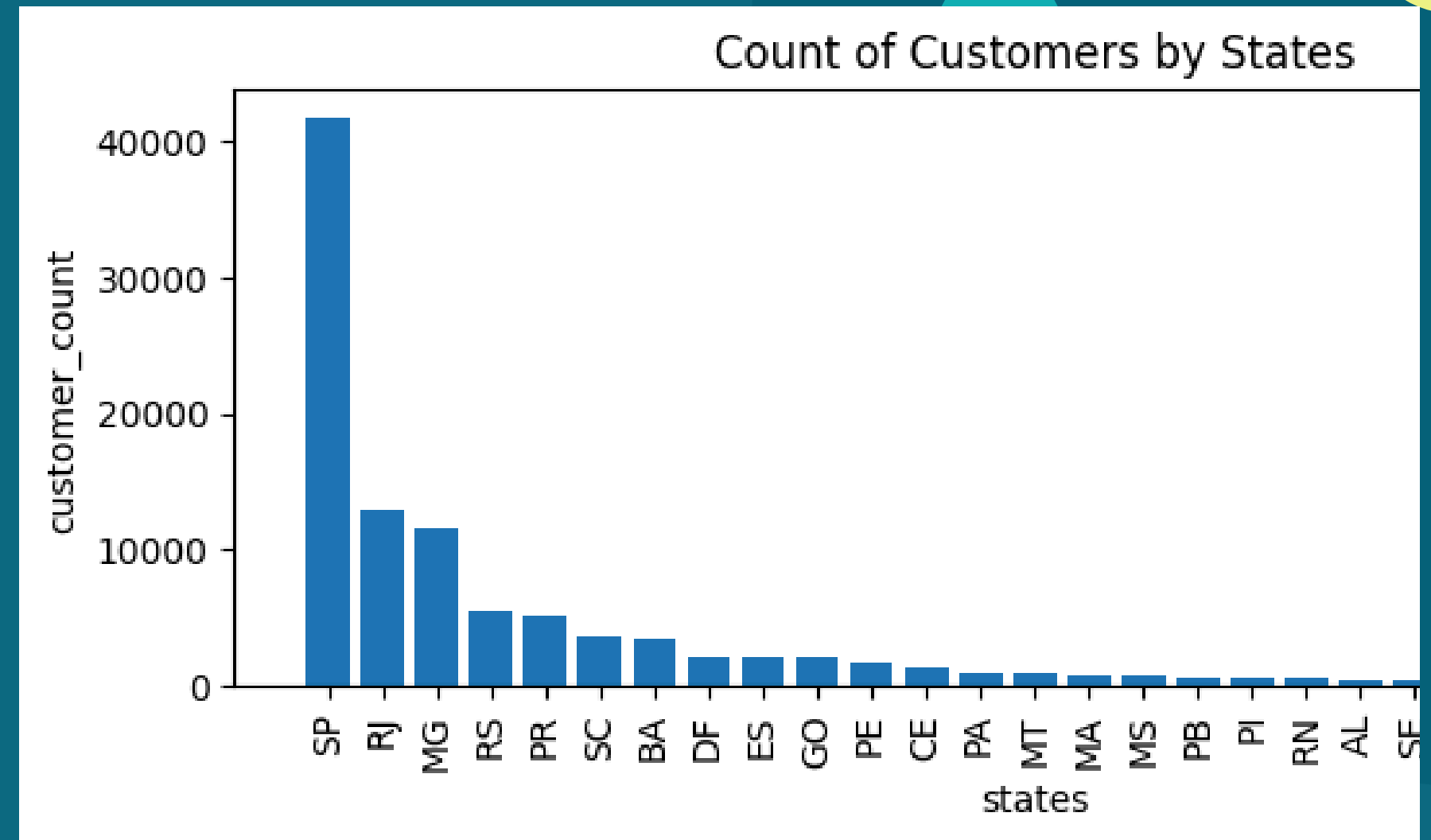
✓ Count the number of customers from each state.

```
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 = 90)
plt.xlabel("states")
plt.ylabel("customer_count")
plt.title("Count of Customers by States")
plt.show()
```



Calculate the number of orders per month in 2018.

QUERY6

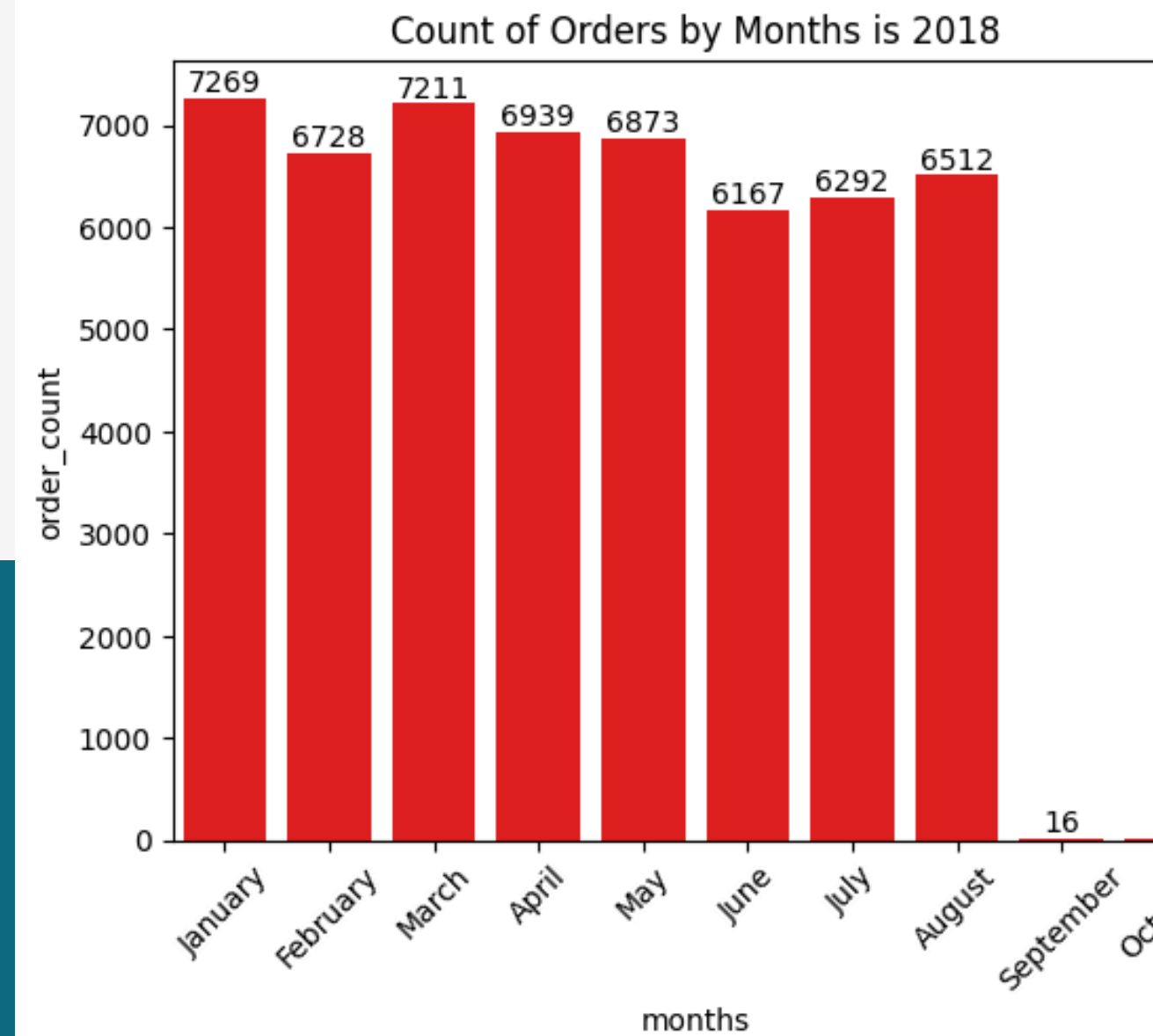
```
query = """ select monthname(order_purchase_timestamp) months, count(order_id) order_count
from orders where year(order_purchase_timestamp) = 2018
group by months
"""

cur.execute(query)

data = cur.fetchall()
df = pd.DataFrame(data, columns = ["months", "order_count"])
o = ["January", "February", "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 is 2018")

plt.show()
```



QUERY7

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) average_orders
from customers join count_per_order
on customers.customer_id = count_per_order.customer_id
group by customers.customer_city order by average_orders desc
"""

cur.execute(query)

data = cur.fetchall()
df = pd.DataFrame(data,columns = ["customer city", "average products/order"])
df.head(10)
```

customer city	average products/order
padre carvalho	7.00
celso ramos	6.50
datas	6.00
candido godoi	6.00
matias olimpiao	5.00
cidelandia	4.00
picarra	4.00
morro de sao paulo	4.00
teixeira soares	4.00
curralinho	4.00

QUERY8

Calculate the percentage of total revenue contributed by each product category

```
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()
```

	Category	percentage distribution
0	BED TABLE BATH	10.70
1	HEALTH BEAUTY	10.35
2	COMPUTER ACCESSORIES	9.90
3	FURNITURE DECORATION	8.93
4	WATCHES PRESENT	8.93

QUERY9

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

```
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 correlation is", a[0][-1])
```

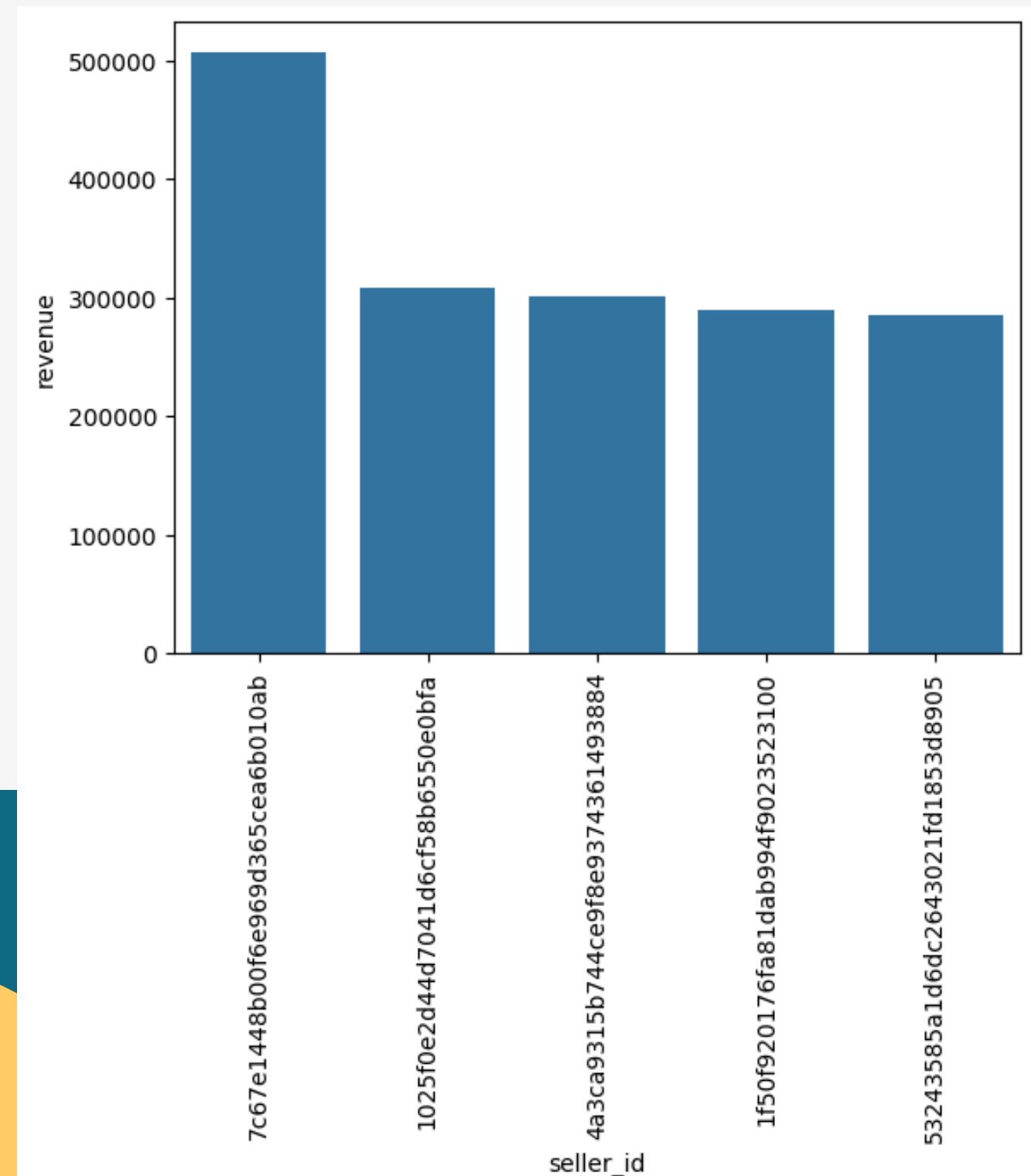
the correlation is -0.10631514167157562

QUERY10

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 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"])
df = df.head()
sns.barplot(x = "seller_id", y = "revenue", data = df)
plt.xticks(rotation = 90)
plt.show()
```



QUERY11

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

```
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 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
```

0	00012a2ce6f8dcda20d059ce98491703	2017-11-14 16:08:26	114.74	114.739998
1	000161a058600d5901f007fab4c27140	2017-07-16 09:40:32	67.41	67.410004
2	0001fd6190edaaf884bcaf3d49edf079	2017-02-28 11:06:43	195.42	195.419998
3	0002414f95344307404f0ace7a26f1d5	2017-08-16 13:09:20	179.35	179.350006
4	000379cdec625522490c315e70c7a9fb	2018-04-02 13:42:17	107.01	107.010002

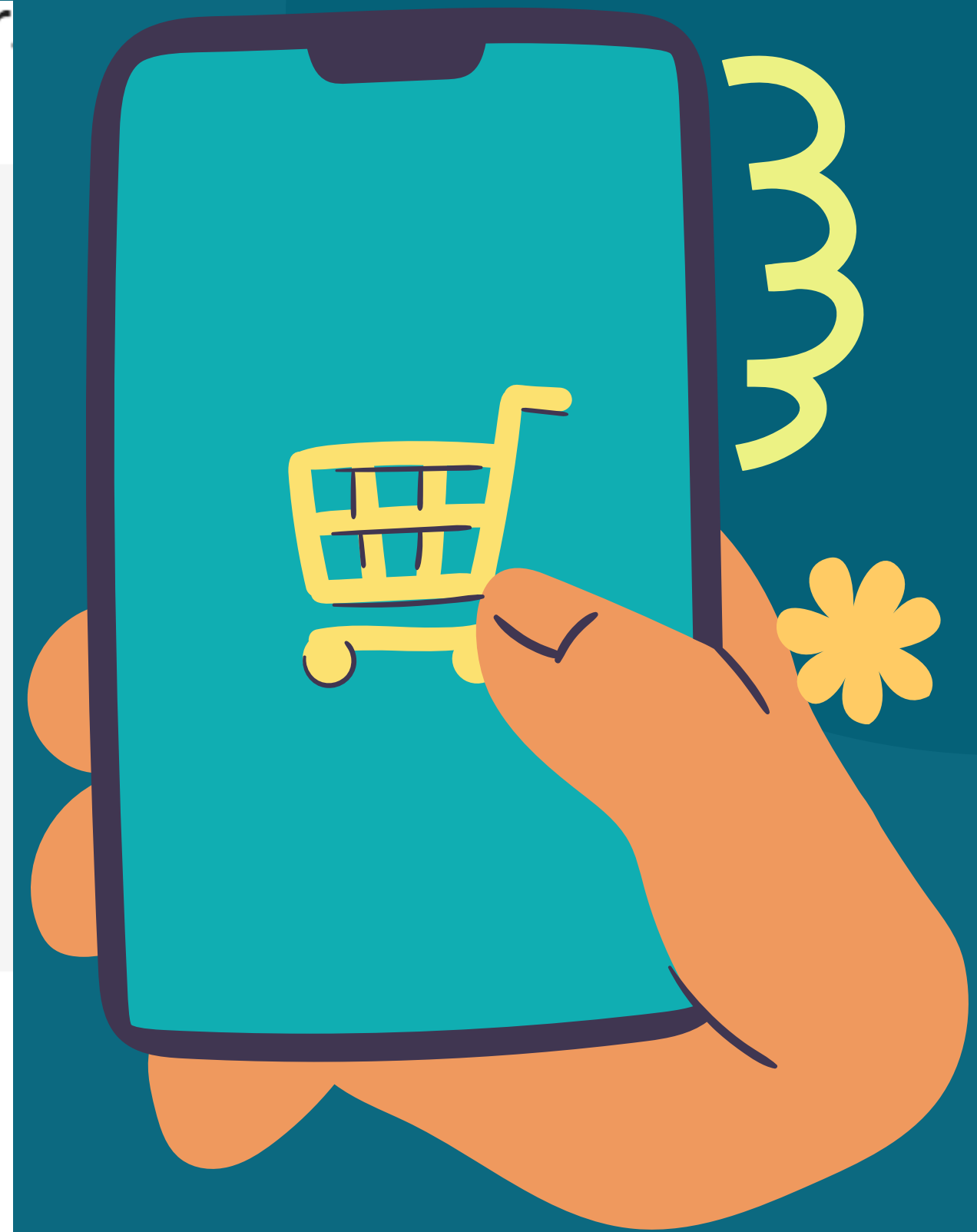
QUERY12

Calculate the cumulative sales per month for each year

```
query = """select years, months , payment, sum(payment)
over(order by years, months) cumulative_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 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
```

	0	1	2	3
0	2016	9	252.24	252.24
1	2016	10	59090.48	59342.72
2	2016	12	19.62	59362.34
3	2017	1	138488.04	197850.38



QUERY13

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

QUERY14

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 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()
```

```
data
```

```
[(None,)]
```

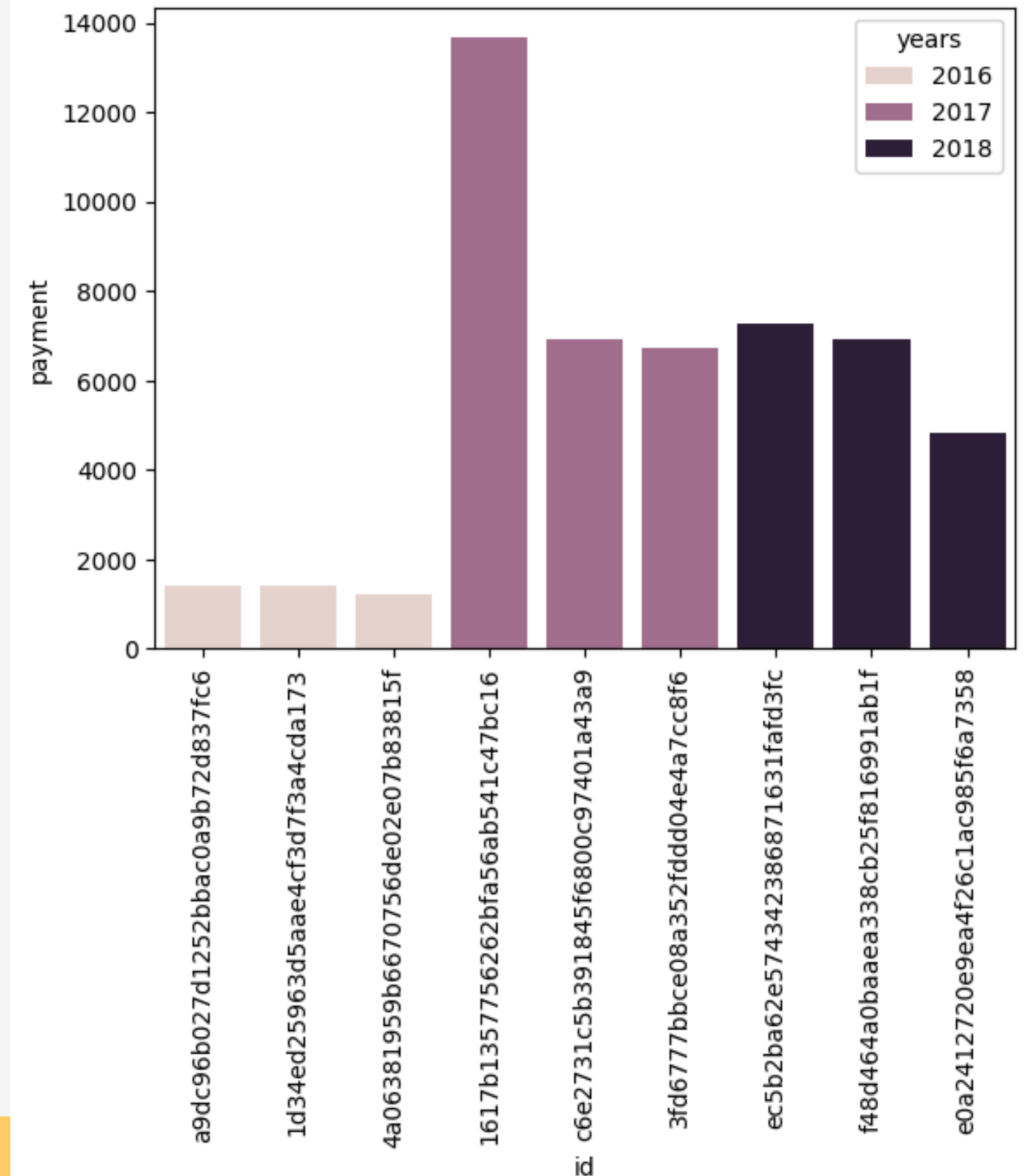

QUERY15

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



```
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()
```



**THANK
YOU**

