```
import pandas as pd
import mysql.connector
import os
# List of CSV files and their corresponding table names
csv files = [
    ('customers.csv', 'customers'),
    ('orders.csv', 'orders'),
    ('orders.csv', 'orders'),
('sellers.csv', 'sellers'),
('products.csv', 'products'),
('geolocation.csv', 'geolocation'),
    ('payments.csv', 'payments'),
    ('order items.csv', 'order items')# Added payments.csv for
specific handling
# Connect to the MySQL database
conn = mysql.connector.connect(
    host='localhost',
    user='root',
    password='Piyush@28',
    database='ecommerce'
)
cursor = conn.cursor()
# Folder containing the CSV files
folder path = 'F:/E-commerce'
def get_sql_type(dtype):
    if pd.api.types.is_integer_dtype(dtype):
         return 'INT'
    elif pd.api.types.is_float_dtype(dtype):
         return 'FLOAT'
    elif pd.api.types.is_bool_dtype(dtype):
         return 'BOOLEAN'
    elif pd.api.types.is datetime64 any dtype(dtype):
        return 'DATETIME'
    else:
         return 'TEXT'
for csv file, table name in csv files:
    file path = os.path.join(folder path, csv file)
    # Read the CSV file into a pandas DataFrame
    df = pd.read csv(file path)
    # Replace NaN with None to handle SQL NULL
    df = df.where(pd.notnull(df), None)
    # Debugging: Check for NaN values
```

```
print(f"Processing {csv file}")
    print(f"NaN values before replacement:\n{df.isnull().sum()}\n")
    # Clean column names
    df.columns = [col.replace(' ', '_').replace('-', '_').replace('.',
' ') for col in df.columns]
    # Generate the CREATE TABLE statement with appropriate data types
    columns = ', '.join([f'`{col}` {get_sql_type(df[col].dtype)}' for
col in df.columns])
    create table query = f'CREATE TABLE IF NOT EXISTS `{table name}`
({columns})'
    cursor.execute(create table query)
    # Insert DataFrame data into the MySQL table
    for , row in df.iterrows():
        # Convert row to tuple and handle NaN/None explicitly
        values = tuple(None if pd.isna(x) else x for x in row)
sql = f"INSERT INTO `{table_name}` ({', '.join(['`' + col +
'`' for col in df.columns])}) VALUES ({', '.join(['%s'] * len(row))})"
        cursor.execute(sql, values)
    # Commit the transaction for the current CSV file
    conn.commit()
# Close the connection
conn.close()
Processing customers.csv
NaN values before replacement:
customer id
customer unique id
                              0
customer zip code prefix
                              0
customer city
                              0
customer state
                              0
dtype: int64
Processing orders.csv
NaN values before replacement:
order id
                                      0
customer id
                                      0
order status
                                      0
order purchase timestamp
                                      0
order approved at
                                    160
order delivered carrier date
                                   1783
order delivered customer_date
                                   2965
order estimated delivery date
                                      0
dtype: int64
Processing sellers.csv
```

```
NaN values before replacement:
seller id
seller_zip_code_prefix
                           0
seller city
                           0
                           0
seller state
dtype: int64
Processing products.csv
NaN values before replacement:
                                 0
product id
product category
                               610
product name length
                               610
product description length
                               610
product photos qty
                               610
product weight q
                                 2
product_length_cm
                                 2
                                 2
product height cm
product width cm
                                 2
dtype: int64
Processing geolocation.csv
NaN values before replacement:
geolocation zip code prefix
                                0
geolocation lat
                                0
                                0
geolocation lng
geolocation city
                                0
                                0
geolocation state
dtype: int64
Processing payments.csv
NaN values before replacement:
order id
                         0
                         0
payment sequential
payment type
                         0
payment installments
                         0
                         0
payment value
dtype: int64
Processing order items.csv
NaN values before replacement:
order_id
                        0
order item id
                        0
product id
                        0
                        0
seller id
                        0
shipping_limit_date
                        0
price
freight value
                        0
dtype: int64
```

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 orders placed in 2017 are", data[0][0]

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

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
                                       Sales
                       Category
0
                      PERFUMERY
                                   506738.66
1
           FURNITURE DECORATION 1430176.39
2
                      TELEPHONY
                                  486882.05
3
                 BED TABLE BATH 1712553.67
4
                     AUTOMOTIVE
                                  852294.33
                 CDS MUSIC DVDS
                                     1199.43
69
70
                     LA CUISINE
                                     2913.53
71
    FASHION CHILDREN'S CLOTHING
                                      785.67
72
                       PC GAMER
                                     2174.43
73
         INSURANCE AND SERVICES
                                      324.51
[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'))
```

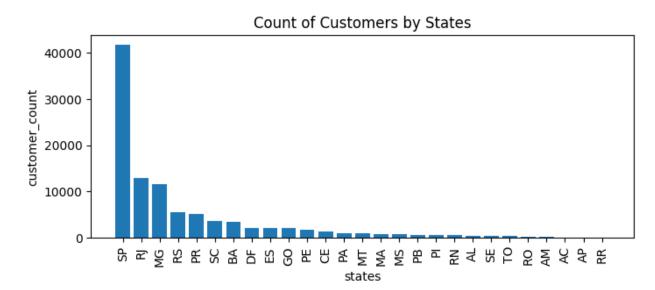
Count the number of customers from each state.

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

cur.execute(query)

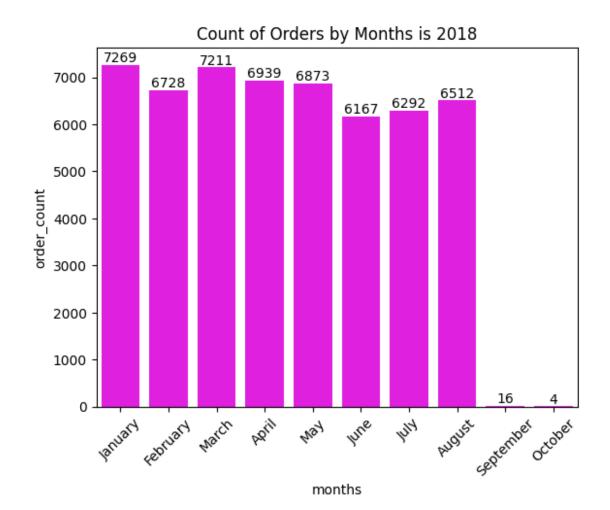
data = cur.fetchall()
df = pd.DataFrame(data, columns = ["state", "customer_count"])
df = df.sort_values(by = "customer_count", ascending= False)

plt.figure(figsize = (8,3))
plt.bar(df["state"], df["customer_count"])
plt.xticks(rotation = 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.

```
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", "O
ctober"1
ax = sns.barplot(x = df["months"],y = df["order_count"], data = df,
order = o, color = "magenta")
plt.xticks(rotation = 45)
ax.bar label(ax.containers[0])
plt.title("Count of Orders by Months is 2018")
plt.show()
```



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
1
                                          6.50
          celso ramos
                datas
                                          6.00
3
        candido godoi
                                          6.00
4
       matias olimpio
                                          5.00
5
           cidelandia
                                          4.00
6
                                          4.00
              picarra
7
   morro de sao paulo
                                          4.00
8
                                          4.00
      teixeira soares
9
           curralinho
                                          4.00
```

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()
                         percentage distribution
               Category
0
         BED TABLE BATH
                                            10.70
          HEALTH BEAUTY
                                            10.35
  COMPUTER ACCESSORIES
                                             9.90
3
                                             8.93
   FURNITURE DECORATION
        WATCHES PRESENT
                                             8.93
```

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"]

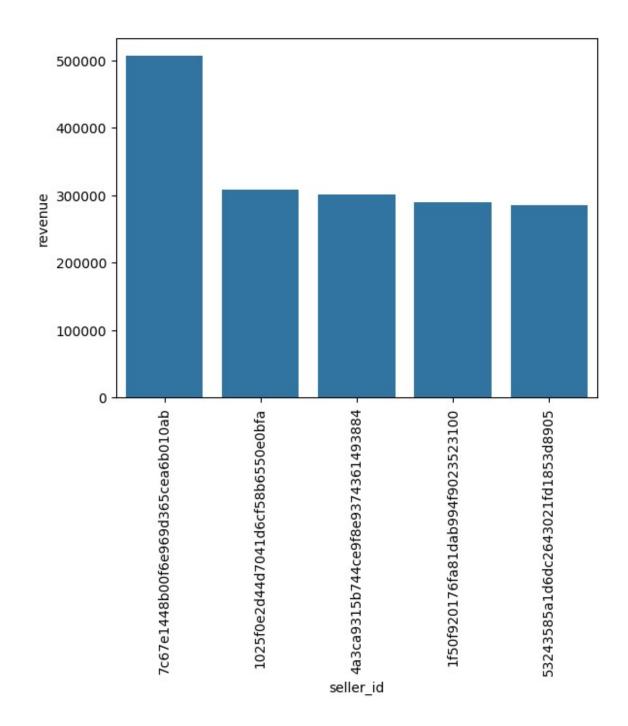
a = np.corrcoef([arr1,arr2])
print("the correlation is", a[0][-1])

the correlation is -0.10631514167157562
```

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()
```



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
        000161a058600d5901f007fab4c27140 2017-07-16 09:40:32
                                                               67.41
2
       0001fd6190edaaf884bcaf3d49edf079 2017-02-28 11:06:43
                                                              195.42
        0002414f95344307404f0ace7a26f1d5 2017-08-16 13:09:20
                                                              179.35
        000379cdec625522490c315e70c7a9fb 2018-04-02 13:42:17
                                                              107.01
103881 fffecc9f79fd8c764f843e9951b11341 2018-03-29 16:59:26
                                                               71.23
103882 fffeda5b6d849fbd39689bb92087f431
                                         2018-05-22 13:36:02
                                                               63.13
103883 ffff42319e9b2d713724ae527742af25 2018-06-13 16:57:05 214.13
103884 ffffa3172527f765de70084a7e53aae8 2017-09-02 11:53:32
                                                               45.50
103885 ffffe8b65bbe3087b653a978c870db99 2017-09-29 14:07:03
                                                               18.37
        114.739998
0
1
        67.410004
2
        195.419998
3
        179.350006
4
        107.010002
103881
        27.120001
        63.130001
103882
103883 214.130005
        45.500000
103884
103885
        18.370001
[103886 rows x 4 columns]
```

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
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
4
    2017
           2
                291908.01
                             489758.39
5
    2017
           3
                449863.60
                             939621.99
6
    2017
           4
                417788.03
                            1357410.02
7
           5
    2017
                592918.82
                            1950328.84
8
    2017
           6
                511276.38
                            2461605.22
9
           7
    2017
                592382.92
                            3053988.14
10
    2017
           8
                674396.32
                            3728384.46
           9
                            4456146.91
11
    2017
                727762.45
12
    2017
          10
                779677.88
                            5235824.79
13
    2017
              1194882.80
                            6430707.59
          11
14
    2017
          12
                878401.48
                            7309109.07
15
    2018
           1
              1115004.18
                            8424113.25
           2
16
    2018
               992463.34
                            9416576.59
17
    2018
           3
              1159652.12
                           10576228.71
18
           4
    2018
              1160785.48
                           11737014.19
           5
19
    2018
              1153982.15
                           12890996.34
20
    2018
           6
              1023880.50
                           13914876.84
           7
21
    2018
              1066540.75
                           14981417.59
22
           8
               1022425.32
    2018
                           16003842.91
           9
23
    2018
                  4439.54
                           16008282.45
24
    2018
                   589.67
                           16008872.12
          10
```

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
   2017 12112.703761
1
2 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 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)</pre>
```

```
b.customer_id))
from a left join b
on a.customer_id = b.customer_id;"""

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

data
[(None,)]
```

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 \leq 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()
```

