Shopify Data Science Internship Test

By: Vedang Anand Waradpande

MS CS student at Rutgers University

Contact: +1 848 437 1822

Email: vedang.waradpande@gmail.com

Solution

Q1.

a. There are several outliers in the data where the order price is much higher than the general range that order amounts fall into (about 90 to 1000).

The value 3145.13 is the mean of all orders in the month, which is highly skewed because of the outliers.

b. There are much better ways to calculate the AOV. For e.g.:

- Median of median of each shop
- By removing the outliers in the following ways:
 - lacktriangle Remove outliers over the 99th percentile and below the 1st percentile ($\mu \pm 3\sigma$)
 - Don't consider shops with very high order amounts in general
 - Don't consider users with very high order amounts in general

```
The best metric would be median of median of each shop
         c. The value of the best metric is 306.0.
In [99]:
           import pandas as pd
           import matplotlib.pyplot as plt
           df = pd.read_csv("data/2019 Winter Data Science Intern Challenge Data Set - Sheet1.csv")
           df.shape
Out[99]: (5000, 7)
In [100...
           df['created_at'] = pd.to_datetime(df['created_at'])
           df['created_at'].min(), df['created_at'].max()
Out[100... (Timestamp('2017-03-01 00:08:09'), Timestamp('2017-03-30 23:55:35'))
In [101...
           df.groupby('shop_id').agg({'order_amount': 'mean'})['order_amount'].mean()
Out[101... 3136.834087887025
In [102...
           df['order_amount'].describe()
                     5000.000000
         count
Out[102...
                     3145.128000
          mean
                     41282.539349
          std
                       90.000000
          min
          25%
                       163.000000
          50%
                       284.000000
          75%
                       390.000000
                   704000.000000
          max
         Name: order_amount, dtype: float64
In [103...
           df['order_amount'].quantile(q=[0.75, 0.80, 0.90, 0.925, 0.95, 0.975, 0.98, 0.985, 0.99, 0.999, 1.0])
                       390.00
         0.750
Out[103...
          0.800
                       444.00
         0.900
                       531.00
         0.925
                       568.00
         0.950
                       640.00
         0.975
                       760.00
         0.980
                       800.08
         0.985
                       948.18
         0.990
                    25725.00
         0.999
                   704000.00
         1.000
                   704000.00
         Name: order_amount, dtype: float64
         The idea in the next few cells is that we can try to remove credit cards as well because the mean amount in credit cards is very high. However, this is not feasible because credit cards constitute roughly
         34.7% of the order traffic.
In [104...
           df.groupby('payment_method').agg({'order_amount': 'mean'})
```

```
order_amount
Out[104...
           payment_method
                              730.353199
                      cash
                credit_card
                              7461.594813
                      debit
                              966.840215
```

```
In [115...
          df['payment method'].value counts(normalize=True)
```

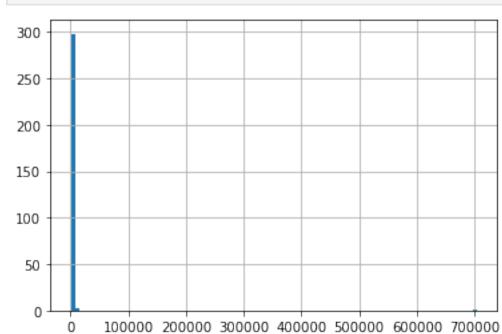
```
credit card
                         0.3470
Out[115...
         debit
                         0.3342
         cash
                         0.3188
         Name: payment_method, dtype: float64
```

Mean of medians of shops

```
df.groupby('shop_id').agg({'order_amount': 'median'})['order_amount'].median()
```

Out[114... 306.0

```
In [107...
          df.groupby('user_id').agg({'order_amount': 'mean'})['order_amount'].hist(bins=100)
          plt.show()
```



```
In [110...
           df['order_amount'].quantile([0.01, 0.99])
Out[110... 0.01
                    101.0
```

```
Mean after outlier removal
```

25725.0

Name: order_amount, dtype: float64

```
In [113...
          first_quantile = df['order_amount'].quantile([0.01, 0.99])[0.01]
          ninety ninth quantile = df['order amount'].quantile([0.01, 0.99])[0.99]
          df[(df['order_amount'] > first_quantile) & (df['order_amount'] < ninety_ninth_quantile)]['order_amount'].mean()</pre>
```

Out[113... 305.0532896085263

0.99

Question 2

2.a.

Code:

```
SELECT
COUNT(Shippers.ShipperName) AS NumOrdersByShipper
FROM
Shippers
LEFT JOIN Orders
ON Shippers.ShipperID = Orders.ShipperID
GROUP BY ShipperName
HAVING Shippers.ShipperName = "Speedy Express";
```

Output:

```
NumOrdersByShipper
-----54
```

2.b.

Code:

```
SELECT Employees.LastName FROM
Employees

LEFT JOIN Orders

ON Employees.EmployeeID = Orders.EmployeeID

GROUP BY Employees.EmployeeID

LIMIT 1;
```

Output:

Question 2

```
LastName
-----
Davolio
```

Count: 29 orders

2.c.

```
SELECT ProductName, COUNT(ProductName) AS ProductCount
FROM
  (
           Products AS pr
            LEFT JOIN OrderDetails AS od
           ON pr.ProductID = od.ProductID
           ) AS pod
          LEFT JOIN Orders AS ord
         ON pod.OrderID = ord.OrderID
        ) AS porders
       LEFT JOIN Customers AS cu
       ON porders.CustomerID = cu.CustomerID
WHERE Country = "Germany"
GROUP BY ProductName
ORDER BY ProductCount DESC
LIMIT 1;
```

Output:

```
ProductName ProductCount
-----Gorgonzola Telino 5
```

Question 2 2