# Fall 2022 Data Science Intern Challenge

### By: Abhishek Kathuria

Please complete the following questions, and provide your thought process/work. You can attach your work in a text file, link, etc. on the application page. Please ensure answers are easily visible for reviewers!

Question 1: Given some sample data, write a program to answer the following: click here to access the required data set

On Shopify, we have exactly 100 sneaker shops, and each of these shops sells only one model of shoe. We want to do some analysis of the average order value (AOV). When we look at orders data over a 30 day window, we naively calculate an AOV of \$3145.13. Given that we know these shops are selling sneakers, a relatively affordable item, something seems wrong with our analysis.

A) Think about what could be going wrong with our calculation. Think about a better way to evaluate this data.

We know that the Average order value (AOV) calculates the average spent every time a customer places an sneaker order.

### Formula:

AOV = sum(order\_amount)/ sum(total\_items)

**Problem:** Using Exploratory Data Analysis and Python libraries such as Pandas, Matplotlib, and Seaborn (which is explained in detail below), I identified the which was an outlier problem. Earlier, AOV was being calculated as the mean of the 'order\_amount' column. But It's value \$3145.12) for a 30 day period was very high for a sneaker store. The explained code shows that an outlier with shop\_id = 78 was committing a fraud by charging \$25725 for every pair of sneakers. Hence, the metrics which was selected as mean to calculate AOV was not appropriate as it was being affected by the values of outliers.

**Better way to evaluate data:** I presented two ways to evaluate the data after handling outlier. The first was either removing outliers and re-calculating the mean of 'order\_amount' column as AOV. However it wouldn't be an optimized approach. Another approach which I would personally choose is using the **median** as the AOV or Average Order Value. Since it is the middle value of the sorted values, it is a central value but large and small outliers do not skew the median as much as the mean.

#### B) What metric would you report for this dataset?

The metric I would use is the Median for the 'order\_amount' column.

### C) What is its value?

The calculated Median value is \$284.0

# **Explanation for Question 1**

## **Import Libraries and Load dataset**

```
In [77]: import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         df=pd.read_csv("Dataset.csv")
         print('Shape of Data ->',df.shape,'\n')
         df.head()
```

Shape of Data -> (5000, 7)

### Out[77]:

	order_id	shop_id	user_id	order_amount	total_items	payment_method	created_at
0	1	53	746	224	2	cash	2017-03-13 12:36:56
1	2	92	925	90	1	cash	2017-03-03 17:38:52
2	3	44	861	144	1	cash	2017-03-14 4:23:56
3	4	18	935	156	1	credit_card	2017-03-26 12:43:37
4	5	18	883	156	1	credit_card	2017-03-01 4:35:11

## **Exploratory Data Analysis or EDA**

```
In [78]: # Check for Nulls and Duplicates
          df.info()
          print('\n\n', 'Checking for duplicates')
          print(df[df.duplicated()])
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 5000 entries, 0 to 4999
          Data columns (total 7 columns):
                           Non-Null Count Dtype
           #
               Column
               ----
                                 -----
                                                  ----
           0
               snop_id 5000 non-null user_id 5000 non-null order_amount 5000 non-null total_items 5000 non-null
               order_id
                                 5000 non-null
                                                   int64
           1
                                                   int64
           2
                                                   int64
           3
                                                   int64
           4
                                                   int64
           5
               payment method 5000 non-null
                                                   object
           6
                                 5000 non-null
               created_at
                                                   object
          dtypes: int64(5), object(2)
          memory usage: 273.6+ KB
           Checking for duplicates
          Empty DataFrame
          Columns: [order_id, shop_id, user_id, order_amount, total_items, payment_metho
          d, created_at]
          Index: []
          Observation: We can see that there are no null values and duplicates in the data. Hence, our
          data looks clean!
```

```
In [79]: # Looking Closely at the order amount column
         df['order_amount'].describe()
Out[79]: count
                    5000.000000
                    3145.128000
         mean
         std
                   41282.539349
         min
                      90.000000
         25%
                     163.000000
         50%
                     284.000000
         75%
                      390.000000
                  704000.000000
         max
         Name: order amount, dtype: float64
```

### **Observation:**

- We notice that the mean of the 'order\_amount' column is \$3145.13 (which is the calculated AOV) which is very high for a shoe store. Hence we can conclude that some **outliers** might be affecting it's value.
- Also, Standard Deviation is also very large with it's value of \$41282.53. This means that data
  points are far from mean and therefore for calculating AOV, mean is not a good statistic.

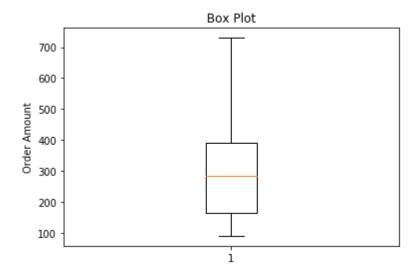
 Given that AOV is \$3145.13, we note that 75% of data is below 390 but the max value is \$704000.0 which is very large. Hence AOV might be definitely affected by some outliers to give such a large Average Order Value.

### **Outlier Detection**

### 1. BoxPlot

```
In [80]: fig1, ax1 = plt.subplots()
    ax1.set_title('Box Plot')
    ax1.boxplot(df.order_amount,showfliers=False)
    ax1.set_ylabel('Order Amount')
```

```
Out[80]: Text(0, 0.5, 'Order Amount')
```



**Observation:** From boxplot it is visible that the maximum amount is very large as compared to the amount for the 3rd quartile hence there are visible chances of outliers.

## 2. Price for each item

We check the price for each item to check if there is some descepancy in the amount charged for a particular shoe.

In [96]: df['price\_per\_item'] = df['order\_amount']/df['total\_items']
df

Out[96]:

	order_id	shop_id	user_id	order_amount	total_items	payment_method	created_at	price_pe
0	1	53	746	224	2	cash	2017-03- 13 12:36:56	
1	2	92	925	90	1	cash	2017-03- 03 17:38:52	
2	3	44	861	144	1	cash	2017-03- 14 4:23:56	
3	4	18	935	156	1	credit_card	2017-03- 26 12:43:37	
4	5	18	883	156	1	credit_card	2017-03- 01 4:35:11	
				•••				
4995	4996	73	993	330	2	debit	2017-03- 30 13:47:17	
4996	4997	48	789	234	2	cash	2017-03- 16 20:36:16	
4997	4998	56	867	351	3	cash	2017-03- 19 5:42:42	
4998	4999	60	825	354	2	credit_card	2017-03- 16 14:51:18	
4999	5000	44	734	288	2	debit	2017-03- 18 15:48:18	

5000 rows × 8 columns

4

```
In [95]: a= list(df['price_per_item'].unique())
    a= sorted(a,reverse=True)
    print('Top 3 prices for each individual pair of sneakers')
    print('Prices ->', a[:3],'\n')
    print('The shop_id of the Outlier (the shop which is charging $25725.0 for a sing print('Shop Id of Outlier-> ',df.loc[df['price_per_item']==25725.0].shop_id.unic
```

Top 3 prices for each individual pair of sneakers Prices -> [25725.0, 352.0, 201.0]

The shop\_id of the Outlier (the shop which is charging \$25725.0 for a single pa ir of sneakers!):
Shop Id of Outlier-> [78]

**Observation**: Yes! We have successfully detected the outlier using Exploratory Data Analysis and Boxplot.

The Shop\_id =78 is a fraud as every price\_per\_item was less than \$395 that means every pair of sneakers sold was less than the value of \$395 except for one shop which sold at \$25,725 which is extraordinarily high!

## **Outlier Handling**

- 1. Remove outliers and re-calculate AOV again
- 2. Use Median as a metric

**My approach**: I would use median as the AOV or Average Order Value. Since it is the middle value of the sorted values, it is a central value but large and small outliers do not skew the median as much as the mean.

```
In [85]: # Calculating median of order amount
print('The median is : $', df.order_amount.median())
```

The median is : \$ 284.0

Question 2: For this question you'll need to use SQL. Follow this link to access the data set required for the challenge. Please use queries to answer the following questions. Paste your queries along with your final numerical answers below.

A) How many orders were shipped by Speedy Express in total?

```
SELECT COUNT(*) FROM Orders

JOIN Shippers ON Shippers.ShipperID = Orders.ShipperID

WHERE Shippers.ShipperName = 'Speedy Express';
```

Answer: 54

### B) What is the last name of the employee with the most orders?

SELECT Employees.LastName, Count() AS num\_orders FROM Orders
JOIN Employees ON Orders.EmployeeID = Employees.EmployeeID
GROUP BY Employees.LastName ORDER BY num\_orders DESC;

Answer: Peacock (number of orders are 40)

### C) What product was ordered the most by customers in Germany?

SELECT Products.ProductName, Sum(OrderDetails.Quantity) as most\_ordered FROM Orders
JOIN OrderDetails ON Orders.OrderID = OrderDetails.OrderID
JOIN Customers ON Orders.CustomerID = Customers.CustomerID
JOIN Products ON OrderDetails.ProductID = Products.ProductID
WHERE Country = 'Germany'
GROUP BY Products.ProductName ORDER BY most\_ordered DESC;

Answer: Boston Crab Meat (with 160 orders)

In [ ]:		