Summer 2022 Data Science Intern Challenge at Shopify

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Question 1: Given some sample data, write a program to answer the following:

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.

Think about what could be going wrong with our calculation. Think about a better way to evaluate this data. What metric would you report for this dataset? What is its value?

Reading data

```
import os
import pandas as pd
import numpy as np
from datetime import datetime
import altair as alt

df = pd.read_csv(os.getcwd() + "/00_source_data/data.csv", parse_dates=["created_at"])
df["date"] = df.created_at.dt.date
# df.info()
df.sample(5)
```

Out[139		order_id	shop_id	user_id	order_amount	total_items	payment_method	created_at	date
	3216	3217	21	868	142	1	credit_card	2017-03-19 00:25:36	2017-03-19
	69	70	58	876	138	1	debit	2017-03-04 10:38:49	2017-03-04
	4563	4564	20	995	254	2	credit_card	2017-03-03 11:48:24	2017-03-03
	1708	1709	98	956	266	2	credit_card	2017-03-20 10:53:58	2017-03-20
	258	259	75	760	512	4	cash	2017-03-18 08:19:53	2017-03-18

Exploring the data

- 1. Sanity checks:
- Check number of unique sneaker shops
- Check for nulls in data
- Check datatypes of columns
- Reproduce AOV of \$3145.13 for 30 day period

```
assert df.shop_id.nunique() == 100 # 100 shops as mentioned in the problem
assert (
         df.date.nunique() == 30
) # There are 30 days in total, we don't need to subset the data
         df.isna().sum() # no null values across columns
         df.dtypes # datatypes are as expected
        print(f"AOV over 30 day period is ${np.round(df.order_amount.mean(),2)}")
```

AOV over 30 day period is \$3145.13

My initial suspicion is that the above statistic is sensitive to outliers, which drives the average value of a sneaker up. Although sneakers are usually an affordable item, there is still quite a lot of variance in the price of sneakers due to limited edition/designer sneakers that are collectibles are hence relatively expensive. Let's examine this distribution further.

1. Summary Statistics

```
In [141...
          df.order amount.describe()
                      5000.000000
          count
Out [141...
                      3145.128000
          mean
                    41282.539349
          std
          min
                        90.000000
          25%
                      163.000000
          50%
                       284.000000
          75%
                       390.000000
                   704000.000000
          max
          Name: order amount, dtype: float64
```

The most expensive shoe is \$704,000!

That's definitely biasing our results. The figures below show the distribution of the prices with a limit on price set at 1000(Ithinkthat's reasonable given that the third quartile is 390). We can see that the prices for most shoes fall between the 200-400 range so its most likely that if we didn't have our outliers, the average price would be around that value. Hence we need a better measure of central tendency.

```
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
%config InlineBackend.figure_format = 'retina'

#Plotting histogram of prices
sns.set_style("whitegrid")
sns.set_context({"figure.figsize": (6, 3)})
hist=sns.histplot(df['order_amount'],bins=2000,color='#95BF47') #Shopify colors!
hist.set( xlim= (0,1000),title = "Histogram of sneaker prices ", ylabel = "Count")
print("")
```



```
In [143... bx = sns.boxplot(x=df.loc[df["order_amount"] <= 1000]["order_amount"], color="#95BF47")
    bx.set(title="Boxplot of sneaker prices ", xlabel="Prices")

Out[143... [Text(0.5, 1.0, 'Boxplot of sneaker prices '), Text(0.5, 0, 'Prices')]</pre>
```



Alternative Metrics

Instead of the mean, an alternative we can use here is the median. The median is less sensitive to outliers, so will give us a more accurate result in this scenario when our data is skewed. This is the metric I would report for this dataset.

```
print(
    f"The median price of shoes sold is ${np.median(df['order_amount'])}. I believe that this is a better estiment)
```

The median price of shoes sold is \$284.0. I believe that this is a better estimate for average prices than the mean.