# $case\_study2$

October 31, 2021

# 0.1 DDA Full Stack Interview

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# 0.1.1 Case Study 2

```
[1]: #import pandas
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np

#downloaded file, had issue with encoding and so downloaded file
file = "casestudy.csv"

df = pd.read_csv(file)

df
```

[1]:	Unnamed: 0	customer_email	net_revenue	year
0	0	${\tt nhknapwsbx@gmail.com}$	249.92	2015
1	1	${ t joiuzbvcpn@gmail.com}$	87.61	2015
2	2	ukkjctepxt@gmail.com	168.38	2015
3	3	${\tt gykatilzrt@gmail.com}$	62.40	2015
4	4	${\tt mmsgsrtxah@gmail.com}$	43.08	2015
•••	•••	•••		
685922	685922	${\tt qzqttwiftu@gmail.com}$	184.58	2017
685923	685923	pjodiifjop@gmail.com	133.03	2017
685924	685924	appaplmgko@gmail.com	200.98	2017
685925	685925	wvkpmwsgck@gmail.com	235.35	2017
685926	685926	aregboumbw@gmail.com	208.43	2017

[685927 rows x 4 columns]

Let's drop the unnamed column and perform summary.

```
[2]: df = df.drop("Unnamed: 0", axis = 1)
#let's investigate our dataframe
```

```
def investigate(data):
    print(data.info())
    print(data.describe())
investigate(df)
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 685927 entries, 0 to 685926
```

```
Data columns (total 3 columns):
```

#	Column	Non-Null Count	Dtype	
0	customer_email	685927 non-null	object	
1	net_revenue	685927 non-null	${\tt float64}$	
2	year	685927 non-null	int64	
<pre>dtypes: float64(1), int64(1), object(1)</pre>				
memo	ry usage: 15.7+ 1	MB		

None

	net_revenue	year
count	685927.000000	685927.000000
mean	125.647755	2016.027252
std	71.854528	0.837203
min	1.000000	2015.000000
25%	63.510000	2015.000000
50%	125.730000	2016.000000
75%	187.820000	2017.000000
max	250.000000	2017.000000

It looks like we have no null or missing values.

# 0.2 For each year we need the following information:

# 0.2.1 1) Total revenue for the current year

```
[3]: #find customers in 2017
     customers_2017 = df[df["year"] == 2017]
     #sum the net revenue for 2017 customers
     customer_revenue_2017 = customers_2017["net_revenue"].sum()
     print(f'Total Revenue from 2017: ${customer_revenue_2017:.2f}')
```

Total Revenue from 2017: \$31417495.03

# 0.2.2 2) New Customer Revenue e.g., new customers not present in previous year only

```
[4]: #create df for customers in 2016
     customers_2016 = df[df["year"] == 2016]
     #remove customers from customer_2017 that appear in customer_2016
     df2 = customers_2017[np.logical_not(customers_2017["customer_email"]
```

```
.isin(customers_2016['customer_email']))]
new_customer_revenue_2017 = df2["net_revenue"].sum()
print(f'New Customer Revenue from 2017: ${new_customer_revenue_2017:.2f}')
```

New Customer Revenue from 2017: \$28776235.04

0.2.3 3) Existing Customer Growth. To calculate this, use the Revenue of existing customers for current year –(minus) Revenue of existing customers from the previous year

Existing Customer Revenue from 2017: \$20611.34

#### 0.2.4 4) Revenue lost from attrition

```
[6]: #to calculate this, I will find total revenue from clients who purchased in
     #2015 but did not purchase in 2016 or 2017, plus total revenue from clients
     #who purchased 2016 and not 2017.
     #make df for customers from 2015
     customers_2015 = df[df["year"] == 2015]
     #make df for 2015 customers who did not purchase in 2016 or 2017
     df5 = customers_2015[np.logical_not(customers_2015["customer_email"]
                                         .isin(customers_2016['customer_email']))]
     df5 = df5[np.logical_not(df5["customer_email"]
                              .isin(customers_2017['customer_email']))]
     #make df for 2016 customers who did not purchase in 2017
     df6 = customers_2016[np.logical_not(customers_2016["customer_email"]
                                         .isin(customers_2017['customer_email']))]
     #sum revenues from two dfs
     revenue_lost_from_2015cohort = df5["net_revenue"].sum()
     revenue_lost_from_2016cohort = df6["net_revenue"].sum()
     #combine revenues
     revenue_lost_from_attrition = revenue_lost_from_2015cohort +__
     →revenue_lost_from_2016cohort
     print(f'Revenue Lost from Attrition: ${revenue_lost_from_attrition:.2f}')
```

Revenue Lost from Attrition: \$44586295.64

#### 0.2.5 5) Existing Customer Revenue Current Year

```
[7]: print(f'Existing Customer Revenue 2017: ${existing_customer_revenue_2017:.2f}')
```

Existing Customer Revenue 2017: \$2641259.99

#### 0.2.6 6) Existing Customer Revenue Prior Year

Existing Customer Revenue 2016: \$7485452.58

## 0.2.7 7) Total Customers Current Year

```
[9]: np.unique(customers_2017["customer_email"]).shape[0]
```

[9]: 249987

#### 0.2.8 8) Total Customers Previous Year

```
[10]: np.unique(customers_2016["customer_email"]).shape[0]
```

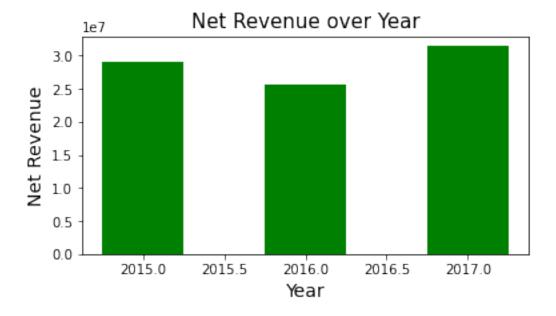
[10]: 204646

# 0.2.9 9) New Customers

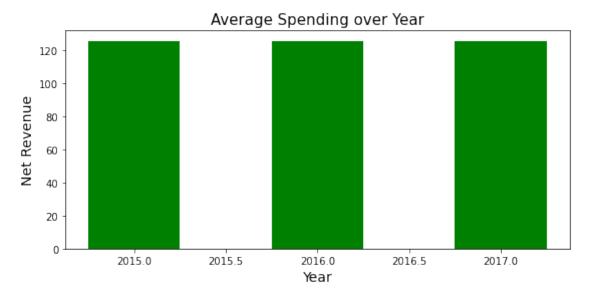
#### **0.2.10 10)** Lost Customers

```
[12]: #assuming this means clients from 2015 that haven't returned
#in 2016 or 2017, and and clients from 2016 that haven't returned in 2017
lost_customers2015 = np.unique(df5["customer_email"])
lost_customers2016 = np.unique(df6["customer_email"])
lost_customers = np.append(lost_customers2015, lost_customers2016)
lost_customers
```

## **0.2.11 11)** Unique Plots



Revenue dipped a little in 2017 before picking back up in 2017.



Average spending did not change from year to year.