Exploratory Data Analysis of Customer Segmentation and marketing campaigns

Problem Statement

While going through the analysis, it is vital to have in mind an objective. We will attempt to answer the following questions through Visualization to see if we can find contributing factors to the success of the past campaigns.

- · Which products are performing best?
- · Which channels are underperforming?
- What does the average customer look like for the Company?
- Which marketing campaign is most successful?
- · Which Regions perform best?
- · Which costumer segment purchase more?
- · What does the costumer segment which accepted the last marketing campaign look like?

Importing Libraries

```
#Importing the Libraries
import numpy as np
import pandas as pd
import datetime
import matplotlib
import matplotlib.pyplot as plt
from matplotlib import colors
import seaborn as sns
import matplotlib.pyplot as plt, numpy as np
from mpl_toolkits.mplot3d import Axes3D
from matplotlib.colors import ListedColormap
```

Reading csv file

dset = pd.read_csv("C:/Users/Lenovo/Desktop/GitHub Projects/njimonda.github.io/Exploratory-Data-Analysis-of-marketing-campaigns/marketing-rint("Number of datapoints:", len(dset))

The Near Birth Education Marital Status - Income Kidhome Teenhome Dt Customer Becency Morthlines - NumbtorePurchase

} ▼		ID	Year_Birth	Education	Marital_Status	Income	Kidhome	Teenhome	Dt_Customer	Recency	MntWines	 NumStorePurchase:
	0	1826	1970	Graduation	Divorced	\$84,835.00	0	0	6/16/14	0	189	 •
	1	1	1961	Graduation	Single	\$57,091.00	0	0	6/15/14	0	464	 7
	2	10476	1958	Graduation	Married	\$67,267.00	0	1	5/13/14	0	134	 ŧ

3 rows × 28 columns

_		ID	Year_Birth	Education	Marital_Status	Income	Kidhome	Teenhome	Dt_Customer	Recency	MntWines	 NumStorePurcha:
	2237	22	1976	Graduation	Divorced	\$46 310 00	1	0	12/3/12	99	185	

 2237
 22
 1976
 Graduation
 Divorced
 \$46,310.00
 1
 0
 12/3/12
 99
 185
 ...

 2238
 528
 1978
 Graduation
 Married
 \$65,819.00
 0
 0
 11/29/12
 99
 267
 ...

 2239
 4070
 1969
 PhD
 Married
 \$94,871.00
 0
 2
 9/1/12
 99
 169
 ...

3 rows × 28 columns

dset.shape

dset.tail(3)

→ (2240, 28)

dset.describe()

 $\overrightarrow{\Rightarrow}$

	ID	Year_Birth	Kidhome	Teenhome	Recency	MntWines	MntFruits	${\tt MntMeatProducts}$	MntFishProducts	Mn ⁻
count	2240.000000	2240.000000	2240.000000	2240.000000	2240.000000	2240.000000	2240.000000	2240.000000	2240.000000	
mean	5592.159821	1968.805804	0.444196	0.506250	49.109375	303.935714	26.302232	166.950000	37.525446	
std	3246.662198	11.984069	0.538398	0.544538	28.962453	336.597393	39.773434	225.715373	54.628979	
min	0.000000	1893.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
25%	2828.250000	1959.000000	0.000000	0.000000	24.000000	23.750000	1.000000	16.000000	3.000000	
50%	5458.500000	1970.000000	0.000000	0.000000	49.000000	173.500000	8.000000	67.000000	12.000000	
75%	8427.750000	1977.000000	1.000000	1.000000	74.000000	504.250000	33.000000	232.000000	50.000000	
max	11191.000000	1996.000000	2.000000	2.000000	99.000000	1493.000000	199.000000	1725.000000	259.000000	

dset.info()

8 rows × 23 columns

<class 'pandas.core.frame.DataFrame'>
 RangeIndex: 2240 entries, 0 to 2239
 Data columns (total 28 columns):

#	Column (total 28 to.	Non-Null Count	Dtype								
0	ID	2240 non-null	int64								
1	Year_Birth	2240 non-null	int64								
2	Education	2240 non-null	object								
3	Marital_Status	2240 non-null	object								
4	Income	2216 non-null	object								
5	Kidhome	2240 non-null	int64								
6	Teenhome	2240 non-null	int64								
7	Dt_Customer	2240 non-null	object								
8	Recency	2240 non-null	int64								
9	MntWines	2240 non-null	int64								
10	MntFruits	2240 non-null	int64								
11	MntMeatProducts	2240 non-null	int64								
12	MntFishProducts	2240 non-null	int64								
13	MntSweetProducts	2240 non-null	int64								
14	MntGoldProds	2240 non-null	int64								
15	NumDealsPurchases	2240 non-null	int64								
16	NumWebPurchases	2240 non-null	int64								
17	NumCatalogPurchases	2240 non-null	int64								
18	NumStorePurchases	2240 non-null	int64								
19	NumWebVisitsMonth	2240 non-null	int64								
20	AcceptedCmp3	2240 non-null	int64								
21	AcceptedCmp4	2240 non-null	int64								
22	AcceptedCmp5	2240 non-null	int64								
23	AcceptedCmp1	2240 non-null	int64								
24	AcceptedCmp2	2240 non-null	int64								
25	Response	2240 non-null	int64								
26	Complain	2240 non-null	int64								
27	Country	2240 non-null	object								
	es: int64(23), object	(5)									
memoi	memory usage: 490.1+ KB										

dset.isnull().sum().sort_values(ascending=False)

→ ▼	Income	24
_	Country	0
	Complain	0
	Year_Birth	0
	Education	0
	Marital_Status	0
	Kidhome	0
	Teenhome	0
	Dt_Customer	0
	Recency	0
	MntWines	0
	MntFruits	0
	MntMeatProducts	0
	MntFishProducts	0
	MntSweetProducts	0
	MntGoldProds	0
	NumDealsPurchases	0
	NumWebPurchases	0
	NumCatalogPurchases	0
	NumStorePurchases	0
	NumWebVisitsMonth	0
	AcceptedCmp3	0
	AcceptedCmp4	0
	AcceptedCmp5	0
	AcceptedCmp1	0
	-	

```
AcceptedCmp2 0
Response 0
ID 0
dtype: int64
```

Clean whitespace and converting datatype

```
# clean up column names that contain whitespace
dset.columns = dset.columns.str.replace(' ', '')
# transform Income column to a numerical
dset['Income'] = dset['Income'].str.replace('$', '')
dset['Income'] = dset['Income'].str.replace(',', '').astype('float')
dset["Dt_Customer"] = pd.to_datetime(dset["Dt_Customer"])
#As it's a date, it's better to change to format of datetime
dates = []
for i in dset["Dt_Customer"]:
   i = i.date()
   dates.append(i)
#Dates of the newest and oldest recorded customer
print("The newest customer's enrolment date in the records:",max(dates))
print("The oldest customer's enrolment date in the records:",min(dates))
The newest customer's enrolment date in the records: 2014-06-29
     The oldest customer's enrolment date in the records: 2012-07-30
```

Handling null values or outliers

```
dset['Income'].fillna(dset['Income'].median())
            84835.0
₹
            57091.0
            67267.0
     3
            32474.0
     4
            21474.0
     2235
            66476.0
     2236
            31056.0
     2237
            46310.0
     2238
            65819.0
            94871.0
     Name: Income, Length: 2240, dtype: float64
#To remove the NA values
dset = dset.dropna()
print("The total number of data-points after removing the rows with missing values are:", len(dset))
The total number of data-points after removing the rows with missing values are: 2216
dset[dset.isnull()["Income"] == True].count()["ID"]
#There are 24 objects, it's better to get rid of them not to corrupt the whole picture and insights of the data
→ 0
Outliers
```

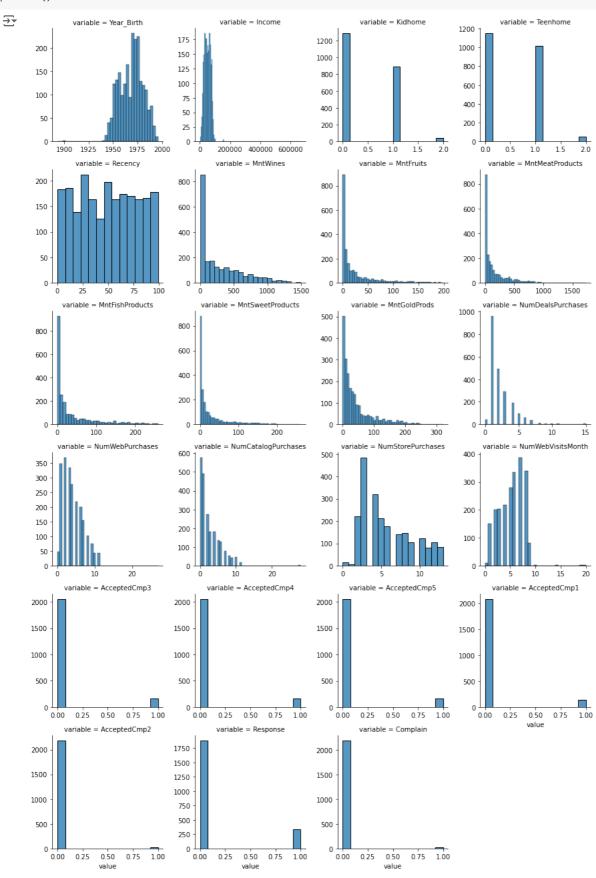
```
'Teenhome', 'Recency', 'MntWines', 'MntFruits',
    'MntMeatProducts', 'MntFishProducts', 'MntSweetProducts',
    'MntGoldProds', 'NumDealsPurchases', 'NumWebPurchases',
    'NumCatalogPurchases', 'NumStorePurchases', 'NumWebVisitsMonth',
    'AcceptedCmp3', 'AcceptedCmp4', 'AcceptedCmp5', 'AcceptedCmp1',
    'AcceptedCmp2', 'Response', 'Complain'])

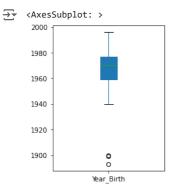
nd = pd.melt(dist, value_vars = dist)

n1 = sns.FacetGrid(nd, col = "variable", col_wrap = 4, sharex = False, sharey = False)

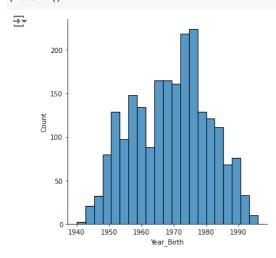
n1 = n1.map(sns.histplot, "value")

plt.show()
```

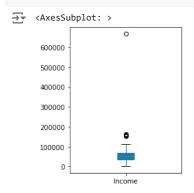




dset.drop(dset[dset['Year_Birth'] <= 1900].index, inplace = True) #Drop customers who were born before 1900.



dset['Income'].plot(kind='box', figsize=(3,4), patch_artist=True)



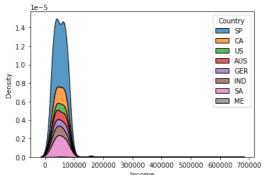
dset["Income"].describe() #Quantiles, mean, std of Income feature

```
→ count

                 2213.000000
     mean
                52236.581563
                25178.603047
     std
                 1730.000000
     min
                35246.000000
     25%
               51373.000000
68487.000000
     50%
     75%
               666666.000000
     max
     Name: Income, dtype: float64
```

sns.kdeplot(data=dset, x="Income", hue="Country", multiple="stack")

<a < AxesSubplot: xlabel='Income', ylabel='Density'>



dset[dset['Income'] > 600000]

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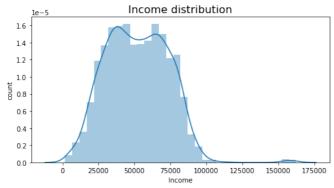
	ID	Year_Birth	Education	Marital_Status	Income	Kidhome	Teenhome	Dt_Customer	Recency	MntWines	 NumStorePurchases
527	9432	1977	Graduation	Together	666666.0	1	0	2013-06-02	23	9	 3

1 rows × 28 columns

dset.drop(dset[dset['Income'] == 666666].index, inplace = True) #Drop one customer, who has 666666 \$ income.

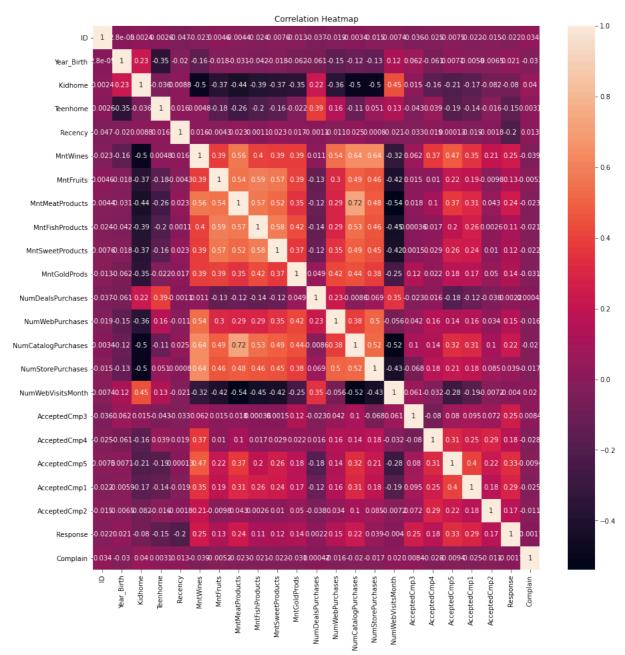
```
plt.figure(figsize=(8,4))
sns.distplot(dset['Income'], kde=True, hist=True)
plt.title('Income distribution', size=16)
plt.ylabel('count');
```

C:\Users\Lenovo\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and wi warnings.warn(msg, FutureWarning)



plt.figure(figsize=(15,15)) #Corellation Heatmap
plt.title(label = "Correlation Heatmap")
sns.heatmap(dset.corr(), annot=True)
plt.show()



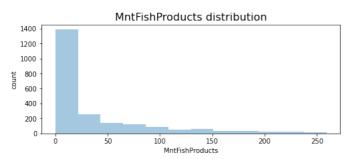


Statistical Analysis

```
plt.figure(figsize=(8,3))
sns.distplot(dset['MntFishProducts'], kde=False, hist=True, bins=12)
plt.title('MntFishProducts distribution', size=16)
plt.ylabel('count');
```

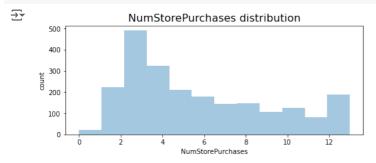
C:\Users\Lenovo\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning:

`distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figur



```
plt.figure(figsize=(8,3))
sns.distplot(dset['NumStorePurchases'], kde=False, hist=True, bins=12)
```

```
plt.title('NumStorePurchases distribution', size=16)
plt.ylabel('count');
```



```
print("Total categories in the feature Education:\n\n")
print(dset["Education"].value_counts())
```

→ Total categories in the feature Education:

```
Graduation 1115
PhD 480
Master 365
2n Cycle 198
Basic 54
```

Name: Education, dtype: int64

```
fig = px.pie(dset, names='Education')
fig.show()
```

Next steps: Explain error

```
print("Total categories in the feature Marital_Status:\n")
print(dset["Marital_Status"].value_counts(), "\n")
```

→ Total categories in the feature Marital_Status:

Married 864 Together 580 Single 480 Divorced 232 Widow 77

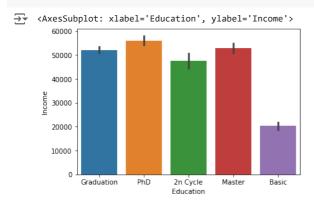
Name: Marital_Status, dtype: int64

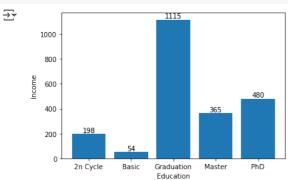
```
dset.drop(dset[dset['Marital_Status'] == "YOLO"].index, inplace = True)
dset.drop(dset[dset['Marital_Status'] == "Absurd"].index, inplace = True)
dset.drop(dset[dset['Marital_Status'] == "Alone"].index, inplace = True)
```

```
fig = px.pie(dset, names='Marital_Status')
fig.show()
```

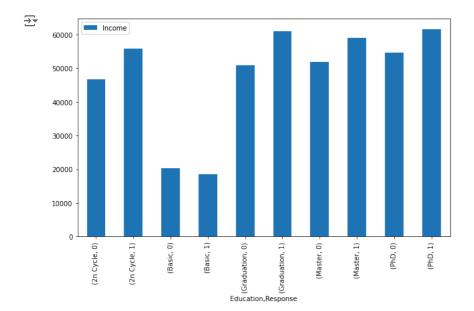
```
 \begin{tabular}{ll} \# \ fig = px.histogram(dset, x=" Income", y="Education", title="Income by Education") \\ \end{tabular} 
# fig.show()
dset["Dt_Customer"][dset["Response"] == 1].dt.month.value_counts()
\hbox{\#Count month when customers enroll with the company}\\
<del>_</del> 8
            41
     10
            38
     11
            33
            30
            28
     3
            26
     5
4
            24
            22
     12
            18
            16
            11
     Name: Dt_Customer, dtype: int64
dset["Dt_Customer"][dset["Response"] == 1].dt.weekday.value_counts()
#Count days of week when customers enroll with the company
₹
     0
           59
           55
     4
     1
           50
           46
           42
           30
     Name: Dt_Customer, dtype: int64
fig = px.histogram(dset, x="Country", y="TotalMnt", title="Total Amount Spent by Country")
# fig.title('Total Amount Spent by Country', size=16)
# fig.ylabel('Amount Spent');
fig.show()
```

sns.barplot(x=dset["Education"],y=dset["Income"])





```
dset.pivot_table(["Income"], ["Education", "Response"], aggfunc="mean").plot.bar()
plt.show()
```



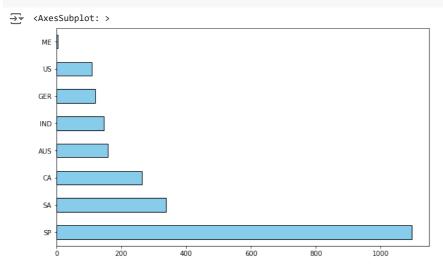
print("Total categories in the feature Countries:\n")
print(dset["Country"].value_counts(), "\n")

 $\overrightarrow{\Rightarrow}$ Total categories in the feature Countries:

SP 1092 SA 335 CA 261 AUS 146 IND 145 GER 116 US 107 ME 3

Name: Country, dtype: int64

 $dset['Country'].value_counts().plot(kind='barh', color='skyblue', edgecolor=(0,0,0))$

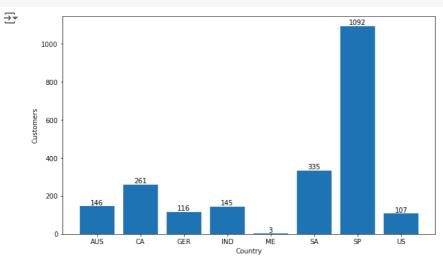


```
# plt.figure(figsize=(5,4))
# dset.groupby('Country')['TotalPurchases'].sum().sort_values(ascending=False).plot(kind='bar')
# plt.title('Total Number of Purchases by Country', size=16)
# plt.ylabel('Number of Purchases');

fig = px.histogram(dset, x="Country", y="TotalPurchases", color="Country")
fig.show()
```

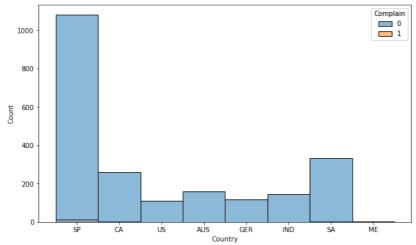
```
dset["Marital_Status"][dset["Response"] == 1].value_counts()
#Count Marital_Status of customers of those who accepted last campaign
```

```
Single 106
Married 98
Together 60
Divorced 48
Widow 18
Name: Marital_Status, dtype: int64
```



sns.histplot(data=dset, x="Country", hue="Complain", multiple="stack")



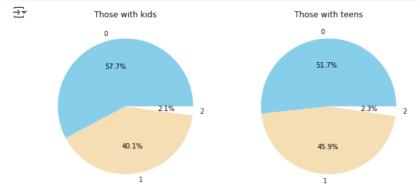


fig, (ax1,ax2) = plt.subplots(1,2,figsize=(10,10)) #ax1,ax2 refer to your two pies
1,2 denotes 1 row, 2 columns - if you want to stack vertically, it would be 2,1
colors= 'skyblue', 'wheat', 'white'

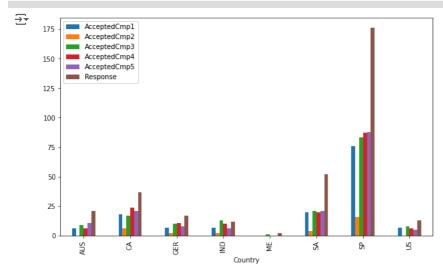
labels = dset['Kidhome'].unique()
values = dset['Kidhome'].value_counts()
ax1.pie(values,labels = labels,colors = colors,autopct = '%1.1f%%') #plot first pie

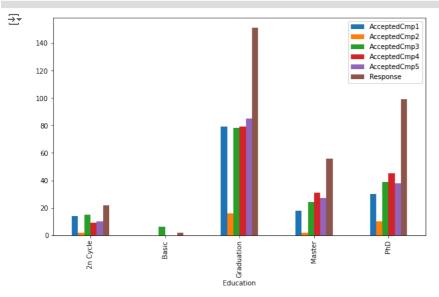
labels = dset['Teenhome'].unique()
values = dset['Teenhome'].value_counts()
ax2.pie(values,labels = labels,colors = colors,autopct = '%1.1f%%') #plot second pie

ax1.set(aspect="equal", title='Those with kids')
ax2.set(aspect="equal", title='Those with teens')
plt.show()



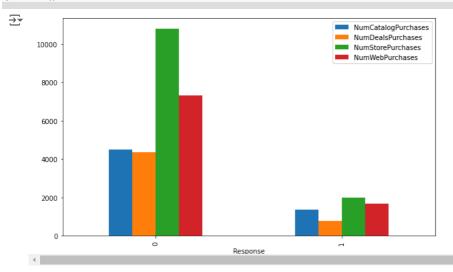
dset.pivot_table(["AcceptedCmp1", "AcceptedCmp2", "AcceptedCmp3", "AcceptedCmp4", "AcceptedCmp5", "Response"], ["Country"], aggfunc="sur plt.show()



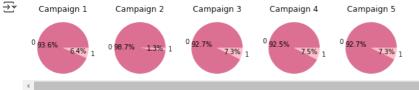


```
# convert country codes to correct nomenclature for choropleth plot
# the dataset doesn't provide information about country codes
\#\# ...so I'm taking my best guess about the largest nations that make sense given the codes provided
dset['Country_code'] = dset['Country'].replace({'SP': 'ESP', 'CA': 'CAN', 'US': 'USA', 'SA': 'ZAF', 'ME': 'MEX'})
\ensuremath{\text{\#}} success of campaigns by country code
df_cam = dset[['Country_code', 'AcceptedCmp1', 'AcceptedCmp2', 'AcceptedCmp3', 'AcceptedCmp4', 'AcceptedCmp5', 'Response']].melt(
         id_vars='Country_code', var_name='Campaign', value_name='Accepted (%)')
df_cam = pd.DataFrame(df_cam.groupby(['Country_code', 'Campaign'])['Accepted (%)'].mean()*100).reset_index(drop=False)
# rename the campaign variables so they're easier to interpret
df_cam['Campaign'] = df_cam['Campaign'].replace({'AcceptedCmp1': '1',
                                                                                                                           'AcceptedCmp2': '2',
                                                                                                                           'AcceptedCmp3': '3',
                                                                                                                          'AcceptedCmp4': '4',
                                                                                                                          'AcceptedCmp5': '5',
                                                                                                                             'Response': 'Most recent'
                                                                                                                          })
# choropleth plot
{\tt import plotly.express as } \ {\tt px}
\label{fig} fig = px.choropleth (df_cam, locationmode='ISO-3', color='Accepted (\%)', facet\_col='Campaign', facet\_col\_wrap=2, figure for the color of the color 
                                                  facet_row_spacing=0.05, facet_col_spacing=0.01, width=700,
                                                  locations='Country_code', projection='natural earth', title='Advertising Campaign Success Rate by Country'
fig.show()
```

__



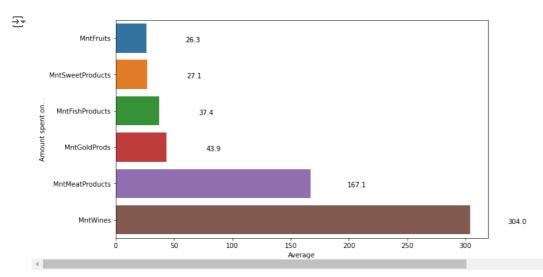
```
fig, (ax1,ax2, ax3, ax4, ax5) = plt.subplots(1,5, figsize=(10,10))
# 1,5 denotes 1 row, 5 columns
colors= 'palevioletred', 'pink'
labels = dset['AcceptedCmp1'].unique()
values = dset['AcceptedCmp1'].value_counts()
ax1.pie(values,labels = labels,colors = colors,autopct = '%1.1f%%')
labels = dset['AcceptedCmp2'].unique()
values = dset['AcceptedCmp2'].value_counts()
ax2.pie(values,labels = labels,colors = colors,autopct = '%1.1f%'')
labels = dset['AcceptedCmp3'].unique()
values = dset['AcceptedCmp3'].value_counts()
ax3.pie(values,labels = labels,colors = colors,autopct = '%1.1f%%')
labels = dset['AcceptedCmp4'].unique()
values = dset['AcceptedCmp4'].value_counts()
ax4.pie(values,labels = labels,colors = colors,autopct = '%1.1f%%')
labels = dset['AcceptedCmp5'].unique()
values = dset['AcceptedCmp5'].value_counts()
ax5.pie(values,labels = labels,colors = colors,autopct = '%1.1f%%')
ax1.set(aspect="equal", title='Campaign 1')
ax2.set(aspect="equal", title='Campaign 2')
ax3.set(aspect="equal", title='Campaign 3')
ax4.set(aspect="equal", title='Campaign 4')
ax5.set(aspect="equal", title='Campaign 5')
plt.show()
```

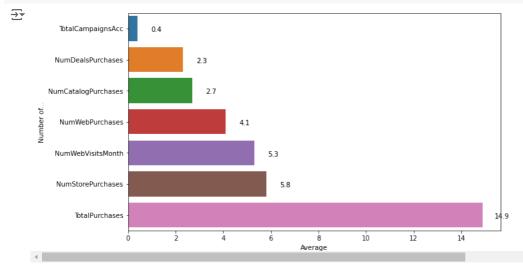




AcceptedCmp2 AcceptedCmp3 AcceptedCmp4 AcceptedCmp4 Response 2 4 6 8 10 12 14 Accepted(%)

```
₹
                     Average
       Year_Birth
                       1968.9
     Year_Customer
                      2013.0
         Income
                      51954.6
       Dependents
                          0.9
        Kidhome
                          0.4
        Teenhome
                          0.5
        Recency
                         49.1
```





```
g = sns.FacetGrid(df_cam2, col='Campaign', col_wrap=3)
g.map(sns.barplot, 'Country', 'Accepted (%)')
```

C:\Users\Lenovo\anaconda3\lib\site-packages\seaborn\axisgrid.py:645: UserWarning:

Using the barplot function without specifying `order` is likely to produce an incorrect plot.

