DATA 119 Final Project

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Project A

```
import numpy as np
import pandas as pd
import plotnine as p9
import statsmodels.api as sm
import sklearn.metrics as metrics

df = pd.read_csv("marketing_campaign.csv", sep="\t")
pd.set_option('display.max_rows', None)
df.head(5)
```

	ID	Year_Birth	Education	Marital_Status	Income	Kidhome	Teenhome	Dt_Customer	Rec
0	5524	1957	Graduation	Single	58138.0	0	0	04-09-2012	58
1	2174	1954	Graduation	Single	46344.0	1	1	08-03-2014	38
2	4141	1965	Graduation	Together	71613.0	0	0	21-08-2013	26
3	6182	1984	Graduation	Together	26646.0	1	0	10-02-2014	26
4	5324	1981	PhD	Married	58293.0	1	0	19-01-2014	94

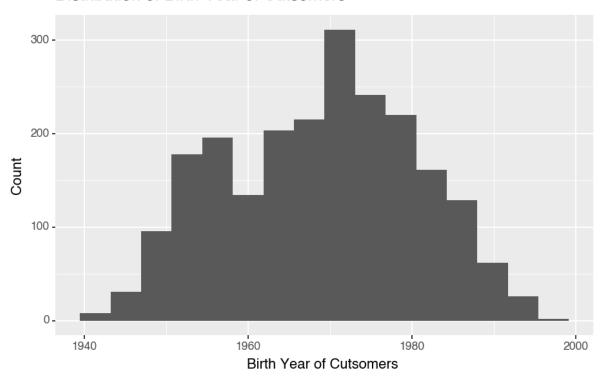
```
df.columns
```

```
'AcceptedCmp3', 'AcceptedCmp4', 'AcceptedCmp5', 'AcceptedCmp1',
       'AcceptedCmp2', 'Complain', 'Z_CostContact', 'Z_Revenue', 'Response'],
      dtype='object')
  #These variables are meaningless
  #so we decide to delete them
  df = df.drop(columns=['ID', 'Z_CostContact','Z_Revenue'])
  df.columns
Index(['Year_Birth', 'Education', 'Marital_Status', 'Income', 'Kidhome',
       'Teenhome', 'Dt_Customer', 'Recency', 'MntWines', 'MntFruits',
       'MntMeatProducts', 'MntFishProducts', 'MntSweetProducts',
       'MntGoldProds', 'NumDealsPurchases', 'NumWebPurchases',
       'NumCatalogPurchases', 'NumStorePurchases', 'NumWebVisitsMonth',
       'AcceptedCmp3', 'AcceptedCmp4', 'AcceptedCmp5', 'AcceptedCmp1',
       'AcceptedCmp2', 'Complain', 'Response'],
      dtype='object')
  # Name the variables in a more meaning way
  df = df.rename(columns={"Response": "AcceptedLastCmp"})
  df.columns
Index(['Year_Birth', 'Education', 'Marital_Status', 'Income', 'Kidhome',
       'Teenhome', 'Dt_Customer', 'Recency', 'MntWines', 'MntFruits',
       'MntMeatProducts', 'MntFishProducts', 'MntSweetProducts',
       'MntGoldProds', 'NumDealsPurchases', 'NumWebPurchases',
       'NumCatalogPurchases', 'NumStorePurchases', 'NumWebVisitsMonth',
       'AcceptedCmp3', 'AcceptedCmp4', 'AcceptedCmp5', 'AcceptedCmp1',
       'AcceptedCmp2', 'Complain', 'AcceptedLastCmp'],
      dtype='object')
  df['Income'].isna().sum() / df.shape[0]
  #only 1% missing data, so let's drop it
  df = df.dropna()
  #Summary statistics for variables
  print(df['Year_Birth'].describe())
```

```
2216.000000
count
         1968.820397
mean
std
          11.985554
min
         1893.000000
25%
         1959.000000
50%
         1970.000000
75%
         1977.000000
max
         1996.000000
Name: Year_Birth, dtype: float64
  print(df[df['Year_Birth'] < 1920]['Dt_Customer'])</pre>
  print("It looks like some data entry error")
  print("because it is quite impossible to be born that early \nand become a customer for the
  print("So we can delete the data")
  index = df[df['Year_Birth'] < 1920].index</pre>
  df = df.drop(index)
192
       26-09-2013
239
       17-05-2014
339
       26-09-2013
Name: Dt_Customer, dtype: object
It looks like some data entry error
because it is quite impossible to be born that early
and become a customer for the first time in 2010s.
So we can delete the data
  #Summary statistics for variables
  print(df['Year_Birth'].describe())
count
         2213.000000
         1968.917307
mean
std
          11.700216
min
        1940.000000
25%
        1959.000000
50%
        1970.000000
75%
         1977.000000
         1996.000000
max
```

Name: Year_Birth, dtype: float64

Distribution of Birth Year of Cutsomers



This histogram shows the distribution of Birth Year of Cutsomers. The distribution is centered at approximately 1970, and roughly ranges from 1940 to 1995. It is an unimodal, left skewed distribution. There are no obvious unusual values.

```
<Figure Size: (460 x 345)>

print(df['Income'].describe())

count 2213.000000
```

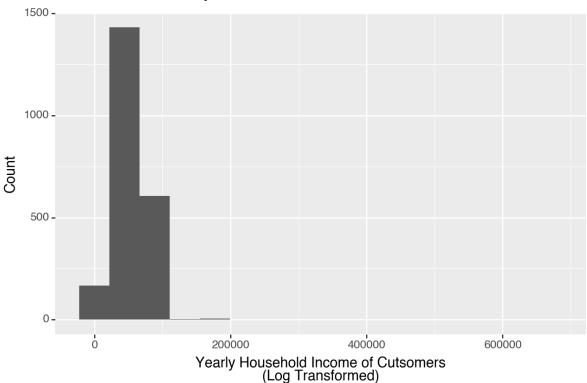
```
1730.000000
min
25%
          35246.000000
50%
          51373.000000
75%
          68487.000000
max
         666666.000000
Name: Income, dtype: float64
  print(df['Income'].describe())
           2213.000000
count
          52236.581563
mean
std
          25178.603047
          1730.000000
min
25%
          35246.000000
50%
          51373.000000
75%
          68487.000000
max
         666666.000000
Name: Income, dtype: float64
  # Visualize the distirbution of variables
  (p9.ggplot(df) +
   p9.aes (x = 'Income') +
   p9.geom_histogram(bins=16)+
   p9.labs(x = "Yearly Household Income of Cutsomers\n(Log Transformed)", y = "Count", title=
           caption = "This histogram shows the distribution of yearly household income of cu
            "The distribution is centered at approximately 50,000, and roughly ranges from 1
            "It is an unimodal, left skewed distribution. There are unusually high values.")
```

52236.581563

25178.603047

mean std

Distribution of Yearly Household Income of Cutsomers



This histogram shows the distribution of yearly household income of cutsomers. The distribution is centered at approximately 50,000, and roughly ranges from 1000 to 700,000. It is an unimodal, left skewed distribution. There are unusually high values.

```
<Figure Size: (640 x 480)>
```

```
print(df['Dt_Customer'].describe())
```

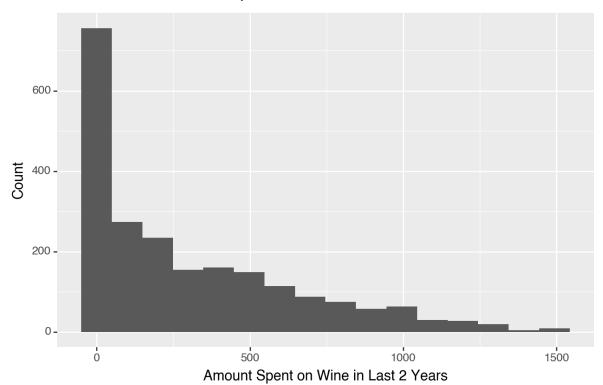
count 2213 unique 662 top 31-08-2012 freq 12

Name: Dt_Customer, dtype: object

```
print(df['Recency'].describe())
```

```
2213.000000
count
           49.007682
mean
std
           28.941864
            0.000000
min
25%
           24.000000
50%
           49.000000
75%
           74.000000
max
           99.000000
Name: Recency, dtype: float64
  print(df['MntWines'].describe())
count
         2213.000000
mean
          305.153638
          337.305490
std
            0.000000
min
25%
           24.000000
50%
          175.000000
75%
          505.000000
max
         1493.000000
Name: MntWines, dtype: float64
  # Visualize the distirbution of variables
  (p9.ggplot(df) +
   p9.aes (x = 'MntWines') +
   p9.geom_histogram(bins=16)+
   p9.labs(x = "Amount Spent on Wine in Last 2 Years", y = "Count", title= "Distribution of A
           caption = "This histogram shows the distribution of amount spent on wine in last
             "The distribution is centered at approximately 0, and roughly ranges from 0 to 1
             "It is an unimodal, right skewed distribution. There are no unusual values."))
```

Distribution of Amount Spent on Wine in Last 2 Years



This histogram shows the distribution of amount spent on wine in last 2 years. The distribution is centered at approximately 0, and roughly ranges from 0 to 1500. It is an unimodal, right skewed distribution. There are no unusual values.

<Figure Size: (640 x 480)>

print(df['MntFruits'].describe())

count	2213.000000
mean	26.323995
std	39.735932
min	0.000000
25%	2.000000
50%	8.000000
75%	33.000000
max	199.000000

Name: MntFruits, dtype: float64

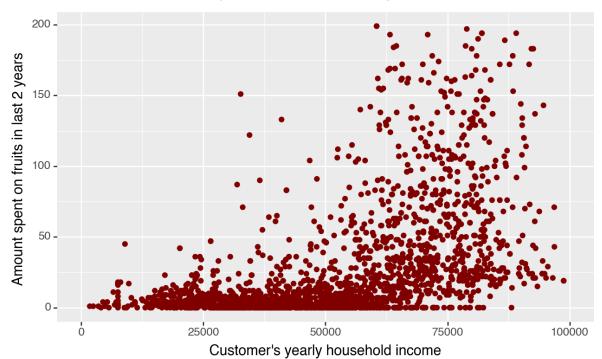
print(df['MntMeatProducts'].describe()) 2213.000000 count 166.962494 mean std 224.226178 0.000000 min 25% 16.000000 50% 68.000000 75% 232.000000 1725.000000 maxName: MntMeatProducts, dtype: float64 print(df['MntFishProducts'].describe()) 2213.000000 count mean37.635337 54.763278 std min 0.000000 25% 3.000000 50% 12.000000 75% 50.000000 259.000000 maxName: MntFishProducts, dtype: float64 print(df['MntSweetProducts'].describe()) 2213.000000 count mean27.034794 std 41.085433 0.000000 min 25% 1.000000 50% 8.000000 75% 33.000000 262.000000 maxName: MntSweetProducts, dtype: float64 print(df['MntGoldProds'].describe())

```
2213.000000
count
           43.911432
mean
std
           51.699746
min
            0.000000
25%
            9.000000
50%
           24.000000
75%
           56.000000
max
          321.000000
Name: MntGoldProds, dtype: float64
  print(df['NumDealsPurchases'].describe())
count
         2213.000000
            2.325350
mean
std
            1.924402
min
            0.000000
25%
            1.000000
50%
            2.000000
75%
            3.000000
max
           15.000000
Name: NumDealsPurchases, dtype: float64
  print(df['NumWebPurchases'].describe())
count
         2213.000000
            4.087664
mean
std
            2.741664
            0.000000
min
25%
            2.000000
50%
            4.000000
75%
            6.000000
           27.000000
max
Name: NumWebPurchases, dtype: float64
  print(df['NumCatalogPurchases'].describe())
         2213.000000
count
            2.671487
mean
```

```
2.927096
std
\min
            0.000000
25%
            0.000000
50%
            2.000000
75%
            4.000000
max
           28.000000
Name: NumCatalogPurchases, dtype: float64
  print(df['NumStorePurchases'].describe())
         2213.000000
count
mean
            5.805242
            3.250752
std
            0.000000
min
25%
            3.000000
50%
            5.000000
75%
            8.000000
max
           13.000000
Name: NumStorePurchases, dtype: float64
  print(df['NumWebVisitsMonth'].describe())
         2213.000000
count
mean
            5.321735
std
            2.425092
min
            0.000000
25%
            3.000000
50%
            6.000000
75%
            7.000000
max
           20.000000
Name: NumWebVisitsMonth, dtype: float64
  # create scatterplots
  (p9.ggplot(df, p9.aes(x = 'Income', y = 'MntFruits')) +
   p9.geom_point(color = 'maroon') +
   p9.xlim(0,100000) +
   p9.labs(x = "Customer's yearly household income", y = "Amount spent on fruits in last 2 y
          title= "Relationship between customer's yearly household \nincome and amount spent
             caption = "This scatterplot displays the relationship between \ncustomer's yearl
```

/opt/homebrew/lib/python3.9/site-packages/plotnine/layer.py:364: PlotnineWarning: geom_point

Relationship between customer's yearly household income and amount spent on fruits in last 2 years

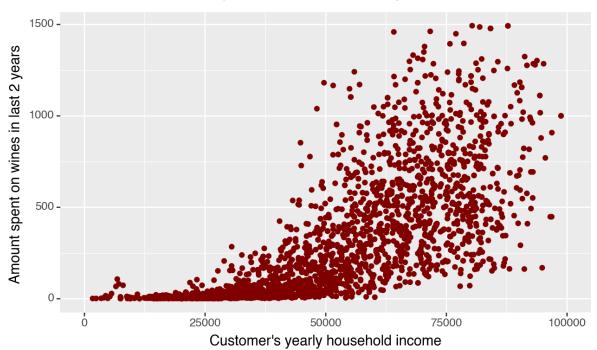


This scatterplot displays the relationship between customer's yearly household income and amount spent on fruits in last 2 years. It has a strong positive linear relationship.

<Figure Size: (640 x 480)>

```
# create scatterplots
(p9.ggplot(df, p9.aes(x = 'Income', y = 'MntWines')) +
    p9.geom_point(color = 'maroon') +
    p9.xlim(0,100000) +
    p9.labs(x = "Customer's yearly household income", y = "Amount spent on wines in last 2 yearly title= "Relationship between customer's yearly household \nincome and amount spent caption = "This scatterplot displays the relationship between \ncustomer's yearly + "It has a \nstrong positive relationship but looks \nlike quadratic relationship
```

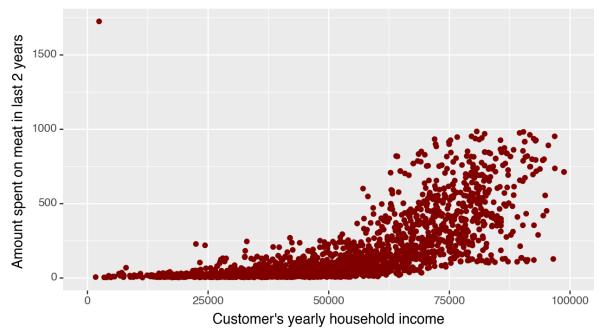
Relationship between customer's yearly household income and amount spent on wines in last 2 years



This scatterplot displays the relationship between customer's yearly household income and amount spent on wines in last 2 years. It has a strong positive relationship but looks like quadratic relationship instead of linear.

<Figure Size: (640 x 480)>

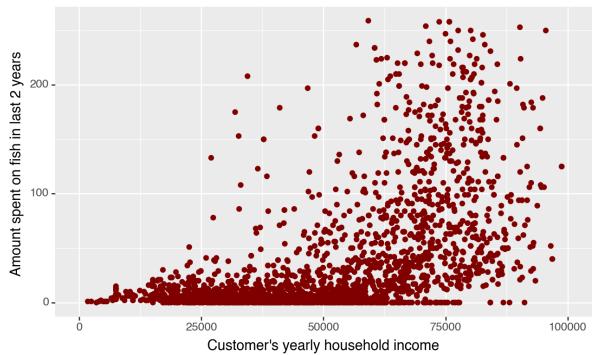
Relationship between customer's yearly household income and amount spent on meat in last 2 years



This scatterplot displays the relationship between customer's yearly household income and amount spent on meat in last 2 years. It has a strong positive relationship but looks like quadratic relationship instead of linear. There is an unusually high value in amount spent against low income.

<Figure Size: (640 x 480)>

Relationship between customer's yearly household income and amount spent on fish in last 2 years

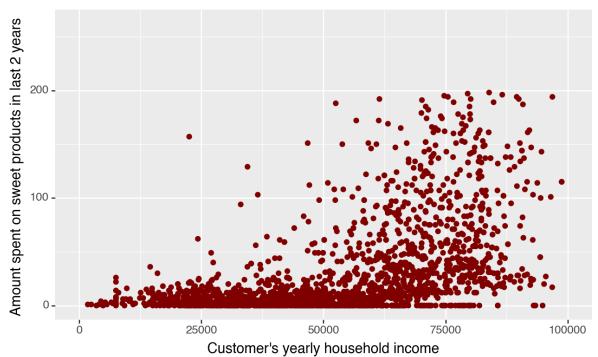


This scatterplot displays the relationship between customer's yearly household income and amount spent on fish in last 2 years. It has a strong positive linear relationship.

<Figure Size: (640 x 480)>

```
# create scatterplots
(p9.ggplot(df, p9.aes(x = 'Income', y = 'MntSweetProducts')) +
p9.geom_point(color = 'maroon') +
p9.xlim(0,100000) +
p9.labs(x = "Customer's yearly household income", y = "Amount spent on sweet products in
    title= "Relationship between customer's yearly household \nincome and amount spent
    caption = "This scatterplot displays the relationship between \ncustomer's yearl
    + "It has a \nstrong positive linear relationship."))
```

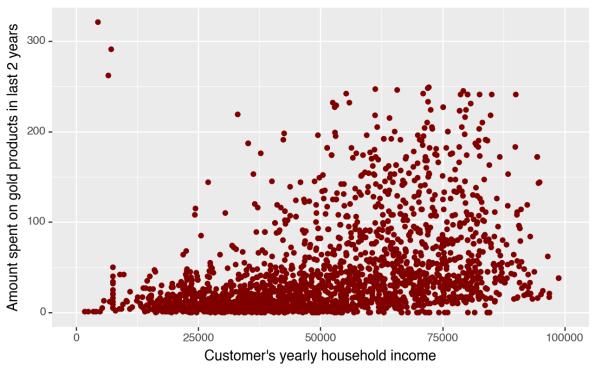
Relationship between customer's yearly household income and amount spent on sweet products in last 2 years



This scatterplot displays the relationship between customer's yearly household income and amount spent on sweet products in last 2 years. It has a strong positive linear relationship.

<Figure Size: (640 x 480)>

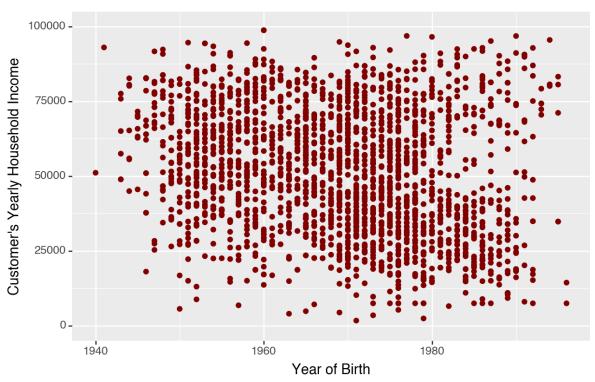
Relationship between customer's yearly household income and amount spent on gold products in last 2 years



This scatterplot displays the relationship between customer's yearly household income and amount spent on gold products in last 2 years.It has a \moderate positive linear relationship.

<Figure Size: (640 x 480)>

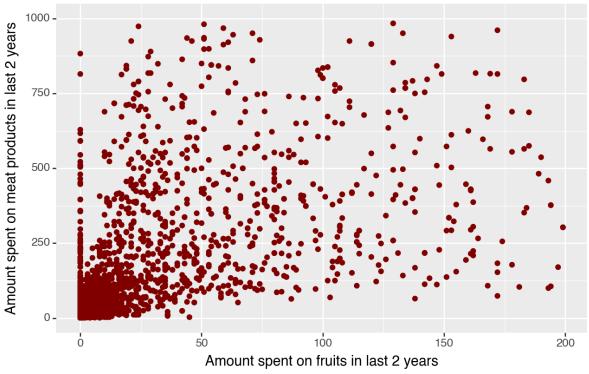
Relationship between year of birth and customer's yearly household income



This scatterplot displays the relationship between year of birth and customer's yearly household income. It has no linear relationship.

<Figure Size: (640 x 480)>

Relationship between amount spent on fruits in last 2 years and amount spent on meat products in last 2 years



This scatterplot displays the relationship between amount spent on fruits in last 2 years and amount spent on meat products in last 2 years. It has no linear relationship.

<Figure Size: (640 x 480)>