```
In [1]: import numpy as np
In [2]: import pandas as pd
```



CMSC 320 / HomeWork4 Published at Mar 27, 2023 Unlisted

```
df = pd.read_csv('Dataset3.csv')
# Count the number of people who got superpowers with the serum
num_super_serum = df[df['SuperSerum'] == 1]['GotPowers'].sum()

# Count the number of people who got superpowers naturally
num_natural = df[df['SuperSerum'] == 0]['GotPowers'].sum()

# Print the results
print(f"Number of people who got superpowers with the serum: {num_super_serum}")
print(f"Number of people who got superpowers naturally: {num_natural}")

Number of people who got superpowers with the serum: 44902
Number of people who got superpowers naturally: 44145
```

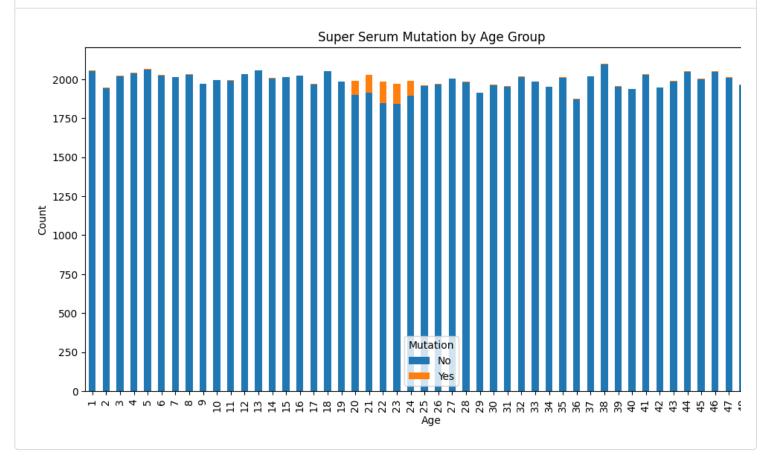
```
from scipy.stats import chi2_contingency
# Load the dataset into a pandas dataframe
df = pd.read_csv('Dataset3.csv')
# Create a contingency table
contingency_table = pd.crosstab(df['GotPowers'], df['SuperSerum'])
from scipy.stats import chi2_contingency
stat, p, dof, expected = chi2_contingency(contingency_table)
print('chi-square statistic:', stat)
print('p-value:', p)
print('degrees of freedom:', dof)
print('expected frequencies:', expected)
chi-square statistic: 102.89321293670491
p-value: 3.53709332538169e-24
degrees of freedom: 1
expected frequencies: [[ 5491.61514 5461.38486]
 [44646.38486 44400.61514]]
```

```
import matplotlib.pyplot as plt
import pandas as pd
df = pd.read_csv('Dataset3.csv')

# create a new column for mutation status
df["Mutation"] = df["GotMutation"].apply(lambda x: "No" if x==0 else "Yes")

# group the data by age and mutation status
age_groups = df.groupby(["Age", "Mutation"]).size().unstack()

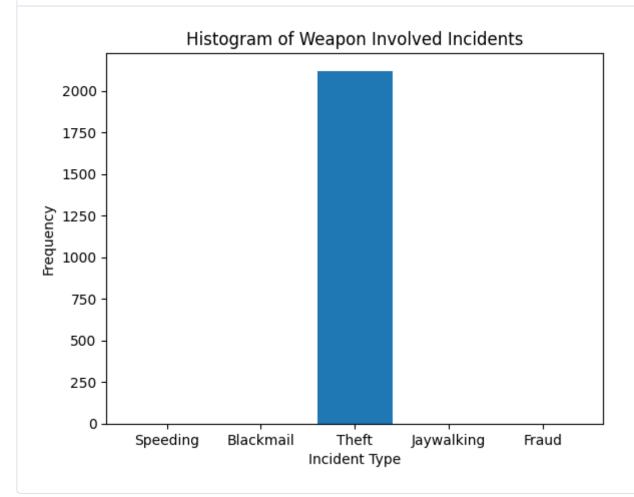
# plot the bar chart
ax = age_groups.plot(kind="bar", stacked=True, figsize=(12,6))
ax.set_xlabel("Age")
ax.set_ylabel("Count")
ax.set_title("Super Serum Mutation by Age Group")
plt.show()
```



```
import matplotlib.pyplot as plt
# Load the dataset into a pandas dataframe
df = pd.read_csv('Dataset1.csv')
# Identify incidents of Speeding, Blackmail, Theft, Jaywalking, and Fraud in the Weapon_Invol
ic = ['Speeding', 'Blackmail', 'Theft', 'Jaywalking', 'Fraud']
weapon_involved = df.loc[df['Weapon_Involved'] == True, 'Type']
```

```
types = weapon_involved.value_counts().reindex(ic, fill_value=0)

# Plot a histogram of the identified incidents
plt.bar(types.index, types.values)
plt.title("Histogram of Weapon Involved Incidents")
plt.xlabel("Incident Type")
plt.ylabel("Frequency")
plt.show()
```



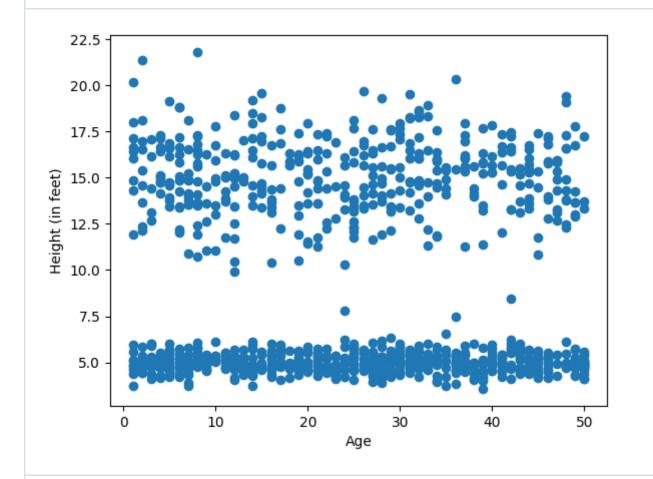
```
import pandas as pd
import matplotlib.pyplot as plt

# Load the dataset
df = pd.read_csv('Dataset2.csv')

# Create a scatter plot of Age vs. Height_in_feet
plt.scatter(df['Age'], df['Height_in_feet'])
plt.xlabel('Age')
plt.ylabel('Height (in feet)')
plt.show()

# Group the data by Age and calculate the mean height for each group
age_groups = df.groupby('Age')['Height_in_feet'].mean().reset_index()
```

Display the results
print(age_groups)



```
0.011072
38
     39
                8.346501
39
     40
                8.140776
                8.259033
40
     41
41
     42
                9.867789
42
     43
                8.076909
43
     44
               10.044014
44
     45
                7.893874
45
     46
                9.348085
46
     47
                12.165970
47
     48
               11.459068
     49
                8.548599
48
49
     50
                6.344203
```

```
import pandas as pd
import matplotlib.pyplot as plt
import statsmodels.api as sm
# create the dataframe
df = pd.read_csv('Dataset4.csv')
# filter for unemployed individuals
unemployed_df = df[df['Profession'] == 'Unemployed']
# check for non-zero values in the income column
non_zero_income = unemployed_df[unemployed_df['Income'] != 0]
print(non_zero_income)
# create histogram of income for unemployed individuals
plt.hist(unemployed_df['Income'])
plt.xlabel('Income')
plt.ylabel('Frequency')
plt.title('Income Distribution for Unemployed Individuals')
plt.show()
     Unnamed: 0
                 Age Gender Profession
                                              Income Job_Sat
3
              3
                  38
                        Male Unemployed 3700.764831
                                                            0
                                                            2
4
              4
                  55
                        Male Unemployed 5555.823704
                                                            2
6
              6
                  44
                        Male Unemployed 7222.011737
17
             17
                  54
                        Male Unemployed 6242.092048
                                                            5
21
             21
                     Female Unemployed 7096.535467
                                                            1
                  31
            . . .
                 . . .
. . .
9976
           9976
                  58
                      Female Unemployed 2198.367847
                                                            0
9979
           9979
                  28
                      Female Unemployed 2189.989185
                                                            3
9980
           9980
                      Female Unemployed 4372.559115
                                                            2
                  51
                                                            4
9981
           9981
                  23
                      Female Unemployed 5788.317711
9985
           9985
                  39
                        Male Unemployed 2627.314924
                                                            4
```

[1512 rows x 6 columns]

