

# Age-Income Dataset

```
In [1]: import pandas as pd
```

```
In [2]: age_df = pd.read_csv('Age-Income.csv')
```

```
In [4]: age_df.head(3)
```

```
Out[4]:
```

	Age	Income
0	Young	25000
1	Middle Age	54000
2	Old	60000

```
In [27]: def calculate_mean(column):
    length = len(column)
    total_sum = 0
    for value in column:
        total_sum += value
    mean = total_sum / length
    print("Mean Value: ", mean)

def calculate_median(column):
    column.sort_values()
    length = len(column)
    if length % 2 == 0:
        first = column[length//2]
        second = column[length//2 - 1]
        median = (first + second)/2
    else:
        median = column[length//2]
    print("Median :",median)

def minimum_value(column):
    min_val = column[0]
    for value in column:
        if value < min_val:
            min_val = value
    print("Minimum value :", min_val)

def maximum_value(column):
    max_val = column[0] # Initialize max_val with the first element of the column
    for value in column:
        if value > max_val:
            max_val = value
    print("Maximum value :", max_val)

def standard_deviation(column):
    length = len(column)
    total_sum = 0
    summation = 0
    for value in column:
        total_sum += value
    mean = total_sum / length
    for i in column:
        summation+= (i-mean)**2
    std_deviation = ((summation)/length)**0.5
    print("Standard Deviation: ", std_deviation)
```

```
In [63]: calculate_mean(age_df['Income'])
         calculate_median(age_df['Income'])
         minimum_value(age_df['Income'])
         maximum_value(age_df['Income'])
         standard_deviation(age_df['Income'])
```

```
Mean Value:  50966.0
Median  : 91500.0
Minimum value : 15000
Maximum value : False
Standard Deviation:  20884.6509187968
```

```
In [66]: age_df.groupby('Age')['Income'].mean()
```

```
Out[66]: Age
Middle Age    52453.333333
Old           53942.105263
Young         46037.500000
Name: Income, dtype: float64
```

```
In [67]: age_df.groupby('Age')['Income'].median()
```

```
Out[67]: Age
Middle Age    53200.0
Old           45300.0
Young         41500.0
Name: Income, dtype: float64
```

```
In [69]: age_df.groupby('Age')['Income'].std()
```

```
Out[69]: Age
Middle Age    20497.800114
Old           20868.165968
Young         22356.859499
Name: Income, dtype: float64
```

## Iris Dataset

```
In [77]: iris_df = pd.read_csv("iris.csv")
         iris_df.drop('Id', axis = 1, inplace = True)
```

```
In [78]: categories = [i for i in iris_df['Species'].unique()]
         categories
```

```
Out[78]: ['Iris-setosa', 'Iris-versicolor', 'Iris-virginica']
```

```
In [79]: features = [feat for feat in iris_df.columns if iris_df[feat].dtype != 'O']
         features
```

```
Out[79]: ['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm']
```

```
In [80]: species_group = iris_df.groupby('Species')
```

```
In [81]: species_group.mean()
```

Out[81]:

Species	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
Iris-setosa	5.006	3.418	1.464	0.244
Iris-versicolor	5.936	2.770	4.260	1.326
Iris-virginica	6.588	2.974	5.552	2.026

In [82]:

```
species_group.std()
```

Out[82]:

Species	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
Iris-setosa	0.352490	0.381024	0.173511	0.107210
Iris-versicolor	0.516171	0.313798	0.469911	0.197753
Iris-virginica	0.635880	0.322497	0.551895	0.274650

In [83]:

```
species_group.var()
```

Out[83]:

Species	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
Iris-setosa	0.124249	0.145180	0.030106	0.011494
Iris-versicolor	0.266433	0.098469	0.220816	0.039106
Iris-virginica	0.404343	0.104004	0.304588	0.075433

In [87]:

```
print('Iris-setosa')
setosa = iris_df['Species'] == 'Iris-setosa'
print(iris_df[setosa].describe())
print('\nIris-versicolor')
versicolor = iris_df['Species'] == 'Iris-versicolor'
print(iris_df[versicolor].describe())
print('\nIris-virginica')
virginica = iris_df['Species'] == 'Iris-virginica'
print(iris_df[virginica].describe())
```

Iris-setosa

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.00000	50.000000	50.000000	50.00000
mean	5.00600	3.418000	1.464000	0.24400
std	0.35249	0.381024	0.173511	0.10721
min	4.30000	2.300000	1.000000	0.10000
25%	4.80000	3.125000	1.400000	0.20000
50%	5.00000	3.400000	1.500000	0.20000
75%	5.20000	3.675000	1.575000	0.30000
max	5.80000	4.400000	1.900000	0.60000

Iris-versicolor

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.000000	50.000000	50.000000	50.000000
mean	5.936000	2.770000	4.260000	1.326000
std	0.516171	0.313798	0.469911	0.197753
min	4.900000	2.000000	3.000000	1.000000
25%	5.600000	2.525000	4.000000	1.200000
50%	5.900000	2.800000	4.350000	1.300000
75%	6.300000	3.000000	4.600000	1.500000
max	7.000000	3.400000	5.100000	1.800000

Iris-virginica

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.00000	50.000000	50.000000	50.00000
mean	6.58800	2.974000	5.552000	2.02600
std	0.63588	0.322497	0.551895	0.27465
min	4.90000	2.200000	4.500000	1.40000
25%	6.22500	2.800000	5.100000	1.80000
50%	6.50000	3.000000	5.550000	2.00000
75%	6.90000	3.175000	5.875000	2.30000
max	7.90000	3.800000	6.900000	2.50000

In [90]:

```
def covariance(x, y):
    mean_x = sum(x)/len(x)
    mean_y = sum(y)/len(y)
    sub_x = [i - mean_x for i in x]
    sub_y = [i - mean_y for i in y]
    numerator = sum([sub_x[i]*sub_y[i] for i in range(len(sub_x))])
    denominator = len(x)-1
    cov = numerator/denominator
    return cov
```

In [94]:

```
for i in [0,1,2,3]:
    for j in [0,1,2,3]:
        if (i < j and i != j):
            val = covariance(iris_df[features[i]],iris_df[features[j]])
            print('Covariance for {} and {} : {}'.format(features[i],features[j], val))
```

Covariance for SepalLengthCm and SepalWidthCm : -0.03926845637583892  
Covariance for SepalLengthCm and PetalLengthCm : 1.2736823266219242  
Covariance for SepalLengthCm and PetalWidthCm : 0.5169038031319912  
Covariance for SepalWidthCm and PetalLengthCm : -0.32171275167785246  
Covariance for SepalWidthCm and PetalWidthCm : -0.11798120805369122  
Covariance for PetalLengthCm and PetalWidthCm : 1.2963874720357946

In [95]:

```
def correlation(x, y):
    mean_x = sum(x)/float(len(x))
    mean_y = sum(y)/float(len(y))
    sub_x = [i-mean_x for i in x]
    sub_y = [i-mean_y for i in y]
    numerator = sum([sub_x[i]*sub_y[i] for i in range(len(sub_x))])
    std_deviation_x = sum([sub_x[i]**2.0 for i in range(len(sub_x))])
    std_deviation_y = sum([sub_y[i]**2.0 for i in range(len(sub_y))])
    denominator = (std_deviation_x*std_deviation_y)**0.5
    cor = numerator/denominator
    return cor
```

In [96]:

```
for i in [0,1,2,3]:
    for j in [0,1,2,3]:
        if (i < j and i != j):
            val = correlation(iris_df[features[i]],iris_df[features[j]])
            print('Correlation coefficient for {} and {} : {}'.format(features[i],features[j],
```

Correlation coefficient for SepalLengthCm and SepalWidthCm : -0.10936924995064935  
Correlation coefficient for SepalLengthCm and PetalLengthCm : 0.8717541573048719  
Correlation coefficient for SepalLengthCm and PetalWidthCm : 0.8179536333691635  
Correlation coefficient for SepalWidthCm and PetalLengthCm : -0.42051609640115484  
Correlation coefficient for SepalWidthCm and PetalWidthCm : -0.3565440896138055  
Correlation coefficient for PetalLengthCm and PetalWidthCm : 0.9627570970509667

In [107...]

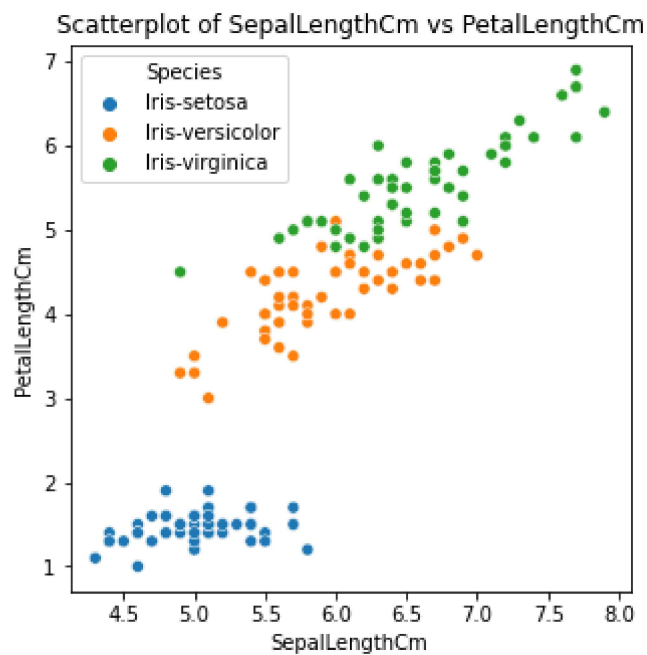
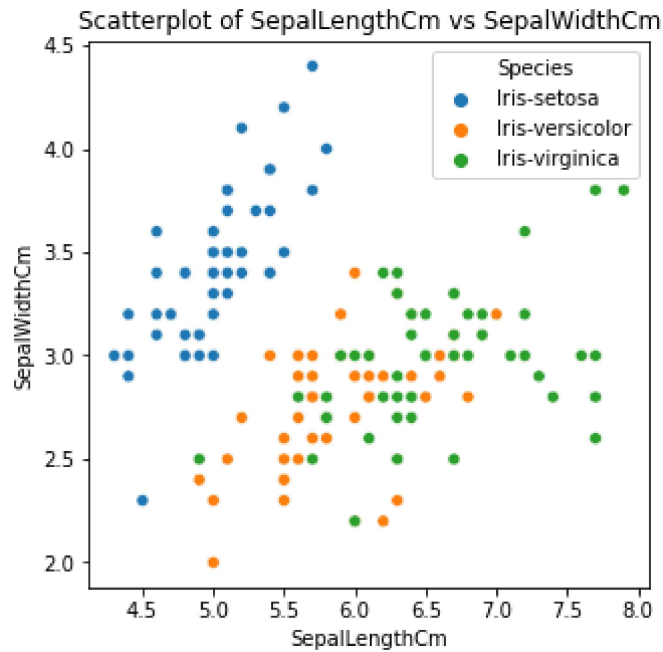
```
import matplotlib.pyplot as plt
import seaborn as sns

features = [feat for feat in iris_df.columns if iris_df[feat].dtype != 'O']
```

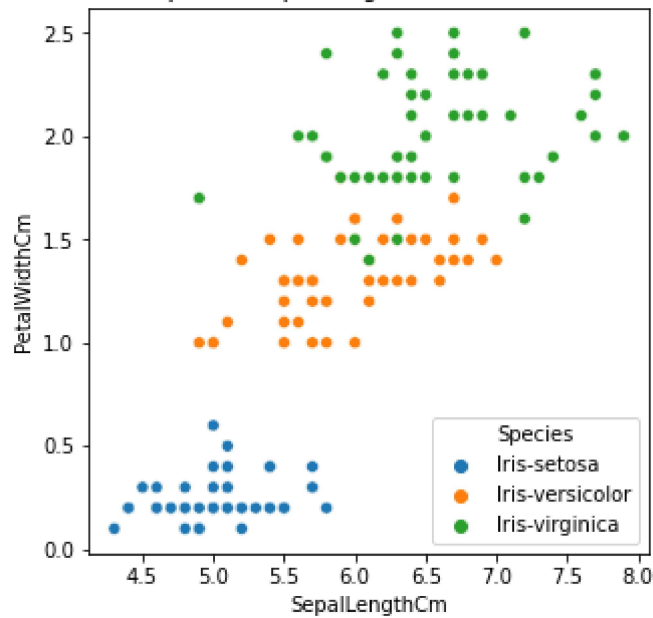
```

for i in range(len(features)):
    for j in range(len(features)):
        if (i < j and i != j):
            fig = plt.figure()
            fig.set_figheight(5)
            fig.set_figwidth(5)
            ax = sns.scatterplot(x=features[i], y=features[j], data=iris_df, hue='Species')
            plt.xlabel(features[i])
            plt.ylabel(features[j])
            plt.title('Scatterplot of {} vs {}'.format(features[i], features[j]))
            plt.show()

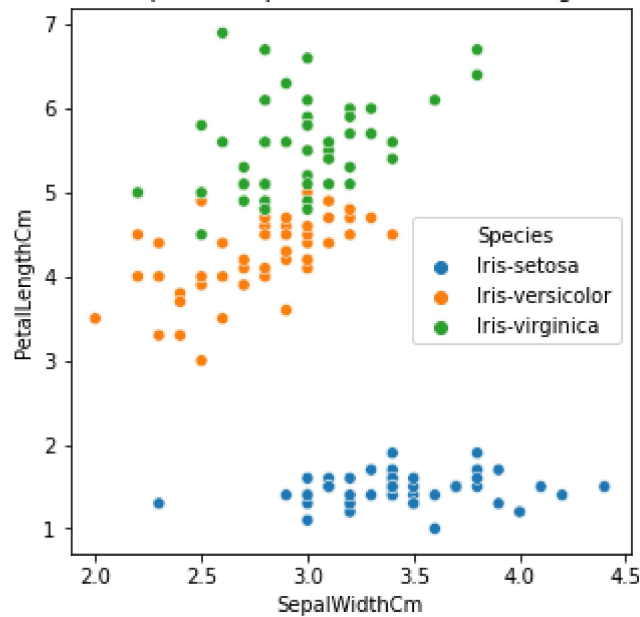
```



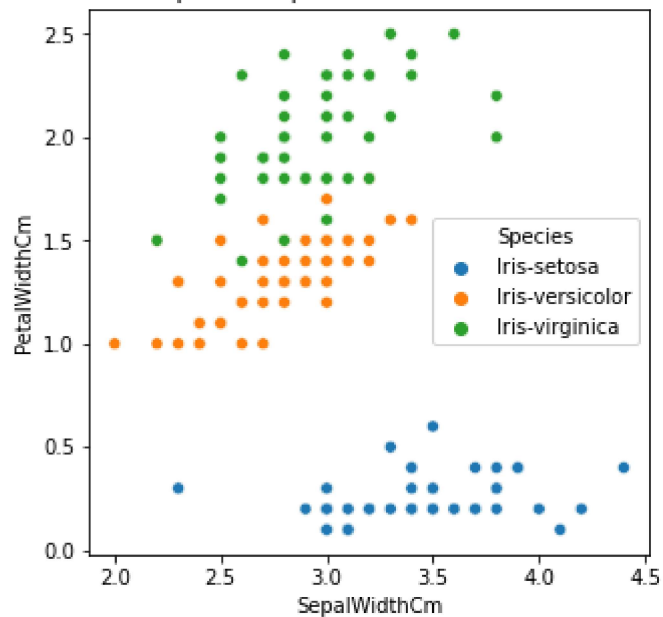
Scatterplot of SepalLengthCm vs PetalWidthCm



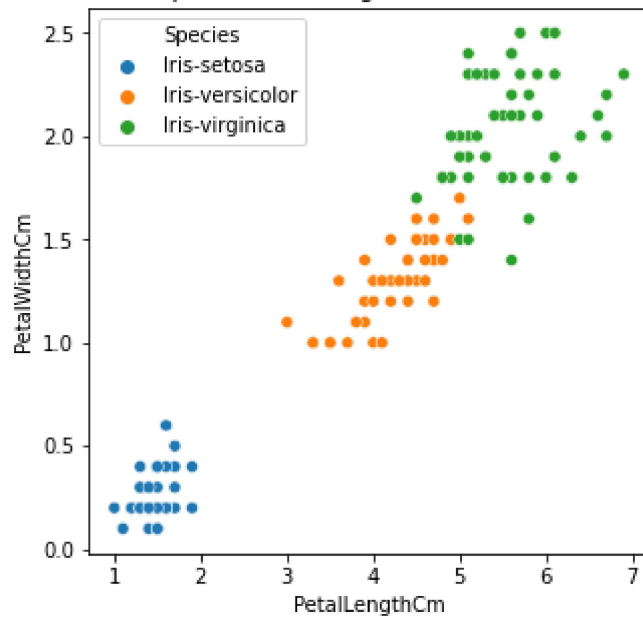
Scatterplot of SepalWidthCm vs PetalLengthCm



Scatterplot of SepalWidthCm vs PetalWidthCm



Scatterplot of PetalLengthCm vs PetalWidthCm



In [108...

```
cormatrix = iris_df.corr()
round(cormatrix,4)
```

Out[108...

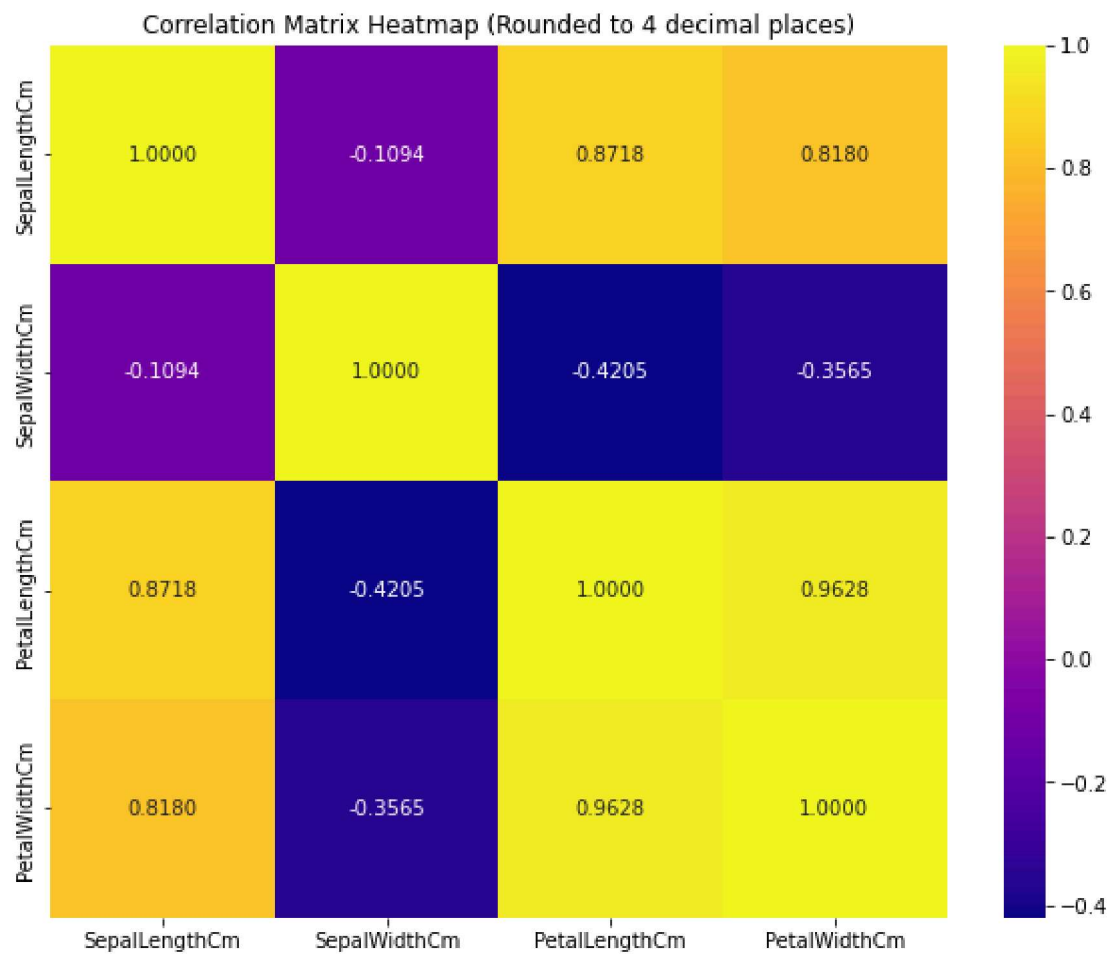
	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
SepalLengthCm	1.0000	-0.1094	0.8718	0.8180
SepalWidthCm	-0.1094	1.0000	-0.4205	-0.3565
PetalLengthCm	0.8718	-0.4205	1.0000	0.9628
PetalWidthCm	0.8180	-0.3565	0.9628	1.0000

In [109...

```
import seaborn as sns
import matplotlib.pyplot as plt

cormatrix = iris_df.corr()
cormatrix_rounded = round(cormatrix, 4)

plt.figure(figsize=(10, 8))
sns.heatmap(cormatrix_rounded, annot=True, cmap='plasma', fmt=".4f", annot_kws={"size": 10})
plt.title('Correlation Matrix Heatmap (Rounded to 4 decimal places)')
plt.show()
```



In [ ]: