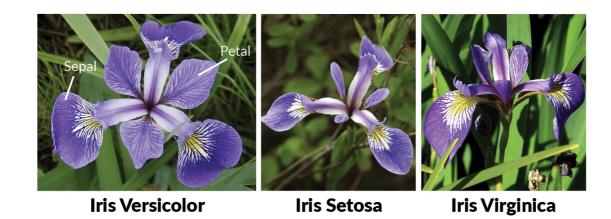
Iris Flower Classification

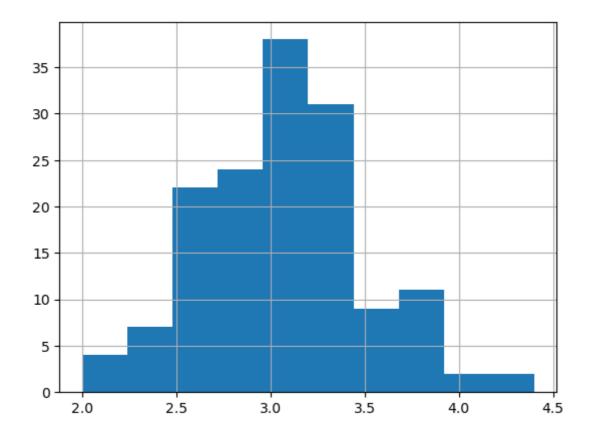


Importing Libraries

```
In [1]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
          import warnings
          warnings.filterwarnings('ignore')
In [4]: iris_data = pd.read_csv("iris.csv", names=['sepal_length', 'sepal_width', 'petal_le
          iris_data.head()
             sepal_length sepal_width petal_length petal_width
Out[4]:
                                                                    species
          0
                      5.1
                                   3.5
                                                 1.4
                                                             0.2 Iris-setosa
          1
                      4.9
                                   3.0
                                                 1.4
                                                             0.2 Iris-setosa
          2
                                   3.2
                      4.7
                                                 1.3
                                                             0.2 Iris-setosa
          3
                      4.6
                                   3.1
                                                 1.5
                                                             0.2 Iris-setosa
                      5.0
                                   3.6
                                                 1.4
                                                             0.2 Iris-setosa
          iris_data.tail()
In [5]:
Out[5]:
               sepal_length sepal_width petal_length petal_width
                                                                        species
          145
                        6.7
                                     3.0
                                                   5.2
                                                                2.3 Iris-virginica
          146
                        6.3
                                     2.5
                                                   5.0
                                                                1.9 Iris-virginica
                        6.5
                                                   5.2
          147
                                     3.0
                                                                2.0 Iris-virginica
          148
                        6.2
                                     3.4
                                                   5.4
                                                                2.3 Iris-virginica
          149
                        5.9
                                     3.0
                                                   5.1
                                                                1.8 Iris-virginica
```

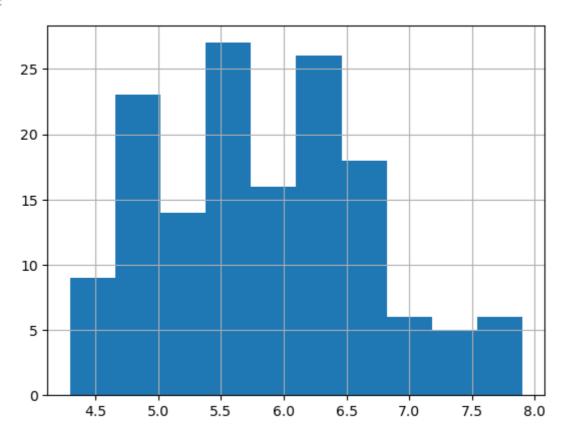
Statistical Data Analysis

```
iris_data.describe()
 In [6]:
 Out[6]:
                sepal_length sepal_width petal_length petal_width
                  150.000000
                             150.000000
                                         150.000000
                                                    150.000000
          count
                                           3.758667
                                                      1.198667
          mean
                   5.843333
                               3.054000
                   0.828066
                               0.433594
            std
                                           1.764420
                                                      0.763161
           min
                   4.300000
                               2.000000
                                           1.000000
                                                      0.100000
           25%
                   5.100000
                               2.800000
                                           1.600000
                                                      0.300000
           50%
                   5.800000
                               3.000000
                                           4.350000
                                                      1.300000
           75%
                   6.400000
                               3.300000
                                           5.100000
                                                      1.800000
                   7.900000
                               4.400000
                                           6.900000
                                                      2.500000
           max
          #Length of Data
 In [7]:
          iris_data.shape
          (150, 5)
 Out[7]:
         summary of a DataFrame
 In [8]:
         iris_data.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 150 entries, 0 to 149
         Data columns (total 5 columns):
                             Non-Null Count Dtype
          #
              Column
              _____
                             -----
          0
              sepal_length 150 non-null
                                              float64
                                              float64
              sepal_width 150 non-null
          1
               petal_length 150 non-null
                                              float64
               petal_width 150 non-null
                                              float64
               species
                             150 non-null
                                              object
          dtypes: float64(4), object(1)
         memory usage: 6.0+ KB
         #Checking null value
 In [9]:
          iris_data.isnull().sum()
         sepal_length
                          0
 Out[9]:
          sepal_width
                          0
         petal_length
                          0
         petal_width
                          0
          species
                          0
         dtype: int64
         iris_data['species'].value_counts()
In [10]:
         Iris-setosa
                             50
Out[10]:
         Iris-versicolor
                             50
         Iris-virginica
                             50
         Name: species, dtype: int64
          iris_data['sepal_width'].hist()
In [11]:
         <Axes: >
Out[11]:
```



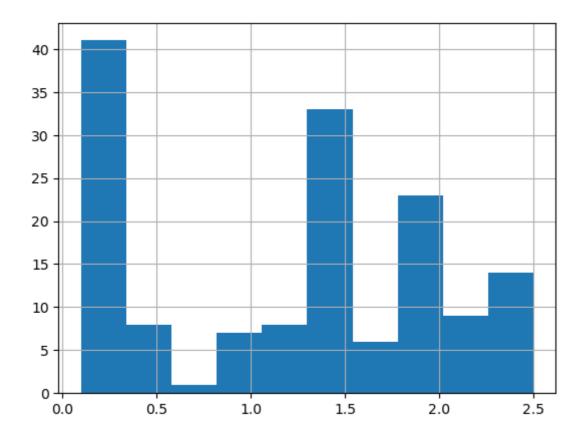
In [12]: iris_data['sepal_length'].hist()

Out[12]: <Axes: >



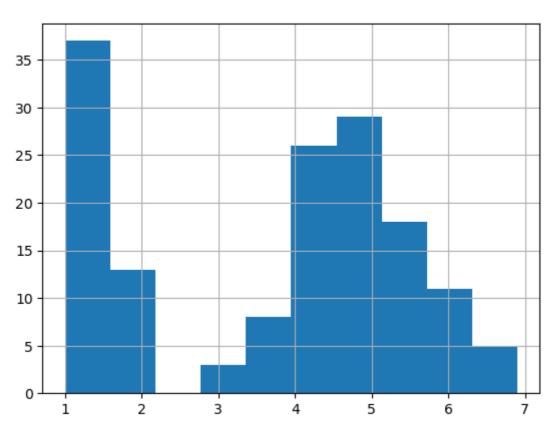
In [13]: iris_data['petal_width'].hist()

Out[13]: <Axes: >

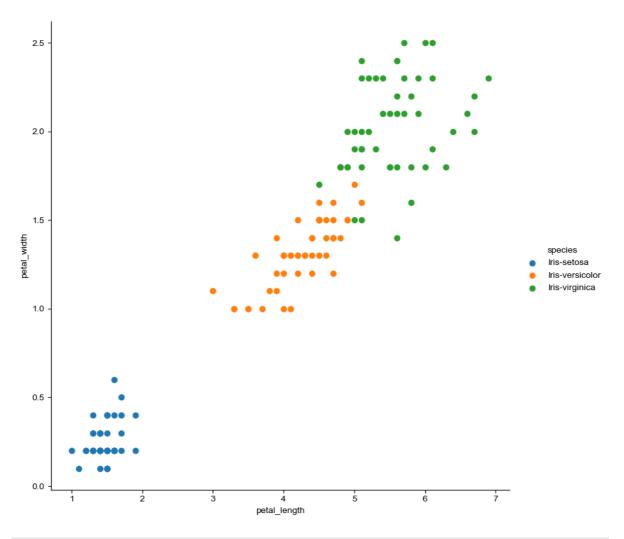


```
In [14]: iris_data['petal_length'].hist()
```

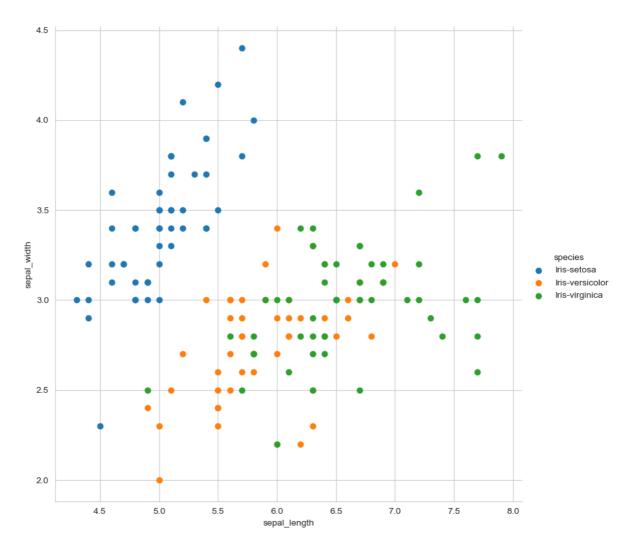
Out[14]: <Axes: >



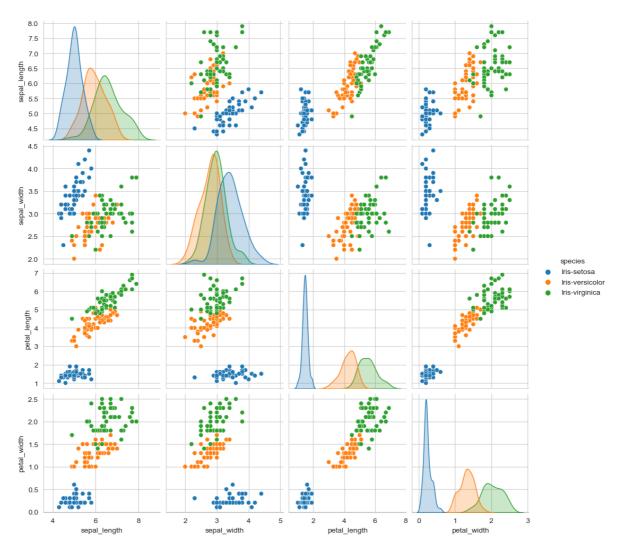
```
In [15]: s = sns.FacetGrid(iris_data, height=8, hue="species")
    s.map(plt.scatter, "petal_length", "petal_width")
    s.add_legend()
    sns.set_style("whitegrid")
    plt.show()
```



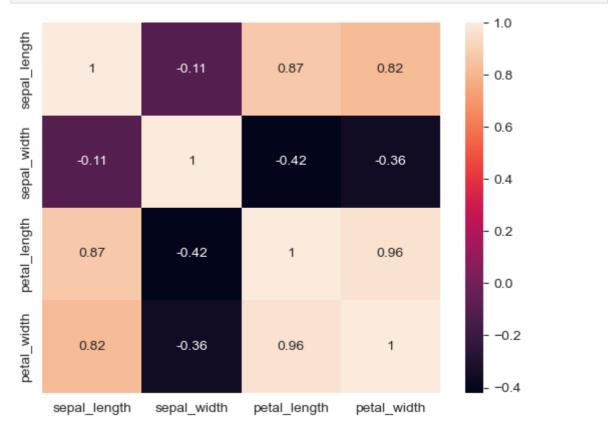
```
In [16]: s = sns.FacetGrid(iris_data, height=8, hue="species")
    s.map(plt.scatter, "sepal_length", "sepal_width")
    s.add_legend()
    sns.set_style("whitegrid")
    plt.show()
```



In [17]: sns.pairplot(iris_data, height=2.5, hue="species")
 plt.show()



In [18]: #Checking Correlation use of Heatmap
sns.heatmap(iris_data.corr(), annot=True)
plt.show()



Split the data into training and testing

```
In [19]: from sklearn.model_selection import train_test_split
          X = iris_data[["sepal_length", "sepal_width", "petal_length", "petal_width"]]
          y = iris_data["species"]
In [20]:
               sepal_length sepal_width petal_length petal_width
Out[20]:
            0
                       5.1
                                   3.5
                                               1.4
                                                           0.2
                                   3.0
            1
                       4.9
                                               1.4
                                                           0.2
                                   3.2
            2
                       4.7
                                               1.3
                                                           0.2
            3
                                   3.1
                       4.6
                                               1.5
                                                           0.2
            4
                       5.0
                                   3.6
                                               1.4
                                                           0.2
          145
                       6.7
                                   3.0
                                               5.2
                                                           2.3
          146
                       6.3
                                   2.5
                                               5.0
                                                           1.9
                       6.5
                                   3.0
                                               5.2
                                                           2.0
          147
          148
                       6.2
                                   3.4
                                               5.4
                                                           2.3
                                               5.1
          149
                       5.9
                                   3.0
                                                           1.8
         150 rows × 4 columns
In [21]:
                     Iris-setosa
Out[21]:
                     Iris-setosa
          2
                     Iris-setosa
          3
                     Iris-setosa
                     Iris-setosa
          145
                 Iris-virginica
          146
                 Iris-virginica
          147
                 Iris-virginica
          148
                 Iris-virginica
          149
                 Iris-virginica
          Name: species, Length: 150, dtype: object
In [22]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, random_st
          Logistic regression model
In [23]:
          from sklearn.linear_model import LogisticRegression
          model=LogisticRegression()
          model.fit(X_train,y_train)
In [24]:
```

```
Out[24]:
         ▼ LogisticRegression
         LogisticRegression()
         #metrics to get performance
In [25]:
         print('Accuracy', model.score(X_test,y_test)*100)
         Accuracy 97.777777777777
         K-Nearest Neighbours model
In [26]:
         from sklearn.neighbors import KNeighborsClassifier
         model=KNeighborsClassifier()
         model.fit(X_train,y_train)
In [27]:
Out[27]: ▼ KNeighborsClassifier
         KNeighborsClassifier()
         #metrics to get performance
In [28]:
         print('Accuracy', model.score(X_test,y_test)*100)
         Accuracy 97.777777777777
         Decision tree model
In [29]:
         from sklearn.tree import DecisionTreeClassifier
         model=DecisionTreeClassifier()
In [30]:
         model.fit(X_train,y_train)
Out[30]:
        ▼ DecisionTreeClassifier
         DecisionTreeClassifier()
In [31]:
         #metrics to get performance
         print('Accuracy', model.score(X_test,y_test)*100)
         Accuracy 95.55555555556
```

You can find this project on GitHub.