Iris_Flower_Classification







Iris Versicolor

Iris Setosa

Iris Virginica

Importing Libraries

In [37]: import numpy as np
import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

import warnings

warnings.filterwarnings('ignore')

Out[38]:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

In [7]: iris_data.tail()

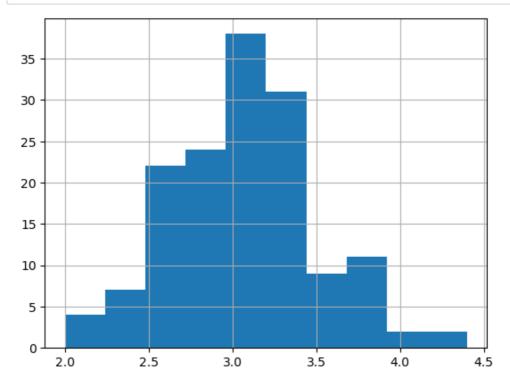
Out[7]:

	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
147	6.5	3.0	5.2	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

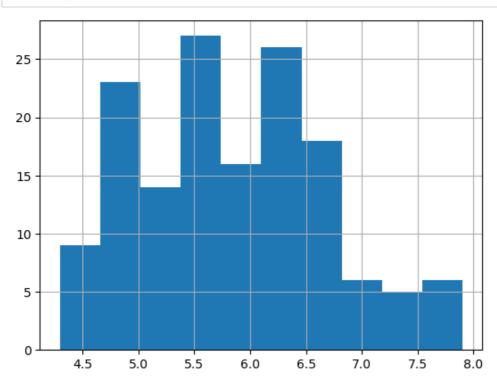
Statistical Data Analysis

```
In [8]: | iris_data.describe()
Out[8]:
                sepal_length sepal_width petal_length petal_width
                 150.000000
                            150.000000
                                       150.000000
                                                  150.000000
          count
                              3.054000
                                                    1.198667
          mean
                   5.843333
                                         3.758667
            std
                   0.828066
                              0.433594
                                         1.764420
                                                    0.763161
                   4.300000
                              2.000000
                                         1.000000
                                                    0.100000
           min
           25%
                   5.100000
                              2.800000
                                         1.600000
                                                    0.300000
           50%
                   5.800000
                              3.000000
                                         4.350000
                                                    1.300000
                   6.400000
                              3.300000
                                         5.100000
                                                    1.800000
           75%
                   7.900000
                              4.400000
                                         6.900000
                                                    2.500000
           max
In [9]: #Length of Data
         iris_data.shape
Out[9]: (150, 5)
         summary of a DataFrame
In [10]: 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
              sepal width 150 non-null
                                             float64
              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
In [11]: #Checking null value
         iris data.isnull().sum()
Out[11]: sepal_length
                          0
         sepal_width
                          0
         petal_length
                          0
         petal_width
                          0
         species
                          0
         dtype: int64
In [12]: | iris_data['species'].value_counts()
Out[12]: Iris-setosa
                             50
                             50
         Iris-versicolor
         Iris-virginica
                             50
         Name: species, dtype: int64
```

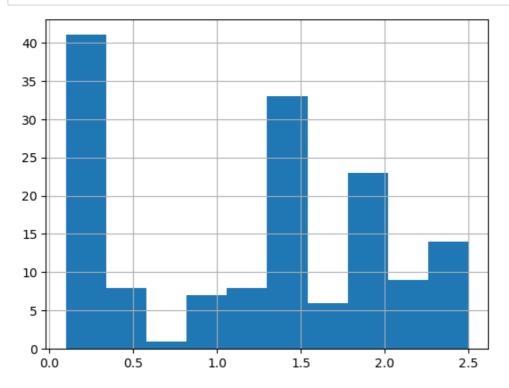
In [15]: iris_data['sepal_width'].hist()
plt.show()



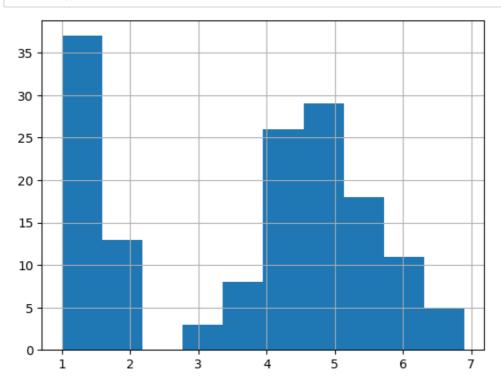
In [16]: iris_data['sepal_length'].hist()
 plt.show()



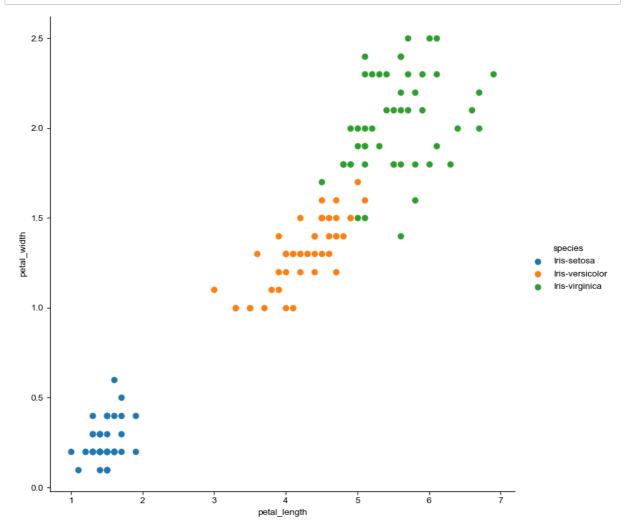
In [17]: iris_data['petal_width'].hist()
 plt.show()



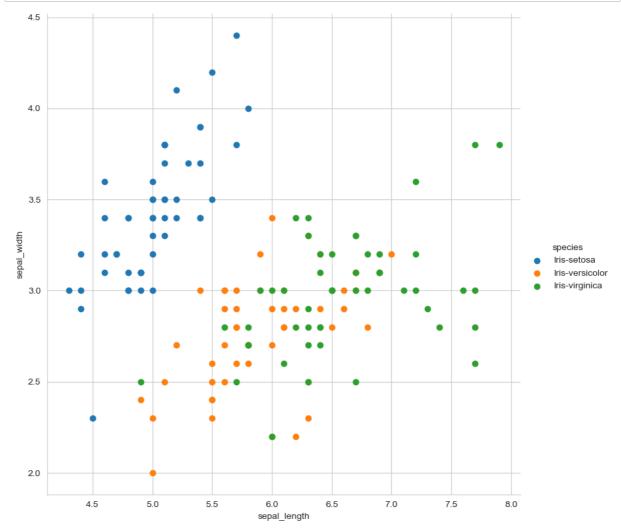
In [18]: iris_data['petal_length'].hist()
plt.show()



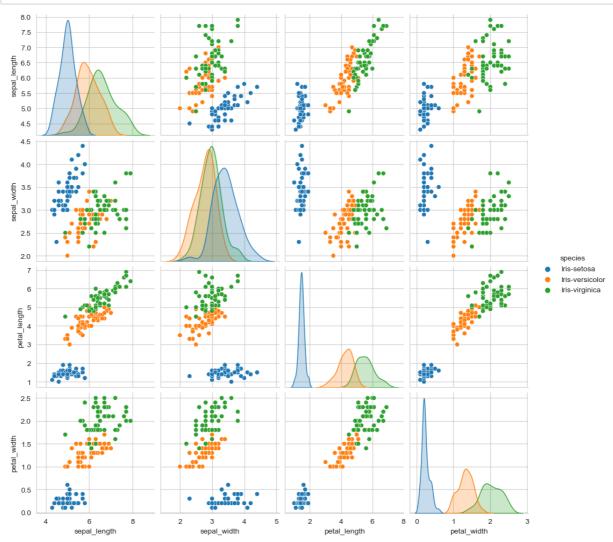
```
In [19]: s = sns.FacetGrid(iris_data, height=8, hue="species")
    s.map(plt.scatter, "petal_length", "petal_width")
    s.add_legend()
    sns.set_style("whitegrid")
    plt.show()
```



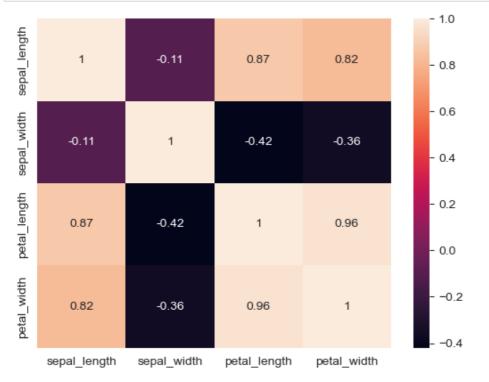
```
In [20]: s = sns.FacetGrid(iris_data, height=8, hue="species")
    s.map(plt.scatter, "sepal_length", "sepal_width")
    s.add_legend()
    sns.set_style("whitegrid")
    plt.show()
```



In [23]: sns.pairplot(iris_data, height=2.5, hue="species")
plt.show()



In [22]: #Checking Correlation use of Heatmap
sns.heatmap(iris_data.corr(), annot=True)
plt.show()



Split the data into training and testing

```
In [24]: | from sklearn.model_selection import train_test_split
          X = iris_data[["sepal_length", "sepal_width", "petal_length", "petal_width"]]
          y = iris_data["species"]
In [25]: X
Out[25]:
               sepal_length sepal_width petal_length petal_width
            0
                       5.1
                                  3.5
                                              1.4
                                                         0.2
                       4.9
                                  3.0
                                              1.4
                                                         0.2
            1
            2
                       4.7
                                  3.2
                                              1.3
                                                         0.2
            3
                       4.6
                                  3.1
                                              1.5
                                                         0.2
            4
                       5.0
                                  3.6
                                              1.4
                                                         0.2
                                                          ...
           145
                       6.7
                                  3.0
                                              5.2
                                                         2.3
                       6.3
                                  2.5
                                              5.0
           146
                                                         1.9
           147
                       6.5
                                  3.0
                                              5.2
                                                         2.0
           148
                       6.2
                                  3.4
                                              5.4
                                                         2.3
           149
                       5.9
                                  3.0
                                              5.1
                                                         1.8
          150 rows × 4 columns
In [26]: y
Out[26]:
          0
                    Iris-setosa
                    Iris-setosa
          1
          2
                    Iris-setosa
                    Iris-setosa
          3
                    Iris-setosa
          145
                 Iris-virginica
          146
                 Iris-virginica
          147
                 Iris-virginica
                 Iris-virginica
          148
          149
                 Iris-virginica
          Name: species, Length: 150, dtype: object
In [27]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, random_state=12
          Logistic regression model
In [28]:
          from sklearn.linear_model import LogisticRegression
          model=LogisticRegression()
In [29]: model.fit(X_train,y_train)
Out[29]:
          LogisticRegression
          LogisticRegression()
```

```
In [30]: #metrics to get performance
         print('Accuracy', model.score(X_test,y_test)*100)
         Accuracy 97.777777777777
         K-Nearest Neighbours model
In [31]: from sklearn.neighbors import KNeighborsClassifier
         model=KNeighborsClassifier()
In [32]: model.fit(X_train,y_train)
Out[32]:
         KNeighborsClassifier
         KNeighborsClassifier()
In [33]: #metrics to get performance
         print('Accuracy', model.score(X_test,y_test)*100)
         Accuracy 97.777777777777
         Decision tree model
In [34]: | from sklearn.tree import DecisionTreeClassifier
         model=DecisionTreeClassifier()
In [35]: model.fit(X_train,y_train)
Out[35]:
         ▼ DecisionTreeClassifier
         DecisionTreeClassifier()
```

Accuracy 97.777777777777

print('Accuracy', model.score(X_test,y_test)*100)

THANK YOU!

In [36]: | #metrics to get performance

GitHub Link: https://github.com/anujtiwari21?tab=repositories)