```
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')

from sklearn.linear_model import LogisticRegression as lr
from sklearn import metrics
from sklearn.metrics import classification_report
from sklearn.metrics import accuracy_score
```

In [2]: irisdf = pd.read_csv("C:/Users/HP/Downloads/iris.csv")
#first 5 dataset
irisdf.head()

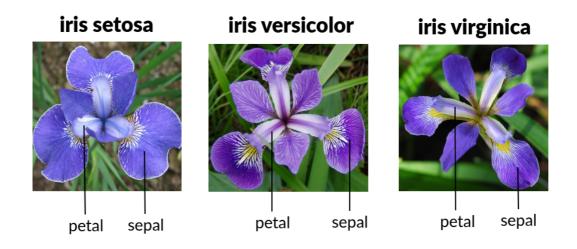
Out[2]:		ld	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
	0	1	5.1	3.5	1.4	0.2	Iris-setosa
	1	2	4.9	3.0	1.4	0.2	Iris-setosa
	2	3	4.7	3.2	1.3	0.2	Iris-setosa
	3	4	4.6	3.1	1.5	0.2	Iris-setosa
	4	5	5.0	3.6	1.4	0.2	Iris-setosa

In [3]: #last 5 dataset
 irisdf.tail()

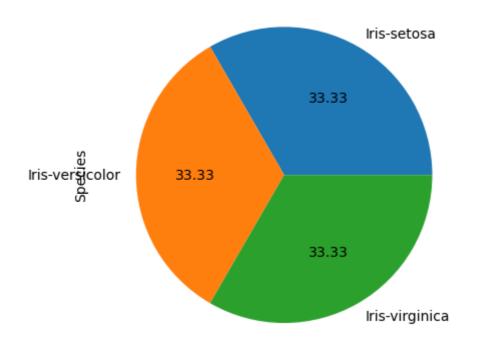
Out[3]:		Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
	145	146	6.7	3.0	5.2	2.3	Iris-virginica
	146	147	6.3	2.5	5.0	1.9	Iris-virginica
	147	148	6.5	3.0	5.2	2.0	Iris-virginica
	148	149	6.2	3.4	5.4	2.3	Iris-virginica
	149	150	5.9	3.0	5.1	1.8	Iris-virginica

In [4]: from IPython import display
 display.Image("C:/Users/HP/Downloads/iris.png")

Out[4]:



```
In [5]: irisdf.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 150 entries, 0 to 149
        Data columns (total 6 columns):
                    Non-Null Count Dtype
            Column
            ----
                          -----
                          150 non-null int64
         0
            Ιd
         1 SepalLengthCm 150 non-null float64
         2 SepalWidthCm 150 non-null float64
         3 PetalLengthCm 150 non-null float64
            PetalWidthCm 150 non-null float64
                           150 non-null object
         5
            Species
        dtypes: float64(4), int64(1), object(1)
        memory usage: 7.2+ KB
        irisdf.shape
In [6]:
        (150, 6)
Out[6]:
In [7]:
        irisdf.size
        900
Out[7]:
In [8]: #check for values of each species whether its balanced or imbalanced
        irisdf['Species'].value_counts()
        Iris-setosa
                          50
Out[8]:
        Iris-versicolor
                          50
        Iris-virginica
                          50
        Name: Species, dtype: int64
In [9]: # Visualization in form of pie chart
        irisdf['Species'].value_counts().plot(kind='pie',autopct='%.2f')
        <AxesSubplot:ylabel='Species'>
Out[9]:
```



```
#Check if any null values is present in give dataset
In [10]:
          irisdf.isnull().sum()
          Ιd
                            0
Out[10]:
          SepalLengthCm
                            0
          SepalWidthCm
                            0
          PetalLengthCm
                            0
          {\tt PetalWidthCm}
                            0
          Species
                            0
          dtype: int64
In [11]:
          #Check for duplicate values
          duplicate_count = irisdf.duplicated().sum()
          print(duplicate_count)
          0
          sns.FacetGrid(irisdf,hue='Species',height=4).map(plt.scatter,"SepalWidthCm" , "Peta
In [12]:
          <seaborn.axisgrid.FacetGrid at 0x23c093c2e20>
Out[12]:
              7
              6
              5
          PetalLengthCm
                                                                        Species
                                                                        Iris-setosa
                                                                        Iris-versicolor
             3
                                                                        Iris-virginica
```

In [13]: plt.figure(figsize =(15,6))
 sns.countplot(x='SepalLengthCm', data=irisdf, hue= irisdf['Species'], palette= 'wir
 plt.show()

3.5

4.0

4.5

2

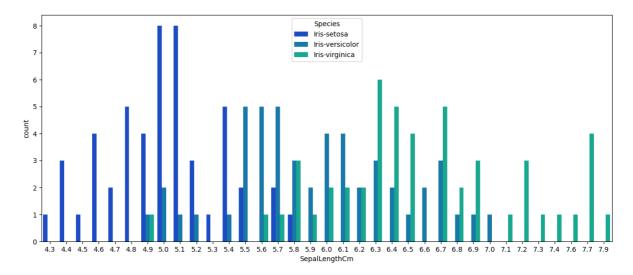
1

2.0

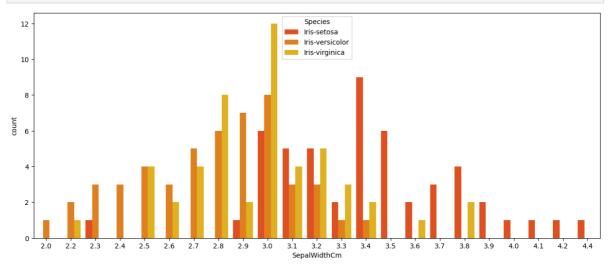
2.5

3.0

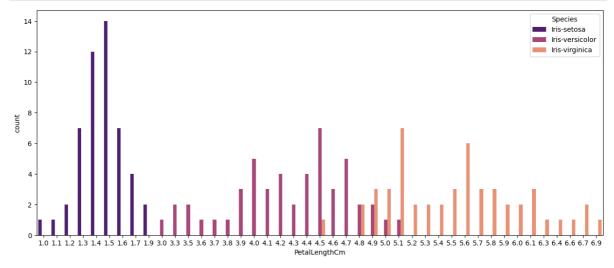
SepalWidthCm



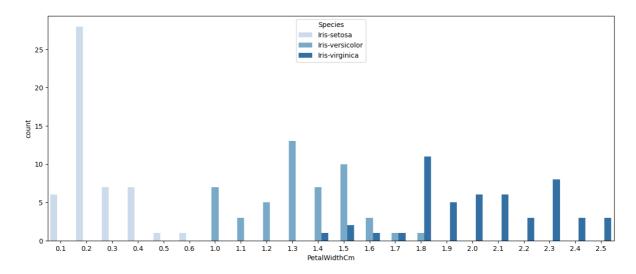
In [14]: plt.figure(figsize =(15,6))
 sns.countplot(x='SepalWidthCm', data=irisdf, hue= irisdf['Species'], palette= 'autu
 plt.show()



In [15]: plt.figure(figsize =(15,6))
 sns.countplot(x='PetalLengthCm', data=irisdf, hue= irisdf['Species'], palette= 'mag
 plt.show()



```
In [16]: plt.figure(figsize =(15,6))
    sns.countplot(x='PetalWidthCm', data=irisdf, hue= irisdf['Species'], palette= 'Blue
    plt.show()
```



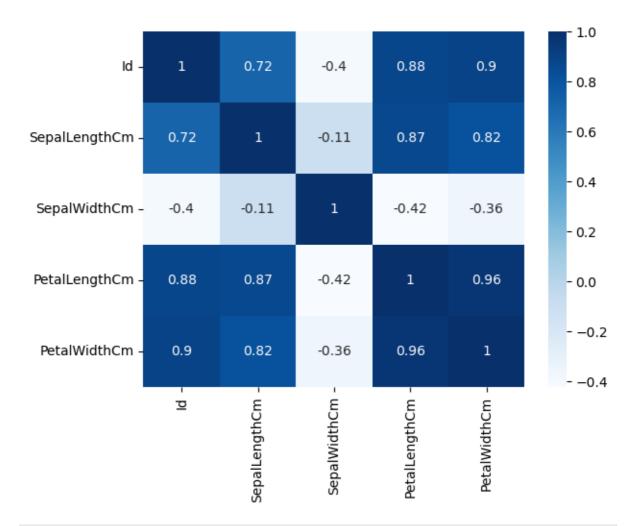
In [17]: irisdf.corr()

Out[17]:

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
Id	1.000000	0.716676	-0.397729	0.882747	0.899759
SepalLengthCm	0.716676	1.000000	-0.109369	0.871754	0.817954
SepalWidthCm	-0.397729	-0.109369	1.000000	-0.420516	-0.356544
PetalLengthCm	0.882747	0.871754	-0.420516	1.000000	0.962757
PetalWidthCm	0.899759	0.817954	-0.356544	0.962757	1.000000

In [18]: corr = irisdf.corr()
 sns.heatmap(corr,annot=True,cmap='Blues')

Out[18]: <AxesSubplot:>



In [19]: flower_mapping= {'Iris-setosa': 0, 'Iris-versicolor' : 1, 'Iris-virginica' :2}
 irisdf['Species'] = irisdf ['Species'].map (flower_mapping)

In [20]: irisdf.head()

Out[20]: Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species 0 1 5.1 3.5 1.4 0.2 0 1 2 4.9 3.0 1.4 0.2 0 2 3 4.7 3.2 1.3 0.2 0 3 4 4.6 3.1 1.5 0.2 0 1.4 0.2 0 4 5 5.0 3.6

In [21]: irisdf.tail()

Out[21]:		Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
	145	146	6.7	3.0	5.2	2.3	2
	146	147	6.3	2.5	5.0	1.9	2
	147	148	6.5	3.0	5.2	2.0	2
	148	149	6.2	3.4	5.4	2.3	2
	149	150	5.9	3.0	5.1	1.8	2

```
In [22]: x=irisdf[['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm']].values
In [23]: from sklearn.linear_model import LinearRegression
    model= LinearRegression()

In [24]: model.fit(x,y)
Out[24]:
LinearRegression()
```