

Task 2- Iris Flower Classification ML Project

```
In [13]: #Load Ddatasets
#DataFlair Iris Flower Classification
# Import Packages
import os
import csv
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
%matplotlib inline
```

```
In [15]: df=pd.read_csv(r'C:\Users\admin\Downloads\iris.data')
print(df)

      5.1  3.5  1.4  0.2      Iris-setosa
0      4.9  3.0  1.4  0.2      Iris-setosa
1      4.7  3.2  1.3  0.2      Iris-setosa
2      4.6  3.1  1.5  0.2      Iris-setosa
3      5.0  3.6  1.4  0.2      Iris-setosa
4      5.4  3.9  1.7  0.4      Iris-setosa
..    ...  ...  ...  ...      ...
144    6.7  3.0  5.2  2.3  Iris-virginica
145    6.3  2.5  5.0  1.9  Iris-virginica
146    6.5  3.0  5.2  2.0  Iris-virginica
147    6.2  3.4  5.4  2.3  Iris-virginica
148    5.9  3.0  5.1  1.8  Iris-virginica

[149 rows x 5 columns]
```

```
In [ ]: columns = ['Sepal length', 'Sepal width', 'Petal length', 'Petal width', 'Class_labels']
# Load the data
df = pd.read_csv('iris.data', names=columns)
df.head()
```

```
In [16]: df.head()
```

Out[16]:

	5.1	3.5	1.4	0.2	Iris-setosa
0	4.9	3.0	1.4	0.2	Iris-setosa
1	4.7	3.2	1.3	0.2	Iris-setosa
2	4.6	3.1	1.5	0.2	Iris-setosa
3	5.0	3.6	1.4	0.2	Iris-setosa
4	5.4	3.9	1.7	0.4	Iris-setosa

```
In [17]: #Describe dataset
df.describe()
```

Out[17]:

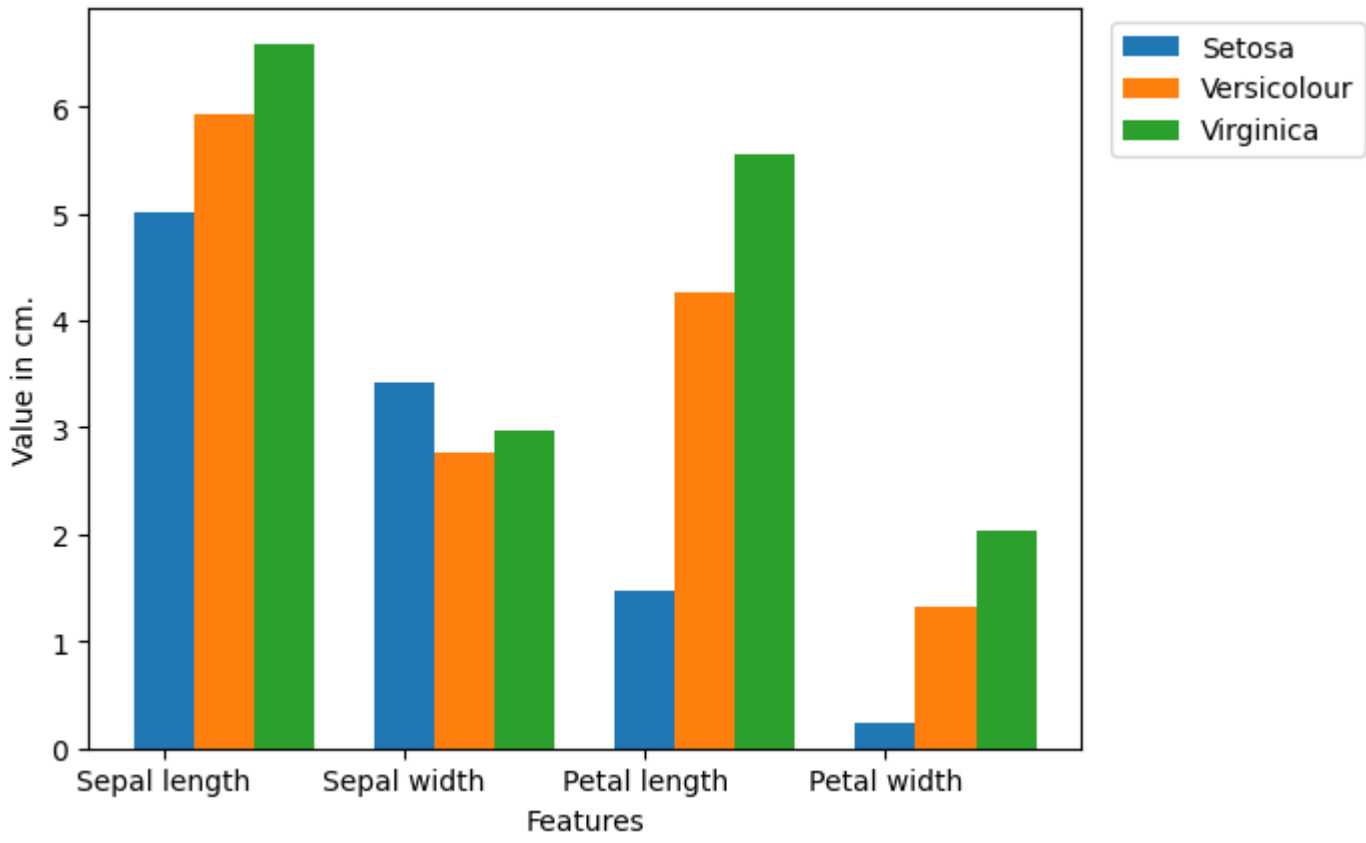
	5.1	3.5	1.4	0.2
count	149.000000	149.000000	149.000000	149.000000
mean	5.848322	3.051007	3.774497	1.205369
std	0.828594	0.433499	1.759651	0.761292
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.400000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

```
In [ ]: # Visualize the whole dataset
sns.pairplot(df, hue='Class_labels')
```

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In [18]: # Separate features and target
data = df.values
X = data[:,0:4]
Y = data[:,4]
```

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In [19]: # Calculate average of each features for all classes
Y_Data = np.array([np.average(X[:, i][Y==j].astype('float32')) for i in range (X.shape[1])
                    for j in (np.unique(Y))])
Y_Data_resaped = Y_Data.reshape(4, 3)
Y_Data_resaped = np.swapaxes(Y_Data_resaped, 0, 1)
X_axis = np.arange(len(columns)-1)
width = 0.25
```

```
In [20]: # Plot the average
plt.bar(X_axis, Y_Data_resaped[0], width, label = 'Setosa')
plt.bar(X_axis+width, Y_Data_resaped[1], width, label = 'Versicolour')
plt.bar(X_axis+width*2, Y_Data_resaped[2], width, label = 'Virginica')
plt.xticks(X_axis, columns[:4])
plt.xlabel("Features")
plt.ylabel("Value in cm.")
plt.legend(bbox_to_anchor=(1.3,1))
plt.show()
```



```
In [22]: #2.Model Training
# Split the data to train and test dataset.
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.2)
```

```
In [ ]: #3.Model Evaluation
# Predict from the test dataset
predictions = svm.predict(X_test)
# Calculate the accuracy
from sklearn.metrics import accuracy_score
accuracy_score(y_test, predictions)
```

```
In [ ]: #4.Testing Model
X_new = np.array([[3, 2, 1, 0.2], [ 4.9, 2.2, 3.8, 1.1 ], [ 5.3, 2.5, 4.6, 1.9 ]])
#Prediction of the species from the input vector
prediction = svm.predict(X_new)
print("Prediction of Species: {}".format(prediction))
```