

▼ TASK 1 - IRIS FLOWER CLASSIFICATION

Data Science Internship  
Batch - February Phase 2 OIBSIP.  
Author - **HIRAL CHOKSI**

▼ 1) Importing Libraries

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

▼ 2) Importing Dataset

```
irisdataset=pd.read_csv("Iris.csv")
```

▼ 3)Inspecting the Data

```
irisdataset
```

	Id	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
...	...	...	...	...	...	...
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

150 rows × 6 columns

```
irisdataset.shape
```

(150, 6)

```
irisdataset.size
```

900

```
irisdataset.head()
```

	Id	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

```
irisdataset.tail()
```

	<b>Id</b>	<b>SepalLengthCm</b>	<b>SepalWidthCm</b>	<b>PetalLengthCm</b>	<b>PetalWidthCm</b>	<b>Species</b>
<b>145</b>	146	6.7	3.0	5.2	2.3	Iris-virginica
<b>146</b>	147	6.3	2.5	5.0	1.9	Iris-virginica
<b>147</b>	148	6.5	3.0	5.2	2.0	Iris-virginica
<b>148</b>	149	6.2	3.4	5.4	2.3	Iris-virginica

```
irisdataset.Species.value_counts()
```

```
Iris-setosa      50
Iris-versicolor  50
Iris-virginica   50
Name: Species, dtype: int64
```

```
irisdataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 6 columns):
 #   Column          Non-Null Count  Dtype
---  ---
 0   Id              150 non-null   int64
 1   SepalLengthCm   150 non-null   float64
 2   SepalWidthCm    150 non-null   float64
 3   PetalLengthCm   150 non-null   float64
 4   PetalWidthCm    150 non-null   float64
 5   Species         150 non-null   object
dtypes: float64(4), int64(1), object(1)
memory usage: 7.2+ KB
```

```
irisdataset.describe()
```

	<b>Id</b>	<b>SepalLengthCm</b>	<b>SepalWidthCm</b>	<b>PetalLengthCm</b>	<b>PetalWidthCm</b>
<b>count</b>	150.000000	150.000000	150.000000	150.000000	150.000000
<b>mean</b>	75.500000	5.843333	3.054000	3.758667	1.198667
<b>std</b>	43.445368	0.828066	0.433594	1.764420	0.763161
<b>min</b>	1.000000	4.300000	2.000000	1.000000	0.100000
<b>25%</b>	38.250000	5.100000	2.800000	1.600000	0.300000
<b>50%</b>	75.500000	5.800000	3.000000	4.350000	1.300000
<b>75%</b>	112.750000	6.400000	3.300000	5.100000	1.800000
<b>max</b>	150.000000	7.900000	4.400000	6.900000	2.500000

4) Data Cleaning

a) Dropping ID Column

```
irisdataset = irisdataset.drop('Id', axis = 1)
irisdataset
```

	<b>SepalLengthCm</b>	<b>SepalWidthCm</b>	<b>PetalLengthCm</b>	<b>PetalWidthCm</b>	<b>Species</b>
<b>0</b>	5.1	3.5	1.4	0.2	Iris-setosa
<b>1</b>	4.9	3.0	1.4	0.2	Iris-setosa
<b>2</b>	4.7	3.2	1.3	0.2	Iris-setosa
<b>3</b>	4.6	3.1	1.5	0.2	Iris-setosa
<b>4</b>	5.0	3.6	1.4	0.2	Iris-setosa
...	...	...	...	...	...
<b>145</b>	6.7	3.0	5.2	2.3	Iris-virginica
<b>146</b>	6.3	2.5	5.0	1.9	Iris-virginica
<b>147</b>	6.5	3.0	5.2	2.0	Iris-virginica
<b>148</b>	6.2	3.4	5.4	2.3	Iris-virginica
<b>149</b>	5.9	3.0	5.1	1.8	Iris-virginica

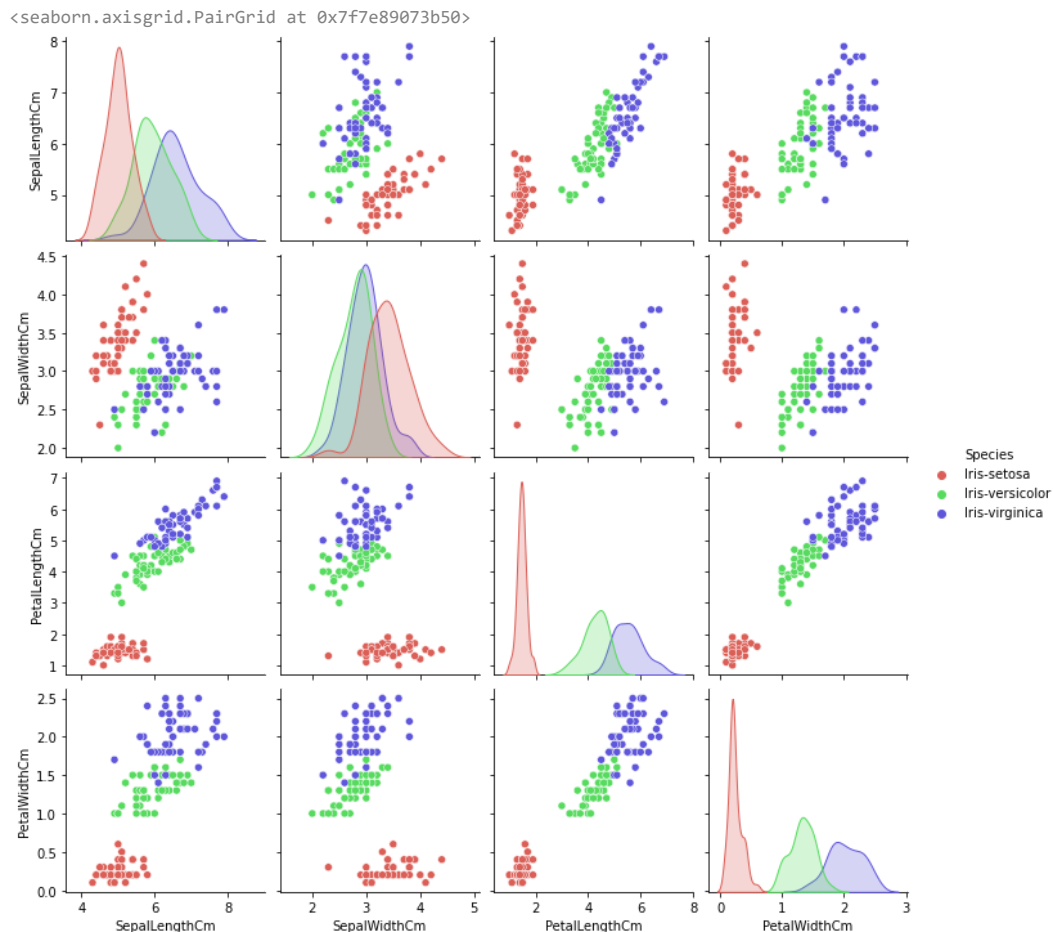
150 rows × 5 columns

```
irisdataset.isnull().sum()
```

```
SepalLengthCm    0
SepalWidthCm      0
PetalLengthCm     0
PetalWidthCm      0
Species           0
dtype: int64
```

## 5) Visualization of the Dataset

```
sns.pairplot(irisdataset, hue="Species", palette='hls')
```



## 6) Separating INPUT COLUMNS and OUTPUT COLUMNS

```
X = irisdataset.drop('Species', axis=1)
y = irisdataset['Species']
```

```
print(X)
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
0	5.1	3.5	1.4	0.2
1	4.9	3.0	1.4	0.2
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
146	6.3	2.5	5.0	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 x 4 columns]
```

```
print(y)
```

```

0      Iris-setosa
1      Iris-setosa
2      Iris-setosa
3      Iris-setosa
4      Iris-setosa
...
145    Iris-virginica
146    Iris-virginica
147    Iris-virginica
148    Iris-virginica
149    Iris-virginica
Name: Species, Length: 150, dtype: object

```

a) Seperate the features and target

```
data = irisdataset.values
```

```
X = data[ :,0:4]
```

```
y = data[:,4]
```

```
print(X)
```

```

[[5.1 3.5 1.4 0.2]
 [4.9 3.0 1.4 0.2]
 [4.7 3.2 1.3 0.2]
 [4.6 3.1 1.5 0.2]
 [5.0 3.6 1.4 0.2]
 [5.4 3.9 1.7 0.4]
 [4.6 3.4 1.4 0.3]
 [5.0 3.4 1.5 0.2]
 [4.4 2.9 1.4 0.2]
 [4.9 3.1 1.5 0.1]
 [5.4 3.7 1.5 0.2]
 [4.8 3.4 1.6 0.2]
 [4.8 3.0 1.4 0.1]
 [4.3 3.0 1.1 0.1]
 [5.8 4.0 1.2 0.2]
 [5.7 4.4 1.5 0.4]
 [5.4 3.9 1.3 0.4]
 [5.1 3.5 1.4 0.3]
 [5.7 3.8 1.7 0.3]
 [5.1 3.8 1.5 0.3]
 [5.4 3.4 1.7 0.2]
 [5.1 3.7 1.5 0.4]
 [4.6 3.6 1.0 0.2]
 [5.1 3.3 1.7 0.5]
 [4.8 3.4 1.9 0.2]
 [5.0 3.0 1.6 0.2]
 [5.0 3.4 1.6 0.4]
 [5.2 3.5 1.5 0.2]
 [5.2 3.4 1.4 0.2]
 [4.7 3.2 1.6 0.2]
 [4.8 3.1 1.6 0.2]
 [5.4 3.4 1.5 0.4]
 [5.2 4.1 1.5 0.1]
 [5.5 4.2 1.4 0.2]
 [4.9 3.1 1.5 0.1]
 [5.0 3.2 1.2 0.2]
 [5.5 3.5 1.3 0.2]
 [4.9 3.1 1.5 0.1]
 [4.4 3.0 1.3 0.2]
 [5.1 3.4 1.5 0.2]
 [5.0 3.5 1.3 0.3]
 [4.5 2.3 1.3 0.3]
 [4.4 3.2 1.3 0.2]
 [5.0 3.5 1.6 0.6]
 [5.1 3.8 1.9 0.4]
 [4.8 3.0 1.4 0.3]
 [5.1 3.8 1.6 0.2]
 [4.6 3.2 1.4 0.2]
 [5.3 3.7 1.5 0.2]
 [5.0 3.3 1.4 0.2]
 [7.0 3.2 4.7 1.4]
 [6.4 3.2 4.5 1.5]
 [6.9 3.1 4.9 1.5]
 [5.5 2.3 4.0 1.3]
 [6.5 2.8 4.6 1.5]
 [5.7 2.8 4.5 1.3]
 [6.3 3.3 4.7 1.6]
 [4.9 2.4 3.3 1.0]

```

```
print(y)
```

```

['Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa'
 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa'
 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa'
 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa'

```

### 6) Splitting the data into Training and Testing

[6.4 3.1 5.5 1.8]  
[5.4 3.0 4.5 1.5]  
[5.2 3.5 1.5 0.2]  
[6.1 3.0 4.9 1.8]  
[6.4 2.8 5.6 2.2]  
[5.2 2.7 3.9 1.4]  
[5.7 3.8 1.7 0.3]  
[6.0 2.7 5.1 1.6]  
[5.9 3.0 4.2 1.5]  
[5.8 2.6 4.0 1.2]  
[6.8 3.0 5.5 2.1]  
[4.7 3.2 1.3 0.2]  
[6.9 3.1 5.1 2.3]  
[5.0 3.5 1.6 0.6]  
[5.4 3.7 1.5 0.2]  
[5.0 2.0 3.5 1.0]  
[6.5 3.0 5.5 1.8]  
[6.7 3.3 5.7 2.5]  
[6.0 2.2 5.0 1.5]  
[6.7 2.5 5.8 1.8]  
[5.6 2.5 3.9 1.1]  
[7.7 3.0 6.1 2.3]  
[6.3 3.3 4.7 1.6]  
[5.5 2.4 3.8 1.1]  
[6.3 2.7 4.9 1.8]  
[6.3 2.8 5.1 1.5]  
[4.9 2.5 4.5 1.7]  
[6.3 2.5 5.0 1.9]  
[7.0 3.2 4.7 1.4]  
[6.5 3.0 5.2 2.0]  
[6.0 3.4 4.5 1.6]  
[4.8 3.1 1.6 0.2]  
[5.8 2.7 5.1 1.9]  
[5.6 2.7 4.2 1.3]  
[5.6 2.9 3.6 1.3]  
[5.5 2.5 4.0 1.3]  
[6.1 3.0 4.6 1.4]  
[7.2 3.2 6.0 1.8]  
[5.3 3.7 1.5 0.2]  
[4.3 3.0 1.1 0.1]  
[6.4 2.7 5.3 1.9]  
[5.7 3.0 4.2 1.2]  
[5.4 3.4 1.7 0.2]  
[5.7 4.4 1.5 0.4]  
[6.9 3.1 4.9 1.5]  
[4.6 3.1 1.5 0.2]  
[5.9 3.0 5.1 1.8]  
[5.1 2.5 3.0 1.1]  
[4.6 3.4 1.4 0.3]

```
[6.2 2.2 4.5 1.5]
[7.2 3.6 6.1 2.5]
[5.7 2.9 4.2 1.3]
[4.8 3.0 1.4 0.1]
[7.1 3.0 5.9 2.1]
[6.9 3.2 5.7 2.3]
[6.5 3.0 5.8 2.2]
[6.4 2.8 5.6 2.1]
[5.1 3.8 1.6 0.2]
```

```
print(y_train)
```

```
['Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica'
'Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-versicolor'
'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica' 'Iris-setosa'
'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor'
'Iris-virginica' 'Iris-virginica' 'Iris-virginica' 'Iris-virginica'
'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor'
'Iris-virginica' 'Iris-virginica' 'Iris-virginica' 'Iris-virginica'
'Iris-versicolor' 'Iris-virginica' 'Iris-setosa' 'Iris-setosa'
'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor'
'Iris-versicolor' 'Iris-virginica' 'Iris-setosa' 'Iris-setosa'
'Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa'
'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor'
'Iris-setosa' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor'
'Iris-setosa' 'Iris-virginica' 'Iris-virginica' 'Iris-virginica'
'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica'
'Iris-virginica' 'Iris-setosa' 'Iris-virginica' 'Iris-setosa'
'Iris-virginica' 'Iris-virginica' 'Iris-setosa' 'Iris-setosa'
'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa'
'Iris-versicolor' 'Iris-virginica' 'Iris-virginica' 'Iris-setosa'
'Iris-setosa' 'Iris-setosa' 'Iris-versicolor' 'Iris-versicolor'
'Iris-setosa' 'Iris-setosa' 'Iris-versicolor' 'Iris-setosa'
'Iris-virginica' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor'
'Iris-setosa' 'Iris-virginica' 'Iris-setosa' 'Iris-virginica'
'Iris-setosa' 'Iris-setosa' 'Iris-virginica' 'Iris-setosa'
'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor'
'Iris-virginica' 'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor'
'Iris-setosa' 'Iris-versicolor' 'Iris-virginica' 'Iris-virginica'
'Iris-setosa' 'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor'
'Iris-versicolor' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa'
'Iris-virginica' 'Iris-versicolor' 'Iris-virginica' 'Iris-setosa']
```

```
print(X_test)
```

```
[[5.8 2.8 5.1 2.4]
[6.0 2.2 4.0 1.0]
[5.5 4.2 1.4 0.2]
[7.3 2.9 6.3 1.8]
[5.0 3.4 1.5 0.2]
[6.3 3.3 6.0 2.5]
[5.0 3.5 1.3 0.3]
[6.7 3.1 4.7 1.5]
[6.8 2.8 4.8 1.4]
[6.1 2.8 4.0 1.3]
[6.1 2.6 5.6 1.4]
[6.4 3.2 4.5 1.5]
[6.1 2.8 4.7 1.2]
[6.5 2.8 4.6 1.5]
[6.1 2.9 4.7 1.4]
[4.9 3.1 1.5 0.1]
[6.0 2.9 4.5 1.5]
[5.5 2.6 4.4 1.2]
[4.8 3.0 1.4 0.3]
[5.4 3.9 1.3 0.4]
[5.6 2.8 4.9 2.0]
[5.6 3.0 4.5 1.5]
[4.8 3.4 1.9 0.2]
[4.4 2.9 1.4 0.2]
[6.2 2.8 4.8 1.8]
[4.6 3.6 1.0 0.2]
[5.1 3.8 1.9 0.4]
[6.2 2.9 4.3 1.3]
[5.0 2.3 3.3 1.0]
[5.0 3.4 1.6 0.4]]
```

```
print(y_test)
```

```
['Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica'
'Iris-setosa' 'Iris-virginica' 'Iris-setosa' 'Iris-versicolor'
'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor'
'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa'
'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa'
'Iris-virginica' 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa'
'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor'
'Iris-versicolor' 'Iris-setosa']
```

## MODEL 1 : Support Vecotor Machine Algorithm

---

```
from sklearn.svm import SVC
model_svc = SVC()
model_svc.fit(X_train, y_train)

SVC()

Prediction1 = model_svc.predict(X_test)
from sklearn.metrics import accuracy_score
print("Accuracy of the Decision Tree Classifier Model :", accuracy_score(y_test,Prediction1)*100)
for i in range(len(Prediction1)):
    print(y_test[i],Prediction1[i])
```

```
Accuracy of the Decision Tree Classifier Model : 100.0
Iris-virginica Iris-virginica
Iris-versicolor Iris-versicolor
Iris-setosa Iris-setosa
Iris-virginica Iris-virginica
Iris-setosa Iris-setosa
Iris-virginica Iris-virginica
Iris-setosa Iris-setosa
Iris-versicolor Iris-versicolor
Iris-versicolor Iris-versicolor
Iris-versicolor Iris-versicolor
Iris-virginica Iris-virginica
Iris-versicolor Iris-versicolor
Iris-versicolor Iris-versicolor
Iris-versicolor Iris-versicolor
Iris-versicolor Iris-versicolor
Iris-setosa Iris-setosa
Iris-versicolor Iris-versicolor
Iris-versicolor Iris-versicolor
Iris-setosa Iris-setosa
Iris-setosa Iris-setosa
Iris-virginica Iris-virginica
Iris-versicolor Iris-versicolor
Iris-setosa Iris-setosa
Iris-setosa Iris-setosa
Iris-virginica Iris-virginica
Iris-setosa Iris-setosa
Iris-setosa Iris-setosa
Iris-versicolor Iris-versicolor
Iris-versicolor Iris-versicolor
Iris-setosa Iris-setosa
```

## MODEL 2 : KNN (K Nearest Neighbor)

---

```
from sklearn.neighbors import KNeighborsClassifier
model_knn = KNeighborsClassifier()

model_knn.fit(X_train, y_train)

KNeighborsClassifier()

y_pred=knc1r.predict(X_test)

from sklearn import metrics
print("Accuracy of the K Nearest Neighbor Model :", metrics.accuracy_score(y_test,y_pred)*100)

Accuracy of the Decision Tree Classifier Model : 100.0
```

## MODEL 3 : Decision Tree Classifier

---

```
from sklearn.tree import DecisionTreeClassifier
model_dtc = DecisionTreeClassifier()
model_dtc.fit(X_train,y_train)

DecisionTreeClassifier()

print("Accuracy of the Decision Tree Classifier Model :", model_dtc.score(X_test, y_test)*100)
```

Accuracy of the Decision Tree Classifier Model : 100.0

THANK YOU!

---

