

```
In [ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn import preprocessing
from scipy import stats
from sklearn.tree import DecisionTreeClassifier
from sklearn.svm import SVC
from sklearn.metrics import classification_report
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
```

```
In [ ]: columns = ["Sepal Length", "Sepal width", "Petal Length", "Petal Width", "Names"]
data = pd.read_csv("iris.data", names= columns)
data.head()
```

```
Out[ ]:
```

	Sepal Length	Sepal width	Petal Length	Petal Width	Names
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 [ ]: data.shape
```

```
Out[ ]: (150, 5)
```

```
In [ ]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  ---
0   Sepal Length    150 non-null    float64
1   Sepal width     150 non-null    float64
2   Petal Length    150 non-null    float64
3   Petal Width     150 non-null    float64
4   Names           150 non-null    object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

Statistical Description

```
In [ ]: data.describe()
```

```
Out[ ]:
```

	Sepal Length	Sepal width	Petal Length	Petal Width
count	150.000000	150.000000	150.000000	150.000000

	Sepal Length	Sepal width	Petal Length	Petal Width
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	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
max	7.900000	4.400000	6.900000	2.500000

Finding Null Values

```
In [ ]: data.isnull().sum()
```

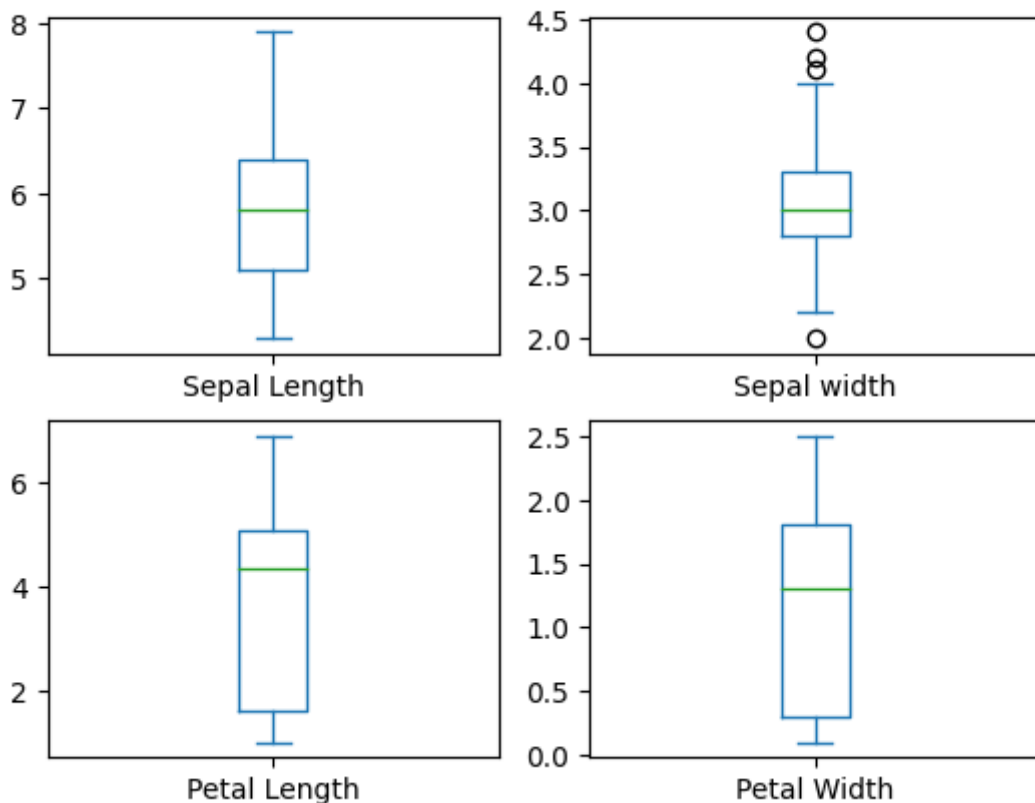
```
Out[ ]: Sepal Length    0
        Sepal width    0
        Petal Length    0
        Petal Width    0
        Names          0
        dtype: int64
```

Checking Outliers

```
In [ ]: data.plot(kind='box',subplots=True,layout=(2,2),title="Whisker Plot For checking Outliers")
```

```
Out[ ]: Sepal Length    Axes(0.125,0.53;0.352273x0.35)
        Sepal width    Axes(0.547727,0.53;0.352273x0.35)
        Petal Length    Axes(0.125,0.11;0.352273x0.35)
        Petal Width    Axes(0.547727,0.11;0.352273x0.35)
        dtype: object
```

Whisker Plot For checking Outliers



Inter Quartile Range

```
In [ ]: Q1 = data['Sepal width'].quantile(0.25)
Q3 = data['Sepal width'].quantile(0.75)
IQR = Q3 - Q1
print("Quartile 1 : ", Q1)
print("Quartile 3 : ", Q3)
print("Inter Quartile Range : ", IQR)
```

```
Quartile 1 : 2.8
Quartile 3 : 3.3
Inter Quartile Range : 0.5
```

Above Upper Bound

Formula : Quartile 3 which is upper side, is Added to (1.5 times the Inter Quartile range)

```
In [ ]: upper = Q3+1.5*IQR
upper_arr = np.where(data['Sepal width']>upper)[0]
print("Upper Bound : ", upper)
# print(upper_arr.sum())
```

```
Upper Bound : 4.05
```

Below Lower Bound

Formula : Quartile 1 which is Lower side, is Subtracted to (1.5 times the Inter Quartile range)

```
In [ ]: lower = Q1-1.5*IQR
lower_arr = np.where(data['Sepal width']<lower)[0]
```

```
print("Lower Bound : ",lower)
# print(lower_arr.sum())
```

Lower Bound : 2.05

Removing Outliers

```
In [ ]: print("Before Removing Outliers : ", data.shape)
data = data.drop(index=upper_arr)
data = data.drop(index=lower_arr)
```

Before Removing Outliers : (150, 5)

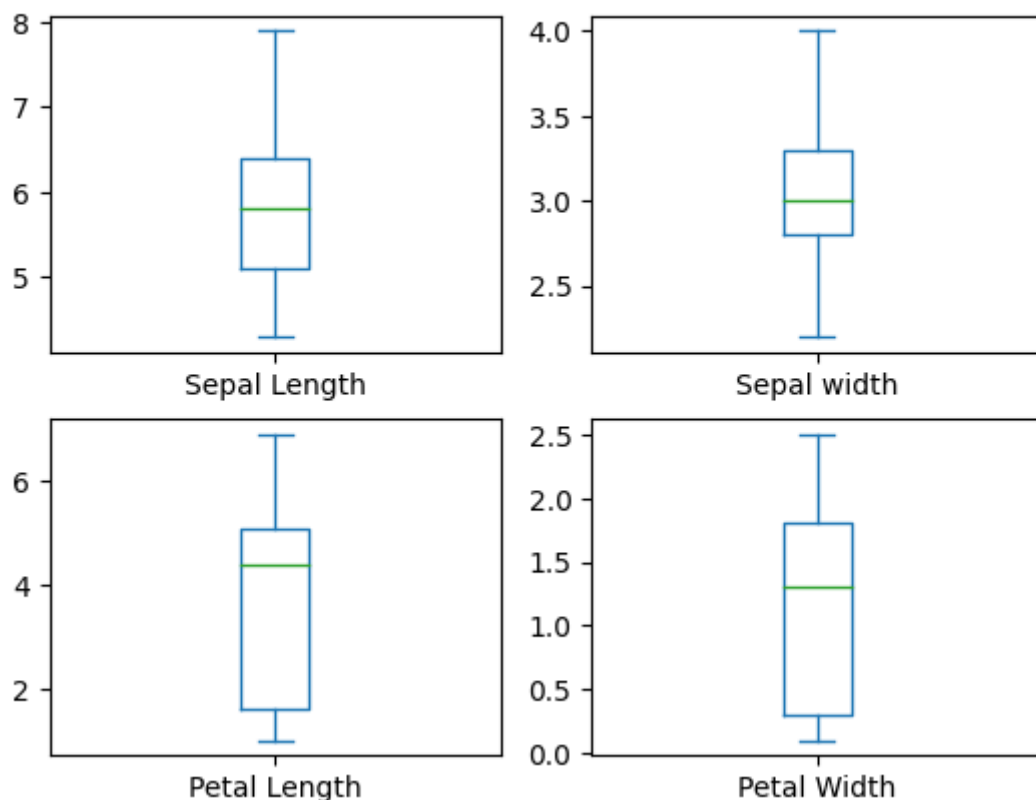
```
In [ ]: print("After Removing Outliers : ", data.shape )
```

After Removing Outliers : (146, 5)

```
In [ ]: data.plot(kind = 'box',subplots = True, layout = (2,2), title = 'After Removing Outl
```

```
Out[ ]: Sepal Length      Axes(0.125,0.53;0.352273x0.35)
Sepal width    Axes(0.547727,0.53;0.352273x0.35)
Petal Length   Axes(0.125,0.11;0.352273x0.35)
Petal Width    Axes(0.547727,0.11;0.352273x0.35)
dtype: object
```

After Removing Outliers



Checking Skewness

```
In [ ]: plt.subplot(2,2,1)
plt.tight_layout()
# The first parameter, 2, specifies the number of rows in the subplot grid.
# The second parameter, 2, specifies the number of columns in the subplot grid.
# The third parameter, 1, specifies the position of the current subplot within the g
```

```
plt.title("Petal Length Distribution")
sns.distplot(data['Petal Length'],)

plt.subplot(2,2,2)
plt.tight_layout()
plt.title("Petal Width Distribution")
sns.distplot(data["Petal Width"])

plt.subplot(2,2,3)
plt.tight_layout()
plt.title("Sepal Length Distribution")
sns.distplot(data["Sepal Length"])

plt.subplot(2,2,4)
plt.tight_layout()
plt.title("Sepal Width Distribution")
sns.distplot(data["Sepal width"])
```

C:\Users\Hunain\AppData\Local\Temp\ipykernel_12196\2321211178.py:7: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(data['Petal Length'],)
```

C:\Users\Hunain\AppData\Local\Temp\ipykernel_12196\2321211178.py:12: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(data["Petal Width"])
```

C:\Users\Hunain\AppData\Local\Temp\ipykernel_12196\2321211178.py:17: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(data["Sepal Length"])
```

C:\Users\Hunain\AppData\Local\Temp\ipykernel_12196\2321211178.py:22: UserWarning:

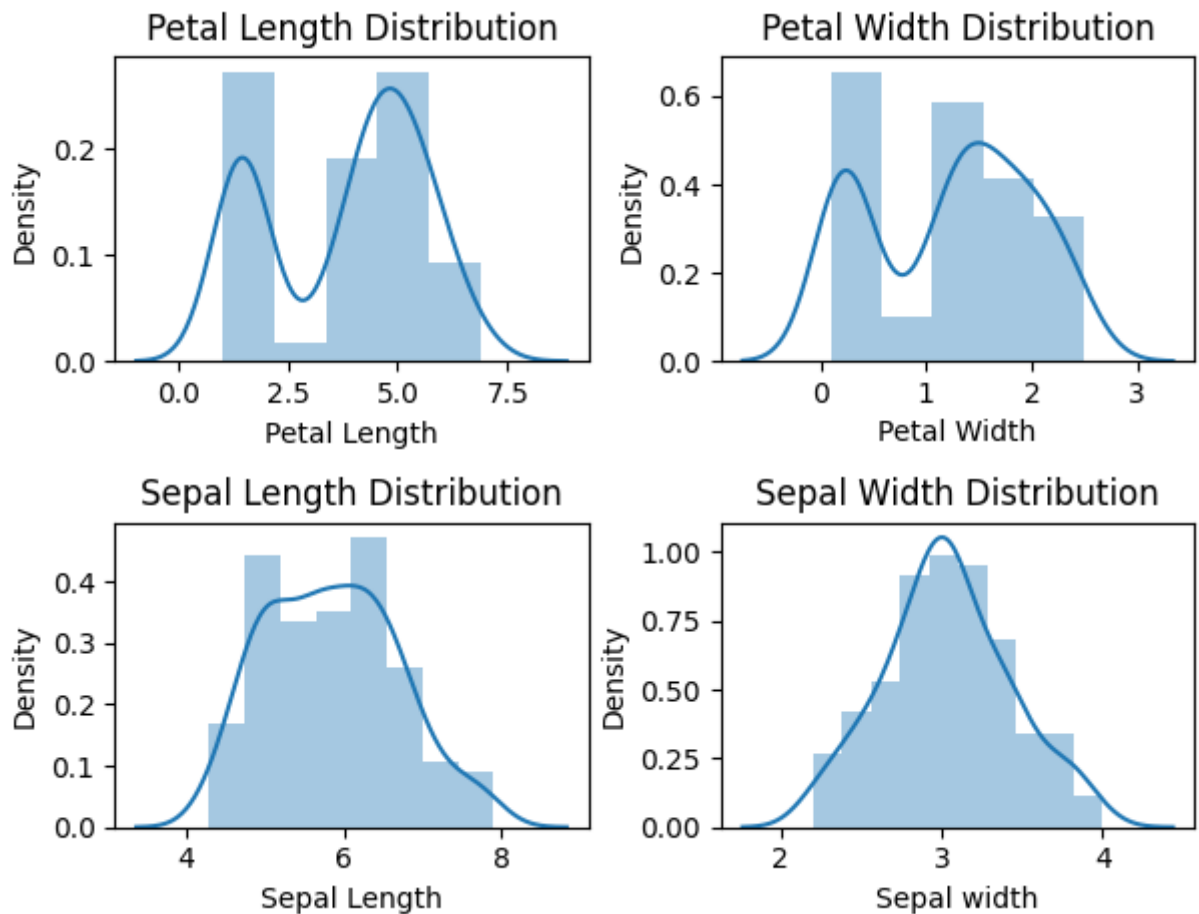
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

```
sns.distplot(data["Sepal width"])
```

Out[]: <Axes: title={'center': 'Sepal Width Distribution'}, xlabel='Sepal width', ylabel='Density'>



Label Encoding

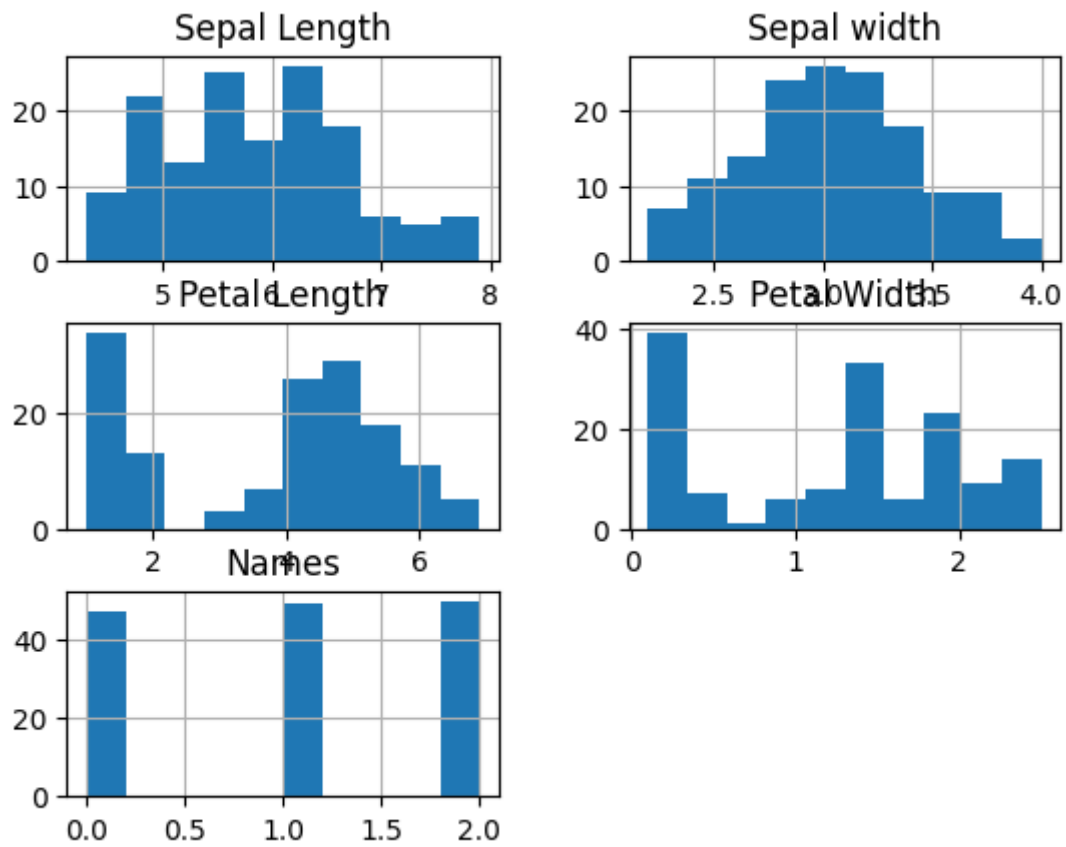
Names Should be converted to labels as it can be recognizable.

```
In [ ]: enc = LabelEncoder()
data['Names'] = enc.fit_transform(data['Names'])
data.head()
```

```
Out[ ]:   Sepal Length  Sepal width  Petal Length  Petal Width  Names
0          5.1         3.5         1.4         0.2         0
1          4.9         3.0         1.4         0.2         0
2          4.7         3.2         1.3         0.2         0
3          4.6         3.1         1.5         0.2         0
4          5.0         3.6         1.4         0.2         0
```

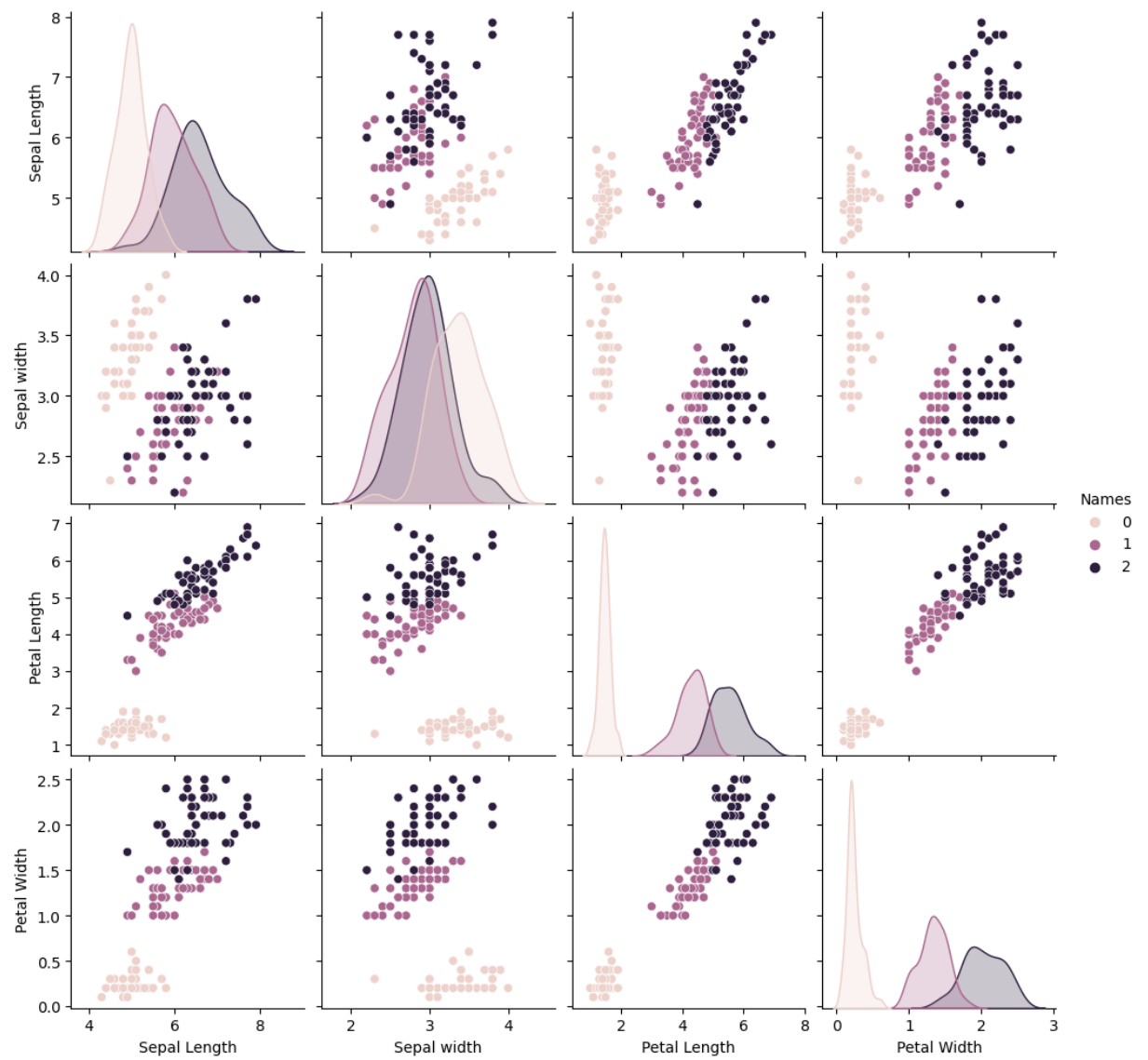
```
In [ ]: data.hist()
```

```
Out[ ]: array([[<Axes: title={'center': 'Sepal Length'}>,
               <Axes: title={'center': 'Sepal width'}>],
               [<Axes: title={'center': 'Petal Length'}>,
               <Axes: title={'center': 'Petal Width'}>],
               [<Axes: title={'center': 'Names'}>, <Axes: >]], dtype=object)
```



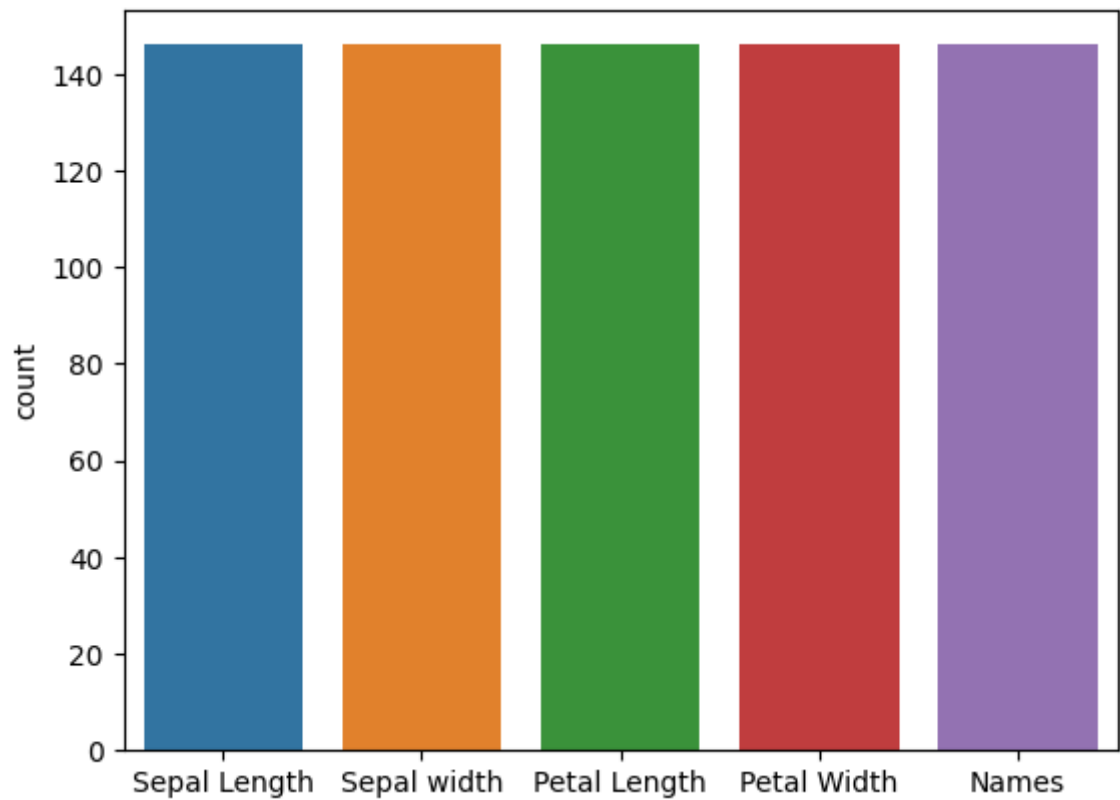
```
In [ ]: sns.pairplot(data,hue='Names')
```

```
Out[ ]: <seaborn.axisgrid.PairGrid at 0x1cb0e966410>
```



```
In [ ]: sns.countplot(data=data)
```

```
Out[ ]: <Axes: ylabel='count'>
```

Correlation

```
In [ ]: data1 = data.drop('Names',axis=1) # Removing Names Because Correlation will not work
        data1.corr()
```

```
Out[ ]:
```

	Sepal Length	Sepal width	Petal Length	Petal Width
Sepal Length	1.000000	-0.113977	0.879374	0.822577
Sepal width	-0.113977	1.000000	-0.397123	-0.329782
Petal Length	0.879374	-0.397123	1.000000	0.961755
Petal Width	0.822577	-0.329782	0.961755	1.000000

```
In [ ]: sns.heatmap(data1.corr(),annot=True)
```

```
Out[ ]: <Axes: >
```



Training Set and Testing Set Selection

```
In [ ]: x = data.drop(columns=['Names'])
        y = data['Names']
```

```
In [ ]: x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3,random_state=0)
```

```
In [ ]: Lg = LogisticRegression()
        Lg.fit(x_train,y_train)
```

```
Out[ ]: ▼ LogisticRegression
        LogisticRegression()
```

```
In [ ]: accuracy = Lg.score(x_test,y_test)*100
        print("accuracy : ",accuracy)
```

```
accuracy : 97.72727272727273
```

```
In [ ]: y_pred = Lg.predict(x_test)
```

Accuracy

```
In [ ]: print(classification_report(y_test,y_pred))
```

```

              precision    recall  f1-score   support

0               1.00      1.00      1.00         16
```

	1	1.00	0.94	0.97	16
	2	0.92	1.00	0.96	12
accuracy				0.98	44
macro avg		0.97	0.98	0.98	44
weighted avg		0.98	0.98	0.98	44

New Values Prediction

```
In [ ]: inp = np.array([[2.5,3.8,4.3,3.3],[3.5,2.8,4,3],[3.2,3.8,3.1,2.6]])
new_check = Lg.predict(inp)
print(new_check)
```

```
[2 2 0]
```

C:\Users\Hunain\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11_qbz5n2kfr8p0\LocalCache\local-packages\Python311\site-packages\sklearn\base.py:464: UserWarning: X does not have valid feature names, but LogisticRegression was fitted with feature names
warnings.warn(

Decision Tree Classifier (Accuracy)

```
In [ ]: model = DecisionTreeClassifier()
```

```
In [ ]: model.fit(x_train,y_train)
```

```
Out[ ]: ▾ DecisionTreeClassifier
DecisionTreeClassifier()
```

```
In [ ]: print(model.score(x_test,y_test)*100,"%")
```

```
97.72727272727273 %
```

Support Vector Classifier (Accuracy)

```
In [ ]: model_svc = SVC()
model_svc.fit(x_train,y_train)
```

```
Out[ ]: ▾ SVC
SVC()
```

```
In [ ]: print(model.score(x_test,y_test)*100,"%")
```

```
97.72727272727273 %
```

```
In [ ]:
```