SUPPORT VECTOR MACHINE

In [1]:

import pandas as pd
import numpy as np
import seaborn as sns
from sklearn import svm

In [3]:

data = pd.read_csv('C:\\Users\\MARVEL SAMUEL JACOB\\Desktop\\marvel.csv')

In [4]: ▶

data

Out[4]:

	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
			•••		
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

150 rows × 5 columns

```
In [5]:
```

```
data.head()
```

Out[5]:

	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 [6]:
```

```
data.isnull().sum()
```

Out[6]:

```
sepal_length 0
sepal_width 0
petal_length 0
petal_width 0
species 0
dtype: int64
```

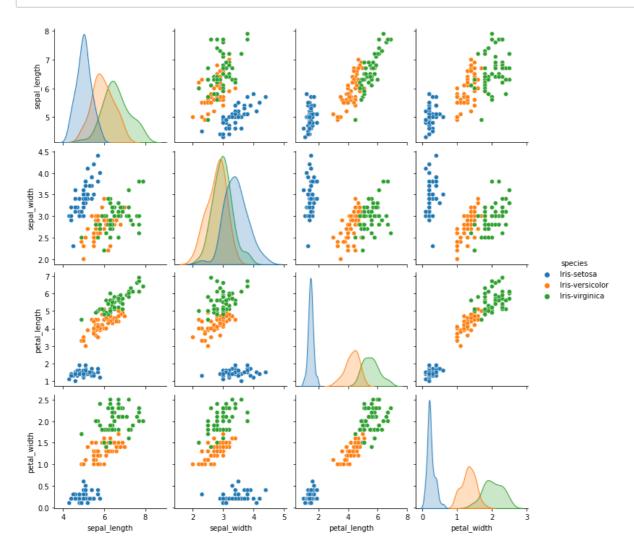
In [8]:

```
x=data.drop('species', axis = 1)
y = data['species']
```

M

In [9]: ▶

```
import seaborn as sns
sns.pairplot(data, hue = 'species');
```



```
In [69]:
                                                                                           H
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=30, random_state= 89)
In [70]:
from sklearn.svm import SVC
model = svm.SVC()
model.fit(x_train,y_train)
Out[70]:
SVC()
In [71]:
                                                                                           M
pred = model.predict(x_test)
In [72]:
                                                                                           H
from sklearn.metrics import classification_report ,confusion_matrix
print("classification_report:\n",classification_report(y_test,pred))
print("confusion_matrix:\n",confusion_matrix(y_test,pred))
classification_report:
               precision
                            recall f1-score
                                                support
                             1.00
           0
                   1.00
                                        1.00
                                                    14
                   0.78
                             1.00
                                        0.88
           1
                                                     7
           2
                   1.00
                             0.78
                                        0.88
                                                     9
                                        0.93
                                                    30
    accuracy
                                        0.92
                                                    30
   macro avg
                   0.93
                             0.93
                   0.95
                             0.93
                                        0.93
                                                    30
weighted avg
confusion_matrix:
 [[14 0 0]
 [070]
 [027]]
In [76]:
                                                                                           H
model.score(x_test,y_test)*100
Out[76]:
93.3333333333333
In [ ]:
                                                                                           H
```

DECISION TREE

```
In [12]:
                                                                                                          M
import pandas as pd
import numpy as np
import seaborn as sns
from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import train_test_split
from sklearn.metrics import classification_report, confusion_matrix
In [14]:
                                                                                                          H
data = pd.read_csv('C:\\Users\\MARVEL SAMUEL JACOB\\Desktop\\marvel.csv')
In [15]:
                                                                                                          H
data
Out[15]:
      sepal_length sepal_width petal_length
                                            petal_width
                                                            species
   0
               5.1
                           3.5
                                        1.4
                                                    0.2
                                                           Iris-setosa
                           3.0
                                                    0.2
                                                           Iris-setosa
   1
               4.9
                                        1.4
               4.7
                           3.2
                                                    0.2
                                                          Iris-setosa
   2
                                        1.3
   3
               4.6
                           3.1
                                        1.5
                                                    0.2
                                                          Iris-setosa
               5.0
                           3.6
                                        1.4
                                                    0.2
                                                          Iris-setosa
   4
               6.7
 145
                           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
                                        5.4
                           3.4
                                                    2.3
                                                         Iris-virginica
 149
               5.9
                           3.0
                                        5.1
                                                     1.8 Iris-virginica
150 rows × 5 columns
                                                                                                          H
In [16]:
X = iris.data
y = iris.target
In [17]:
                                                                                                          H
class_names = iris.target_names
```

```
In [18]:
                                                                                            H
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30)
In [19]:
                                                                                            H
DTM = DecisionTreeClassifier()
In [20]:
y_final = DTM.fit(X_train, y_train).predict(X_test)
In [21]:
                                                                                            H
print()
print(classification_report(y_test, y_final, target_names=class_names))
              precision
                            recall f1-score
                                                support
      setosa
                   1.00
                              1.00
                                        1.00
                                                     10
  versicolor
                   0.94
                              1.00
                                        0.97
                                                     16
                   1.00
                              0.95
                                        0.97
                                                     19
   virginica
                                                     45
                                        0.98
    accuracy
                   0.98
                              0.98
                                        0.98
                                                     45
   macro avg
                                        0.98
                                                     45
weighted avg
                   0.98
                              0.98
In [37]:
                                                                                            M
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.3,random_state=90)
In [38]:
                                                                                            H
DTM.fit(X_train,y_train)
Out[38]:
DecisionTreeClassifier()
In [39]:
                                                                                            H
y_predict=DTM.predict(X_test)
In [40]:
                                                                                            H
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
In [41]:
                                                                                            H
y_final = DTM.fit(X_train, y_train).predict(X_test)
```

```
In [42]:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30)

In [43]:

y_final = DTM.fit(X_train, y_train).predict(X_test)

In [44]:

print("Accuracy of the model by DT is :-",accuracy_score(y_test,y_predict)*100)
```

Accuracy of the model by DT is :- 35.5555555555556

K NEAREST NEIGHBOURS

In [1]: ▶

import pandas as pd
import numpy as np
import seaborn as sns
from sklearn import svm
import missingno as msn

In [2]:

data = pd.read_csv('C:\\Users\\MARVEL SAMUEL JACOB\\Desktop\\marvel.csv')

In [3]: ▶

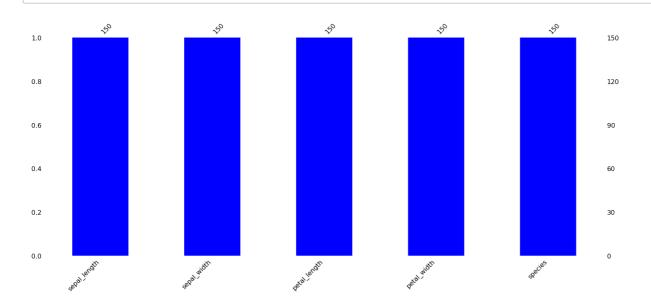
data

Out[3]:

	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
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

150 rows × 5 columns





```
In [5]:

data['species'].unique()

Out[5]:
    array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)

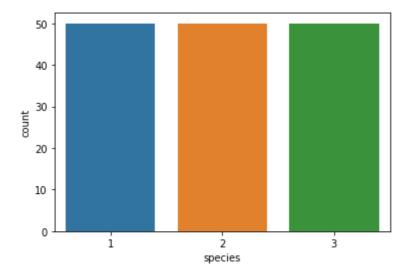
In [6]:

data['species']=data['species'].map({'Iris-setosa':1, 'Iris-versicolor':2, 'Iris-virginica'})
```

In [7]: ▶

sns.countplot(data['species']);

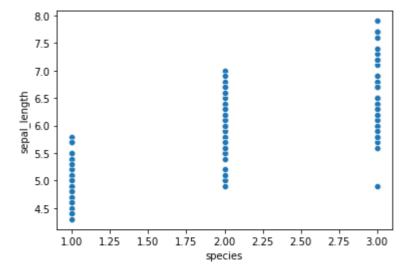
C:\Users\MARVEL SAMUEL JACOB\anaconda3\lib\site-packages\seaborn_decorator s.py:36: FutureWarning: Pass the following variable as a keyword arg: x. Fro m version 0.12, the only valid positional argument will be `data`, and passi ng other arguments without an explicit keyword will result in an error or mi sinterpretation.



In [8]: ▶

```
sns.scatterplot(data['species'],data['sepal_length']);
```

C:\Users\MARVEL SAMUEL JACOB\anaconda3\lib\site-packages\seaborn_decorator s.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and pa ssing other arguments without an explicit keyword will result in an error or misinterpretation.

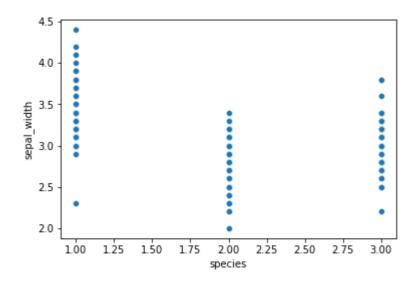


In [9]: ▶

```
sns.scatterplot(data['species'],data['sepal_width']);
```

C:\Users\MARVEL SAMUEL JACOB\anaconda3\lib\site-packages\seaborn_decorator s.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and pa ssing other arguments without an explicit keyword will result in an error or misinterpretation.

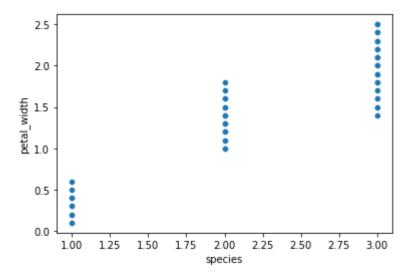
warnings.warn(



In [10]:

```
sns.scatterplot(data['species'],data['petal_width']);
```

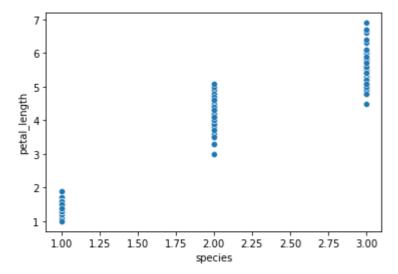
C:\Users\MARVEL SAMUEL JACOB\anaconda3\lib\site-packages\seaborn_decorator s.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and pa ssing other arguments without an explicit keyword will result in an error or misinterpretation.



In [11]: ▶

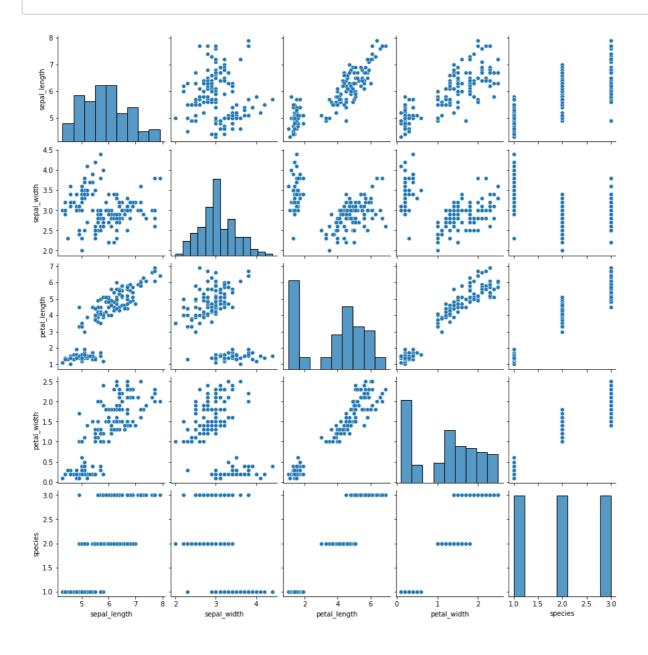
```
sns.scatterplot(data['species'],data['petal_length']);
```

C:\Users\MARVEL SAMUEL JACOB\anaconda3\lib\site-packages\seaborn_decorator s.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and pa ssing other arguments without an explicit keyword will result in an error or misinterpretation.



In [12]:

sns.pairplot(data);

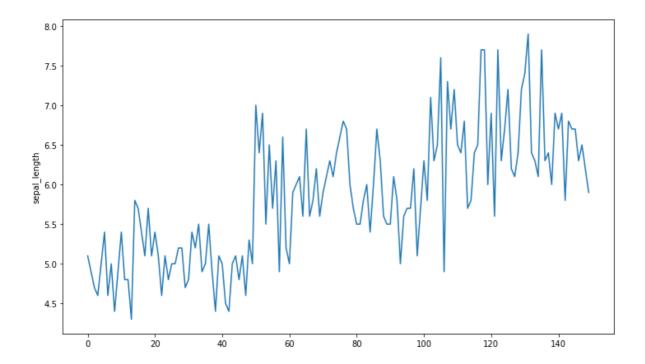


In [13]: ▶

import matplotlib.pyplot as plt

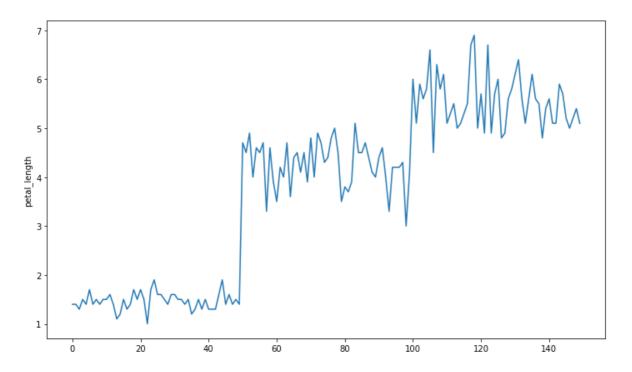
In [14]: ▶

fig,ax=plt.subplots(figsize=(12,7));
sns.lineplot(data=data['sepal_length']);



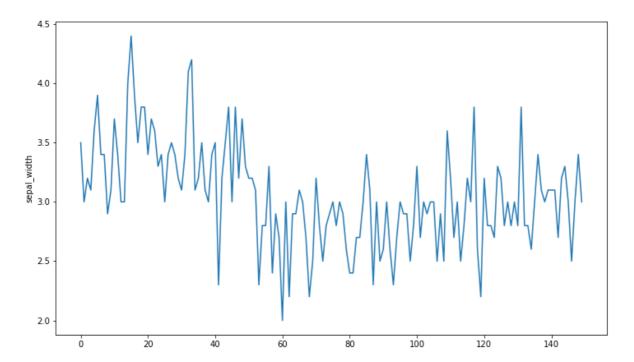
In [15]: ▶

```
fig,ax=plt.subplots(figsize=(12,7));
sns.lineplot(data=data['petal_length']);
```



In [16]: ▶

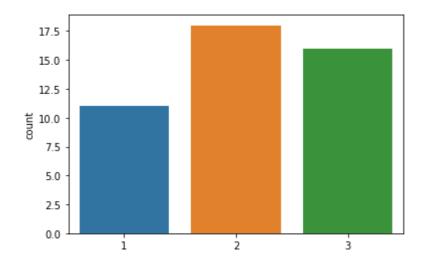
```
fig,ax=plt.subplots(figsize=(12,7));
sns.lineplot(data=data['sepal_width'],);
```



```
In [17]:
                                                                                            H
data.columns
Out[17]:
Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
       'species'],
      dtype='object')
In [18]:
                                                                                            M
X=data[['sepal_length', 'sepal_width', 'petal_length', 'petal_width']]
In [19]:
y=data['species']
In [61]:
                                                                                            H
from sklearn.model selection import train test split
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.3,random_state=78)
In [62]:
                                                                                            H
from sklearn.neighbors import KNeighborsClassifier
model=KNeighborsClassifier()
In [63]:
                                                                                            H
model.fit(X_train,y_train)
Out[63]:
KNeighborsClassifier()
In [64]:
                                                                                            H
y_predict=model.predict(X_test)
```

```
In [65]:
sns.countplot(y_predict);
```

C:\Users\MARVEL SAMUEL JACOB\anaconda3\lib\site-packages\seaborn_decorator s.py:36: FutureWarning: Pass the following variable as a keyword arg: x. Fro m version 0.12, the only valid positional argument will be `data`, and passi ng other arguments without an explicit keyword will result in an error or mi sinterpretation.



```
In [66]:
                                                                                           M
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
In [68]:
                                                                                           H
class names = data.species
In [69]:
KNN = KNeighborsClassifier()
In [70]:
y_final = model.fit(X_train, y_train).predict(X_test)
In [71]:
                                                                                           H
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30)
                                                                                           H
In [72]:
y_final = KNN.fit(X_train, y_train).predict(X_test)
```

In [73]:	M
<pre>print("Accuracy of the model by KNN is :-",accuracy_score(y_test,y_predict)*100)</pre>	
Accuracy of the model by KNN is :- 35.55555555556	
In []:	K

LOGISTIC REGRESSION

```
In [8]:

import pandas as pd
import numpy as np
import seaborn as sns
from sklearn import svm
In [9]:
```

In [9]:

data = pd.read_csv('C:\\Users\\MARVEL SAMUEL JACOB\\Desktop\\marvel.csv')

In [10]: ▶

data

Out[10]:

	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
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

150 rows × 5 columns

```
In [15]:
X=data[['sepal_length', 'sepal_width', 'petal_length', 'petal_width']]
```

```
In [16]:
y=data['species']
```

```
In [37]:
                                                                                          H
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.3,random_state=68)
In [38]:
from sklearn.linear_model import LogisticRegression
model=LogisticRegression()
model.fit(X_train,y_train)
C:\Users\MARVEL SAMUEL JACOB\anaconda3\lib\site-packages\sklearn\linear_mode
1\_logistic.py:814: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max_iter) or scale the data as shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html (https://scik
it-learn.org/stable/modules/preprocessing.html)
Please also refer to the documentation for alternative solver options:
    https://scikit-learn.org/stable/modules/linear_model.html#logistic-regre
ssion (https://scikit-learn.org/stable/modules/linear model.html#logistic-re
gression)
  n_iter_i = _check_optimize_result(
Out[38]:
LogisticRegression()
                                                                                          H
In [39]:
y_pre=model.predict(X_test)
In [40]:
                                                                                          H
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
In [41]:
print("Accuracy of the model by Logistic Regression is :-",accuracy_score(y_test,y_pre)*100
```

MODELS	<u>ACCURACY</u>
SUPPORT VECTOR MACHINE	93.333333333333
KNN	42.22222222222
DECISION TREE	35.55555555556
LOGISTIC REGRESSION	97.77777777777

FROM HERE, WE CAME TO AN CONCLUSION THAT THE MODEL LOGISTIC REGRESSION HAS THE HIGHEST ACCURACY AMONG SVM, KNN AND DECISION TREE WHIL USING IRIS FLOWERS DATASET