	Name = Maalika Maini Internship at Let's grow more
[_0].	# Loading Libraries import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns import sklearn from sklearn.model_selection import train_test_split from sklearn.metrics import classification_report from sklearn.svm import SVC from sklearn.metrics import accuracy_score
In [16]:	Loadind datasets  data_link="http://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data" iris_data=pd.read_csv("http://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data")  Reading the dataset
In [17]: Out[17]:	ITIS_uatarnead() # Starting 5 values
In [18]: Out[18]:	ITIS_uata.tall() #last 3 values
	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
In [19]: Out[19]: In [20]:	Index(['5.1', '3.5', '1.4', '0.2', 'Iris-setosa'], dtype='object')
Out[20]:	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
In [21]:	4 5.4 3.9 1.7 0.4 Iris-setosa  Explorartory Data Analysis on Dataset  iris_data.shape # To get no.of columns  (149, 5)
In [22]: Out[22]:	1115_444446561156()
	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 [24]: Out[24]:	0FalseFalseFalseFalse1FalseFalseFalseFalseFalse2FalseFalseFalseFalseFalse
	3FalseFalseFalseFalseFalse4FalseFalseFalseFalseFalse144FalseFalseFalseFalseFalse145FalseFalseFalseFalseFalse146FalseFalseFalseFalseFalse
In [10]:	147 False False False False False  148 False False False False False  149 rows × 5 columns  # Checking missing values iris_data.isnull().sum()
Out[10]: In [12]:	<pre>sepal_length 0 sepal_width 0 petal_length 0 petal_width 0 class 0 dtype: int64  iris_data.value_counts() # To get the count of each features in the columns</pre>
Out[12]:	sepal_length         sepal_width         petal_length         class           4.9         3.1         1.5         0.1         Iris-setosa         3           5.8         2.7         5.1         1.9         Iris-virginica         2           6.2         3.4         5.4         2.3         Iris-virginica         1           6.3         2.3         4.4         1.3         Iris-versicolor         1           2.5         4.9         1.5         Iris-versicolor         1           5.5         2.4         3.7         1.0         Iris-versicolor         1           3.8         1.1         Iris-versicolor         1           2.5         4.0         1.3         Iris-versicolor         1
In [13]: Out[13]:	2.6
In [22]:	petal_width 2.5 class Iris-virginica dtype: object  Visualization  plt.figure(figsize=(12,8)) sns.boxplot(data=iris_data,width=0.5,fliersize=5) plt.show()
	3-
In [23]: Out[23]:	<b>sepal_length</b> 1.000000 -0.103784 0.871283 0.816971
In [24]:	sepal_width       -0.103784       1.000000       -0.415218       -0.350733         petal_length       0.871283       -0.415218       1.000000       0.962314         petal_width       0.816971       -0.350733       0.962314       1.000000         plt.figure(figsize=(12,8))         sns.heatmap(corr, annot=True, cmap='RdGy_r')         plt.show()
	1 -0.1 0.87 0.82 -0.8 -0.6
	- 0.0  - 0.82  - 0.82  - 0.35  - 0.96  1  - 0.0  - 0.0
In [25]:	<pre># Viol in Plot sns.violinplot(y='class', x='sepal_length', data=iris_data, inner='quartitle') plt.show() sns.violinplot(y='class', x='sepal_width', data=iris_data, inner='quartitle') plt.show() sns.violinplot(y='class', x='sepal_width', data=iris_data, inner='quartitle') plt.show() sns.violinplot(y='class', x='petal_length', data=iris_data, inner='quartitle')</pre>
	plt.show() sns.violinplot(y='class',x='petal_width',data=iris_data,inner='quartitle') plt.show()  Iris-setosa
	Iris-virginica - 4 5 6 7 8 sepal_length
	Iris-setosa - Se Iris-versicolor - Se Iris-versicol
	Iris-virginica - 2.0 2.5 3.0 3.5 4.0 4.5 sepal_width
	Iris-setosa - W Iris-versicolor - W Iris-versi
	Iris-virginica 1 2 3 4 5 6 7 petal_length  Iris-setosa -
	Iris-virginica -
In [26]: Out[26]:	# Box and Whisker plots iris_data.plot(kind='box', subplots=True, layout=(3,2), figsize=(10,12))  sepal_length
	petal_width dtype: object  8.0 7.5 7.0 6.5 6.0  AxesSubplot(0.547727,0.391471;0.352273x0.222059)  4.5 8 8 4.0 3.5 3.0
	5.5 -
In [27]:	iris_data.hist(figsize=(12,12)) ##Drawing a histogram to see the distribution of data plt.show()  sepal_length sepal_width
	25 20 25 20 15
	10 4.5 5.0 5.5 6.0 6.5 7.0 7.5 8.0 2.0 2.5 3.0 3.5 4.0 4.5
	petal_length
In [28]: Out[28]:	sns.pairplot(iris_data)
In [33]:	Sepal_length Sepal_width Petal_length Petal_width  Building Model  x=iris_data.drop(['class'], axis=1) y=iris_data['class']
In [34]: In [37]:	<pre>print(f'x shape:{x.shape}  y shape:{y.shape}')  x shape:(149, 4)  y shape:(149,)  x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.10,random_state=1)  model=[] model.append(('SVC',SVC(gamma='auto')))</pre>
In [45]: In [47]:	<pre>model.append(('SVC',SVC(gamma='auto')))  model=SVC(gamma='auto') model.fit(x_train,y_train) predict=model.predict(x_test)  print(f'Test Accuracy:{accuracy_score(y_test,predict)}') print(f'Classification_report:{classification_report(y_test,predict)}')</pre>
	Test Accuracy:0.9333333333333333333333333333333333333
	macro avg 0.97 0.83 0.87 15 weighted avg 0.94 0.93 0.92 15