

```
In [2]: #Step1:)Import required libraries
import pandas
from pandas.plotting import scatter_matrix
import matplotlib.pyplot as plt
from sklearn import model_selection
from sklearn.metrics import classification_report
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
from sklearn.linear_model import LogisticRegression #LR
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
from sklearn.naive_bayes import GaussianNB #NB
from sklearn.svm import SVC
```

```
In [5]: #step2)Loading dataset
url = "https://raw.githubusercontent.com/jbrownlee/Datasets/master/iris.csv"
names = ['sepal-length', 'sepal-width', 'petal-length', 'petal-width', 'class']
dataset = pandas.read_csv(url, names=names) #{url,name=var where labels are}
dataset.head()
```

Out[5]:

	sepal-length	sepal-width	petal-length	petal-width	class
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 [4]: #shape->no of rows and columns
print(dataset.shape)
```

(150, 5)

```
In [5]: print(dataset.head(50))
```

	sepal-length	sepal-width	petal-length	petal-width	class
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
5	5.4	3.9	1.7	0.4	Iris-setosa
6	4.6	3.4	1.4	0.3	Iris-setosa
7	5.0	3.4	1.5	0.2	Iris-setosa
8	4.4	2.9	1.4	0.2	Iris-setosa
9	4.9	3.1	1.5	0.1	Iris-setosa
10	5.4	3.7	1.5	0.2	Iris-setosa
11	4.8	3.4	1.6	0.2	Iris-setosa
12	4.8	3.0	1.4	0.1	Iris-setosa
13	4.3	3.0	1.1	0.1	Iris-setosa
14	5.8	4.0	1.2	0.2	Iris-setosa
15	5.7	4.4	1.5	0.4	Iris-setosa
16	5.4	3.9	1.3	0.4	Iris-setosa
17	5.1	3.5	1.4	0.3	Iris-setosa
18	5.7	3.8	1.7	0.3	Iris-setosa
19	5.1	3.8	1.5	0.3	Iris-setosa
20	5.4	3.4	1.7	0.2	Iris-setosa
21	5.1	3.7	1.5	0.4	Iris-setosa
22	4.6	3.6	1.0	0.2	Iris-setosa
23	5.1	3.3	1.7	0.5	Iris-setosa
24	4.8	3.4	1.9	0.2	Iris-setosa
25	5.0	3.0	1.6	0.2	Iris-setosa
26	5.0	3.4	1.6	0.4	Iris-setosa
27	5.2	3.5	1.5	0.2	Iris-setosa
28	5.2	3.4	1.4	0.2	Iris-setosa
29	4.7	3.2	1.6	0.2	Iris-setosa
30	4.8	3.1	1.6	0.2	Iris-setosa
31	5.4	3.4	1.5	0.4	Iris-setosa
32	5.2	4.1	1.5	0.1	Iris-setosa
33	5.5	4.2	1.4	0.2	Iris-setosa
34	4.9	3.1	1.5	0.1	Iris-setosa
35	5.0	3.2	1.2	0.2	Iris-setosa
36	5.5	3.5	1.3	0.2	Iris-setosa
37	4.9	3.1	1.5	0.1	Iris-setosa
38	4.4	3.0	1.3	0.2	Iris-setosa
39	5.1	3.4	1.5	0.2	Iris-setosa
40	5.0	3.5	1.3	0.3	Iris-setosa
41	4.5	2.3	1.3	0.3	Iris-setosa
42	4.4	3.2	1.3	0.2	Iris-setosa
43	5.0	3.5	1.6	0.6	Iris-setosa
44	5.1	3.8	1.9	0.4	Iris-setosa
45	4.8	3.0	1.4	0.3	Iris-setosa
46	5.1	3.8	1.6	0.2	Iris-setosa
47	4.6	3.2	1.4	0.2	Iris-setosa
48	5.3	3.7	1.5	0.2	Iris-setosa
49	5.0	3.3	1.4	0.2	Iris-setosa

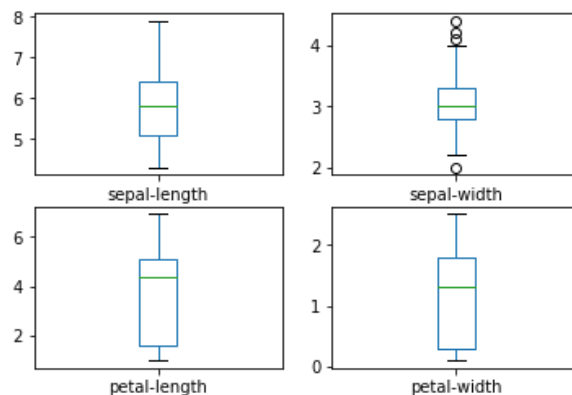
```
In [8]: #Step3:Summerization of dataset
#descriptions
print(dataset.describe())
```

	sepal-length	sepal-width	petal-length	petal-width
count	150.000000	150.000000	150.000000	150.000000
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

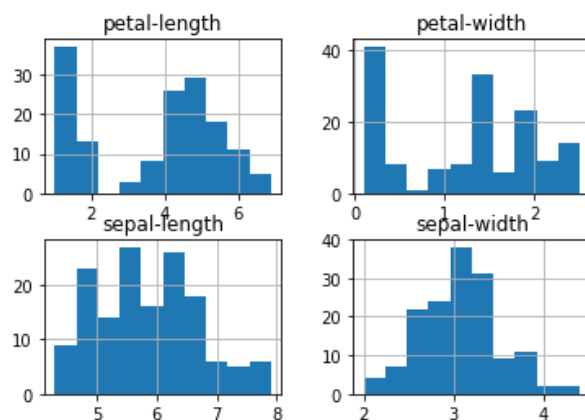
```
In [7]: #class distribution
print(dataset.groupby('class').size())
```

```
class
Iris-setosa      50
Iris-versicolor  50
Iris-virginica   50
dtype: int64
```

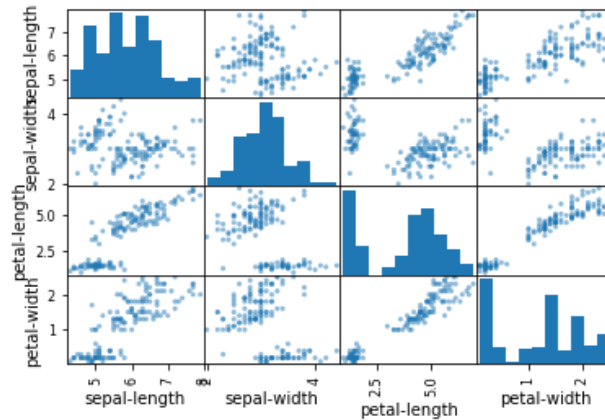
```
In [8]: # box and whisker plots
#univariate_plot (diff attr->diff plotting)
dataset.plot(kind='box', subplots=True, layout=(2,2), sharex=True, sharey=False)
plt.show()
```



```
In [9]: #Step:4) Visualization of dataset
#histograms #univariate_plot
dataset.hist()
plt.show()
```



```
In [11]: #multivariate_plot
scatter_matrix(dataset)
plt.show()
```



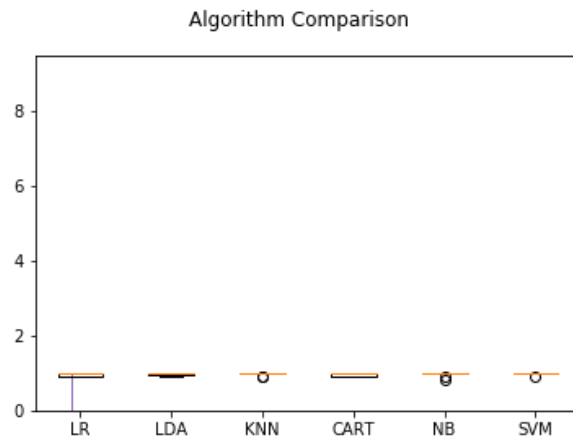
```
In [9]: #dividing dataset for training(with-label) and testing(without-label) datasets
# Split-out validation dataset
array = dataset.values
X = array[:,0:4] #training-data column 1 to 4
Y = array[:,4] #training-answer column 5
validation_size = 0.20 #20% of dataset
seed = 7 #value for random number generation
X_train, X_validation, Y_train, Y_validation = model_selection.train_test_split
(X, Y, test_size=validation_size, random_state=seed #based on seed)
```

```
In [10]: #Step4)Evaluating algorithms
seed = 7
scoring = 'accuracy'
models = []
models.append(('LR', LogisticRegression()))
models.append(('LDA', LinearDiscriminantAnalysis()))
models.append(('KNN', KNeighborsClassifier()))
models.append(('CART', DecisionTreeClassifier()))
models.append(('NB', GaussianNB()))
models.append(('SVM', SVC()))
# evaluate each model in turn
results = []
names = []
for name, model in models:
    kfold = model_selection.KFold(n_splits=10, random_state=seed) #dividing 120 in 10 10 epochs
    cv_results = model_selection.cross_val_score(model, X_train, Y_train, cv=kfold, scoring=scoring)
    results.append(cv_results)
    names.append(name)
    msg = "%s: %f (%f)" % (name, cv_results.mean(), cv_results.std())
    print(msg)
```

LR: 0.966667 (0.040825)
LDA: 0.975000 (0.038188)
KNN: 0.983333 (0.033333)
CART: 0.975000 (0.038188)
NB: 0.975000 (0.053359)
SVM: 0.991667 (0.025000)

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In [12]: # Compare Algorithms
fig = plt.figure()
fig.suptitle('Algorithm Comparison')
#plt.hist(results)
ax = fig.add_subplot(111)
plt.boxplot(results)
plt.hist(results)
ax.set_xticklabels(names)
plt.show()
```



```
In [14]: #Step:5) Making predictions
# Make predictions on validation dataset
knn = KNeighborsClassifier()
knn.fit(X_train, Y_train) #fitting values for generalized model
predictions = knn.predict(X_validation) #prediction on remaining 20% of data
set that is in X_validation
print("accuracy score",accuracy_score(Y_validation, predictions)) #comparin
g with y-valid for accuracy
print("confusion matrix \n",confusion_matrix(Y_validation, predictions))
#print(classification_report(Y_validation, predictions))
```

```
accuracy score 0.9
confusion matrix
[[ 7  0  0]
 [ 0 11  1]
 [ 0  2  9]]
```

```
In [ ]: #MATRIX:
#      classes
# predi -----
# cted  |
# data  |
#7 are from class 1 (iris-sentosa) 7-right predicted in class 1
#12 are from class 2 (iris-versicolor) 11 in class 2 and 1 false predicted i
n class 3
#11 are from class 3 (iris-) 2 false predicted in class 2 and 9 is true pred
iction of class 3
```

```
In [ ]:
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