

```
In [101]: 1 from sklearn.datasets import load_iris  
2 import pandas as pd  
3 import numpy as np
```

```
In [102]: 1 iris = load_iris()
```

```
In [103]: 1 data = pd.DataFrame(iris.data, columns = iris.feature_names)
```

```
In [104]: 1 data.head()
```

Out[104]:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
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

```
In [105]: 1 from sklearn.preprocessing import StandardScaler
```

In [106]:

```
1 data
```

Out[106]:

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
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 × 4 columns

In [107]:

```
1 scaler = StandardScaler()
```

In [108]:

```
1 data = scaler.fit_transform(data)
```

In [109]:

```
1 data = pd.DataFrame(data, columns = iris.feature_names)
```

```
In [110]: 1 data.describe()
```

```
Out[110]:
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
count	1.500000e+02	1.500000e+02	1.500000e+02	1.500000e+02
mean	-1.690315e-15	-1.842970e-15	-1.698641e-15	-1.409243e-15
std	1.003350e+00	1.003350e+00	1.003350e+00	1.003350e+00
min	-1.870024e+00	-2.433947e+00	-1.567576e+00	-1.447076e+00
25%	-9.006812e-01	-5.923730e-01	-1.226552e+00	-1.183812e+00
50%	-5.250608e-02	-1.319795e-01	3.364776e-01	1.325097e-01
75%	6.745011e-01	5.586108e-01	7.627583e-01	7.906707e-01
max	2.492019e+00	3.090775e+00	1.785832e+00	1.712096e+00

```
In [111]: 1 from sklearn.decomposition import PCA
```

```
In [112]: 1 pca = PCA(n_components = 2)
```

```
In [113]: 1 principal_components = pca.fit_transform(data)
```

```
In [114]: 1 data = pd.DataFrame(principal_components, columns = ["PC1", "PC2"])
```

In [115]:

```
1 data
```

Out[115]:

	PC1	PC2
0	-2.264703	0.480027
1	-2.080961	-0.674134
2	-2.364229	-0.341908
3	-2.299384	-0.597395
4	-2.389842	0.646835
...
145	1.870503	0.386966
146	1.564580	-0.896687
147	1.521170	0.269069
148	1.372788	1.011254
149	0.960656	-0.024332

150 rows × 2 columns

In [116]:

```
1 pca.explained_variance_ratio_.sum()
```

Out[116]: 0.9581320720000164

In [117]:

```
1 data['Target'] = iris.target
```

In [118]: 1 data

Out[118]:

	PC1	PC2	Target
0	-2.264703	0.480027	0
1	-2.080961	-0.674134	0
2	-2.364229	-0.341908	0
3	-2.299384	-0.597395	0
4	-2.389842	0.646835	0
...
145	1.870503	0.386966	2
146	1.564580	-0.896687	2
147	1.521170	0.269069	2
148	1.372788	1.011254	2
149	0.960656	-0.024332	2

150 rows × 3 columns

In [119]: 1 from sklearn.model_selection import train_test_split

In [126]: 1 X_train,X_test,y_train,y_test = train_test_split(data.drop(['Target'],1),data['Target'],random_state=1)

In [127]: 1 from sklearn.linear_model import LogisticRegression

In [128]: 1 model1 = LogisticRegression()

```
In [129]: 1 model1.fit(X_train,y_train)
```

```
Out[129]: LogisticRegression()
```

```
In [130]: 1 from sklearn.metrics import classification_report
```

```
In [131]: 1 print(classification_report(model1.predict(X_test),y_test))
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	13
1	0.94	1.00	0.97	15
2	1.00	0.90	0.95	10
accuracy			0.97	38
macro avg	0.98	0.97	0.97	38
weighted avg	0.98	0.97	0.97	38

Without Principal Components

```
In [132]: 1 data = pd.DataFrame(iris.data,columns = iris.feature_names)
```

```
In [133]: 1 data['Target'] = iris.target
```

```
In [134]: 1 from sklearn.model_selection import train_test_split
```

```
In [135]: 1 X_train,X_test,y_train,y_test = train_test_split(data.drop('Target',1),data['Target'],random_state=1)
```

In [136]: 1 model1.fit(X_train,y_train)

C:\Users\yashm\anaconda3\lib\site-packages\sklearn\linear_model_logistic.py:765: 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://scikit-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-regression (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)

extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG)

Out[136]: LogisticRegression()

In [137]: 1 print(classification_report(model1.predict(X_test),y_test))

	precision	recall	f1-score	support
0	1.00	1.00	1.00	13
1	0.94	1.00	0.97	15
2	1.00	0.90	0.95	10
accuracy			0.97	38
macro avg	0.98	0.97	0.97	38
weighted avg	0.98	0.97	0.97	38

In []: 1

In []: 1

