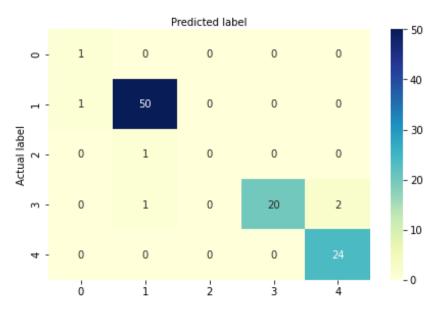
Confusion Matrix:

```
## Assignment 2 Machine learning
      import matplotlib.pyplot as plt
      import numpy as np
      import pandas as pd
      from sklearn.linear_model import LogisticRegression
      from sklearn.metrics import classification_report, confusion_matrix
      data = pd.read_csv("auto-mpg.csv")
      data.head()
      feature_cols = ['mpg', 'displacement', 'horsepower', 'weight','acceleration','model year']
X = data[feature_cols] # Features
      y = data.cylinders # Target variable
      from sklearn.model_selection import train_test_split
      X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.25,random_state=0) from sklearn.linear_model import LogisticRegression
      logreg = LogisticRegression()
      # fit the model with data
      logreg.fit(X_train,y_train)
      y_pred=logreg.predict(X_test)
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      from sklearn import metrics
      cnf_matrix = metrics.confusion_matrix(y_test, y_pred)
      import numpy as np
      import matplotlib.pyplot as plt
      import seaborn as sns
      class_names=[0,1] # name of classes
      fig, ax = plt.subplots()
      tick_marks = np.arange(len(class_names))
      plt.xticks(tick_marks, class_names)
      plt.yticks(tick_marks, class_names)
      sns.heatmap(pd.DataFrame(cnf_matrix), annot=True, cmap="YlGnBu" ,fmt='g')
      ax.xaxis.set_label_position("top")
      plt.tight_layout()
      plt.title('Confusion matrix', y=1.1)
plt.ylabel('Actual label')
plt.xlabel('Predicted label')
```

OUTPUT:





PCA Analysis:

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

# importing or loading the dataset
data = pd.read_csv('auto-mpg.csv')

feature_cols = ['mpg','displacement', 'horsepower', 'weight','acceleration','model year']

X = data[feature_cols] # Features
y = data.cylinders # Target variable
from sklearn.model_selection import train_test_split
X_train_X_test_y_train_y_test=train_test_split(X,y,test_size=0.25,random_state=0)
# performing preprocessing part
from sklearn.preprocessing import StandardScaler
sc StandardScaler()

X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_train)
X_test = sc.transform(X_test)

# Applying PCA function on training
and testing set of X component
from sklearn.decomposition import PCA

pca = PCA(n_components=2)

X_train = pca.fit_transform(X_train)
X_test = pca.transform(X_test)

# Fitting Logistic Regression To the training set
# from sklearn.ensemble import RandomForestClassifier

# classifier = RandomForestClassifier(max_depth=2, random_state=0)
# classifier.fit(X_train, y_train)

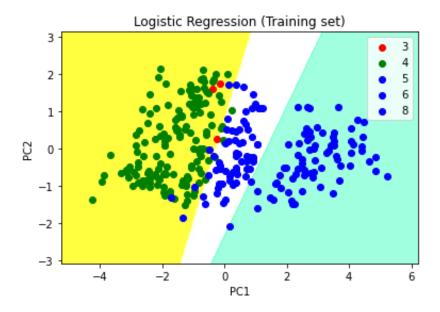
# Predicting the Test set results
# y_pred = classifier.predict(X_test)
# from sklearn.metrics import confusion_matrix
```

```
# cm = confusion_matrix(y_test, y_pred)
     # print(cm)
     # print('Accuracy' + accuracy_score(y_test, y_pred))
     # Fitting Logistic Regression To the training set
     from sklearn.linear_model import LogisticRegression
     classifier = LogisticRegression(random_state = 0)
     classifier.fit(X_train, y_train)
     # Predicting the test set result using
     # predict function under LogisticRegression
     y_pred = classifier.predict(X_test)
     # making confusion matrix between
     # test set of Y and predicted value.
     from sklearn.metrics import confusion matrix
     cm = confusion_matrix(y_test, y_pred)
     # Predicting the training set
     from matplotlib.colors import ListedColormap
     X_set, y_set = X_train, y_train
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     X1, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1,
                         stop = X_set[:, 0].max() + 1, step = 0.01),
np.arange(start = X_set[:, 1].min() - 1,
stop = X_set[:, 1].max() + 1, step = 0.01))
     plt.xlim(X1.min(), X1.max())
plt.ylim(X2.min(), X2.max())
     for i, j in enumerate(np.unique(y_set)):
```

```
plt.title('Logistic Regression (Training set)')
plt.xlabel('PC1') # for Xlabel
plt.ylabel('PC2') # for Ylabel
plt.legend() # to show legend

# show scatter plot
plt.show()
```

OUTPUT:



Variables:

