Script reads from fft and mfcc files and trains using logistic regression and knn

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- Evaluation Environment: Ubuntu 18.04 , Anaconda3 (202002), Python 2.7
- IN: Paths to directories consisting of FFT files, and MFCC files.
- OUT: Splits dataset as per code into train and test sets, performs training and tests. Displays classification accuracy along with confusion matrix.

In [1]:

```
import sklearn
from sklearn import linear_model
from sklearn.neighbors import KNeighborsClassifier
# from sklearn.cross_validation import train_test_split
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report
from sklearn.metrics import accuracy_score
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import StandardScaler
from sklearn.feature_selection import VarianceThreshold
from sklearn.feature_selection import SelectFromModel
import lightgbm as lgbm
from sklearn.svm import SVC
from sklearn.model_selection import GridSearchCV

import matplotlib.pyplot as plt
import scipy
import os
import sys
import glob
import numpy as np
```

Reads FFT-files and prepares X_train and y_train. genre_list must consist of names of folders/genres consisting of the required FFT-files. base_dir must contain genre_list of directories

In [2]:

```
def read_fft(genre_list, base_dir):
    X = []
    y = []
    for label, genre in enumerate(genre_list):
        # create UNIX pathnames to id FFT-files.
        genre_dir = os.path.join(base_dir, genre, "*.fft.npy")
        # get path names that math genre-dir
        file_list = glob.glob(genre_dir)
        for file in file_list:
            fft_features = np.load(file)
            X.append(fft_features)
            y.append(label)

    # print(X)
    # print(y)
    # print(len(X))
    # print(len(y))

    return np.array(X), np.array(y)
```

Rreads MFCC-files and prepares X_train and y_train. genre_list must consist of names of folders/genres consisting of the required MFCC-files base_dir must contain genre_list of directories

In [3]:

```
def read_ceps(genre_list, base_dir):
    X, y = [], []
    for label, genre in enumerate(genre_list):
        for fn in glob.glob(os.path.join(base_dir, genre, "*.ceps.npy")):
            ceps = np.load(fn)
            num_ceps = len(ceps)
            X.append(np.mean(ceps[int(num_ceps*1/10):int(num_ceps*9/10)], axis=0))
            y.append(label)

    return np.array(X), np.array(y)
```

In [4]:

```
def train_score(classifier, Xtrain, Xtest, ytrain, ytest):
        train_acc = classifier.score(Xtrain, ytrain)
        test_acc = classifier.score(Xtest, ytest)
        print("Training Data Accuracy: %0.2f" % (train_acc))
        print("Test Data Accuracy:     %0.2f" % (test_acc))
        ypred = classifier.predict(Xtest)
        conf = confusion_matrix(ytest, ypred)
        precision = (conf[0, 0] / (conf[0, 0] + conf[1, 0]))
        recall = (conf[0, 0] / (conf[0, 0] + conf[0, 1]))
        f1_score = 2 * ((precision * recall)/(precision + recall))
        print("Precision:              %0.2f" % precision)
        print("Recall:                 %0.2f" % recall)
        print("F1 Score:                  %0.2f" % f1_score)
        print('\n')
```

In [5]:

```
def plot_confusion_matrix(cm, title, genre_list, cmap=plt.cm.Blues):
        plt.imshow(cm, interpolation='nearest', cmap=cmap)
        plt.title(title)
        plt.colorbar()
        tick_marks = np.arange(len(genre_list))
        plt.xticks(tick_marks, genre_list, rotation=45)
        plt.yticks(tick_marks, genre_list)
        plt.tight_layout()
        plt.ylabel('True label')
        plt.xlabel('Predicted label')
        plt.show()
```

From now on, the program will be started actually. a main routine is as follows.

In [6]:

```
# List of genres (these must be folder names consisting .wav of respective genre in the base_dir) Change list if ne
# For example, IF YOU WANT TO CLASSIFY ONLY CLASSICAL AND JAZZ,  specify genre_list = ["classical", "jazz"]
genre_list = ["classical", "hiphop", "jazz", "pop", "rock", "metal"]
target_names = genre_list
```

In [7]:

```
# use FFT
# base_dir_fft  = "genres.FFT/"
#X, y = read_fft(genre_list, base_dir_fft)
#X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = .25)
#print('\n******USING FFT******')

# print("X_train = " + str(len(X_train)), "y_train = " + str(len(y_train)), "X_test = " + str(len(X_test)), "y_test = " + s
```

In [8]:

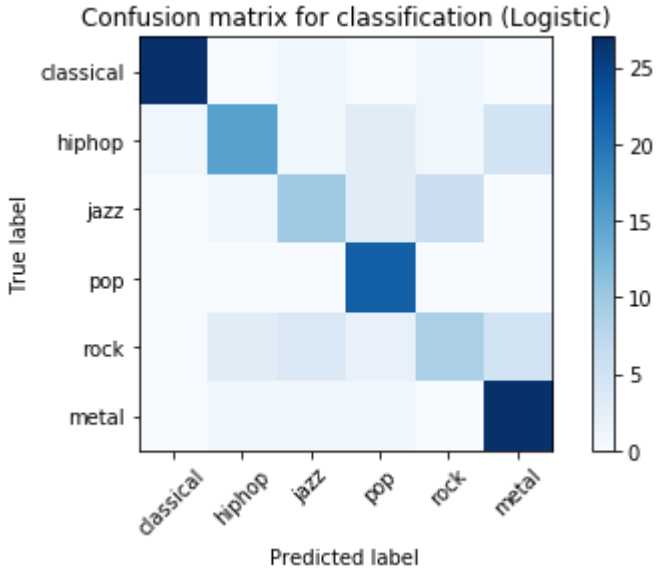
```
# use MFCC
base_dir_mfcc = "genres.MFCC/"
X, y = read_ceps(genre_list, base_dir_mfcc)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.25)
print('******USING MFCC******')
```

```
******USING MFCC******
```

In [9]:

```
# print("X_train = " + str(len(X_train)), "y_train = " + str(len(y_train)), "X_test = " + str(len(X_test)), "y_test = " + st

print("######## CLASSIFICATION REPORT with Logistic Regression ########")
logistic_classifier = linear_model.LogisticRegression()
logistic_classifier.fit(X_train, y_train)

logistic_predictions = logistic_classifier.predict(X_test)
logistic_accuracy = accuracy_score(y_test, logistic_predictions)
logistic_cm = confusion_matrix(y_test, logistic_predictions)
# print("logistic accuracy (validation set)= " + str(logistic_classifier.best_score_))
print("logistic accuracy (test set)= " + str(logistic_accuracy))
print("logistic confusion matrix:")
print(logistic_cm)

print(classification_report(y_test, logistic_predictions,target_names=target_names))

plot_confusion_matrix(logistic_cm, "Confusion matrix for classification (Logistic)", genre_list)
```

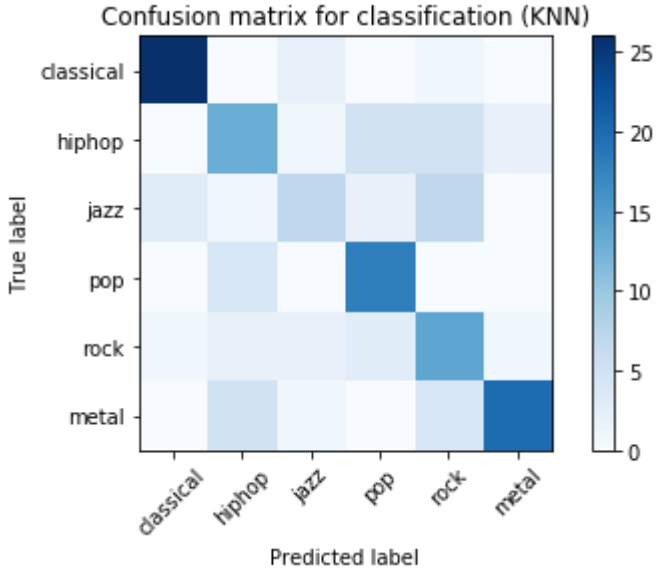
```
######## CLASSIFICATION REPORT with Logistic Regression ########
logistic accuracy (test set)= 0.7333333333333333
logistic confusion matrix:
[[27  0  1  0  1  0]
 [ 1 15  1  3  1  5]
 [ 0  1 10  3  6  0]
 [ 0  0  0 22  0  0]
 [ 0  3  4  2  9  5]
 [ 0  1  1  1  0 27]]
              precision    recall  f1-score   support

   classical       0.96      0.93      0.95        29
      hiphop       0.75      0.58      0.65        26
        jazz       0.59      0.50      0.54        20
         pop       0.71      1.00      0.83        22
        rock       0.53      0.39      0.45        23
       metal       0.73      0.90      0.81        30

   micro avg       0.73      0.73      0.73       150
   macro avg       0.71      0.72      0.70       150
weighted avg       0.73      0.73      0.72       150


/var/www/invain/anaconda3/envs/python27/lib/python2.7/site-packages/sklearn/linear_mod
el/logistic.py:433: FutureWarning: Default solver will be changed to 'lbfgs' in 0.22. Specify a sol
ver to silence this warning.
  FutureWarning)
/var/www/invain/anaconda3/envs/python27/lib/python2.7/site-packages/sklearn/linear_mod
el/logistic.py:460: FutureWarning: Default multi_class will be changed to 'auto' in 0.22. Specify
the multi_class option to silence this warning.
  "this warning.", FutureWarning)
```


Confusion matrix for classification (Logistic)
classical
hiphop
jazz
pop
rock
metal
True label
Predicted label
25
20
15
10
5
0

In [10]:

```
print("######## CLASSIFICATION REPORT with KNeighbors Classifier ########")
knn_classifier = KNeighborsClassifier()
knn_classifier.fit(X_train, y_train)
knn_predictions = knn_classifier.predict(X_test)
knn_accuracy = accuracy_score(y_test, knn_predictions)
knn_cm = confusion_matrix(y_test, knn_predictions)

#print("knn accuracy (validation set)= " + str(knn_classifier.best_score_))
print("knn accuracy (test set)= " + str(knn_accuracy))
print("knn confusion matrix:")
print(knn_cm)

print(classification_report(y_test, knn_predictions,target_names=target_names))

plot_confusion_matrix(knn_cm, "Confusion matrix for classification (KNN)", genre_list)
```

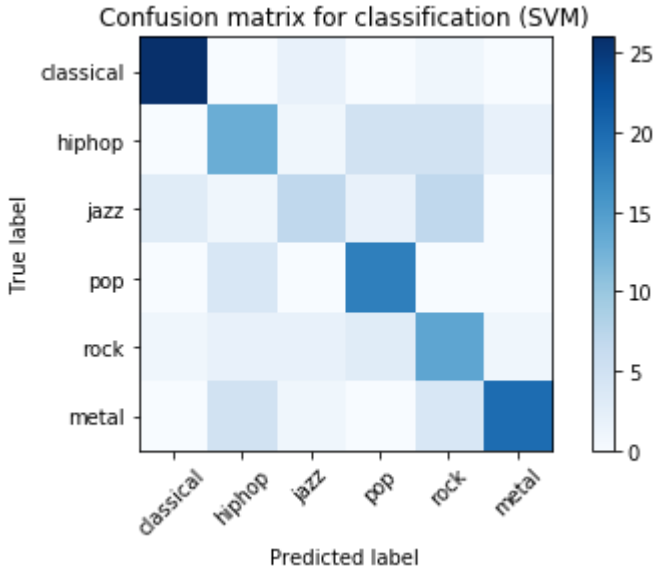
```
######## CLASSIFICATION REPORT with KNeighbors Classifier ########
knn accuracy (test set)= 0.6533333333333333
knn confusion matrix:
[[26  0  2  0  1  0]
 [ 0 13  1  5  5  2]
 [ 3  1  7  2  7  0]
 [ 0  4  0 18  0  0]
 [ 1  2  2  3 14  1]
 [ 0  5  1  0  4 20]]
              precision    recall  f1-score   support

   classical       0.87      0.90      0.88        29
     hiphop        0.52      0.50      0.51        26
        jazz       0.54      0.35      0.42        20
         pop       0.64      0.82      0.72        22
        rock       0.45      0.61      0.52        23
       metal       0.87      0.67      0.75        30

   micro avg       0.65      0.65      0.65       150
   macro avg       0.65      0.64      0.63       150
weighted avg       0.67      0.65      0.65       150
```



In [11]:

```
print("######## CLASSIFICATION REPORT with SVM (Support Vector Machin)########")
params = {
    "cls__C": [0.5, 1, 2, 5],
    "cls__kernel": ['rbf', 'linear', 'sigmoid'],
}

pipe_svm = Pipeline([
    ('scale', StandardScaler()),
    ('var_tresh', VarianceThreshold(threshold=(.8 * (1 - .8)))),
    ('feature_selection', SelectFromModel(lgbm.LGBMClassifier())),
    ('cls', SVC())
])

svm_classifier = GridSearchCV(pipe_svm, params, scoring='accuracy', n_jobs=6, cv=5)
svm_classifier.fit(X_train, y_train)
svm_predictions = svm_classifier.predict(X_test)
svm_cm = confusion_matrix(y_test, svm_predictions)
svm_accuracy = accuracy_score(y_test, svm_predictions)
#print("svm accuracy (validation set)= " + str(svm_classifier.best_score_))
print("svm accuracy (test set)= " + str(svm_accuracy))
print("svm confusion matrix:")
print(svm_cm)

print(classification_report(y_test, svm_predictions,target_names=target_names))

plot_confusion_matrix(knn_cm, "Confusion matrix for classification (SVM)", genre_list)
```

```
######## CLASSIFICATION REPORT with SVM (Support Vector Machin)########

/var/www/invain/anaconda3/envs/python27/lib/python2.7/site-packages/sklearn/model_sel
ection/_search.py:841: DeprecationWarning: The default of the `iid` parameter will change from
True to False in version 0.22 and will be removed in 0.24. This will change numeric results whe
n test-set sizes are unequal.
  DeprecationWarning)

svm accuracy (test set)= 0.7533333333333333
svm confusion matrix:
[[29  0  0  0  0  0]
 [ 1 10  0  4  3  8]
 [ 2  2 15  1  0  0]
 [ 0  2  0 20  0  0]
 [ 0  3  3  3 12  2]
 [ 0  2  0  0  1 27]]
              precision    recall  f1-score   support

   classical       0.91      1.00      0.95        29
      hiphop       0.53      0.38      0.44        26
        jazz       0.83      0.75      0.79        20
         pop       0.71      0.91      0.80        22
        rock       0.75      0.52      0.62        23
       metal       0.73      0.90      0.81        30

   micro avg       0.75      0.75      0.75       150
   macro avg       0.74      0.74      0.73       150
weighted avg       0.74      0.75      0.74       150
```


Confusion matrix for classification (SVM)
classical
hiphop
jazz
pop
rock
metal
True label
Predicted label
25
20
15
10
5
0