

```
In [1]: import os
import torch
import librosa

import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline

from datasets import AudioDataset

import sklearn.metrics as met
# hyper parameters
MAX_EPOCH = 100
```

```
In [2]: # path to urban sound 8k
data_root = "/home/wew016/UrbanSound8K/"
# path to label
label_path = "/home/wew016/UrbanSound8K/metadata/UrbanSound8K.csv"
```

```
In [3]: # initialize dataset (feature can be "mfcc" or "spec")
audio_dataset = AudioDataset(3, DataRoot=data_root, LabelPath=label_path
, feature="mfcc_mean", mode="train")

verify mfcc_mean feature success
```

```
In [4]: # initialize dataloader
data_loader = torch.utils.data.DataLoader(audio_dataset, batch_size=len(
audio_dataset), shuffle=True, num_workers=1)

for idx, data in enumerate(data_loader):
    x_train, y_train = data
    print(x_train.shape)
    print(y_train.shape)

torch.Size([7807, 40])
torch.Size([7807])
```

```
In [5]: x_train=x_train.numpy()
x_train.shape
```

```
Out[5]: (7807, 40)
```

```
In [6]: y_train=y_train.numpy()
y_train.shape
```

```
Out[6]: (7807,)
```

```
In [7]: audio_dataset_test = AudioDataset(3, DataRoot=data_root, LabelPath=label
_path, feature="mfcc_mean", mode="test")

verify mfcc_mean feature success
```

```
In [8]: # initialize dataloader
data_loader_test = torch.utils.data.DataLoader(audio_dataset_test, batch
_size=len(audio_dataset_test), shuffle=True, num_workers=1)
```

```

for idx, data in enumerate(data_loader_test):
    x_test, y_test = data
    print(x_test.shape)
    print(y_test.shape)

```

```

torch.Size([925, 40])
torch.Size([925])

```

```

In [9]: x_test=x_test.numpy()
        x_test.shape

```

```

Out[9]: (925, 40)

```

```

In [10]: y_test=y_test.numpy()
         y_test.shape

```

```

Out[10]: (925,)

```

XGBOOST Model

```

In [12]: from xgboost import XGBClassifier

```

```

In [13]: model_xgboost = XGBClassifier(learning_rate =0.2,
                                       n_estimators=150,
                                       max_depth=4,
                                       num_class=10,
                                       min_child_weight=5,
                                       gamma=0.1,
                                       subsample=0.5,
                                       colsample_bytree=0.5,
                                       objective='multi:softmax',
                                       seed=50)

```

```

In [14]: model_xgboost.fit(x_train,y_train)

```

```

Out[14]: XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1,
                       colsample_bynode=1, colsample_bytree=0.5, gamma=0.1, gpu_i
d=-1,
                       importance_type='gain', interaction_constraints='',
                       learning_rate=0.2, max_delta_step=0, max_depth=4,
                       min_child_weight=5, missing=nan, monotone_constraints='()'
,
                       n_estimators=150, n_jobs=0, num_class=10, num_parallel_tre
e=1,
                       objective='multi:softprob', random_state=50, reg_alpha=0,
                       reg_lambda=1, scale_pos_weight=None, seed=50, subsample=0.
5,
                       tree_method='exact', validate_parameters=1, verbosity=None
)

```

```

In [15]: pred_train_xgboost = model_xgboost.predict(x_train)
         met.accuracy_score(y_train,pred_train_xgboost)

```

```

Out[15]: 0.9964134750864608

```

```

In [16]: pred_test_xgboost = model_xgboost.predict(x_test)

```

```

In [17]: y_test

```

```

Out[17]: array([[5, 9, 7, 5, 7, 4, 0, 9, 2, 8, 2, 8, 4, 0, 0, 8, 2, 4, 2, 7, 7, 3,
8, 9, 7, 5, 9, 8, 7, 7, 5, 5, 1, 7, 8, 8, 1, 5, 3, 3, 9, 7, 4, 0,
9, 7, 3, 4, 8, 3, 2, 2, 4, 0, 9, 9, 7, 5, 4, 5, 8, 8, 9, 3, 8, 4,
8, 1, 2, 6, 2, 4, 0, 5, 0, 7, 5, 1, 9, 5, 3, 4, 8, 3, 2, 3, 8, 7,
4, 6, 3, 5, 0, 5, 7, 4, 5, 2, 0, 4, 3, 2, 5, 0, 9, 8, 0, 6, 8, 2,
7, 5, 5, 5, 0, 7, 9, 3, 5, 8, 8, 0, 8, 9, 8, 9, 1, 0, 7, 5, 3, 4,
0, 9, 8, 3, 5, 2, 4, 2, 6, 0, 3, 0, 7, 8, 8, 8, 4, 9, 7, 3, 8, 9,
4, 8, 3, 6, 2, 3, 3, 3, 3, 3, 1, 9, 0, 9, 5, 4, 2, 6, 0, 4, 0, 0,
2, 5, 3, 8, 8, 5, 0, 8, 4, 0, 1, 0, 4, 5, 4, 5, 0, 2, 1, 8, 7, 9,
8, 7, 2, 8, 7, 4, 9, 7, 1, 6, 7, 9, 0, 2, 5, 4, 7, 7, 4, 0, 2, 9,
4, 2, 4, 1, 9, 3, 2, 8, 8, 4, 7, 3, 8, 2, 7, 5, 0, 4, 0, 7, 2, 6,
1, 7, 7, 8, 7, 2, 7, 2, 8, 9, 7, 8, 2, 6, 5, 0, 3, 3, 3, 1, 1, 6,
8, 2, 8, 7, 5, 4, 4, 5, 7, 8, 4, 5, 4, 3, 4, 4, 8, 1, 8, 0, 4, 5,
2, 7, 4, 8, 2, 3, 4, 9, 1, 3, 9, 7, 8, 4, 4, 9, 6, 0, 8, 5, 9, 4,
7, 2, 0, 4, 2, 8, 9, 2, 5, 1, 1, 2, 6, 0, 7, 9, 4, 7, 4, 3, 9, 1,
2, 1, 8, 8, 9, 6, 4, 7, 2, 0, 9, 8, 5, 3, 5, 2, 3, 4, 2, 5, 2, 4,
7, 7, 7, 5, 4, 3, 4, 8, 2, 8, 3, 2, 5, 7, 2, 8, 2, 5, 2, 4, 2, 5,
9, 7, 6, 9, 3, 0, 9, 4, 9, 3, 4, 2, 9, 7, 3, 6, 1, 7, 4, 1, 4, 2,
9, 5, 9, 4, 9, 3, 6, 7, 8, 0, 2, 5, 7, 8, 3, 7, 6, 0, 5, 5, 7, 5,
2, 0, 7, 7, 0, 0, 4, 8, 3, 3, 5, 9, 8, 5, 7, 8, 5, 6, 8, 0, 3, 3,
7, 2, 3, 5, 8, 3, 6, 8, 1, 8, 8, 9, 2, 2, 0, 7, 1, 8, 5, 0, 0, 3,
7, 8, 0, 2, 8, 7, 3, 8, 5, 8, 9, 3, 0, 0, 5, 3, 3, 7, 4, 4, 8, 2,
0, 8, 2, 4, 7, 5, 8, 9, 4, 5, 0, 9, 8, 7, 7, 3, 2, 9, 1, 0, 9, 3,
4, 5, 9, 8, 8, 2, 7, 7, 6, 6, 1, 8, 5, 4, 5, 4, 6, 9, 8, 0, 7, 4,
4, 2, 3, 1, 5, 9, 3, 0, 4, 1, 1, 9, 5, 7, 2, 3, 4, 5, 0, 9, 3, 7,
2, 4, 0, 2, 2, 2, 8, 6, 0, 2, 7, 0, 4, 0, 5, 5, 5, 0, 3, 7, 2, 5,
2, 8, 0, 8, 2, 4, 2, 1, 3, 8, 1, 7, 7, 5, 2, 0, 3, 7, 9, 3, 0, 0,
9, 8, 9, 4, 1, 1, 0, 7, 8, 7, 0, 3, 3, 4, 7, 7, 2, 4, 8, 3, 3, 3,
5, 1, 5, 0, 4, 5, 5, 4, 8, 3, 3, 0, 8, 7, 9, 9, 2, 9, 2, 2, 8, 5,
1, 5, 5, 9, 7, 4, 3, 7, 0, 3, 5, 9, 2, 7, 5, 5, 2, 8, 0, 3, 7, 7,
5, 7, 0, 9, 0, 3, 0, 7, 2, 7, 5, 0, 5, 2, 9, 6, 3, 9, 0, 5, 4, 7,
6, 9, 4, 7, 5, 0, 5, 3, 2, 1, 3, 0, 3, 5, 8, 8, 0, 8, 6, 9, 8, 4,
5, 7, 6, 4, 5, 7, 7, 4, 8, 8, 1, 9, 3, 0, 8, 5, 1, 7, 4, 0, 2, 7,
3, 3, 2, 5, 1, 7, 2, 8, 5, 4, 0, 8, 6, 8, 0, 0, 8, 0, 7, 0, 2, 0,
0, 5, 4, 8, 3, 0, 8, 3, 8, 7, 3, 9, 9, 3, 9, 1, 3, 6, 3, 9, 3, 7,
6, 9, 9, 7, 4, 2, 4, 8, 9, 7, 7, 9, 5, 3, 2, 2, 9, 3, 5, 7, 8, 3,
1, 9, 2, 4, 2, 2, 2, 2, 3, 7, 2, 0, 7, 2, 8, 5, 9, 9, 1, 7, 8, 9,
5, 4, 8, 6, 2, 8, 0, 9, 9, 5, 9, 9, 8, 4, 7, 0, 9, 0, 5, 7, 7, 9,
4, 0, 5, 9, 5, 2, 8, 6, 8, 5, 8, 9, 7, 0, 5, 0, 0, 8, 4, 0, 8, 3,
8, 7, 0, 4, 8, 3, 9, 5, 7, 5, 7, 9, 4, 9, 2, 3, 0, 9, 3, 4, 2, 5,
3, 7, 7, 1, 3, 4, 8, 3, 2, 2, 6, 9, 5, 5, 2, 4, 8, 9, 4, 9, 9, 6,
5, 4, 0, 7, 6, 8, 3, 8, 7, 9, 7, 7, 4, 9, 1, 6, 9, 5, 4, 3, 7, 7,
0])

```

```
In [18]: pred_test_xgboost
```

```

Out[18]: array([[7, 9, 9, 7, 2, 4, 9, 5, 2, 9, 2, 8, 3, 3, 3, 8, 2, 4, 2, 9, 4, 3,
8, 1, 5, 7, 9, 8, 7, 9, 9, 7, 1, 5, 8, 8, 1, 2, 3, 3, 9, 9, 3, 2,
9, 9, 3, 4, 8, 3, 2, 9, 3, 3, 9, 2, 9, 9, 9, 7, 8, 8, 9, 3, 8, 4,
8, 1, 9, 6, 2, 4, 5, 4, 5, 7, 0, 1, 3, 7, 3, 4, 8, 2, 2, 3, 9, 5,
9, 6, 2, 5, 9, 4, 9, 3, 2, 2, 9, 4, 6, 9, 7, 9, 9, 7, 9, 6, 8, 2,
9, 7, 5, 9, 9, 9, 4, 2, 9, 8, 8, 5, 8, 9, 8, 1, 1, 0, 4, 9, 3, 4,
2, 9, 8, 9, 0, 8, 4, 2, 6, 3, 3, 5, 9, 8, 8, 8, 4, 9, 9, 3, 8, 2,
4, 8, 3, 6, 2, 3, 9, 2, 2, 3, 1, 1, 2, 2, 9, 0, 2, 2, 2, 4, 9, 3,
2, 2, 2, 8, 8, 7, 9, 8, 4, 2, 1, 3, 3, 9, 9, 2, 9, 2, 1, 9, 2, 9,
8, 0, 2, 2, 9, 3, 9, 9, 1, 6, 9, 2, 9, 2, 4, 9, 9, 5, 4, 2, 2, 3,
3, 2, 2, 1, 9, 3, 2, 8, 8, 3, 7, 2, 8, 2, 9, 3, 3, 4, 7, 9, 3, 6,
1, 9, 4, 8, 5, 2, 0, 9, 8, 9, 9, 9, 2, 6, 7, 2, 3, 3, 3, 1, 1, 3,
8, 2, 8, 7, 5, 3, 9, 4, 7, 8, 4, 2, 3, 2, 3, 4, 8, 1, 8, 7, 9, 9,
9, 9, 4, 8, 3, 3, 4, 2, 1, 3, 0, 9, 2, 3, 4, 2, 6, 2, 8, 2, 9, 4,

```

```

7, 2, 5, 9, 2, 8, 9, 9, 9, 1, 1, 3, 4, 3, 9, 9, 4, 9, 4, 3, 9, 1,
6, 1, 8, 8, 4, 2, 4, 9, 2, 9, 9, 8, 9, 8, 7, 2, 3, 4, 2, 7, 2, 3,
9, 9, 9, 9, 9, 3, 8, 8, 2, 2, 2, 3, 5, 9, 3, 8, 2, 7, 9, 9, 2, 2,
2, 5, 6, 9, 3, 7, 2, 3, 5, 9, 4, 2, 4, 2, 3, 6, 1, 7, 2, 1, 3, 2,
9, 7, 9, 4, 9, 3, 9, 0, 8, 3, 9, 2, 9, 8, 9, 0, 6, 5, 0, 5, 0, 2,
2, 3, 7, 4, 9, 9, 4, 8, 3, 3, 2, 2, 8, 7, 9, 8, 0, 6, 3, 3, 3, 5,
7, 2, 3, 9, 8, 3, 6, 8, 1, 8, 8, 9, 2, 2, 5, 7, 1, 8, 9, 2, 9, 9,
2, 8, 5, 9, 8, 8, 3, 8, 7, 8, 9, 3, 3, 7, 0, 3, 2, 0, 4, 9, 8, 3,
2, 8, 3, 4, 9, 9, 8, 2, 4, 9, 3, 4, 2, 9, 4, 3, 3, 2, 1, 2, 9, 3,
3, 7, 9, 9, 8, 3, 7, 2, 6, 3, 1, 8, 5, 4, 7, 9, 6, 5, 8, 3, 7, 3,
3, 2, 2, 1, 2, 9, 3, 7, 9, 1, 5, 1, 2, 9, 2, 3, 4, 9, 7, 3, 3, 5,
2, 4, 8, 6, 2, 2, 8, 3, 3, 4, 9, 3, 9, 3, 2, 2, 5, 5, 2, 9, 3, 5,
3, 8, 2, 8, 2, 4, 2, 1, 8, 8, 1, 9, 7, 5, 9, 3, 3, 9, 2, 3, 2, 5,
2, 8, 9, 3, 1, 1, 5, 0, 8, 9, 2, 2, 8, 9, 9, 9, 9, 4, 8, 9, 3, 9,
2, 1, 2, 9, 3, 7, 9, 3, 8, 3, 3, 3, 8, 9, 2, 2, 9, 9, 9, 2, 8, 4,
1, 7, 5, 2, 9, 4, 3, 2, 3, 9, 5, 9, 2, 7, 9, 5, 9, 2, 9, 3, 9, 2,
9, 9, 7, 9, 9, 9, 2, 7, 2, 9, 5, 9, 9, 2, 9, 3, 9, 2, 5, 2, 5, 9,
6, 4, 9, 7, 9, 5, 2, 3, 2, 1, 3, 9, 3, 2, 8, 8, 3, 2, 6, 4, 8, 9,
9, 0, 3, 3, 7, 9, 2, 3, 8, 8, 1, 9, 3, 2, 8, 7, 5, 7, 4, 9, 3, 8,
2, 9, 2, 9, 1, 9, 9, 8, 9, 4, 2, 8, 3, 8, 2, 3, 8, 9, 4, 2, 2, 3,
3, 9, 9, 8, 3, 2, 5, 3, 9, 7, 3, 9, 9, 3, 9, 1, 3, 6, 2, 2, 3, 0,
6, 2, 2, 9, 4, 2, 9, 8, 9, 9, 9, 9, 7, 2, 2, 3, 2, 3, 7, 9, 8, 2,
1, 9, 2, 3, 3, 3, 2, 9, 3, 9, 2, 3, 9, 2, 8, 9, 9, 9, 1, 7, 0, 9,
3, 3, 9, 2, 3, 8, 7, 9, 2, 9, 9, 5, 8, 8, 9, 2, 9, 2, 9, 7, 4, 9,
9, 2, 7, 9, 7, 2, 8, 9, 8, 4, 8, 0, 9, 3, 9, 7, 3, 8, 3, 5, 8, 3,
0, 9, 2, 9, 2, 3, 2, 2, 9, 7, 7, 2, 4, 9, 2, 6, 7, 9, 3, 4, 2, 9,
2, 9, 9, 1, 8, 3, 8, 2, 2, 9, 3, 9, 2, 7, 2, 3, 8, 2, 9, 9, 9, 9,
7, 4, 5, 9, 6, 8, 8, 5, 9, 2, 9, 5, 4, 9, 1, 6, 9, 5, 3, 3, 9, 4,
5])

```

```
In [21]: met.accuracy_score(y_test, pred_test_xgboost)
```

```
Out[21]: 0.4562162162162162
```

```
In [20]: met.confusion_matrix(y_test, pred_test_xgboost)
```

```
Out[20]: array([[ 1,  0, 24, 27,  0, 16,  0, 10,  1, 21],
 [ 0, 41,  0,  0,  0,  2,  0,  0,  0,  0],
 [ 0,  0, 63, 16,  1,  0,  2,  0,  1, 17],
 [ 0,  0, 19, 62,  0,  1,  2,  0,  5, 11],
 [ 1,  0,  2, 29, 44,  1,  0,  0,  2, 21],
 [ 5,  0, 21,  2,  6, 14,  0, 29,  0, 30],
 [ 0,  0,  3,  7,  1,  0, 22,  0,  0,  3],
 [ 9,  0,  8,  0,  8,  8,  0, 21,  2, 64],
 [ 2,  0,  7,  1,  0,  2,  0,  1, 99,  7],
 [ 2,  4, 26,  3,  6,  4,  0,  0,  0, 55]])

```

SVM Model

```
In [11]: import sklearn
from sklearn.svm import SVC
from sklearn.model_selection import GridSearchCV
```

```
In [36]: import warnings
warnings.filterwarnings('ignore')
```

```
In [ ]: svr = SVC()
parameters = {'kernel':('linear', 'rbf'), 'C':[1, 2, 4], 'gamma':[0.125,
```

```

0.25, 0.5 ,1, 2, 4]]
clf = GridSearchCV(svr, parameters, scoring='f1_samples')
clf.fit(x_train, y_train)
print('The parameters of the best model are: ')
print(clf.best_params_)

```

```

In [12]: model_svm = sklearn.svm.SVC(C=0.75,
                                     kernel='rbf',
                                     degree=3,
                                     gamma='auto',
                                     coef0=0.0,
                                     shrinking=True,
                                     probability=False,
                                     tol=0.001,
                                     cache_size=200,
                                     class_weight=None,
                                     verbose=False,
                                     max_iter=-1,
                                     decision_function_shape=None,
                                     random_state=None)

```

```

In [13]: model_svm.fit(x_train,y_train)

```

```

Out[13]: SVC(C=0.75, cache_size=200, class_weight=None, coef0=0.0,
             decision_function_shape=None, degree=3, gamma='auto', kernel='rbf',
             max_iter=-1, probability=False, random_state=None, shrinking=True,
             tol=0.001, verbose=False)

```

```

In [14]: pred_train_svm = model_svm.predict(x_train)
met.accuracy_score(y_train,pred_train_svm)

```

```

Out[14]: 0.9984629178941975

```

```

In [15]: pred_test_svm = model_svm.predict(x_test)
met.accuracy_score(y_test,pred_test_svm)

```

```

Out[15]: 0.10918918918918918

```

RandomForest Model

```

In [16]: from sklearn.ensemble import RandomForestClassifier

```

```

In [17]: model_RF = RandomForestClassifier(n_estimators=100 ,criterion = "entropy
", bootstrap= False)

```

```

In [18]: model_RF.fit(x_train,y_train)

```

```

Out[18]: RandomForestClassifier(bootstrap=False, class_weight=None, criterion='en
tropy',
                                max_depth=None, max_features='auto', max_leaf_nod
es=None,
                                min_impurity_decrease=0.0, min_impurity_split=Non
e,
                                min_samples_leaf=1, min_samples_split=2,
                                min_weight_fraction_leaf=0.0, n_estimators=100,
                                n_jobs=None, oob_score=False, random_state=None,
                                verbose=0, warm_start=False)

```

```
In [19]: pred_train_RF = model_RF.predict(x_train)
         met.accuracy_score(y_train,pred_train_RF)
```

```
Out[19]: 1.0
```

```
In [20]: pred_test_RF = model_RF.predict(x_test)
         met.accuracy_score(y_test,pred_test_RF)
```

```
Out[20]: 0.4648648648648649
```

KNN Model

```
In [21]: from sklearn.neighbors import KNeighborsClassifier
```

```
In [22]: model_KNN = KNeighborsClassifier(n_neighbors=10,
                                         weights='uniform',
                                         algorithm='auto',
                                         leaf_size=30,
                                         p=2,
                                         metric='minkowski',
                                         metric_params=None,
                                         n_jobs=None)
```

```
In [23]: model_KNN.fit(x_train,y_train)
```

```
Out[23]: KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
                             metric_params=None, n_jobs=None, n_neighbors=10, p=
                             2,
                             weights='uniform')
```

```
In [24]: pred_train_KNN = model_KNN.predict(x_train)
         met.accuracy_score(y_train,pred_train_KNN)
```

```
Out[24]: 0.8523120276674779
```

```
In [25]: pred_test_KNN = model_KNN.predict(x_test)
         met.accuracy_score(y_test,pred_test_KNN)
```

```
Out[25]: 0.3848648648648649
```

KNN Model

```
In [ ]:
```