```
1
 1 from google.colab import drive
 2 drive. mount ('/content/drive')
С→
 1 cd /content/drive/My Drive/soundClasstion
\Box
 1 !1s
Гэ
 1 import os
 2 import keras
 3 import librosa
 4 import numpy as np
 5 import matplotlib.pyplot as plt
 6 from keras import Sequential
 7 from keras.utils import to_categorical
 8 from keras. layers import Dense
 9 from sklearn.model_selection import train_test_split
10
11 DATA = 'data.npy'
12 TARGET = 'target.npy'
13
14
15 # 加载标签
16 def load_label(label_path):
17
     label = os.listdir(label path)
18
     return label
19
20
21 # 提取 mfcc 参数
   def wav2mfcc(path, max pad size=11):
     y, sr = librosa.load(path=path, sr=None, mono=1)
23
24
     y = y[::3] #每三个点选用一个
25
     audio mac = librosa. feature. mfcc (y=y, sr=16000)
26
     y shape = audio mac.shape[1]
27
     if y shape < max pad size:
28
          pad_size = max_pad_size - y_shape
          audio_mac = np.pad(audio_mac, ((0, 0), (0, pad size)), mode='constant')
29
30
     else:
31
          audio_mac = audio_mac[:, :max_pad_size]
32
     return audio_mac
33
34
35 # 存储处理过的数据,方便下一次的使用
36 | def save_data_to_array(label_path, max_pad_size=11):
37
     mfcc vectors = []
38
     target = []
     labels = load_label(label_path=label_path)
39
     for i, label in enumerate(labels):
    path = label_path + '/' + label
    wavfiles = [path + '/' + file for file in os.listdir(path)]
40
41
42
```

```
43
         for wavfile in wavfiles:
44
             wav = wav2mfcc(wavfile, max pad size=max pad size)
45
             mfcc vectors.append(wav)
46
             target.append(i)
47
     np. save (DATA, mfcc vectors)
48
     np. save (TARGET, target)
49
     # return mfcc vectors, target
50
51
52 # 获取训练集与测试集
53 def get train test(split ratio=.6, random state=42):
     X = np. 1oad(DATA)
54
     y = np. 1oad (TARGET)
55
56
     assert X. shape[0] == y. shape[0]
     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=(1 - split|ratio), random
57
58
                                                            shuffle=True)
59
     return X_train, X_test, y_train, y_test
60
61
62 def main():
63
64
     x_train, x_test, y_train, y_test = get_train_test()
     x_{train} = x_{train}. reshape (-1, 220)
65
66
     x_{test} = x_{test}. reshape (-1, 220)
     y_train_hot = to_categorical(y train)
67
     y_test_hot = to_categorical(y_test)
68
69
     model = Sequential()
     model.add(Dense(64, activation='relu', input_shape=(220,)))
70
71
     model.add(Dense(64, activation='relu'))
     model.add(Dense(64, activation='relu'))
72
73
     model.add(Dense(2, activation='softmax'))
74
75
     model.compile(loss=keras.losses.categorical crossentropy,
                    optimizer=keras.optimizers.RMSprop(),
76
                    metrics=['accuracy'])
77
78
     history = model.fit(x train, y train hot, batch size=100, epochs=100, verbose=1,
79
                          validation data=(x test, y test hot))
80
     plot history (history)
81
     model. save ("classaud. h5")
82
83
84
85
86 def plot history (history):
     plt. plot (history. history['acc'], label='train')
87
     plt.plot(history.history['val_acc'], label='validation')
88
89
     plt.legend()
90
     plt.show()
91
92
93 if __name__ == "__main__":
     #save_data_to_array("./data/", max_pad_size=11)
95
     main()
96
```

Гэ

```
625/625 [=============] - 0s 77us/step - loss: 0.0258 - acc: 0.9984 - val los
Epoch 81/100
625/625 [===
                            =======] - Os 71us/step - loss: 0.0258 - acc: 0.9984 - val los
Epoch 82/100
                            =======] - Os 67us/step - loss: 0.0258 - acc: 0.9984 - val_los
625/625 [====
Epoch 83/100
                                ======] - Os 72us/step - loss: 0.0258 - acc: 0.9984 - val los
625/625 [===
Epoch 84/100
                                ======] - Os 69us/step - loss: 0.0258 - acc: 0.9984 - val_los
625/625 [===
Epoch 85/100
625/625 [===
                                ======] - Os 66us/step - loss: 0.0258 - acc: 0.9984 - val los
Epoch 86/100
625/625 [===
                            ========] - Os 78us/step - loss: 0.0258 - acc: 0.9984 - val_los
Epoch 87/100
625/625 [===
                               =======] - Os 65us/step - loss: 0.0258 - acc: 0.9984 - val_los
Epoch 88/100
625/625 [===
                                  ====] - 0s 67us/step - loss: 0.0258 - acc: 0.9984 - val_los
Epoch 89/100
625/625 [===
                                     ==] - Os 68us/step - loss: 0.2672 - acc: 0.9696 - val los
Epoch 90/100
625/625 [====
                                ======] - Os 66us/step - loss: 0.0258 - acc: 0.9984 - val_los
Epoch 91/100
                               ======] - Os 71us/step - loss: 0.0258 - acc: 0.9984 - val los
625/625 [===
Epoch 92/100
625/625 [===
                                  =====] - Os 68us/step - loss: 0.0258 - acc: 0.9984 - val_los
Epoch 93/100
625/625 [===
                                  ====] - Os 69us/step - loss: 0.0258 - acc: 0.9984 - val los
Epoch 94/100
                                     ==] - Os 68us/step - loss: 0.0258 - acc: 0.9984 - val los
625/625 [===
Epoch 95/100
625/625 [====
                            =======] - Os 64us/step - loss: 0.0258 - acc: 0.9984 - val los
Epoch 96/100
                                ======] - Os 67us/step - loss: 0.0258 - acc: 0.9984 - val los
625/625 [===
Epoch 97/100
                               ======] - Os 73us/step - loss: 0.0258 - acc: 0.9984 - val_los
625/625 [====
Epoch 98/100
                            =======] - Os 69us/step - loss: 0.0258 - acc: 0.9984 - val_los
625/625 [====
Epoch 99/100
                         ========] - Os 73us/step - loss: 0.0258 - acc: 0.9984 - val los
625/625 [=====
Epoch 100/100
                                ======] - Os 77us/step - loss: 0.0258 - acc: 0.9984 - val los
625/625 [===
 1.00
 0.95
 0.90
 0.85
 0.80
 0.75
                                             train
 0.70
                                             validation
               20
                        40
                                 60
                                          80
                                                  100
```

```
1 import wave
 2 import numpy as np
 3
 4
   from keras. models import load model
 5
 6
 7
 8
   def get_wav_mfcc(wav_path):
        f = wave. open (wav_path, 'rb')
9
       params = f. getparams()

# print("params:", params)

nchannels, sampwidth, framerate, nframes = params[:4]

strData = f. readframes(nframes) #读取音频, 字符串格式
10
11
12
13
        waveData = np. fromstring(strData, dtype=np. int16)#将字符串转化为int
14
        waveData = waveData*1.0/(max(abs(waveData)))#wave幅值归一化
15
16
        waveData = np. reshape(waveData, [nframes, nchannels]). T
17
        f.close()
18
        ### 对音频数据进行长度大小的切割,保证每一个的长度都是一样的【因为训练文件全部是1秒钟长度,
19
        data = list(np.array(waveData[0]))
20
```