CNN_4

April 26, 2017

1 CS109B Project Group 26 - Deep Learning

Main Specifications - Ver3
 ** Data preprocessing:**

- Channel rearrangement
- 14 labels multi label classification
- Data augmentation by shift and zoom for 8000 training sample and 2000 test sample
- Centered features
- ** Main Architecture **
- Multi layer CNN:
- Conv2D Relu depth 32 and 7x7 kernel He uniform initialization MaxPool 2x2 Conv2D Relu depth 32 and 5x5 kernel He uniform initialization MaxPool 2x2 Conv2D Relu depth 64 and 3x3 kernel He uniform initialization MaxPool 2x2 Conv2D Relu depth 64 and 3x3 kernel He uniform initialization FC 128, Relu, He uniform initialization FC 64, Relu, He uniform initialization
- Adam optimzer with 1e-6 learning rate Binary cross entropy loss function Batch size 64 Training convergence after 30 epochs

1.0.1 Import Modules

```
In [1]: import requests
    import json
    import time
    import itertools
    import wget
    import os
    import pickle
    import numpy as np

import random
    import matplotlib
    import matplotlib.pyplot as plt
    %matplotlib inline
```

```
import seaborn as sns
        from sklearn.cluster.bicluster import SpectralCoclustering
        from sklearn.metrics import precision_recall_curve
        import scipy
        sns.set style('white')
        import tensorflow as tf
        import pandas as pd
        import keras
        from keras.optimizers import SGD, Adam
        from keras.models import Sequential
        from keras.layers import Dense, Dropout, Flatten
        from keras.layers import Conv2D, MaxPooling2D
        from keras.layers.normalization import BatchNormalization
        import keras.initializers as init
        from keras.preprocessing.image import ImageDataGenerator
        from keras import backend as K
        from keras.models import load_model
Using TensorFlow backend.
In [2]: x_train_dict = pickle.load(open('training_num.pik' , 'rb'))
In [3]: train split = 8000
In [4]: x_train_raw = x_train_dict['images']
        x_train = np.array(x_train_raw)[:train_split,: ,: , :]
        x_test = np.array(x_train_raw)[train_split:,: ,: ,:]
        print x_train.shape
(8000, 128, 85, 3)
In [5]: img_rows = x_train.shape[1]
        img_cols = x_train.shape[2]
        if K.image_data_format() == 'channels_first':
            x train = x train.reshape(x train.shape[0], 3, img rows, img cols)
            input_shape = (3, img_rows, img_cols)
        else:
            x_train = x_train.reshape(x_train.shape[0], img_rows, img_cols, 3)
            input_shape = (img_rows, img_cols, 3)
In [6]: x_train = x_train.astype('float32')
        x_test = x_test.astype('float32')
```

```
x_train /= 255.0
       x_{test} /= 255.0
       print 'x_train shape:', x_train.shape
       print x_train.shape[0], 'train samples'
       print 'x_test shape:', x_test.shape
       print x_test.shape[0], 'test samples'
x_train shape: (8000, 128, 85, 3)
8000 train samples
x_test shape: (1988, 128, 85, 3)
1988 test samples
In [7]: y_raw = pd.read_csv('Genres_labels_All_cleaned.csv')
       y_train = y_raw.iloc[:, 1:-1].values[:train_split, :]
       y_test = y_raw.iloc[:, 1:-1].values[train_split:, :]
       num_classes = y_train.shape[1]
In [8]: datagen = ImageDataGenerator(
           featurewise_center=True,
           featurewise std normalization=True,
           width_shift_range=0.2,
           height_shift_range=0.2,
            zoom_range = 0.5,
           fill_mode = 'wrap')
        datagen.fit(x_train)
        datagen.fit(x_test)
In [10]: # create an empty network model
         model3 = Sequential()
         # --- input layer ---
         model3.add(Conv2D(32, kernel_size=(7, 7), activation='relu', input_shape=input_shape
                          kernel_initializer = init.he_normal(109)))
         # --- max pool ---
         model3.add(MaxPooling2D(pool_size=(2, 2)))
         # ---- Conv Layer ---
         model3.add(Conv2D(32, kernel_size=(5, 5), activation='relu',
                         kernel_initializer = init.he_normal(109)))
```

```
model3.add(MaxPooling2D(pool_size=(2, 2)))
       # --- Conv layer ---
       model3.add(Conv2D(64, kernel_size=(3, 3), activation='relu',
                   kernel_initializer = init.he_normal(109)))
       # --- max pool ---
       model3.add(MaxPooling2D(pool_size=(2, 2)))
       # --- Conv layer ---
       model3.add(Conv2D(64, kernel_size=(3, 3), activation='relu' ,
                   kernel_initializer = init.he_normal(109)))
       # --- max pool ---
       model3.add(MaxPooling2D(pool_size=(2, 2)))
       # flatten for fully connected classification layer
       model3.add(Flatten())
       # --- fully connected layer ---
       model3.add(Dense(128, activation='relu',
                  kernel_initializer = init.he_normal(109)))
       # --- fully connected layer ---
       model3.add(Dense(64, activation='relu' ,
                   kernel_initializer = init.he_normal(109)))
       # --- classification ---
       model3.add(Dense(num_classes, activation='sigmoid'))
       # prints out a summary of the model architecture
       model3.summary()
______
Layer (type) Output Shape Param #
______
                       (None, 122, 79, 32) 4736
conv2d 5 (Conv2D)
max_pooling2d_5 (MaxPooling2 (None, 61, 39, 32) 0
_____
              (None, 57, 35, 32) 25632
conv2d_6 (Conv2D)
max_pooling2d_6 (MaxPooling2 (None, 28, 17, 32) 0
conv2d_7 (Conv2D) (None, 26, 15, 64) 18496
```

--- max pool ---