## CNN\_BN

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## 1 CS109B Project Group 26 - Deep Learning

## Main Specifications - Ver4
\*\* Data preprocessing:\*\*

- Channel rearrangement
- 14 labels multi label classification
- data augmentation by shift and zoom for 8000 training samples and 2000 test samples
- Centered features
- \*\* Main Architecture \*\*
- Multi layer CNN with Batch Normalization Layers :
- Conv2D Relu depth 32 and 7x7 kernel He uniform initialization Batch Norm MaxPool 2x2 Conv2D Relu depth 32 and 5x5 kernel He uniform initialization Batch Norm MaxPool 2x2 Conv2D Relu depth 64 and 3x3 kernel He uniform initialization Batch Norm MaxPool 2x2 Conv2D Relu depth 64 and 3x3 kernel He uniform initialization Batch Norm FC 128 , Relu , He uniform initialization Batch Norm FC 64 , Relu , He uniform initialization Batch Norm
- SGD optimzer with 1e-4 learning rate and 0.99 momentum Binary cross entropy loss function Batch size 128 Training convergence after 50 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 \text{ 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 [59]: 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]
         print 'number of classes: ' , num_classes
number of classes:
                     14
In [60]: 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 [74]: # create an empty network model
         modela = Sequential()
         # --- input layer ---
         modela.add(Conv2D(32, kernel_size=(7, 7), activation='relu', input_shape=input_shape
                          kernel_initializer = init.he_normal(109)))
```

```
# ----Batch Normalization ----
modela.add(BatchNormalization())
# --- max pool ---
modela.add(MaxPooling2D(pool size=(2, 2)))
# ---- Conv Layer ---
modela.add(Conv2D(32, kernel_size=(5, 5), activation='relu',
                kernel_initializer = init.he_normal(109)))
# ----Batch Normalization ----
modela.add(BatchNormalization())
# --- max pool ---
modela.add(MaxPooling2D(pool_size=(2, 2)))
# --- Conv layer ---
modela.add(Conv2D(64, kernel_size=(3, 3), activation='relu',
               kernel_initializer = init.he_normal(109)))
# ----Batch Normalization ----
modela.add(BatchNormalization())
# --- max pool ---
modela.add(MaxPooling2D(pool_size=(2, 2)))
# --- Conv layer ---
modela.add(Conv2D(64, kernel_size=(3, 3), activation='relu' ,
               kernel initializer = init.he normal(109)))
# ----Batch Normalization ----
modela.add(BatchNormalization())
# --- max pool ---
modela.add(MaxPooling2D(pool_size=(2, 2)))
# flatten for fully connected classification layer
modela.add(Flatten())
# --- fully connected layer ---
```

Layer (type)	Output Shape	Param #
conv2d_48 (Conv2D)	(None, 122, 79, 32)	4736
batch_normalization_63 (Batc	(None, 122, 79, 32)	128
max_pooling2d_48 (MaxPooling	(None, 61, 39, 32)	0
conv2d_49 (Conv2D)	(None, 57, 35, 32)	25632
batch_normalization_64 (Batc	(None, 57, 35, 32)	128
max_pooling2d_49 (MaxPooling	(None, 28, 17, 32)	0
conv2d_50 (Conv2D)	(None, 26, 15, 64)	18496
batch_normalization_65 (Batc	(None, 26, 15, 64)	256
max_pooling2d_50 (MaxPooling	(None, 13, 7, 64)	0
conv2d_51 (Conv2D)	(None, 11, 5, 64)	36928
batch_normalization_66 (Batc	(None, 11, 5, 64)	256