FCN5

June 24, 2020

0.0.1 -- coding: utf-8 --

MusicTaggerCNN model for Keras.

0.0.2 Reference:

- Automatic tagging using deep convolutional neural networks
- Music-auto_tagging-keras

```
[1]: from __future__ import print_function
from __future__ import absolute_import

import keras
from keras import backend as K
from keras.layers import Input, Dense
from keras.models import Model
from keras.layers import Dense, Dropout, Flatten
from keras.layers.convolutional import Convolution2D
from keras.layers.convolutional import MaxPooling2D, ZeroPadding2D
from keras.layers.normalization import BatchNormalization
from keras.layers.advanced_activations import ELU
from keras.utils.data_utils import get_file
from keras.layers import Input, Dense
```

Using Theano backend.

WARNING (theano.tensor.blas): Using NumPy C-API based implementation for BLAS functions.

```
[2]: TH_WEIGHTS_PATH = 'https://github.com/keunwoochoi/music-auto_tagging-keras/blob/

→master/data/music_tagger_cnn_weights_theano.h5'

TF_WEIGHTS_PATH = 'https://github.com/keunwoochoi/music-auto_tagging-keras/blob/

→master/data/music_tagger_cnn_weights_tensorflow.h5'
```

1 MusicTaggerCNN

Instantiate the MusicTaggerCNN architecture, optionally loading weights pre-trained on Million Song Dataset. Note that when using TensorFlow, for best performance you should set image_dim_ordering="tf" in your Keras config at ~/.keras/keras.json.

The model and the weights are compatible with both TensorFlow and Theano. The dimension ordering convention used by the model is the one specified in your Keras config file.

For preparing mel-spectrogram input, see audio_conv_utils.py

in applications. You will need to install Librosa to use it.

Arguments

weights: one of `None` (random initialization) or "msd" (pre-training on ImageNet).

input_tensor: optional Keras tensor (i.e. output of `layers.Input()`) to use as image input for include_top: whether to include the 1 fully-connected layer (output layer) at the top of the name of t

Returns

A Keras model instance.

```
[3]: def MusicTaggerCNN(weights='msd', input_tensor=None,
                                                                    include_top=True):
                           '''Instantiate the MusicTaggerCNN architecture,
                          optionally loading weights pre-trained
                          on Million Song Dataset. Note that when using TensorFlow,
                         for best performance you should set
                          `image_dim_ordering="tf"` in your Keras config
                         at ~/.keras/keras.json.
                         The model and the weights are compatible with both
                         TensorFlow and Theano. The dimension ordering
                         convention used by the model is the one
                         specified in your Keras config file.
                         For preparing mel-spectrogram input, see
                          `audio\_conv\_utils.py` in \ [applications] (https://github.com/fchollet/keras/line) and in the convergence of the convergence 
                 → tree/master/keras/applications).
                          You will need to install [Librosa] (http://librosa.github.io/librosa/)
                          to use it.
                         # Arguments
                                     weights: one of `None` (random initialization)
                                                 or "msd" (pre-training on ImageNet).
                                     input_tensor: optional Keras tensor (i.e. output of `layers.Input()`)
                                                 to use as image input for the model.
                                     include_top: whether to include the 1 fully-connected
                                                 layer (output layer) at the top of the network.
                                                If False, the network outputs 256-dim features.
```

```
# Returns
    A Keras model instance.
if weights not in {'msd', None}:
    raise ValueError('The `weights` argument should be either '
                     '`None` (random initialization) or `msd` '
                     '(pre-training on Million Song Dataset).')
# Determine proper input shape
if keras.backend.image_data_format() == 'channels_first':
    input_shape = (1, 96, 1366)
else:
    input_shape = (96, 1366, 1)
if input_tensor is None:
    melgram_input = Input(shape=input_shape)
else:
    if not K.is_keras_tensor(input_tensor):
        melgram_input = Input(tensor=input_tensor, shape=input_shape)
    else:
        melgram_input = input_tensor
# Determine input axis
if keras.backend.image data format() == 'channels first':
    channel_axis = 1
    freq axis = 2
    time_axis = 3
else:
    channel_axis = 3
    freq_axis = 1
    time_axis = 2
# Input block
x = BatchNormalization(axis=freq_axis, name='bn_0_freq')(melgram_input)
# Conv block 1
x = Convolution2D(64, 3, 3, border_mode='same', name='conv1')(x)
x = BatchNormalization(axis=channel_axis, mode=0, name='bn1')(x)
x = ELU()(x)
x = MaxPooling2D(pool_size=(2, 4), name='pool1')(x)
# Conv block 2
x = Convolution2D(128, 3, 3, border_mode='same', name='conv2')(x)
x = BatchNormalization(axis=channel_axis, mode=0, name='bn2')(x)
x = ELU()(x)
x = MaxPooling2D(pool_size=(2, 4), name='pool2')(x)
```

```
# Conv block 3
  x = Convolution2D(128, 3, 3, border_mode='same', name='conv3')(x)
  x = BatchNormalization(axis=channel_axis, mode=0, name='bn3')(x)
  x = MaxPooling2D(pool_size=(2, 4), name='pool3')(x)
  # Conv block 4
  x = Convolution2D(128, 3, 3, border_mode='same', name='conv4')(x)
  x = BatchNormalization(axis=channel_axis, mode=0, name='bn4')(x)
  x = ELU()(x)
  x = MaxPooling2D(pool_size=(3, 5), name='pool4')(x)
  # Conv block 5
  x = Convolution2D(64, 3, 3, border_mode='same', name='conv5')(x)
  x = BatchNormalization(axis=channel_axis, mode=0, name='bn5')(x)
  x = ELU()(x)
  x = MaxPooling2D(pool_size=(4, 4), name='pool5')(x)
  # Output
  x = Flatten()(x)
  if include_top:
       x = Dense(10, activation='sigmoid', name='output')(x)
   # Create model
  model = Model(melgram_input, x)
  if weights is None:
      return model
  else:
       # Load input
      if keras.backend.image_data_format() == 'channels_last':
          raise RuntimeError("Please set keras.backend.image_data_format() ==_
"You can set it at ~/.keras/keras.json")
      model.load_weights('data/music_tagger_cnn_weights_%s.h5' % K._BACKEND,
                         by_name=True)
      return model
```

```
[4]: from IPython.display import Image
Image(filename='tf_th_keras_v2.png')
```

[4]:

Keras is ignoring the image_dim_ordering setting, because with **Keras v2** (v2.0.4 being the latest at the time of this writing) the name of the parameter has been changed.

Now you set it using the image_data_format parameter.

image_data_format can be set to "channels_last" or "channels_first", which corresponds to the TensorFlow or Theano dimension orders respectively.

i.e. When using TensorFlow, you now set your keras.json like this,

```
{
   "image_data_format": "channels_last",
   "epsilon": 1e-07,
   "floatx": "float32",
   "backend": "tensorflow"
}
```

For Theano, you need to set it like this,

```
{
   "image_data_format": "channels_first",
   "epsilon": le-07,
   "floatx": "float32",
   "backend": "theano"
}
```

```
if keras.backend.image_data_format() == 'channels_last':
    print("here backend is tensorflow" , keras.backend.image_data_format())
elif keras.backend.image_data_format() == 'channels_first':
    print("here backend is theano" , keras.backend.image_data_format())

#print(K.image_dim_ordering())
```

here backend is theano channels_first

TRY TO MAKE A MODEL USER-FRIENDLY IN NEAR FUTURE.

```
[6]: model = MusicTaggerCNN(weights=None)
```

```
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:66:
UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(64, (3, 3), name="conv1", padding="same")`
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:67:
UserWarning: Update your `BatchNormalization` call to the Keras 2 API:
`BatchNormalization(axis=1, name="bn1")`
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:72:
UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(128, (3, 3),
```

```
name="conv2", padding="same")`
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:73:
UserWarning: Update your `BatchNormalization` call to the Keras 2 API:
`BatchNormalization(axis=1, name="bn2")`
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel launcher.py:78:
UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(128, (3, 3),
name="conv3", padding="same")`
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:79:
UserWarning: Update your `BatchNormalization` call to the Keras 2 API:
`BatchNormalization(axis=1, name="bn3")`
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:84:
UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(128, (3, 3),
name="conv4", padding="same")`
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:85:
UserWarning: Update your `BatchNormalization` call to the Keras 2 API:
`BatchNormalization(axis=1, name="bn4")`
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:90:
UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(64, (3, 3),
name="conv5", padding="same")`
/home/user/anaconda3/lib/python3.7/site-packages/ipykernel launcher.py:91:
UserWarning: Update your `BatchNormalization` call to the Keras 2 API:
`BatchNormalization(axis=1, name="bn5")`
```

[7]: model.summary()

Layer (type)	Output	Shape	Param #
input_1 (InputLayer)	(None,	1, 96, 1366)	0
bn_0_freq (BatchNormalizatio	(None,	1, 96, 1366)	384
conv1 (Conv2D)	(None,	64, 96, 1366)	640
bn1 (BatchNormalization)	(None,	64, 96, 1366)	256
elu_1 (ELU)	(None,	64, 96, 1366)	0
pool1 (MaxPooling2D)	(None,	64, 48, 341)	0
conv2 (Conv2D)	(None,	128, 48, 341)	73856
bn2 (BatchNormalization)	(None,	128, 48, 341)	512
elu_2 (ELU)	(None,	128, 48, 341)	0
pool2 (MaxPooling2D)	(None,	128, 24, 85)	0

conv3 (Conv2D)	(None, 128, 24, 85)	147584
bn3 (BatchNormalization)	(None, 128, 24, 85)	512
elu_3 (ELU)	(None, 128, 24, 85)	0
pool3 (MaxPooling2D)	(None, 128, 12, 21)	0
conv4 (Conv2D)	(None, 128, 12, 21)	147584
bn4 (BatchNormalization)	(None, 128, 12, 21)	512
elu_4 (ELU)	(None, 128, 12, 21)	0
pool4 (MaxPooling2D)	(None, 128, 4, 4)	0
conv5 (Conv2D)	(None, 64, 4, 4)	73792
bn5 (BatchNormalization)	(None, 64, 4, 4)	256
elu_5 (ELU)	(None, 64, 4, 4)	0
pool5 (MaxPooling2D)	(None, 64, 1, 1)	0
flatten_1 (Flatten)	(None, 64)	0
output (Dense)	(None, 10)	650
Total params: 446.538		

Total params: 446,538 Trainable params: 445,322 Non-trainable params: 1,216

1.0.1 Compiling the model

```
[2]: #compile model using accuracy to measure model performance
     model.compile(optimizer='adam',
                  loss='categorical_crossentropy',
                  metrics=['accuracy'])
```

ш

NameError →last)

Traceback (most recent call⊔

<ipython-input-2-5b0921aaafdd> in <module>

NameError: name 'model' is not defined

1.0.2 Load data

```
[9]: import numpy as np
[10]: concat_x = np.load('concat_x.npy')
    concat_y = np.load('concat_y.npy')
[11]: print(len(concat_x),' ',(len(concat_y)) )
    1000
           1000
[12]: train_x = concat_x[0:800]
    train_y = concat_y[0:800]
[13]: test_x = concat_x[800:1000]
    test_y = concat_y[800:1000]
    valid_x = concat_x[950:1000] valid_y = concat_y[950:1000]
    1.0.3 Training the model
    #train the model model.fit(X_train, y_train, validation_data=(X_test, y_test),
    epochs=1000)
[14]: model.fit(train_x, train_y, validation_data=(test_x, test_y), epochs=12)
    Train on 800 samples, validate on 200 samples
    Epoch 1/12
    0.1725 - val_loss: 4.6778 - val_acc: 0.0850
    Epoch 2/12
    0.4138 - val_loss: 3.3504 - val_acc: 0.1100
    Epoch 3/12
    800/800 [=============== ] - 1193s 1s/step - loss: 1.3535 - acc:
    0.5587 - val_loss: 2.7192 - val_acc: 0.1100
    Epoch 4/12
    0.6375 - val_loss: 2.4091 - val_acc: 0.1750
    Epoch 5/12
```

```
0.6763 - val_loss: 2.3538 - val_acc: 0.2950
   Epoch 6/12
   0.7375 - val_loss: 2.4140 - val_acc: 0.2550
   Epoch 7/12
   0.7725 - val_loss: 2.3458 - val_acc: 0.2600
   Epoch 8/12
   0.8100 - val_loss: 2.1673 - val_acc: 0.2900
   800/800 [============ - 1137s 1s/step - loss: 0.5068 - acc:
   0.8475 - val_loss: 1.9019 - val_acc: 0.4050
   Epoch 10/12
   0.8762 - val_loss: 1.9972 - val_acc: 0.3900
   Epoch 11/12
   0.9113 - val_loss: 1.7700 - val_acc: 0.4350
   Epoch 12/12
   0.9413 - val_loss: 2.1803 - val_acc: 0.3350
[14]: <keras.callbacks.History at 0x7ff343327c90>
[15]: from keras.models import load_model
   model.save('fcn.h5') # creates a HDF5 file 'my_model.h5'
   del model # deletes the existing model
   # returns a compiled model
   # identical to the previous one
   model = load_model('fcn.h5')
[]:
```