

Task-1-keras-Xception

April 3, 2023

This notebook contains the Tensorflow Keras implementation of the Xception model. Refer this paper for Xception model specification: <https://arxiv.org/pdf/1610.02357.pdf>

It's worth noting this keras implementation is same as the pytorch implementation of Xception model, except for here we output only one scalar probability. In case of pytorch implementation we output two scalar probabilities (softmax). This small difference is due to unavailability of multi class AUC metric in tf/keras (<https://github.com/tensorflow/addons/issues/265>).

```
[1]: import tensorflow as tf
import tensorflow.keras

from tensorflow.keras import models, layers
from tensorflow.keras.models import Model, model_from_json, Sequential
from tensorflow.keras.layers import (
    Dense,
    Activation,
    Conv2D,
    Lambda,
    Resizing,
    MaxPooling2D,
    SeparableConv2D,
    BatchNormalization,
    Input,
    GlobalAveragePooling2D
)
from tensorflow.keras.optimizers import SGD
import os
import h5py
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
```

```
[2]: # extracted this code snippet from tensorflow documentation
# this will enable GPU memory growth
gpus = tf.config.list_physical_devices('GPU')

if gpus:
    try:
```

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for gpu in gpus:
    tf.config.experimental.set_memory_growth(gpu, True)
logical_gpus = tf.config.list_logical_devices('GPU')
print(len(gpus), "Physical GPUs,", len(logical_gpus), "Logical GPUs")
except RuntimeError as e:
    # Memory growth must be set before GPUs have been initialized
    print(e)

```

1 Physical GPUs, 1 Logical GPUs

```

[3]: # dataset directory
# this directory contains all the datasets related for ML4SCI tests.
os.listdir("../dataset")

```

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[3]: ['QCDToGGQQ_IMGjet_RH1all_jet0_run0_n36272',
      'QCDToGGQQ_IMGjet_RH1all_jet0_run0_n36272.test.snappy.parquet',
      'QCDToGGQQ_IMGjet_RH1all_jet0_run1_n47540',
      'QCDToGGQQ_IMGjet_RH1all_jet0_run1_n47540.test.snappy.parquet',
      'QCDToGGQQ_IMGjet_RH1all_jet0_run2_n55494',
      'QCDToGGQQ_IMGjet_RH1all_jet0_run2_n55494.test.snappy.parquet',
      'quark-gluon_data-set_n139306.hdf5',
      'SingleElectronPt50_IMGCROPS_n249k_RHv1.hdf5',
      'SinglePhotonPt50_IMGCROPS_n249k_RHv1.hdf5']

```

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[4]: # import dataset

# importing electron dataset and seperating images and labels
electron_dataset = h5py.File("../dataset/SingleElectronPt50_IMGCROPS_n249k_RHv1.
    ↪hdf5", "r")
electron_imgs=np.array(electron_dataset["X"])
electron_labels=np.array(electron_dataset["y"],dtype=np.int64)

# importing photon dataset and seperating images and labels
photon_dataset = h5py.File("../dataset/SinglePhotonPt50_IMGCROPS_n249k_RHv1.
    ↪hdf5", "r")
photon_imgs=np.array(photon_dataset["X"])
photon_labels=np.array(photon_dataset["y"],dtype=np.int64)

```

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[5]: # concatenate electron and photon images/labels
img_arrs = np.vstack((photon_imgs,electron_imgs))
labels = np.hstack((photon_labels,electron_labels)).astype(np.int64)

```

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[6]: # random split sizes are set to 70%:20%:10%

num_train = int(img_arrs.shape[0]*0.7) if img_arrs.shape[0]%10==0 else
    ↪int(img_arrs.shape[0]*0.7)+1
num_val_test = img_arrs.shape[0] - num_train

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num_val = int(num_val_test*(2/3)) if num_val_test%3==0 else int(num_val_test*(2/
↳3))+1
num_test = num_val_test - num_val

# random split of train, validation, tests set
# seed it set to 42 for reproducability of results
split_seed = 42

X_train, X_val_test, y_train, y_val_test = train_test_split(img_arrs, labels,
↳test_size=num_val_test, train_size = num_train,
↳random_state=split_seed)
X_val, X_test, y_val, y_test = train_test_split(X_val_test, y_val_test,
↳test_size=num_test, train_size=
↳num_val,
↳random_state=split_seed)

```

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[7]: def preprocess(inputs, mean=0.5, std=0.5, size=96):
    """
    This module contains preprocess layers.
    These layers will standardize and resize images.
    """
    x = Lambda(lambda inputs: (inputs - mean) / std)(inputs)
    x = Resizing(size, size)(x)
    return x

def forward_pass(inputs, num_middle_blocks = 8, num_classes=2):
    """
    This module declares the forward pass of the Xception model
    Xception model, as specified in
    https://arxiv.org/pdf/1610.02357.pdf
    """

    # Begin of entry flow
    x = Conv2D(32, 3, strides = 2, padding='same')(inputs)
    x = BatchNormalization()(x)
    x = Activation('relu')(x)

    x = Conv2D(64, 3, padding='same')(x)
    x = BatchNormalization()(x)
    x = Activation('relu')(x)

    Z = x

    for size in [128, 256, 728]:
        x = Activation('relu')(x)

```

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x = SeparableConv2D(size, 3, padding='same')(x)
x = BatchNormalization()(x)

x = Activation('relu')(x)
x = SeparableConv2D(size, 3, padding='same')(x)
x = BatchNormalization()(x)

x = MaxPooling2D(3, strides=2, padding='same')(x)

# skip connection
residual = Conv2D(size, 1, strides=2, padding='same')(Z)
residual = BatchNormalization()(residual)
x += residual

Z = x

# Begin of middle flow
for _ in range(num_middle_blocks) :
    x = Activation('relu')(x)
    x = SeparableConv2D(728, 3, padding='same')(x)
    x = BatchNormalization()(x)

    x = Activation('relu')(x)
    x = SeparableConv2D(728, 3, padding='same')(x)
    x = BatchNormalization()(x)

    x = Activation('relu')(x)
    x = SeparableConv2D(728, 3, padding='same')(x)
    x = BatchNormalization()(x)

    x += Z
    Z = x

# Begin of exist flow
x = SeparableConv2D(728, 3, padding='same')(x)
x = BatchNormalization()(x)

x = Activation('relu')(x)
x = SeparableConv2D(1024, 3, padding='same')(x)
x = BatchNormalization()(x)

x = MaxPooling2D(3, strides=2, padding='same')(x)

# skip connection
residual = Conv2D(1024, 1, strides=2, padding='same')(Z)
residual = BatchNormalization()(residual)
x += residual

```

```

x = Activation('relu')(x)
x = SeparableConv2D(728, 3, padding='same')(x)
x = BatchNormalization()(x)

x = Activation('relu')(x)
x = SeparableConv2D(1024, 3, padding='same')(x)
x = BatchNormalization()(x)

x = GlobalAveragePooling2D()(x)
act = 'softmax'
if num_classes == 1:
    act = 'sigmoid'
output = Dense(num_classes, activation=act)(x)
return output

```

```

[8]: # defining the input layer
inputs = Input(shape=(32, 32, 2))

# defining the output
outputs = forward_pass(
    preprocess(inputs,
                mean=0.5, std=0.5, size=96),
    num_middle_blocks=8, num_classes=1
)

# declaring the model with input and output
model = Model(inputs, outputs, name="xception")

```

```

[9]: # setup for saving and loading model checkpoints
initial_epoch = 0
checkpoint_dir = "../models/tf-ckpts/{model}".format(model=model.name)
checkpoints = os.listdir(checkpoint_dir)
checkpoint = list(filter(lambda i: ".ckpt" in i, checkpoints))

checkpoint_path = checkpoint_dir+"/cp-{epoch}.ckpt"

cp_callback = tf.keras.callbacks.ModelCheckpoint(filepath=checkpoint_path,
                                                  save_weights_only=True,
                                                  verbose=1)

if len(checkpoint)>0:
    initial_epoch = sorted([int(ck.split(".")[0].split("-")[1]) for ck in
↪checkpoint])[-1]
    latest = tf.train.latest_checkpoint(checkpoint_dir)
    model.load_weights(latest)

```

0.0.1 Training and evaluating the Xception model.

Refer the [readme](#) for performance analysis

```
[10]: # training the model
model.compile(optimizer='adam', loss=tf.keras.losses.BinaryCrossentropy(),
             metrics=[tf.keras.metrics.AUC()])

history = model.fit(X_train, y_train, epochs=23, batch_size=16,
                  validation_data=(X_val, y_val),
                  initial_epoch=initial_epoch, callbacks=cp_callback)
```

Epoch 3/23

21788/21788 [=====] - ETA: 0s - loss: 0.5680 - auc:
0.7768

Epoch 3: saving model to ../models/tf-ckpts/xception\cp-3.ckpt

21788/21788 [=====] - 7277s 333ms/step - loss: 0.5680 -
auc: 0.7768 - val_loss: 0.6010 - val_auc: 0.7616