model-training-kaggle-notebook

October 14, 2024

[ ]:
```
import os
import numpy as np
import tensorflow as tf
import keras
from matplotlib import pyplot as plt
import glob
import random
```

[ ]:
```
import os
import numpy as np


def load_img(img_dir, img_list):
    images=[]
    for i, image_name in enumerate(img_list):
        if (image_name.split('.')[1] == 'npy'):

            image = np.load(img_dir+image_name)

            images.append(image)
    images = np.array(images)

    return(images)




def imageLoader(img_dir, img_list, mask_dir, mask_list, batch_size):

    L = len(img_list)

    #keras needs the generator infinite, so we will use while true
    while True:

        batch_start = 0
        batch_end = batch_size
```

```
        while batch_start < L:
            limit = min(batch_end, L)

            X = load_img(img_dir, img_list[batch_start:limit])
            Y = load_img(mask_dir, mask_list[batch_start:limit])

            yield (X,Y) #a tuple with two numpy arrays with batch_size samples ␣
↪

            batch_start += batch_size
            batch_end += batch_size
```

[ ]:
```
train_img_dir = "BraTS2020_TrainingData/input_data_128/train/images/"
train_mask_dir = "BraTS2020_TrainingData/input_data_128/train/masks/"

val_img_dir = "BraTS2020_TrainingData/input_data_128/val/images/"
val_mask_dir = "BraTS2020_TrainingData/input_data_128/val/masks/"

train_img_list=os.listdir(train_img_dir)
train_mask_list = os.listdir(train_mask_dir)

val_img_list=os.listdir(val_img_dir)
val_mask_list = os.listdir(val_mask_dir)
```

[ ]:
```
batch_size = 2

train_img_datagen = imageLoader(train_img_dir, train_img_list,
                                train_mask_dir, train_mask_list, batch_size)

val_img_datagen = imageLoader(val_img_dir, val_img_list,
                                val_mask_dir, val_mask_list, batch_size)
```

[ ]:
```
img, msk = train_img_datagen.__next__()

img_num = random.randint(0,img.shape[0]-1)
test_img=img[img_num]
test_mask=msk[img_num]
test_mask=np.argmax(test_mask, axis=3)

n_slice=random.randint(0, test_mask.shape[2])
```

[ ]:
```
plt.figure(figsize=(12, 8))

plt.subplot(221)
plt.imshow(test_img[:,:,n_slice, 0], cmap='gray')
plt.title('Image flair')
plt.subplot(222)
```

```
plt.imshow(test_img[:,:,n_slice, 1], cmap='gray')
plt.title('Image t1ce')
plt.subplot(223)
plt.imshow(test_img[:,:,n_slice, 2], cmap='gray')
plt.title('Image t2')
plt.subplot(224)
plt.imshow(test_mask[:,:,n_slice])
plt.title('Mask')
plt.show()
```

1 model train

[ ]:
```
wt0, wt1, wt2, wt3 = 0.25,0.25,0.25,0.25
LR = 0.0001
optim = keras.optimizers.Adam(LR)
```

[ ]:
```
def dice_loss(y_true, y_pred):
    numerator = 2 * tf.reduce_sum(y_true * y_pred, axis=(1, 2, 3))
    denominator = tf.reduce_sum(y_true + y_pred, axis=(1, 2, 3))
    return 1 - tf.reduce_mean(numerator / denominator)


def categorical_focal_loss(y_true, y_pred, alpha=0.25, gamma=2.0):
    epsilon = 1e-7
    y_pred = tf.clip_by_value(y_pred, epsilon, 1 - epsilon)
    loss = -tf.reduce_sum(alpha * tf.pow(1 - y_pred, gamma) * y_true * tf.math.
 ↪log(y_pred), axis=-1)
    return tf.reduce_mean(loss)


def total_loss(y_true, y_pred):
    wt0, wt1, wt2, wt3 = 0.25, 0.25, 0.25, 0.25
    dice = dice_loss(y_true, y_pred)
    focal = categorical_focal_loss(y_true, y_pred)
    return dice + focal
def accuracy(y_true, y_pred):
    correct_predictions = tf.equal(tf.argmax(y_true, axis=-1), tf.
 ↪argmax(y_pred, axis=-1))
    return tf.reduce_mean(tf.cast(correct_predictions, tf.float32))
```

[ ]:
```
steps_per_epoch = len(train_img_list)//batch_size
val_steps_per_epoch = len(val_img_list)//batch_size
```

[ ]:
```
from keras.models import Model
from keras.layers import Input, Conv3D, MaxPooling3D, concatenate,␣
 ↪Conv3DTranspose, BatchNormalization, Dropout, Lambda
```

```
from keras.optimizers import Adam
from keras.metrics import MeanIoU

kernel_initializer =  'he_uniform' #Try others if you want


################################################################
def simple_unet_model(IMG_HEIGHT, IMG_WIDTH, IMG_DEPTH, IMG_CHANNELS,␣
 ↪num_classes):
#Build the model
    inputs = Input((IMG_HEIGHT, IMG_WIDTH, IMG_DEPTH, IMG_CHANNELS))
    #s = Lambda(lambda x: x / 255)(inputs)   #No need for this if we normalize␣
 ↪our inputs beforehand
    s = inputs

    #Contraction path
    c1 = Conv3D(16, (3, 3, 3), activation='relu',␣
 ↪kernel_initializer=kernel_initializer, padding='same')(s)
    c1 = Dropout(0.1)(c1)
    c1 = Conv3D(16, (3, 3, 3), activation='relu',␣
 ↪kernel_initializer=kernel_initializer, padding='same')(c1)
    p1 = MaxPooling3D((2, 2, 2))(c1)

    c2 = Conv3D(32, (3, 3, 3), activation='relu',␣
 ↪kernel_initializer=kernel_initializer, padding='same')(p1)
    c2 = Dropout(0.1)(c2)
    c2 = Conv3D(32, (3, 3, 3), activation='relu',␣
 ↪kernel_initializer=kernel_initializer, padding='same')(c2)
    p2 = MaxPooling3D((2, 2, 2))(c2)

    c3 = Conv3D(64, (3, 3, 3), activation='relu',␣
 ↪kernel_initializer=kernel_initializer, padding='same')(p2)
    c3 = Dropout(0.2)(c3)
    c3 = Conv3D(64, (3, 3, 3), activation='relu',␣
 ↪kernel_initializer=kernel_initializer, padding='same')(c3)
    p3 = MaxPooling3D((2, 2, 2))(c3)

    c4 = Conv3D(128, (3, 3, 3), activation='relu',␣
 ↪kernel_initializer=kernel_initializer, padding='same')(p3)
    c4 = Dropout(0.2)(c4)
    c4 = Conv3D(128, (3, 3, 3), activation='relu',␣
 ↪kernel_initializer=kernel_initializer, padding='same')(c4)
    p4 = MaxPooling3D(pool_size=(2, 2, 2))(c4)

    c5 = Conv3D(256, (3, 3, 3), activation='relu',␣
 ↪kernel_initializer=kernel_initializer, padding='same')(p4)
```

```
    c5 = Dropout(0.3)(c5)
    c5 = Conv3D(256, (3, 3, 3), activation='relu',␣
↪kernel_initializer=kernel_initializer, padding='same')(c5)

    #Expansive path
    u6 = Conv3DTranspose(128, (2, 2, 2), strides=(2, 2, 2), padding='same')(c5)
    u6 = concatenate([u6, c4])
    c6 = Conv3D(128, (3, 3, 3), activation='relu',␣
↪kernel_initializer=kernel_initializer, padding='same')(u6)
    c6 = Dropout(0.2)(c6)
    c6 = Conv3D(128, (3, 3, 3), activation='relu',␣
↪kernel_initializer=kernel_initializer, padding='same')(c6)

    u7 = Conv3DTranspose(64, (2, 2, 2), strides=(2, 2, 2), padding='same')(c6)
    u7 = concatenate([u7, c3])
    c7 = Conv3D(64, (3, 3, 3), activation='relu',␣
↪kernel_initializer=kernel_initializer, padding='same')(u7)
    c7 = Dropout(0.2)(c7)
    c7 = Conv3D(64, (3, 3, 3), activation='relu',␣
↪kernel_initializer=kernel_initializer, padding='same')(c7)

    u8 = Conv3DTranspose(32, (2, 2, 2), strides=(2, 2, 2), padding='same')(c7)
    u8 = concatenate([u8, c2])
    c8 = Conv3D(32, (3, 3, 3), activation='relu',␣
↪kernel_initializer=kernel_initializer, padding='same')(u8)
    c8 = Dropout(0.1)(c8)
    c8 = Conv3D(32, (3, 3, 3), activation='relu',␣
↪kernel_initializer=kernel_initializer, padding='same')(c8)

    u9 = Conv3DTranspose(16, (2, 2, 2), strides=(2, 2, 2), padding='same')(c8)
    u9 = concatenate([u9, c1])
    c9 = Conv3D(16, (3, 3, 3), activation='relu',␣
↪kernel_initializer=kernel_initializer, padding='same')(u9)
    c9 = Dropout(0.1)(c9)
    c9 = Conv3D(16, (3, 3, 3), activation='relu',␣
↪kernel_initializer=kernel_initializer, padding='same')(c9)

    outputs = Conv3D(num_classes, (1, 1, 1), activation='softmax')(c9)

    model = Model(inputs=[inputs], outputs=[outputs])
    #compile model outside of this function to make it flexible.
    model.summary()

    return model
```

[ ]:
```
model = simple_unet_model(IMG_HEIGHT=128,
                          IMG_WIDTH=128,
                          IMG_DEPTH=128,
                          IMG_CHANNELS=3,
                          num_classes=4)
```

[ ]:
```
model.compile(optimizer = optim, loss=total_loss, metrics=[accuracy, tf.keras.
↪metrics.MeanIoU(num_classes=4)])
```

[ ]:
```
import tensorflow as tf
import keras
import keras.backend as K
from keras.callbacks import CSVLogger
from tensorflow.keras.models import *
from tensorflow.keras.layers import *
from tensorflow.keras.optimizers import *
from tensorflow.keras.callbacks import ModelCheckpoint
```

[ ]:
```
# callback
```

[ ]:
```
history=model.fit(train_img_datagen,
          steps_per_epoch=steps_per_epoch,
          epochs=100,
          verbose=1,
          validation_data=val_img_datagen,
          validation_steps=val_steps_per_epoch,
          )
```

[ ]:
```
model.save('brats_3d.h5')
```

2 GRAPH

[ ]:
```
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs = range(1, len(loss) + 1)
```

[ ]:
```
plt.plot(epochs, loss, 'y', label='Training loss')
plt.plot(epochs, val_loss, 'r', label='Validation loss')
plt.title('Training and validation loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.show()
```

[ ]:
```
acc = history.history['accuracy']
val_acc = history.history['val_accuracy']
```

[ ]:
```
plt.plot(epochs, acc, 'y', label='Training accuracy')
plt.plot(epochs, val_acc, 'r', label='Validation accuracy')
plt.title('Training and validation accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()
```

3 MOU

[ ]:
```
from keras.metrics import MeanIoU

batch_size=8 #Check IoU for a batch of images
test_img_datagen = imageLoader(val_img_dir, val_img_list,
                               val_mask_dir, val_mask_list, batch_size)

#Verify generator.... In python 3 next() is renamed as __next__()
test_image_batch, test_mask_batch = test_img_datagen.__next__()

test_mask_batch_argmax = np.argmax(test_mask_batch, axis=4)
test_pred_batch = model.predict(test_image_batch)
test_pred_batch_argmax = np.argmax(test_pred_batch, axis=4)

n_classes = 4
IOU_keras = MeanIoU(num_classes=n_classes)
IOU_keras.update_state(test_pred_batch_argmax, test_mask_batch_argmax)
print("Mean IoU =", IOU_keras.result().numpy())
```

4 PREDICT

[ ]:
```
img_num = 82

test_img = np.load("BraTS2020_TrainingData/input_data_128/val/images/
 ↪image_"+str(img_num)+".npy")

test_mask = np.load("BraTS2020_TrainingData/input_data_128/val/masks/
 ↪mask_"+str(img_num)+".npy")
test_mask_argmax=np.argmax(test_mask, axis=3)

test_img_input = np.expand_dims(test_img, axis=0)
test_prediction = my_model.predict(test_img_input)
test_prediction_argmax=np.argmax(test_prediction, axis=4)[0,:,:,:]
```

[ ]:
```
#Plot individual slices from test predictions for verification
from matplotlib import pyplot as plt
```

```
import random

#n_slice=random.randint(0, test_prediction_argmax.shape[2])
n_slice = 55
plt.figure(figsize=(12, 8))
plt.subplot(231)
plt.title('Testing Image')
plt.imshow(test_img[:,:,n_slice,1], cmap='gray')
plt.subplot(232)
plt.title('Testing Label')
plt.imshow(test_mask_argmax[:,:,n_slice])
plt.subplot(233)
plt.title('Prediction on test image')
plt.imshow(test_prediction_argmax[:,:, n_slice])
plt.show()
```