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# -*- coding: utf-8 -*-
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''''
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```
copyright: Quing Zhu, Optical and Ultrasound Imaging Laboratory
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Email: zhu.q@wustl.edu
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```
# TrainingUSPAM.py
```

```
# Import modules
```

```
#####
```

```
import numpy as np
```

```
import scipy as sp
```

```
from math import ceil
```

```
from tensorflow.keras.utils import to_categorical
```

```
from UtilitiesDL import createModelCNN, trainModelCNN, plotAUCWithError
```

```
#####
```

```
# User Chosen Parameters
```

```
#####
```

```
USorPAT = 1 # 0 for US, 1 for PA
```

```
enableAUCPlot = 1 #Must be set to 1 to plot the AUC
```

```
total_iter = 20 #How many times the training will run, performance will be  
averaged over all runs
```

```
validation_percentage = 19.5 #percentage of total data used for validation
```

```
training_percentage = 70 # percentage of total data used for training
```

```
#####
```

```
# Data loading for normal and cancer data
```

```
#####
```

```
if USorPAT==0:
```

```

mat_fname_N = "TrainingImages_US_N_mixed_90by50_10032020.mat" #file containing normal data
mat_fname_M = "TrainingImages_US_C_mixed_90by50_10032020.mat" #file containing normal data
else:
mat_fname_N = "TrainingImages_PA_N_mixed_90by50_10032020.mat"
mat_fname_M = "TrainingImages_PA_C_mixed_90by50_10032020.mat"

mat_contents = sp.io.loadmat(mat_fname_N)
X_b = np.array(mat_contents['All_images'])
mat_contents = sp.io.loadmat(mat_fname_M)
X_m = np.array(mat_contents['All_images'])

#####
# Reshape data (add extra dimension to be used in CNN)
# and assign labels
# 0 is normal, 1 is cancer
#####
X_b = X_b.reshape(X_b.shape[0],X_b.shape[1],X_b.shape[2],1)
X_m = X_m.reshape(X_m.shape[0],X_m.shape[1],X_m.shape[2],1)

no_of_benigns = X_b.shape[0]
no_of_malignants = X_m.shape[0]

no_of_cases = no_of_benigns + no_of_malignants
nTrain = ceil(2*no_of_malignants/3)
num_classes = 2
labels = np.zeros((no_of_cases,))
labels[no_of_benigns:no_of_cases] = 1
labels = to_categorical(labels)
y_b = labels[:no_of_benigns]

```

```
y_m = labels[no_of_benigns:]
```

```
#####
```

```
# model architecture
```

```
# not compiled every time, saved as empty model
```

```
#####
```

```
model_cnn = createModelCNN(X_b.shape[1:],num_classes)
```

```
model_cnn.save('EmptyModel.h5')
```

```
#####
```

```
# training and validation can be run for different training sizes
```

```
# to determine if we have enough data for reliable classification
```

```
# based on validation accuracy
```

```
#####
```

```
tprs = []
```

```
aucs = []
```

```
base_fpr = np.linspace(0, 1, 101)
```

```
training_size = int(min(no_of_benigns,no_of_malignants)*training_percentage/100)
```

```
validation_size = int(min(no_of_benigns, no_of_malignants)*validation_percentage/100)
```

```
for rand_num in range(0, total_iter):
```

```
    roc_auc, tpr = trainModelCNN(X_b,y_b,X_m,y_m, training_size, validation_size, rand_num, USorPAT,  
    base_fpr, training_percentage)
```

```
    aucs.append(roc_auc)
```

```
    tprs.append(tpr)
```

```
if enableAUCPlot:
    tprs = np.array(tprs)
    aucs = np.array(aucs)
    plotAUCWithError(base_fpr,tprs,auCs)

#####
#      End of TrainingUSPAM.py
#####

# -*- coding: utf-8 -*-
"""
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Email: zhu.q@wustl.edu
"""

#####
# UtilitiesDL.py
# Import modules
#####

import os
import numpy as np
from tensorflow.keras.models import Sequential,load_model
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import Activation,Conv2D,MaxPooling2D,Flatten
from tensorflow.keras.layers import Dropout
from tensorflow.keras.optimizers import RMSprop
from tensorflow.keras.backend import clear_session
from scipy import interp
```

```

import matplotlib.pyplot as plt

import scipy as sp

from sklearn.model_selection import train_test_split
from sklearn.metrics import roc_curve, auc
from tensorflow.keras.callbacks import EarlyStopping

def createModelCNN(inputShape, numClasses, kernelSize=3, poolSize=4, hiddenNodes=512):
    # Hyperparameters to tune for users:
    # 1. learning rate (and decay)
    # 2. number of filters
    # 3. convolution kernel size
    # 4. pooling kernel size
    # 5. number of convolution layers
    # 6. Number of dense layers
    # 7. Number of hidden nodes in the dense layers
    # 8. Dropout probability
    opt = RMSprop(lr=0.0001, decay=1e-6)
    model_cnn = Sequential()
    #Feature extraction layer 1
    model_cnn.add(Conv2D(32,(kernelSize,kernelSize),padding='same',input_shape=inputShape))
    model_cnn.add(Activation('relu'))
    model_cnn.add(Conv2D(32,(kernelSize,kernelSize)))
    model_cnn.add(Activation('relu'))
    model_cnn.add(MaxPooling2D(pool_size=(poolSize,poolSize)))
    model_cnn.add(Dropout(0.25))
    #Feature Extraction Layer 2
    model_cnn.add(Conv2D(64,(kernelSize,kernelSize), padding='same'))
    model_cnn.add(Activation('relu'))

```

```

model_cnn.add(Conv2D(64,(kernelSize, kernelSize)))
model_cnn.add(Activation('relu'))
model_cnn.add(MaxPooling2D(pool_size=(poolSize,poolSize)))
model_cnn.add(Dropout(0.25))
#Classification
model_cnn.add(Flatten())
model_cnn.add(Dense(hiddenNodes))
model_cnn.add(Activation('relu'))
model_cnn.add(Dropout(0.5))
model_cnn.add(Dense(numClasses))
model_cnn.add(Activation('softmax'))
model_cnn.compile(loss='categorical_crossentropy',optimizer=opt,metrics=['accuracy'])
return model_cnn

```

```

def trainModelCNN(dataNormal,labelNormal,dataMalignant,labelMalignant, training_size,
validation_size, random_seed, USorPAT, base_fpr, training_percentage):

    #labelNormal: vector of zeros

    #labelMalignant: vector of ones

    #both training and validation sizes are percentages of the total data,
    #for example, training_size=0.5 means use 50% of total data for training

    #####

    # training data

    #####

    X_b_train, X_b_test, y_b_train, y_b_test = train_test_split( dataNormal, labelNormal, test_size=1-
training_size/dataNormal.shape[0], random_state=random_seed)

    X_m_train, X_m_test, y_m_train, y_m_test = train_test_split( dataMalignant, labelMalignant,
test_size=1-training_size/dataMalignant.shape[0], random_state=random_seed)

    X_train = np.concatenate((X_b_train,X_m_train))

    y_train = np.concatenate((y_b_train,y_m_train))

```

```

#####

# get validation data from remaining data (preliminary test set in this case)

#####

validation_size -= 0.005 #to avoid edge cases

X_b_valid, X_b_test, y_b_valid, y_b_test = train_test_split(X_b_test, y_b_test, test_size=1-
validation_size/X_b_test.shape[0], random_state=random_seed)

X_m_valid, X_m_test, y_m_valid, y_m_test = train_test_split(X_m_test, y_m_test, test_size=1-
validation_size/X_m_test.shape[0], random_state=random_seed)

X_valid = np.concatenate((X_b_valid,X_m_valid))
y_valid = np.concatenate((y_b_valid,y_m_valid))

#####

# load an empty model and train

#####

model_cnn = load_model('EmptyModel.h5')
early_stopping_monitor = EarlyStopping(patience=2)

History_metrics =
model_cnn.fit(X_train,y_train,batch_size=20,epochs=20,validation_data=(X_valid,y_valid),callbacks=[ear
ly_stopping_monitor],verbose=1)

y_pred_prob = model_cnn.predict_proba(X_valid.astype('float64'))
y_pred_prob = y_pred_prob[:, 1] # 1 means cancer, 0 is normal.
fpr, tpr, _ = roc_curve(y_valid.argmax(axis=1), y_pred_prob)
roc_auc = auc(fpr, tpr)
tpr = interp(base_fpr, fpr, tpr)
tpr[0] = 0.0

#####

```

```

# Save Data and model

# you can comment model saving part if you don't want to save model

#####

if USorPAT==0:

    save_directory =
os.getcwd()+'\savedResultUS'+str(training_percentage)+'\run_'+str(random_seed+1)

    else:

        save_directory =
os.getcwd()+'\savedResultPA'+str(training_percentage)+'\run_'+str(random_seed+1)

if not os.path.exists(save_directory):

    os.makedirs(save_directory)

sp.io.savemat(save_directory+'\'+'ROC_DATA.mat',{tpr:tpr,'fpr':base_fpr,'loss':History_metrics.history
['loss'],'accuracy':History_metrics.history['accuracy'],'val_loss':History_metrics.history['val_loss'],'val_ac
curacy':History_metrics.history['val_accuracy']})

    model_cnn.save(save_directory+'\'+'TrainedModel.h5')

return roc_auc, tpr

def testModelCNN(testData, testLabels, training_percentage, random_seed, USorPAT, base_fpr):

    #training_percentage:0 to 100%, intergers

    clear_session()

    if USorPAT==0:

        save_directory =
os.getcwd()+'\savedResultUS'+str(training_percentage)+'\run_'+str(random_seed+1)

        else:

            save_directory =
os.getcwd()+'\savedResultPA'+str(training_percentage)+'\run_'+str(random_seed+1)

```

```

#####
# load trained model and test
#####
model_cnn = load_model(save_directory+'\\'+TrainedModel.h5')

y_pred_prob = model_cnn.predict_proba(testData.astype('float64'))
y_pred_prob = y_pred_prob[:, 1]
fpr, tpr, _ = roc_curve(testLabels, y_pred_prob)
roc_auc = auc(fpr, tpr)
tpr = interp(base_fpr, fpr, tpr)
tpr[0] = 0.0

#####
# Save testing results
#####
if not os.path.exists(save_directory):
    os.makedirs(save_directory)

    sp.io.savemat(save_directory+'\\'+ROC_DATA_TEST.mat',{tpr':tpr,'fpr':base_fpr,'probability':y_
pred_prob})

return roc_auc, tpr

def plotAUCWithError(base_fpr,tprs,aucs):
    mean_tprs = tprs.mean(axis=0)
    std = tprs.std(axis=0)

    mean_auc = auc(base_fpr, mean_tprs)
    std_auc = np.std(aucs)

```

```

tprs_upper = np.minimum(mean_tprs + std, 1)
tprs_lower = mean_tprs - std

plt.rc('legend', fontsize = 20)

plt.figure(figsize = (12, 8))
plt.plot(base_fpr, mean_tprs, 'b', alpha = 0.8, label = r'Mean ROC (AUC = %0.3f  $\pm$  %0.3f)' %
(mean_auc, std_auc))
plt.fill_between(base_fpr, tprs_lower, tprs_upper, color = 'blue', alpha = 0.2)
plt.plot([0, 1], [0, 1], linestyle = '--', lw = 2, color = 'r', label = 'Luck', alpha = 0.8)
plt.xlim([-0.01, 1.01])
plt.ylim([-0.01, 1.01])
plt.ylabel('True Positive Rate')
plt.xlabel('False Positive Rate')
plt.legend(loc = "lower right")
plt.title('Receiver operating characteristic (ROC) curve')
plt.show()

#####
#      End of UtilitiesDL.py
#####

# -*- coding: utf-8 -*-
"""

@Quing Zhu, Optical and Ultrasound Imaging Laboratory,
Email: zhu.q@wustl.edu
"""

```

```

#####

# TesingUSPAM.py

# Import modules

#####

import numpy as np

from tensorflow.keras.backend import clear_session

import scipy as sp

from UtilitiesDL import testModelCNN, plotAUCWithError

#####

# User Chosen Parameters

#####

USorPAT = 1                #0 for US, 1 for PA

enableAUCPlot = 1;        #Must be set to 1 to plot the AUC

total_iter = 20           #Should be same as Training

training_percentage = 70  #should be same as Training

#####

# Data loading for normal and cancer data

#####

if USorPAT==0:

    mat_fname_test = "TestingImages_US_90by50_190_6th_208_209_210.mat"

else:

    mat_fname_test = "TestingImages_PA_90by50_190_6th_208_209_210.mat"

mat_contents = sp.io.loadmat(mat_fname_test)

X_test = np.array(mat_contents['All_test_images'])

y_test = np.array(mat_contents['All_test_labels'])

```

```
#####  
# Reshape data  
#####  
X_test = X_test.reshape(X_test.shape[0],X_test.shape[1],X_test.shape[2],1)  
  
tprs = []  
aucs = []  
base_fpr = np.linspace(0, 1, 101)  
  
for rand_num in range(0,total_iter):  
    clear_session()  
    roc_auc, tpr = testModelCNN(X_test, y_test, training_percentage, rand_num, USorPAT, base_fpr)  
    aucs.append(roc_auc)  
    tprs.append(tpr)  
  
if enableAUCPlot:  
    tprs = np.array(tprs)  
    aucs = np.array(aucs)  
    plotAUCWithError(base_fpr,tprs,aucs)  
  
#####  
#     TesingUSPAM.py  
#####
```