

# NASNetLarge-Final Working-V2

February 13, 2020

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In [ ]: import matplotlib.pyplot as plt
        from keras import applications
        from keras.preprocessing.image import ImageDataGenerator
        from keras import optimizers
        from keras.models import Sequential
        from keras.layers import Dropout, Flatten, Dense
        from keras.applications.inception_v3 import InceptionV3
        from keras.preprocessing import image
        from keras.models import Model
        from keras.layers import Dense, Flatten
        from keras import backend as K
        import numpy as np
        import pandas as pd
        import os
        from sklearn.metrics import classification_report, confusion_matrix
        import sklearn.metrics as metrics
        import sklearn
        from sklearn.metrics import roc_auc_score
        from sklearn.metrics import roc_curve
        import matplotlib.pyplot as plt
        %matplotlib inline

In [ ]: # create the base pre-trained model
        # build the VGG16 network
        base_model = applications.nasnet.NASNetMobile(weights='imagenet', include_top=False,
                                                         input_shape=(150,150,3))
        print('Model loaded.')
        base_model.summary()

In [ ]: # this is the model we will train
        model = Sequential()
        model.add(base_model)
        model.add(Flatten())
        model.add(Dense(256,activation='relu'))
        model.add(Dense(1, activation='sigmoid'))

        model.summary()
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In [ ]: print('Number of trainable weights before freezing: ', len(model.trainable_weights))
        ## to freeze all convolutional layers in pretrained network method 1
        # base_model.trainable=False

In [ ]: # def recall_m(y_true, y_pred):
        #     true_positives = K.sum(K.round(K.clip(y_true * y_pred,0,1)))
        #     possible_positives = K.sum(K.round(K.clip(y_true,0,1)))
        #     recall = true_positives / (possible_positives + K.epsilon())
        #     return recall
        # def precision_m(y_true, ypred):
        #     true_positives = K.sum(K.round(K.clip(y_true * y_pred,0,1)))
        #     predicted_positives = K.sum(K.round(K.clip(y_pred,0,1)))
        #     precision = true_positives/(predicted_positives+K.epsilon())
        #     return precision

        # first: train only the top layers (which were randomly initialized)
        # i.e. freeze all convolutional pretrained layers method 2
        for layer in base_model.layers:
            layer.trainable = False
        print('After freezing: ', len(model.trainable_weights))
        # compile the model (should be done *after* setting layers to non-trainable)
        model.compile(optimizer=optimizers.Adam(lr=1e-4),metrics=['acc'], loss='binary_crossentropy')

In [ ]: train_data_dir = '/home/mlab/Documents/brats_hl_data/train'
        validation_data_dir = '/home/mlab/Documents/brats_hl_data/val'
        # 44938
        # 5616
        nb_train_samples = 44938
        nb_validation_samples = 5616
        epochs = 8
        batch_size = 128
        # prepare data augmentation configuration
        train_datagen = ImageDataGenerator(
            rescale=1. / 255,
            shear_range=0.2,
            zoom_range=0.2,
            horizontal_flip=True)

        test_datagen = ImageDataGenerator(rescale=1. / 255)

        train_generator = train_datagen.flow_from_directory(
            train_data_dir,
            target_size=(150, 150),
            batch_size=batch_size,
            class_mode='binary')

        validation_generator = test_datagen.flow_from_directory(
            validation_data_dir,

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target_size=(150, 150),
batch_size=batch_size,
class_mode='binary')
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In [ ]: true_classes = train_generator.classes
        print(true_classes)
        class_labels = list(train_generator.class_indices.keys())
        print(class_labels)
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In [ ]: # train the model on the new data for a few epochs
        history = model.fit_generator(train_generator,
                                     steps_per_epoch=nb_train_samples//batch_size,
                                     epochs=epochs,
                                     validation_data=validation_generator,
                                     validation_steps=nb_validation_samples//batch_size)
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In [ ]: true_classes_1 = validation_generator.classes
        print(true_classes)
        class_labels_1 = list(validation_generator.class_indices.keys())
        print(class_labels_1)
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In [ ]: #Confusion Matrix and Classification Report
        Y_pred = model.predict_generator(validation_generator, nb_validation_samples // batch_size+1)
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In [ ]: # y_pred = np.argmax(Y_pred, axis=1)
        y_pred = (Y_pred<0.475).astype(np.int)

        # print('Confusion Matrix')
        # print(confusion_matrix(true_classes_1, y_pred))
        # print('Classification Report')
        # print(classification_report(validation_generator.classes, y_pred,
        #                             target_names=class_labels_1))
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In [ ]: # print(validation_generator.classes)
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In [ ]: confusion_matrix = metrics.confusion_matrix(true_classes_1,y_pred)
        print(confusion_matrix)
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In [ ]: report= sklearn.metrics.classification_report(true_classes_1, y_pred,
                                                       target_names = class_labels_1)
        print(report)
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In [ ]: fpr, tpr, thresholds = roc_curve(validation_generator.classes, y_pred)

        # create plot
        plt.plot(fpr, tpr, label='ROC curve')
        plt.plot([0, 1], [0, 1], 'k--', label='Random guess')
        _ = plt.xlabel('False Positive Rate')
        _ = plt.ylabel('True Positive Rate')
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_ = plt.title('ROC Curve')
_ = plt.xlim([-0.02, 1])
_ = plt.ylim([0, 1.02])
_ = plt.legend(loc="lower right")

In [ ]: roc_auc_score(validation_generator.classes, y_pred)

In [ ]: batchX, batchy = train_generator.next()
_ , accuracy = model.evaluate(batchX, batchy)
print('Accuracy training: %.2f' % (accuracy*100))
batchXv, batchyv = validation_generator.next()
_ , accuracy = model.evaluate(batchXv, batchyv)
print('Accuracy val: %.2f' % (accuracy*100))

In [ ]: #plot the train and val curve
#get the details from the history object
acc = history.history['acc']
val_acc=history.history['val_acc']
loss = history.history['loss']
val_loss = history.history['val_loss']

epochs = range(1,len(acc)+1)

#train and validation accuracy
plt.plot(epochs,acc,'b',label='Training accuracy')
plt.plot(epochs,val_acc,'r',label='Validation accuracy')
plt.title('Training and Validation accuracy')
plt.legend()

In [ ]: #train and validation loss
plt.plot(epochs, loss, 'b',label='Training loss')
plt.plot(epochs, val_loss, 'r',label='Validation loss')
plt.title('Training and Validation loss')
plt.legend()
plt.show()

In [ ]: test_generator = test_datagen.flow_from_directory('/home/mlab/Documents/brats_hl_data/test',
                                                         class_mode='binary',
                                                         batch_size=batch_size,
                                                         target_size=(150,150))
scores = model.evaluate_generator(test_generator, steps=nb_validation_samples//batch_size)

In [ ]: print("%s: %.2f%%" % (model.metrics_names[1], scores[1]*100))

In [ ]: #Confution Matrix and Classification Report
# Y_pred = model.predict_generator(val_generator, 5616 // batch_size)
nb_test_samples=5619
Y_pred1 = model.predict_generator(test_generator,nb_test_samples//batch_size+1)
# y_pred = np.argmax(Y_pred,axis=1)

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In [ ]: true_classes_2 = test_generator.classes
        print(true_classes_2)
        class_labels_2 = list(test_generator.class_indices.keys())
        print(class_labels_2)

In [ ]: # y_pred1 = (Y_pred1<0.5).astype(np.int)
        y_pred1 = (Y_pred1 < 0.475).astype(np.int)
        # print(y_pred)
        # print('Confusion Matrix')
        # print(confusion_matrix(true_classes_2, y_pred1))
        # print('Classification Report')
        # print(classification_report(true_classes_2, y_pred1, target_names=class_labels_2))

In [15]: confusion_matrix1 = metrics.confusion_matrix(true_classes_2,y_pred1)
        print(confusion_matrix1)

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NameError                                Traceback (most recent call last)

<ipython-input-15-60cf3edce08b> in <module>
----> 1 confusion_matrix1 = metrics.confusion_matrix(true_classes_2,y_pred1)
      2 print(confusion_matrix1)

NameError: name 'true_classes_2' is not defined

In [ ]: report1= sklearn.metrics.classification_report(true_classes_2, y_pred1,
                                                    target_names = class_labels_2)
        print(report1)

In [ ]: print(" Loss: ", scores[0],"\n","Accuracy: ", scores[1])

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