

ResNet15V2-Final Working-V2

February 15, 2020

```
In [1]: 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.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
```

Using TensorFlow backend.

```
In [2]: help(applications)
        # import sys
        # !{sys.executable} -m pip install keras.applications
```

Help on package keras.applications in keras:

NAME

keras.applications

PACKAGE CONTENTS

densenet
imagenet_utils
inception_resnet_v2
inception_v3

mobilenet
mobilenet_v2
mobilenetv2
nasnet
resnet50
vgg16
vgg19
xception

FUNCTIONS

DenseNet121 = wrapper(*args, **kwargs)

DenseNet169 = wrapper(*args, **kwargs)

DenseNet201 = wrapper(*args, **kwargs)

InceptionResNetV2 = wrapper(*args, **kwargs)

InceptionV3 = wrapper(*args, **kwargs)

MobileNet = wrapper(*args, **kwargs)

MobileNetV2 = wrapper(*args, **kwargs)

NASNetLarge = wrapper(*args, **kwargs)

NASNetMobile = wrapper(*args, **kwargs)

ResNet50 = wrapper(*args, **kwargs)

VGG16 = wrapper(*args, **kwargs)

VGG19 = wrapper(*args, **kwargs)

Xception = wrapper(*args, **kwargs)

keras_modules_injection(base_fun)

DATA

absolute_import = _Feature((2, 5, 0, 'alpha', 1), (3, 0, 0, 'alpha', 0)...
division = _Feature((2, 2, 0, 'alpha', 2), (3, 0, 0, 'alpha', 0), 8192...
print_function = _Feature((2, 6, 0, 'alpha', 2), (3, 0, 0, 'alpha', 0)...

FILE

/home/mlab/anaconda3/lib/python3.7/site-packages/keras/applications/__init__.py

```
In [3]: ## create the base pre-trained model
        # build the VGG16 network
        base_model = applications.ResNet50(weights='imagenet', include_top=False,
                                             input_shape=(150,150,3))
        print('Model loaded.')
        base_model.summary()
```

WARNING: Logging before flag parsing goes to stderr.

W0215 16:32:21.521459 139777111516992 deprecation_wrapper.py:119] From /home/mlab/anaconda3/lib/python

W0215 16:32:21.532786 139777111516992 deprecation_wrapper.py:119] From /home/mlab/anaconda3/lib/python

W0215 16:32:21.536382 139777111516992 deprecation_wrapper.py:119] From /home/mlab/anaconda3/lib/python

W0215 16:32:21.554043 139777111516992 deprecation_wrapper.py:119] From /home/mlab/anaconda3/lib/python

W0215 16:32:21.554604 139777111516992 deprecation_wrapper.py:119] From /home/mlab/anaconda3/lib/python

W0215 16:32:21.672361 139777111516992 deprecation_wrapper.py:119] From /home/mlab/anaconda3/lib/python

W0215 16:32:21.724693 139777111516992 deprecation_wrapper.py:119] From /home/mlab/anaconda3/lib/python

/home/mlab/anaconda3/lib/python3.7/site-packages/keras_applications/resnet50.py:265: UserWarning: The output shape of `ResNet50(include_top=False)` is (None, 1000, 3) instead of (None, 1000, 1).
warnings.warn('The output shape of `ResNet50(include_top=False)` is (None, 1000, 3) instead of (None, 1000, 1).')

Model loaded.

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	(None, 150, 150, 3)	0	
conv1_pad (ZeroPadding2D)	(None, 156, 156, 3)	0	input_1[0][0]
conv1 (Conv2D)	(None, 75, 75, 64)	9472	conv1_pad[0][0]
bn_conv1 (BatchNormalization)	(None, 75, 75, 64)	256	conv1[0][0]
activation_1 (Activation)	(None, 75, 75, 64)	0	bn_conv1[0][0]
pool1_pad (ZeroPadding2D)	(None, 77, 77, 64)	0	activation_1[0][0]
max_pooling2d_1 (MaxPooling2D)	(None, 38, 38, 64)	0	pool1_pad[0][0]
res2a_branch2a (Conv2D)	(None, 38, 38, 64)	4160	max_pooling2d_1[0][0]
bn2a_branch2a (BatchNormalization)	(None, 38, 38, 64)	256	res2a_branch2a[0][0]

activation_2 (Activation)	(None, 38, 38, 64)	0	bn2a_branch2a[0][0]
res2a_branch2b (Conv2D)	(None, 38, 38, 64)	36928	activation_2[0][0]
bn2a_branch2b (BatchNormalizati	(None, 38, 38, 64)	256	res2a_branch2b[0][0]
activation_3 (Activation)	(None, 38, 38, 64)	0	bn2a_branch2b[0][0]
res2a_branch2c (Conv2D)	(None, 38, 38, 256)	16640	activation_3[0][0]
res2a_branch1 (Conv2D)	(None, 38, 38, 256)	16640	max_pooling2d_1[0][0]
bn2a_branch2c (BatchNormalizati	(None, 38, 38, 256)	1024	res2a_branch2c[0][0]
bn2a_branch1 (BatchNormalizatio	(None, 38, 38, 256)	1024	res2a_branch1[0][0]
add_1 (Add)	(None, 38, 38, 256)	0	bn2a_branch2c[0][0] bn2a_branch1[0][0]
activation_4 (Activation)	(None, 38, 38, 256)	0	add_1[0][0]
res2b_branch2a (Conv2D)	(None, 38, 38, 64)	16448	activation_4[0][0]
bn2b_branch2a (BatchNormalizati	(None, 38, 38, 64)	256	res2b_branch2a[0][0]
activation_5 (Activation)	(None, 38, 38, 64)	0	bn2b_branch2a[0][0]
res2b_branch2b (Conv2D)	(None, 38, 38, 64)	36928	activation_5[0][0]
bn2b_branch2b (BatchNormalizati	(None, 38, 38, 64)	256	res2b_branch2b[0][0]
activation_6 (Activation)	(None, 38, 38, 64)	0	bn2b_branch2b[0][0]
res2b_branch2c (Conv2D)	(None, 38, 38, 256)	16640	activation_6[0][0]
bn2b_branch2c (BatchNormalizati	(None, 38, 38, 256)	1024	res2b_branch2c[0][0]
add_2 (Add)	(None, 38, 38, 256)	0	bn2b_branch2c[0][0] activation_4[0][0]
activation_7 (Activation)	(None, 38, 38, 256)	0	add_2[0][0]
res2c_branch2a (Conv2D)	(None, 38, 38, 64)	16448	activation_7[0][0]
bn2c_branch2a (BatchNormalizati	(None, 38, 38, 64)	256	res2c_branch2a[0][0]
activation_8 (Activation)	(None, 38, 38, 64)	0	bn2c_branch2a[0][0]

res2c_branch2b (Conv2D)	(None, 38, 38, 64)	36928	activation_8[0][0]
bn2c_branch2b (BatchNormalizati	(None, 38, 38, 64)	256	res2c_branch2b[0][0]
activation_9 (Activation)	(None, 38, 38, 64)	0	bn2c_branch2b[0][0]
res2c_branch2c (Conv2D)	(None, 38, 38, 256)	16640	activation_9[0][0]
bn2c_branch2c (BatchNormalizati	(None, 38, 38, 256)	1024	res2c_branch2c[0][0]
add_3 (Add)	(None, 38, 38, 256)	0	bn2c_branch2c[0][0] activation_7[0][0]
activation_10 (Activation)	(None, 38, 38, 256)	0	add_3[0][0]
res3a_branch2a (Conv2D)	(None, 19, 19, 128)	32896	activation_10[0][0]
bn3a_branch2a (BatchNormalizati	(None, 19, 19, 128)	512	res3a_branch2a[0][0]
activation_11 (Activation)	(None, 19, 19, 128)	0	bn3a_branch2a[0][0]
res3a_branch2b (Conv2D)	(None, 19, 19, 128)	147584	activation_11[0][0]
bn3a_branch2b (BatchNormalizati	(None, 19, 19, 128)	512	res3a_branch2b[0][0]
activation_12 (Activation)	(None, 19, 19, 128)	0	bn3a_branch2b[0][0]
res3a_branch2c (Conv2D)	(None, 19, 19, 512)	66048	activation_12[0][0]
res3a_branch1 (Conv2D)	(None, 19, 19, 512)	131584	activation_10[0][0]
bn3a_branch2c (BatchNormalizati	(None, 19, 19, 512)	2048	res3a_branch2c[0][0]
bn3a_branch1 (BatchNormalizatio	(None, 19, 19, 512)	2048	res3a_branch1[0][0]
add_4 (Add)	(None, 19, 19, 512)	0	bn3a_branch2c[0][0] bn3a_branch1[0][0]
activation_13 (Activation)	(None, 19, 19, 512)	0	add_4[0][0]
res3b_branch2a (Conv2D)	(None, 19, 19, 128)	65664	activation_13[0][0]
bn3b_branch2a (BatchNormalizati	(None, 19, 19, 128)	512	res3b_branch2a[0][0]
activation_14 (Activation)	(None, 19, 19, 128)	0	bn3b_branch2a[0][0]
res3b_branch2b (Conv2D)	(None, 19, 19, 128)	147584	activation_14[0][0]

bn3b_branch2b (BatchNormalizati	(None, 19, 19, 128)	512	res3b_branch2b[0][0]
activation_15 (Activation)	(None, 19, 19, 128)	0	bn3b_branch2b[0][0]
res3b_branch2c (Conv2D)	(None, 19, 19, 512)	66048	activation_15[0][0]
bn3b_branch2c (BatchNormalizati	(None, 19, 19, 512)	2048	res3b_branch2c[0][0]
add_5 (Add)	(None, 19, 19, 512)	0	bn3b_branch2c[0][0] activation_13[0][0]
activation_16 (Activation)	(None, 19, 19, 512)	0	add_5[0][0]
res3c_branch2a (Conv2D)	(None, 19, 19, 128)	65664	activation_16[0][0]
bn3c_branch2a (BatchNormalizati	(None, 19, 19, 128)	512	res3c_branch2a[0][0]
activation_17 (Activation)	(None, 19, 19, 128)	0	bn3c_branch2a[0][0]
res3c_branch2b (Conv2D)	(None, 19, 19, 128)	147584	activation_17[0][0]
bn3c_branch2b (BatchNormalizati	(None, 19, 19, 128)	512	res3c_branch2b[0][0]
activation_18 (Activation)	(None, 19, 19, 128)	0	bn3c_branch2b[0][0]
res3c_branch2c (Conv2D)	(None, 19, 19, 512)	66048	activation_18[0][0]
bn3c_branch2c (BatchNormalizati	(None, 19, 19, 512)	2048	res3c_branch2c[0][0]
add_6 (Add)	(None, 19, 19, 512)	0	bn3c_branch2c[0][0] activation_16[0][0]
activation_19 (Activation)	(None, 19, 19, 512)	0	add_6[0][0]
res3d_branch2a (Conv2D)	(None, 19, 19, 128)	65664	activation_19[0][0]
bn3d_branch2a (BatchNormalizati	(None, 19, 19, 128)	512	res3d_branch2a[0][0]
activation_20 (Activation)	(None, 19, 19, 128)	0	bn3d_branch2a[0][0]
res3d_branch2b (Conv2D)	(None, 19, 19, 128)	147584	activation_20[0][0]
bn3d_branch2b (BatchNormalizati	(None, 19, 19, 128)	512	res3d_branch2b[0][0]
activation_21 (Activation)	(None, 19, 19, 128)	0	bn3d_branch2b[0][0]
res3d_branch2c (Conv2D)	(None, 19, 19, 512)	66048	activation_21[0][0]

bn3d_branch2c (BatchNormalizati	(None, 19, 19, 512)	2048	res3d_branch2c[0][0]
add_7 (Add)	(None, 19, 19, 512)	0	bn3d_branch2c[0][0] activation_19[0][0]
activation_22 (Activation)	(None, 19, 19, 512)	0	add_7[0][0]
res4a_branch2a (Conv2D)	(None, 10, 10, 256)	131328	activation_22[0][0]
bn4a_branch2a (BatchNormalizati	(None, 10, 10, 256)	1024	res4a_branch2a[0][0]
activation_23 (Activation)	(None, 10, 10, 256)	0	bn4a_branch2a[0][0]
res4a_branch2b (Conv2D)	(None, 10, 10, 256)	590080	activation_23[0][0]
bn4a_branch2b (BatchNormalizati	(None, 10, 10, 256)	1024	res4a_branch2b[0][0]
activation_24 (Activation)	(None, 10, 10, 256)	0	bn4a_branch2b[0][0]
res4a_branch2c (Conv2D)	(None, 10, 10, 1024)	263168	activation_24[0][0]
res4a_branch1 (Conv2D)	(None, 10, 10, 1024)	525312	activation_22[0][0]
bn4a_branch2c (BatchNormalizati	(None, 10, 10, 1024)	4096	res4a_branch2c[0][0]
bn4a_branch1 (BatchNormalizatio	(None, 10, 10, 1024)	4096	res4a_branch1[0][0]
add_8 (Add)	(None, 10, 10, 1024)	0	bn4a_branch2c[0][0] bn4a_branch1[0][0]
activation_25 (Activation)	(None, 10, 10, 1024)	0	add_8[0][0]
res4b_branch2a (Conv2D)	(None, 10, 10, 256)	262400	activation_25[0][0]
bn4b_branch2a (BatchNormalizati	(None, 10, 10, 256)	1024	res4b_branch2a[0][0]
activation_26 (Activation)	(None, 10, 10, 256)	0	bn4b_branch2a[0][0]
res4b_branch2b (Conv2D)	(None, 10, 10, 256)	590080	activation_26[0][0]
bn4b_branch2b (BatchNormalizati	(None, 10, 10, 256)	1024	res4b_branch2b[0][0]
activation_27 (Activation)	(None, 10, 10, 256)	0	bn4b_branch2b[0][0]
res4b_branch2c (Conv2D)	(None, 10, 10, 1024)	263168	activation_27[0][0]
bn4b_branch2c (BatchNormalizati	(None, 10, 10, 1024)	4096	res4b_branch2c[0][0]

add_9 (Add)	(None, 10, 10, 1024) 0	bn4b_branch2c[0][0] activation_25[0][0]
activation_28 (Activation)	(None, 10, 10, 1024) 0	add_9[0][0]
res4c_branch2a (Conv2D)	(None, 10, 10, 256) 262400	activation_28[0][0]
bn4c_branch2a (BatchNormalizati	(None, 10, 10, 256) 1024	res4c_branch2a[0][0]
activation_29 (Activation)	(None, 10, 10, 256) 0	bn4c_branch2a[0][0]
res4c_branch2b (Conv2D)	(None, 10, 10, 256) 590080	activation_29[0][0]
bn4c_branch2b (BatchNormalizati	(None, 10, 10, 256) 1024	res4c_branch2b[0][0]
activation_30 (Activation)	(None, 10, 10, 256) 0	bn4c_branch2b[0][0]
res4c_branch2c (Conv2D)	(None, 10, 10, 1024) 263168	activation_30[0][0]
bn4c_branch2c (BatchNormalizati	(None, 10, 10, 1024) 4096	res4c_branch2c[0][0]
add_10 (Add)	(None, 10, 10, 1024) 0	bn4c_branch2c[0][0] activation_28[0][0]
activation_31 (Activation)	(None, 10, 10, 1024) 0	add_10[0][0]
res4d_branch2a (Conv2D)	(None, 10, 10, 256) 262400	activation_31[0][0]
bn4d_branch2a (BatchNormalizati	(None, 10, 10, 256) 1024	res4d_branch2a[0][0]
activation_32 (Activation)	(None, 10, 10, 256) 0	bn4d_branch2a[0][0]
res4d_branch2b (Conv2D)	(None, 10, 10, 256) 590080	activation_32[0][0]
bn4d_branch2b (BatchNormalizati	(None, 10, 10, 256) 1024	res4d_branch2b[0][0]
activation_33 (Activation)	(None, 10, 10, 256) 0	bn4d_branch2b[0][0]
res4d_branch2c (Conv2D)	(None, 10, 10, 1024) 263168	activation_33[0][0]
bn4d_branch2c (BatchNormalizati	(None, 10, 10, 1024) 4096	res4d_branch2c[0][0]
add_11 (Add)	(None, 10, 10, 1024) 0	bn4d_branch2c[0][0] activation_31[0][0]
activation_34 (Activation)	(None, 10, 10, 1024) 0	add_11[0][0]
res4e_branch2a (Conv2D)	(None, 10, 10, 256) 262400	activation_34[0][0]

bn4e_branch2a (BatchNormalizati	(None, 10, 10, 256)	1024	res4e_branch2a[0][0]
activation_35 (Activation)	(None, 10, 10, 256)	0	bn4e_branch2a[0][0]
res4e_branch2b (Conv2D)	(None, 10, 10, 256)	590080	activation_35[0][0]
bn4e_branch2b (BatchNormalizati	(None, 10, 10, 256)	1024	res4e_branch2b[0][0]
activation_36 (Activation)	(None, 10, 10, 256)	0	bn4e_branch2b[0][0]
res4e_branch2c (Conv2D)	(None, 10, 10, 1024)	263168	activation_36[0][0]
bn4e_branch2c (BatchNormalizati	(None, 10, 10, 1024)	4096	res4e_branch2c[0][0]
add_12 (Add)	(None, 10, 10, 1024)	0	bn4e_branch2c[0][0] activation_34[0][0]
activation_37 (Activation)	(None, 10, 10, 1024)	0	add_12[0][0]
res4f_branch2a (Conv2D)	(None, 10, 10, 256)	262400	activation_37[0][0]
bn4f_branch2a (BatchNormalizati	(None, 10, 10, 256)	1024	res4f_branch2a[0][0]
activation_38 (Activation)	(None, 10, 10, 256)	0	bn4f_branch2a[0][0]
res4f_branch2b (Conv2D)	(None, 10, 10, 256)	590080	activation_38[0][0]
bn4f_branch2b (BatchNormalizati	(None, 10, 10, 256)	1024	res4f_branch2b[0][0]
activation_39 (Activation)	(None, 10, 10, 256)	0	bn4f_branch2b[0][0]
res4f_branch2c (Conv2D)	(None, 10, 10, 1024)	263168	activation_39[0][0]
bn4f_branch2c (BatchNormalizati	(None, 10, 10, 1024)	4096	res4f_branch2c[0][0]
add_13 (Add)	(None, 10, 10, 1024)	0	bn4f_branch2c[0][0] activation_37[0][0]
activation_40 (Activation)	(None, 10, 10, 1024)	0	add_13[0][0]
res5a_branch2a (Conv2D)	(None, 5, 5, 512)	524800	activation_40[0][0]
bn5a_branch2a (BatchNormalizati	(None, 5, 5, 512)	2048	res5a_branch2a[0][0]
activation_41 (Activation)	(None, 5, 5, 512)	0	bn5a_branch2a[0][0]
res5a_branch2b (Conv2D)	(None, 5, 5, 512)	2359808	activation_41[0][0]

bn5a_branch2b (BatchNormalizati	(None, 5, 5, 512)	2048	res5a_branch2b[0][0]
activation_42 (Activation)	(None, 5, 5, 512)	0	bn5a_branch2b[0][0]
res5a_branch2c (Conv2D)	(None, 5, 5, 2048)	1050624	activation_42[0][0]
res5a_branch1 (Conv2D)	(None, 5, 5, 2048)	2099200	activation_40[0][0]
bn5a_branch2c (BatchNormalizati	(None, 5, 5, 2048)	8192	res5a_branch2c[0][0]
bn5a_branch1 (BatchNormalizatio	(None, 5, 5, 2048)	8192	res5a_branch1[0][0]
add_14 (Add)	(None, 5, 5, 2048)	0	bn5a_branch2c[0][0] bn5a_branch1[0][0]
activation_43 (Activation)	(None, 5, 5, 2048)	0	add_14[0][0]
res5b_branch2a (Conv2D)	(None, 5, 5, 512)	1049088	activation_43[0][0]
bn5b_branch2a (BatchNormalizati	(None, 5, 5, 512)	2048	res5b_branch2a[0][0]
activation_44 (Activation)	(None, 5, 5, 512)	0	bn5b_branch2a[0][0]
res5b_branch2b (Conv2D)	(None, 5, 5, 512)	2359808	activation_44[0][0]
bn5b_branch2b (BatchNormalizati	(None, 5, 5, 512)	2048	res5b_branch2b[0][0]
activation_45 (Activation)	(None, 5, 5, 512)	0	bn5b_branch2b[0][0]
res5b_branch2c (Conv2D)	(None, 5, 5, 2048)	1050624	activation_45[0][0]
bn5b_branch2c (BatchNormalizati	(None, 5, 5, 2048)	8192	res5b_branch2c[0][0]
add_15 (Add)	(None, 5, 5, 2048)	0	bn5b_branch2c[0][0] activation_43[0][0]
activation_46 (Activation)	(None, 5, 5, 2048)	0	add_15[0][0]
res5c_branch2a (Conv2D)	(None, 5, 5, 512)	1049088	activation_46[0][0]
bn5c_branch2a (BatchNormalizati	(None, 5, 5, 512)	2048	res5c_branch2a[0][0]
activation_47 (Activation)	(None, 5, 5, 512)	0	bn5c_branch2a[0][0]
res5c_branch2b (Conv2D)	(None, 5, 5, 512)	2359808	activation_47[0][0]
bn5c_branch2b (BatchNormalizati	(None, 5, 5, 512)	2048	res5c_branch2b[0][0]

activation_48 (Activation)	(None, 5, 5, 512)	0	bn5c_branch2b[0][0]
res5c_branch2c (Conv2D)	(None, 5, 5, 2048)	1050624	activation_48[0][0]
bn5c_branch2c (BatchNormalizati	(None, 5, 5, 2048)	8192	res5c_branch2c[0][0]
add_16 (Add)	(None, 5, 5, 2048)	0	bn5c_branch2c[0][0] activation_46[0][0]
activation_49 (Activation)	(None, 5, 5, 2048)	0	add_16[0][0]

Total params: 23,587,712
 Trainable params: 23,534,592
 Non-trainable params: 53,120

```

In [4]: # 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()
  
```

Layer (type)	Output Shape	Param #
resnet50 (Model)	(None, 5, 5, 2048)	23587712
flatten_1 (Flatten)	(None, 51200)	0
dense_1 (Dense)	(None, 256)	13107456
dense_2 (Dense)	(None, 1)	257

Total params: 36,695,425
 Trainable params: 36,642,305
 Non-trainable params: 53,120

```

In [5]: 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

Number of trainable weights before freezing: 216
  
```

```

In [6]: # 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, y_pred):
#     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')

```

W0215 16:32:30.244230 139777111516992 deprecation_wrapper.py:119] From /home/mlab/anaconda3/lib/python

W0215 16:32:30.250418 139777111516992 deprecation.py:323] From /home/mlab/anaconda3/lib/python3.7/site-p

Instructions for updating:

Use tf.where in 2.0, which has the same broadcast rule as np.where

After freezing: 4

```

In [7]: 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,
    target_size=(150, 150),
    batch_size=batch_size,
    class_mode='binary')
```

Found 44938 images belonging to 2 classes.

Found 5616 images belonging to 2 classes.

```
In [8]: true_classes = train_generator.classes
        print(true_classes)
        class_labels = list(train_generator.class_indices.keys())
        print(class_labels)
```

```
[0 0 0 ... 1 1 1]
['high', 'low']
```

```
In [9]: # 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)
```

```
Epoch 1/8
351/351 [=====] - 1117s 3s/step - loss: 0.5311 - acc: 0.7401 - val_loss: 0.3625
Epoch 2/8
351/351 [=====] - 1082s 3s/step - loss: 0.3625 - acc: 0.8261 - val_loss: 0.3120
Epoch 3/8
351/351 [=====] - 1085s 3s/step - loss: 0.3120 - acc: 0.8561 - val_loss: 0.2779
Epoch 4/8
351/351 [=====] - 1086s 3s/step - loss: 0.2779 - acc: 0.8760 - val_loss: 0.2471
Epoch 5/8
351/351 [=====] - 1087s 3s/step - loss: 0.2471 - acc: 0.8895 - val_loss: 0.2231
Epoch 6/8
351/351 [=====] - 1095s 3s/step - loss: 0.2231 - acc: 0.9041 - val_loss: 0.2082
Epoch 7/8
351/351 [=====] - 1114s 3s/step - loss: 0.2082 - acc: 0.9117 - val_loss: 0.1936
Epoch 8/8
351/351 [=====] - 1121s 3s/step - loss: 0.1936 - acc: 0.9172 - val_loss: 0.1936
```

```
In [10]: true_classes_1 = validation_generator.classes
         print(true_classes)
```

```

class_labels_1 = list(validation_generator.class_indices.keys())
print(class_labels_1)

[0 0 0 ... 1 1 1]
['high', 'low']

In [11]: #Confution Matrix and Classification Report
Y_pred = model.predict_generator(validation_generator, nb_validation_samples // batch_size+1)

In [12]: # 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))

In [13]: # print(validation_generator.classes)

In [14]: confusion_matrix = metrics.confusion_matrix(true_classes_1,y_pred)
print(confusion_matrix)

[[1432 1343]
 [1473 1368]]

In [15]: report= sklearn.metrics.classification_report(true_classes_1, y_pred,
                                                    target_names = class_labels_1)
print(report)

      precision    recall  f1-score   support

 high         0.49         0.52         0.50         2775
 low          0.50         0.48         0.49         2841

 micro avg         0.50         0.50         0.50         5616
 macro avg         0.50         0.50         0.50         5616
 weighted avg         0.50         0.50         0.50         5616


In [16]: fpr, tpr, thresholds = roc_curve(validation_generator.classes, y_pred)

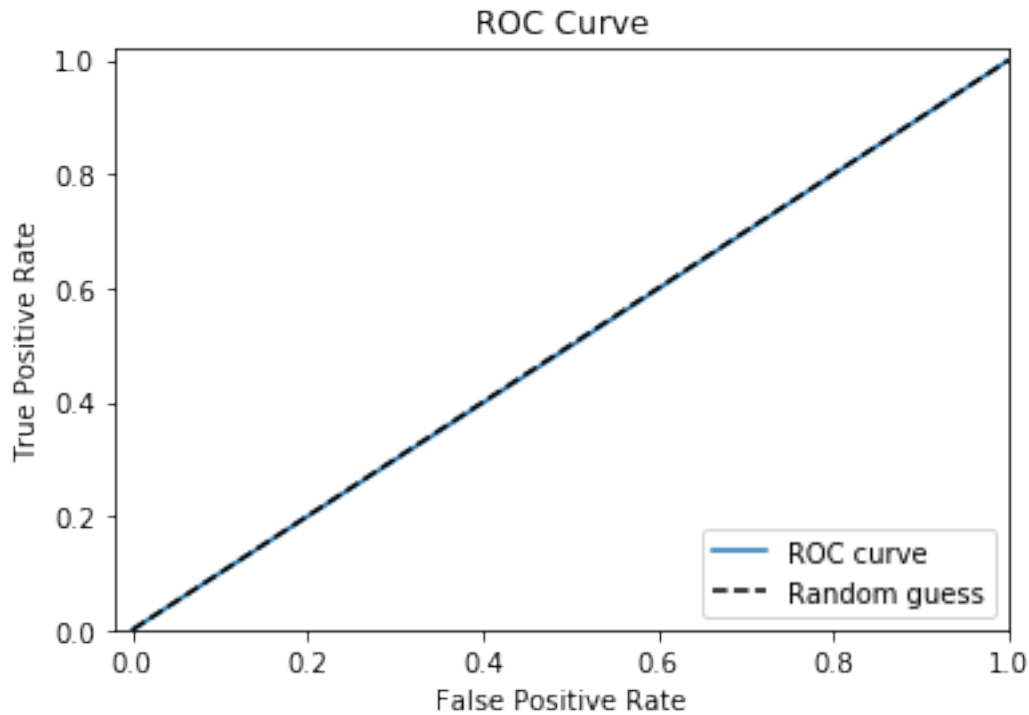
# create plot
plt.plot(fpr, tpr, label='ROC curve')
plt.plot([0, 1], [0, 1], 'k--', label='Random guess')
_ = plt.xlabel('False Positive Rate')

```

```

_ = plt.ylabel('True Positive Rate')
_ = plt.title('ROC Curve')
_ = plt.xlim([-0.02, 1])
_ = plt.ylim([0, 1.02])
_ = plt.legend(loc="lower right")

```



```
In [17]: roc_auc_score(validation_generator.classes, y_pred)
```

```
Out[17]: 0.49877831368855663
```

```

In [18]: 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))

```

```

128/128 [=====] - 3s 26ms/step
Accuracy training: 92.97
128/128 [=====] - 3s 26ms/step
Accuracy val: 87.50

```

```

In [19]: #plot the train and val curve
         #get the details from the history object

```

```

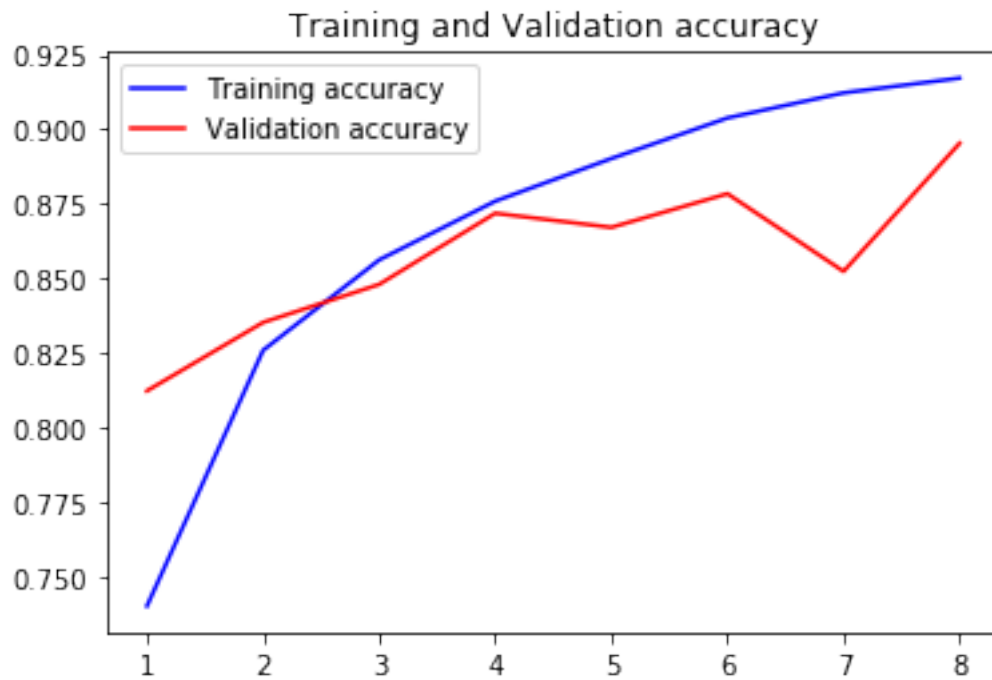
acc = history.history['acc']
val_acc=history.history['val_acc']
loss = history.history['loss']
val_loss = history.history['val_loss']

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()

```

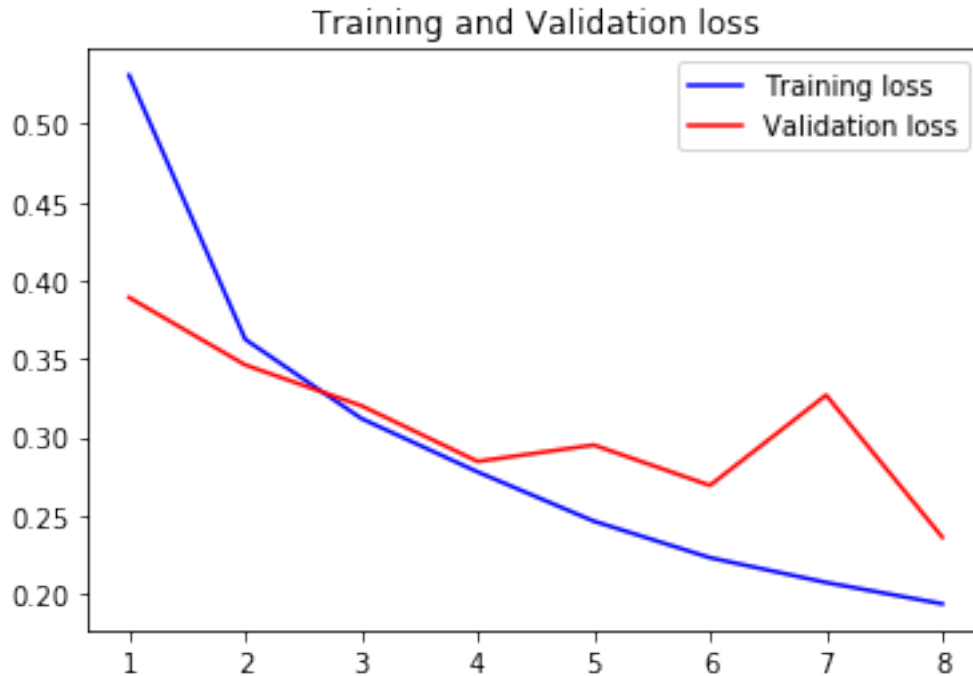
Out[19]: <matplotlib.legend.Legend at 0x7f1f4bfb4358>



```

In [20]: #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 [21]: 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)
```

Found 5619 images belonging to 2 classes.

```
In [22]: print("%s: %.2f%%" % (model.metrics_names[1], scores[1]*100))
```

acc: 88.24%

```
In [23]: #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)
```

```
In [24]: true_classes_2 = test_generator.classes
print(true_classes_2)
class_labels_2 = list(test_generator.class_indices.keys())
print(class_labels_2)
```

```
[0 0 0 ... 1 1 1]
['high', 'low']
```

```
In [25]: # 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 [26]: confusion_matrix1 = metrics.confusion_matrix(true_classes_2,y_pred1)
        print(confusion_matrix1)
```

```
[[1484 1292]
 [1495 1348]]
```

```
In [27]: report1= sklearn.metrics.classification_report(true_classes_2, y_pred1,
                                                         target_names = class_labels_2)
        print(report1)
```

	precision	recall	f1-score	support
high	0.50	0.53	0.52	2776
low	0.51	0.47	0.49	2843
micro avg	0.50	0.50	0.50	5619
macro avg	0.50	0.50	0.50	5619
weighted avg	0.50	0.50	0.50	5619

```
In [28]: print(" Loss: ", scores[0],"\n","Accuracy: ", scores[1])
```

```
Loss: 0.26137544769187304
Accuracy: 0.8824491279069767
```