```
# Processes: Using AI to improve diseases detection and location process
        # Build and train a segmentation Res-U-Net model to localise brain tumor in images
        # Talk about AI in health care
        # Deep learning
        # for the training data, I have two images; the brain MRI (to show if it has cancer or not) and a mask associat
        # First stage we are going to train a ResNet deeplearning classifier (Residual Neural Network) model
        # Second stage, if a tumor has been detected after classification, then we need to localize the tumor using Res
        # Function of the image segmentation is to understand and extract information from images at the pixel-level
In [2]: pip install opency-python
        Requirement already satisfied: opencv-python in c:\users\olayi\anaconda3\lib\site-packages (4.9.0.80)
        Requirement already satisfied: numpy>=1.21.2 in c:\users\olayi\anaconda3\lib\site-packages (from opencv-python)
        (1.24.3)
        Note: you may need to restart the kernel to use updated packages.
In [3]: # Importing important libraries
        import pandas as pd
        import numpy as np
        import seaborn as sns
        import matplotlib.pyplot as plt
        import zipfile
        import cv2
        import tensorflow as tf
        from skimage import io
        from tensorflow.python.keras import Sequential
        from tensorflow.keras import layers, optimizers
        from tensorflow.keras.applications import DenseNet121
        from tensorflow.keras.applications.resnet50 import ResNet50
        from tensorflow.keras.layers import
        from tensorflow.keras.models import Model, load_model
        from tensorflow.keras.initializers import glorot_uniform
        from tensorflow.keras.utils import plot_model
        from tensorflow.keras.callbacks import ReduceLROnPlateau, EarlyStopping, ModelCheckpoint, LearningRateScheduler
        from IPython.display import display
        from tensorflow.keras import backend as k
        from sklearn.preprocessing import StandardScaler, normalize
        import os
        import glob
        import random
        %matplotlib inline
        from warnings import filterwarnings
        filterwarnings('ignore')
        WARNING:tensorflow:From C:\Users\olayi\anaconda3\Lib\site-packages\keras\src\losses.py:2976: The name tf.losses
        .sparse softmax cross entropy is deprecated. Please use tf.compat.v1.losses.sparse softmax cross entropy instea
        d.
        C:\Users\olayi\anaconda3\Lib\site-packages\paramiko\transport.py:219: CryptographyDeprecationWarning: Blowfish
        has been deprecated
          "class": algorithms.Blowfish,
In [4]: # Define the path to the downloads directory
        downloads_path = os.path.join(os.path.expanduser('~'), 'Downloads')
        # Check if the directory exists
        if os.path.exists(r"C:\Users\olayi\Downloads\Projects\Health AI\Brain Tumor\Healthcare+AI+Datasets\Healthcare A
            print("Downloads directory found at:", downloads_path)
        else:
            print("Downloads directory not found!")
        # Change directory to the downloads folder
        os.chdir(r"C:\Users\olayi\Downloads\Projects\Health AI\Brain Tumor\Healthcare+AI+Datasets\Healthcare AI Dataset
        Downloads directory found at: C:\Users\olayi\Downloads
In [5]: # Loading the dataset that contain Brain MRI and their corresponding mask
        # loaded the mask path
        # mask can be represented by associating pixel values with their coordinates
        brain df = pd.read csv(r"C:\Users\olayi\Downloads\Projects\Health AI\Brain Tumor\Healthcare+AI+Datasets\Healthc
In [6]: brain df.info()
```

In [1]: # Task: Improve the speed and accuracy of detecting and localising Brain tumor based on MRI scans

You have been provided with 3900 brain MRI scans along with their brain tumor location

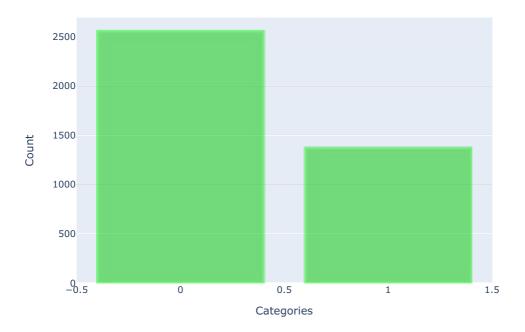
develop a model to detect and localise Brain tumor

Which will drastically remove the cost of cancer diagnosis and help in early diagnoses of brain tumors

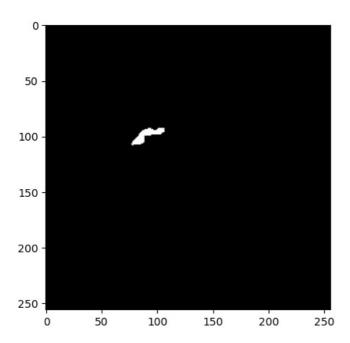
```
<class 'pandas.core.frame.DataFrame'>
                    RangeIndex: 3929 entries, 0 to 3928
                    Data columns (total 4 columns):
                      # Column
                                                       Non-Null Count Dtype
                              patient_id 3929 non-null
                      0
                                                                                           object
                            image_path 3929 non-null
                                                                                           object
                      2
                                                        3929 non-null
                            mask_path
                                                                                           obiect
                      3
                            mask
                                                        3929 non-null
                                                                                           int64
                    dtypes: int64(1), object(3)
                    memory usage: 122.9+ KB
  In [7]: brain_df.head()
                                                                                                                                                                                                                                                mask_path
                                                   patient_id
                                                                                                                                               image_path
                    0 TCGA_CS_5395_19981004 TCGA_CS_5395_19981004/TCGA_CS_5395_19981004_1.tif TCGA_CS_5395_19981004/TCGA_CS_5395_19981004_1...
                    1 TCGA_CS_5395_19981004 TCGA_CS_4944_20010208/TCGA_CS_4944_20010208_1.tif TCGA_CS_4944_20010208/TCGA_CS_4944_20010208_1....
                    2 TCGA_CS_5395_19981004 TCGA_CS_4941_19960909/TCGA_CS_4941_19960909_1.tif TCGA_CS_4941_19960909/TCGA_CS_4941_19960909_1...
                    3 TCGA_CS_5395_19981004 TCGA_CS_4943_20000902/TCGA_CS_4943_20000902_1.tif TCGA_CS_4943_20000902/TCGA_CS_4943_20000902_1...
                    4 TCGA_CS_5395_19981004 TCGA_CS_5396_20010302/TCGA_CS_5396_20010302_1.tif TCGA_CS_5396_20010302/TCGA_CS_5396_20010302_1...
  In [8]: # Percentage of healthy sampples( without tumor)
                    # which means we have unbalanced dataset
                    len(brain_df[brain_df['mask'] == 0])/len(brain_df)
                   0.6505472130313057
  Out[8]:
  In [9]: # counts of healthy samples and unhealthy samples
                    brain_df['mask'].value counts()
                               2556
                    0
  Out[9]:
                              1373
                    Name: mask, dtype: int64
                    Performing Data Exploration
                    # To know the number of categories we have in the mask
In [10]:
                    brain_df['mask'].value_counts().index
                    Int64Index([0, 1], dtype='int64')
In [11]:
                    # using plotly to show the categories
                    # plotly shows interactive bar chart
                    import plotly.graph_objects as go
                     fig = go.Figure([go.Bar(x = brain\_df['mask'].value\_counts().index, y = brain\_df['mask'].value\_counts())]) \\ fig.update\_traces(marker\_color = 'rgb(0,200,0)', marker\_line\_color = 'rgb(0,255,0)', marker\_line\_width = 3, opairs | fig. (0,200,0) |
                    fig.update_layout(title='Distribution of Categories',
                                                           xaxis title='Categories',
                                                           yaxis title='Count')
```

fig.show()

Distribution of Categories

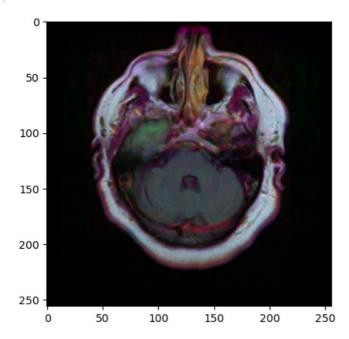


```
In [12]: # mask path
         brain_df['mask_path']
                  TCGA_CS_5395_19981004/TCGA_CS_5395_19981004_1_...
Out[12]:
                  TCGA_CS_4944_20010208/TCGA_CS_4944_20010208_1_...
                  TCGA_CS_4941_19960909/TCGA_CS_4941_19960909_1_...
                  TCGA_CS_4943_20000902/TCGA_CS_4943_20000902_1_...
TCGA_CS_5396_20010302/TCGA_CS_5396_20010302_1_...
         3
         4
         3924
                  TCGA_HT_A61B_19991127/TCGA_HT_A61B_19991127_86...
                  TCGA_HT_A61A_20000127/TCGA_HT_A61A_20000127_87...
         3925
                  TCGA_HT_A61B_19991127/TCGA_HT_A61B_19991127_87...
         3926
         3927
                  TCGA HT A61A 20000127/TCGA HT A61A 20000127 88...
                  TCGA HT A61B 19991127/TCGA HT A61B 19991127 88...
         3928
         Name: mask path, Length: 3929, dtype: object
In [13]: # Image path
         brain_df['image_path']
                  TCGA CS 5395 19981004/TCGA CS 5395 19981004 1.tif
                  TCGA_CS_4944_20010208/TCGA_CS_4944_20010208_1.tif
         2
                  TCGA_CS_4941_19960909/TCGA_CS_4941_19960909_1.tif
         3
                  TCGA_CS_4943_20000902/TCGA_CS_4943_20000902_1.tif
                  TCGA_CS_5396_20010302/TCGA_CS_5396_20010302_1.tif
         4
                  TCGA HT A61B 19991127/TCGA HT A61B 19991127 86...
         3924
                  TCGA HT A61A 20000127/TCGA HT A61A 20000127 87...
         3925
                  TCGA_HT_A61B_19991127/TCGA_HT_A61B_19991127_87...
         3926
         3927
                  TCGA_HT_A61A_20000127/TCGA_HT_A61A_20000127_88...
         3928
                  TCGA HT A61B 19991127/TCGA HT A61B 19991127 88...
         Name: image_path, Length: 3929, dtype: object
In [14]: # using openCV to read the image and plt to show the image
         plt.imshow(cv2.imread(brain_df.mask_path[623]))
         <matplotlib.image.AxesImage at 0x169bd1a2ed0>
```



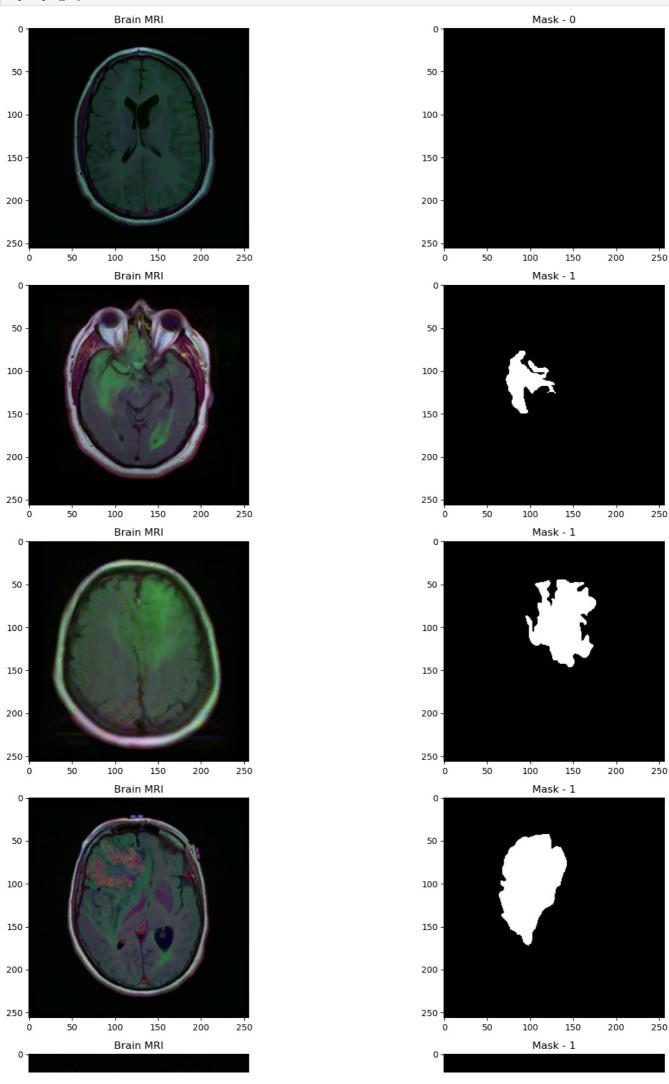
In [15]: plt.imshow(cv2.imread(brain_df.image_path[623]))

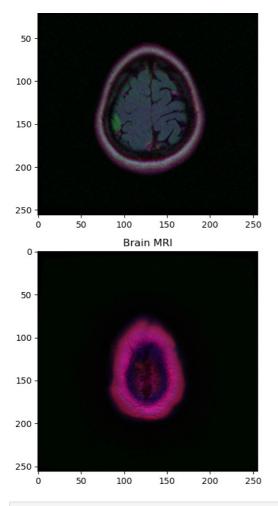
Out[15]: <matplotlib.image.AxesImage at 0x169bd1d3310>

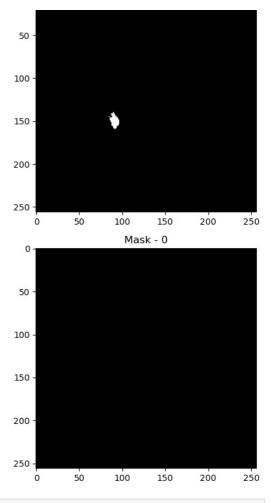


```
In [16]: # visualizing the images(MRI and mask) in the dataset

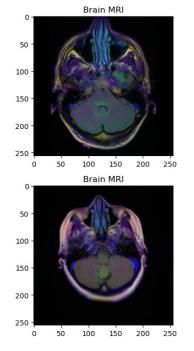
fig, axis = plt.subplots(6, 2, figsize = (16,25))
    count = 0
    for i in range(6):
        i = random.randint(0, len(brain_df)) # Select a random index
        axis[count][0].title.set_text('Brain_MRI') # Set title
        axis[count][0].imshow(cv2.imread(brain_df.image_path[i])) # MRI
        axis[count][1].title.set_text('Mask - ' + str(brain_df['mask'][i])) # Plot title on the mask (0 or 1)
        axis[count][1].imshow(cv2.imread(brain_df.mask_path[i]))
        count += 1
```

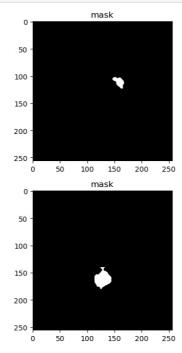


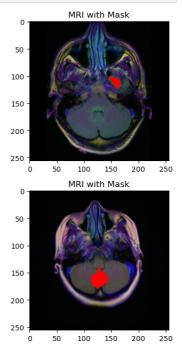


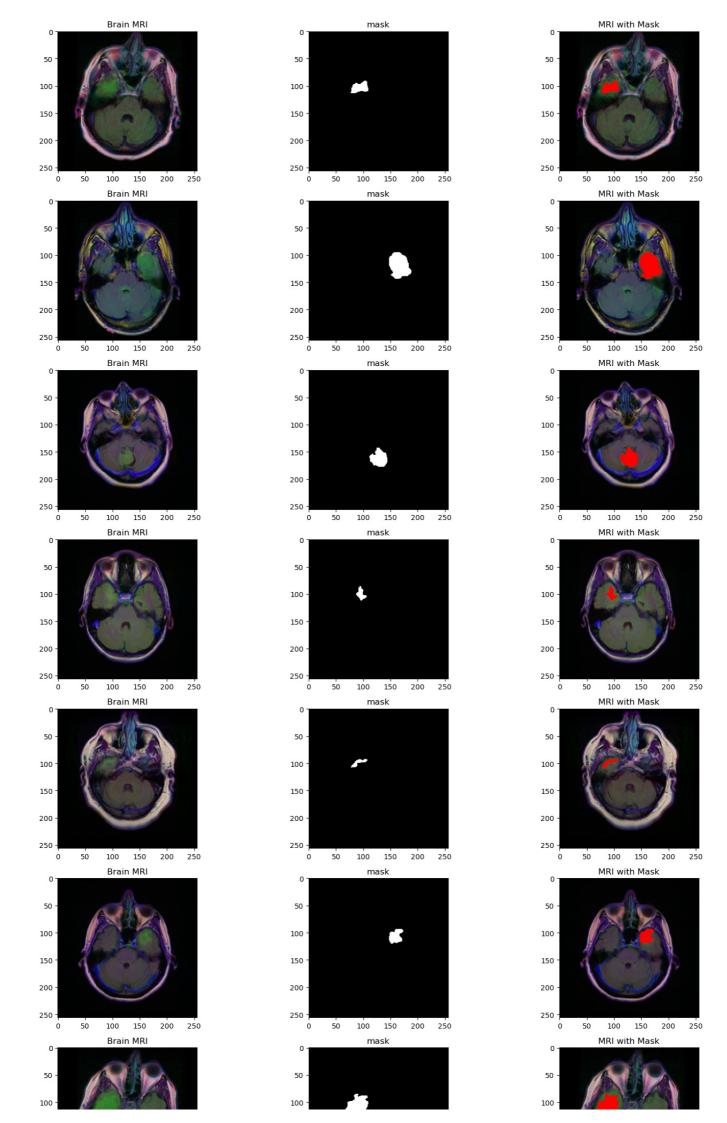


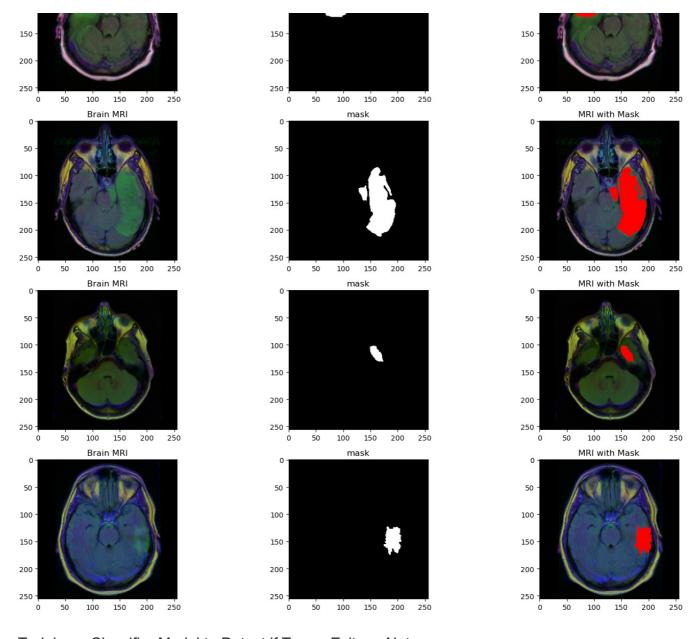
visualizing the 12 randomly selected MRI and it correspong Mask from only unhealthy patients # showing the MRI and the ,mask on top of each other (showing the mask in red color) fig, axis = plt.subplots(12, 3, figsize = (16,40)) for i in range(len(brain_df)): if brain_df['mask'][i] == 1 and count <12:</pre> img = io.imread(brain df.image path[i]) axis[count][0].title.set_text('Brain MRI') axis[count][0].imshow(img) mask = io.imread(brain_df.mask_path[i]) axis[count][1].title.set_text('mask') axis[count][1].imshow(mask, cmap = 'gray') img[mask == 255] = (255,0,0)axis[count][2].title.set_text('MRI with Mask') axis[count][2].imshow(img) count += 1fig.tight_layout()











Training a Classifier Model to Detect if Tumor Exits or Not

```
In [18]:
         # Drop columns not needed for the training
         brain_df_train = brain_df.drop(['patient_id'], axis =1)
         brain df train.shape
         (3929, 3)
Out[18]:
         # Convert the mask column to a categorical format.
         brain df train['mask'] = brain df train['mask'].apply(lambda x : str(x))
In [20]: brain df train.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 3929 entries, 0 to 3928
         Data columns (total 3 columns):
          #
                          Non-Null Count Dtype
              Column
          0
              image_path 3929 non-null
                                          object
                          3929 non-null
                                          object
          1
              mask_path
              mask
                          3929 non-null
                                          object
         dtypes: object(3)
         memory usage: 92.2+ KB
In [21]: # Split the data into train and test data
         from sklearn.model selection import train test split
         train, test = train_test_split(brain_df_train, test_size = 0.15)
         # Create a image generator to generate images in batch format and do some processing on the data
In [22]:
         from keras.preprocessing.image import ImageDataGenerator
         # Create a data generator which scales the data from 0 to 1 and makes validation split of 0.15
         datagen = ImageDataGenerator(rescale = 1./255., validation\_split = 0.15)
In [23]: train_generator = datagen.flow_from_dataframe(
```

```
dataframe = train,
directory = './',
x_col = 'image_path',
y_col = 'mask',
subset='training',
batch_size = 16,
shuffle = True,
class_mode ='categorical',
target_size = (256, 256))
val generator = datagen.flow from dataframe(
dataframe = train,
directory = './'
x_col = 'image_path',
y_col = 'mask',
subset='validation',
batch size = 16,
shuffle = True,
class_mode ='categorical',
target_size = (256, 256))
# Create a data generator for the test images
test datagen = ImageDataGenerator(rescale = 1./255.)
test_generator = datagen.flow_from_dataframe(
dataframe = test,
directory = './'
x_col = 'image_path',
y col = 'mask',
batch size = 16,
shuffle = False,
class mode ='categorical',
target_size = (256, 256))
```

Found 2839 validated image filenames belonging to 2 classes. Found 500 validated image filenames belonging to 2 classes. Found 590 validated image filenames belonging to 2 classes.

```
In [24]: # Using an already pre-trained model (Transfer learning)
# creating the ResNet50 base model
basemodel = ResNet50(weights = 'imagenet', include_top = False, input_tensor= Input(shape=(256, 256, 3)))
```

WARNING:tensorflow:From C:\Users\olayi\anaconda3\Lib\site-packages\keras\src\backend.py:1398: The name tf.execu ting_eagerly_outside_functions is deprecated. Please use tf.compat.v1.executing_eagerly_outside_functions inste ad.

WARNING:tensorflow:From C:\Users\olayi\anaconda3\Lib\site-packages\keras\src\layers\normalization\batch_normalization.py:979: The name tf.nn.fused_batch_norm is deprecated. Please use tf.compat.v1.nn.fused_batch_norm instead

In [25]: basemodel.summary()

Model: "resnet50"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 256, 256, 3)]	0	[]
<pre>conv1_pad (ZeroPadding2D)</pre>	(None, 262, 262, 3)	0	['input_1[0][0]']
conv1_conv (Conv2D)	(None, 128, 128, 64)	9472	['conv1_pad[0][0]']
<pre>conv1_bn (BatchNormalizati on)</pre>	(None, 128, 128, 64)	256	['conv1_conv[0][0]']
<pre>conv1_relu (Activation)</pre>	(None, 128, 128, 64)	0	['conv1_bn[0][0]']
<pre>pool1_pad (ZeroPadding2D)</pre>	(None, 130, 130, 64)	0	['conv1_relu[0][0]']
<pre>pool1_pool (MaxPooling2D)</pre>	(None, 64, 64, 64)	0	['pool1_pad[0][0]']
<pre>conv2_block1_1_conv (Conv2 D)</pre>	(None, 64, 64, 64)	4160	['pool1_pool[0][0]']
<pre>conv2_block1_1_bn (BatchNo rmalization)</pre>	(None, 64, 64, 64)	256	['conv2_block1_1_conv[0][0]']
<pre>conv2_block1_1_relu (Activ ation)</pre>	(None, 64, 64, 64)	0	['conv2_block1_1_bn[0][0]']
conv2_block1_2_conv (Conv2 D)	(None, 64, 64, 64)	36928	['conv2_block1_1_relu[0][0]']
<pre>conv2_block1_2_bn (BatchNo rmalization)</pre>	(None, 64, 64, 64)	256	['conv2_block1_2_conv[0][0]']
conv2_block1_2_relu (Activ	(None, 64, 64, 64)	0	['conv2_block1_2_bn[0][0]']

ation)				
<pre>conv2_block1_0_conv (Conv2 D)</pre>	(None, 64, 64,	256)	16640	['pool1_pool[0][0]']
conv2_block1_3_conv (Conv2 D)	(None, 64, 64,	256)	16640	['conv2_block1_2_relu[0][0]']
<pre>conv2_block1_0_bn (BatchNo rmalization)</pre>	(None, 64, 64,	256)	1024	['conv2_block1_0_conv[0][0]']
<pre>conv2_block1_3_bn (BatchNo rmalization)</pre>	(None, 64, 64,	256)	1024	['conv2_block1_3_conv[0][0]']
conv2_block1_add (Add)	(None, 64, 64,	256)	0	['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']
<pre>conv2_block1_out (Activati on)</pre>	(None, 64, 64,	256)	0	['conv2_block1_add[0][0]']
<pre>conv2_block2_1_conv (Conv2 D)</pre>	(None, 64, 64,	64)	16448	['conv2_block1_out[0][0]']
<pre>conv2_block2_1_bn (BatchNo rmalization)</pre>	(None, 64, 64,	64)	256	['conv2_block2_1_conv[0][0]']
<pre>conv2_block2_1_relu (Activ ation)</pre>	(None, 64, 64,	64)	0	['conv2_block2_1_bn[0][0]']
conv2_block2_2_conv (Conv2 D)	(None, 64, 64,	64)	36928	['conv2_block2_1_relu[0][0]']
<pre>conv2_block2_2_bn (BatchNo rmalization)</pre>	(None, 64, 64,	64)	256	['conv2_block2_2_conv[0][0]']
<pre>conv2_block2_2_relu (Activ ation)</pre>	(None, 64, 64,	64)	0	['conv2_block2_2_bn[0][0]']
conv2_block2_3_conv (Conv2 D)	(None, 64, 64,	256)	16640	['conv2_block2_2_relu[0][0]']
<pre>conv2_block2_3_bn (BatchNo rmalization)</pre>	(None, 64, 64,	256)	1024	['conv2_block2_3_conv[0][0]']
conv2_block2_add (Add)	(None, 64, 64,	256)	Θ	['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']
<pre>conv2_block2_out (Activati on)</pre>	(None, 64, 64,	256)	0	['conv2_block2_add[0][0]']
conv2_block3_1_conv (Conv2 D)	(None, 64, 64,	64)	16448	['conv2_block2_out[0][0]']
<pre>conv2_block3_1_bn (BatchNo rmalization)</pre>	(None, 64, 64,	64)	256	['conv2_block3_1_conv[0][0]']
<pre>conv2_block3_1_relu (Activ ation)</pre>	(None, 64, 64,	64)	0	['conv2_block3_1_bn[0][0]']
conv2_block3_2_conv (Conv2 D)	(None, 64, 64,	64)	36928	['conv2_block3_1_relu[0][0]']
<pre>conv2_block3_2_bn (BatchNo rmalization)</pre>	(None, 64, 64,	64)	256	['conv2_block3_2_conv[0][0]']
<pre>conv2_block3_2_relu (Activ ation)</pre>	(None, 64, 64,	64)	0	['conv2_block3_2_bn[0][0]']
conv2_block3_3_conv (Conv2 D)	(None, 64, 64,	256)	16640	['conv2_block3_2_relu[0][0]']
<pre>conv2_block3_3_bn (BatchNo rmalization)</pre>	(None, 64, 64,	256)	1024	['conv2_block3_3_conv[0][0]']
conv2_block3_add (Add)	(None, 64, 64,	256)	0	['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']
<pre>conv2_block3_out (Activati on)</pre>	(None, 64, 64,	256)	0	['conv2_block3_add[0][0]']
conv3_block1_1_conv (Conv2 D)	(None, 32, 32,	128)	32896	['conv2_block3_out[0][0]']
<pre>conv3_block1_1_bn (BatchNo rmalization)</pre>	(None, 32, 32,	128)	512	['conv3_block1_1_conv[0][0]']
<pre>conv3_block1_1_relu (Activ ation)</pre>	(None, 32, 32,	128)	0	['conv3_block1_1_bn[0][0]']

conv3_block1_2_conv (Conv2 D)	(None, 32, 32, 128)	147584	['conv3_block1_1_relu[0][0]']
<pre>conv3_block1_2_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block1_2_conv[0][0]']
<pre>conv3_block1_2_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block1_2_bn[0][0]']
<pre>conv3_block1_0_conv (Conv2 D)</pre>	(None, 32, 32, 512)	131584	['conv2_block3_out[0][0]']
conv3_block1_3_conv (Conv2 D)	(None, 32, 32, 512)	66048	['conv3_block1_2_relu[0][0]']
<pre>conv3_block1_0_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block1_0_conv[0][0]']
<pre>conv3_block1_3_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block1_3_conv[0][0]']
conv3_block1_add (Add)	(None, 32, 32, 512)	0	['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']
<pre>conv3_block1_out (Activati on)</pre>	(None, 32, 32, 512)	0	['conv3_block1_add[0][0]']
conv3_block2_1_conv (Conv2 D)	(None, 32, 32, 128)	65664	['conv3_block1_out[0][0]']
<pre>conv3_block2_1_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block2_1_conv[0][0]']
<pre>conv3_block2_1_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block2_1_bn[0][0]']
conv3_block2_2_conv (Conv2 D)	(None, 32, 32, 128)	147584	['conv3_block2_1_relu[0][0]']
<pre>conv3_block2_2_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block2_2_conv[0][0]']
<pre>conv3_block2_2_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block2_2_bn[0][0]']
conv3_block2_3_conv (Conv2 D)	(None, 32, 32, 512)	66048	['conv3_block2_2_relu[0][0]']
<pre>conv3_block2_3_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block2_3_conv[0][0]']
conv3_block2_add (Add)	(None, 32, 32, 512)	0	['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']
<pre>conv3_block2_out (Activati on)</pre>	(None, 32, 32, 512)	0	['conv3_block2_add[0][0]']
conv3_block3_1_conv (Conv2 D)	(None, 32, 32, 128)	65664	['conv3_block2_out[0][0]']
<pre>conv3_block3_1_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block3_1_conv[0][0]']
<pre>conv3_block3_1_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block3_1_bn[0][0]']
conv3_block3_2_conv (Conv2 D)	(None, 32, 32, 128)	147584	['conv3_block3_1_relu[0][0]']
<pre>conv3_block3_2_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block3_2_conv[0][0]']
<pre>conv3_block3_2_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block3_2_bn[0][0]']
conv3_block3_3_conv (Conv2 D)	(None, 32, 32, 512)	66048	['conv3_block3_2_relu[0][0]']
<pre>conv3_block3_3_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block3_3_conv[0][0]']
conv3_block3_add (Add)	(None, 32, 32, 512)	0	['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']
<pre>conv3_block3_out (Activati on)</pre>	(None, 32, 32, 512)	0	['conv3_block3_add[0][0]']
conv3_block4_1_conv (Conv2 D)	(None, 32, 32, 128)	65664	['conv3_block3_out[0][0]']

<pre>conv3_block4_1_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block4_1_conv[0][0]']
<pre>conv3_block4_1_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block4_1_bn[0][0]']
conv3_block4_2_conv (Conv2 D)	(None, 32, 32, 128)	147584	['conv3_block4_1_relu[0][0]']
<pre>conv3_block4_2_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block4_2_conv[0][0]']
<pre>conv3_block4_2_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block4_2_bn[0][0]']
conv3_block4_3_conv (Conv2 D)	(None, 32, 32, 512)	66048	['conv3_block4_2_relu[0][0]']
<pre>conv3_block4_3_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block4_3_conv[0][0]']
conv3_block4_add (Add)	(None, 32, 32, 512)	0	['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']
<pre>conv3_block4_out (Activati on)</pre>	(None, 32, 32, 512)	0	['conv3_block4_add[0][0]']
<pre>conv4_block1_1_conv (Conv2 D)</pre>	(None, 16, 16, 256)	131328	['conv3_block4_out[0][0]']
<pre>conv4_block1_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block1_1_conv[0][0]']
<pre>conv4_block1_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block1_1_bn[0][0]']
conv4_block1_2_conv (Conv2 D)	(None, 16, 16, 256)	590080	['conv4_block1_1_relu[0][0]']
<pre>conv4_block1_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block1_2_conv[0][0]']
<pre>conv4_block1_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block1_2_bn[0][0]']
conv4_block1_0_conv (Conv2 D)	(None, 16, 16, 1024)	525312	['conv3_block4_out[0][0]']
conv4_block1_3_conv (Conv2 D)	(None, 16, 16, 1024)	263168	['conv4_block1_2_relu[0][0]']
<pre>conv4_block1_0_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block1_0_conv[0][0]']
<pre>conv4_block1_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block1_3_conv[0][0]']
conv4_block1_add (Add)	(None, 16, 16, 1024)	0	['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']
<pre>conv4_block1_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block1_add[0][0]']
<pre>conv4_block2_1_conv (Conv2 D)</pre>	(None, 16, 16, 256)	262400	['conv4_block1_out[0][0]']
<pre>conv4_block2_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block2_1_conv[0][0]']
<pre>conv4_block2_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block2_1_bn[0][0]']
conv4_block2_2_conv (Conv2 D)	(None, 16, 16, 256)	590080	['conv4_block2_1_relu[0][0]']
<pre>conv4_block2_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block2_2_conv[0][0]']
<pre>conv4_block2_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block2_2_bn[0][0]']
conv4_block2_3_conv (Conv2 D)	(None, 16, 16, 1024)	263168	['conv4_block2_2_relu[0][0]']
<pre>conv4_block2_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block2_3_conv[0][0]']
conv4_block2_add (Add)	(None, 16, 16, 1024)	0	['conv4_block1_out[0][0]',

			'conv4_block2_3_bn[0][0]']
<pre>conv4_block2_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block2_add[0][0]']
<pre>conv4_block3_1_conv (Conv2 D)</pre>	(None, 16, 16, 256)	262400	['conv4_block2_out[0][0]']
<pre>conv4_block3_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block3_1_conv[0][0]']
<pre>conv4_block3_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block3_1_bn[0][0]']
conv4_block3_2_conv (Conv2 D)	(None, 16, 16, 256)	590080	['conv4_block3_1_relu[0][0]']
<pre>conv4_block3_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block3_2_conv[0][0]']
<pre>conv4_block3_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block3_2_bn[0][0]']
conv4_block3_3_conv (Conv2 D)	(None, 16, 16, 1024)	263168	['conv4_block3_2_relu[0][0]']
<pre>conv4_block3_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block3_3_conv[0][0]']
conv4_block3_add (Add)	(None, 16, 16, 1024)	0	['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']
<pre>conv4_block3_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block3_add[0][0]']
conv4_block4_1_conv (Conv2 D)	(None, 16, 16, 256)	262400	['conv4_block3_out[0][0]']
<pre>conv4_block4_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block4_1_conv[0][0]']
<pre>conv4_block4_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block4_1_bn[0][0]']
conv4_block4_2_conv (Conv2 D)	(None, 16, 16, 256)	590080	['conv4_block4_1_relu[0][0]']
<pre>conv4_block4_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block4_2_conv[0][0]']
<pre>conv4_block4_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block4_2_bn[0][0]']
conv4_block4_3_conv (Conv2 D)	(None, 16, 16, 1024)	263168	['conv4_block4_2_relu[0][0]']
<pre>conv4_block4_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block4_3_conv[0][0]']
conv4_block4_add (Add)	(None, 16, 16, 1024)	0	['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']
<pre>conv4_block4_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block4_add[0][0]']
conv4_block5_1_conv (Conv2 D)	(None, 16, 16, 256)	262400	['conv4_block4_out[0][0]']
<pre>conv4_block5_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block5_1_conv[0][0]']
<pre>conv4_block5_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block5_1_bn[0][0]']
conv4_block5_2_conv (Conv2 D)	(None, 16, 16, 256)	590080	['conv4_block5_1_relu[0][0]']
<pre>conv4_block5_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block5_2_conv[0][0]']
<pre>conv4_block5_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block5_2_bn[0][0]']
conv4_block5_3_conv (Conv2 D)	(None, 16, 16, 1024)	263168	['conv4_block5_2_relu[0][0]']
<pre>conv4_block5_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block5_3_conv[0][0]']

conv4_block5_add (Add)	(None, 16, 16, 1024)	Θ	['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']
<pre>conv4_block5_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block5_add[0][0]']
<pre>conv4_block6_1_conv (Conv2 D)</pre>	(None, 16, 16, 256)	262400	['conv4_block5_out[0][0]']
<pre>conv4_block6_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block6_1_conv[0][0]']
<pre>conv4_block6_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block6_1_bn[0][0]']
conv4_block6_2_conv (Conv2 D)	(None, 16, 16, 256)	590080	['conv4_block6_1_relu[0][0]']
<pre>conv4_block6_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block6_2_conv[0][0]']
<pre>conv4_block6_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block6_2_bn[0][0]']
conv4_block6_3_conv (Conv2 D)	(None, 16, 16, 1024)	263168	['conv4_block6_2_relu[0][0]']
<pre>conv4_block6_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block6_3_conv[0][0]']
conv4_block6_add (Add)	(None, 16, 16, 1024)	Θ	['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']
<pre>conv4_block6_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block6_add[0][0]']
conv5_block1_1_conv (Conv2 D)	(None, 8, 8, 512)	524800	['conv4_block6_out[0][0]']
<pre>conv5_block1_1_bn (BatchNo rmalization)</pre>	(None, 8, 8, 512)	2048	['conv5_block1_1_conv[0][0]']
<pre>conv5_block1_1_relu (Activ ation)</pre>	(None, 8, 8, 512)	0	['conv5_block1_1_bn[0][0]']
conv5_block1_2_conv (Conv2 D)	(None, 8, 8, 512)	2359808	['conv5_block1_1_relu[0][0]']
<pre>conv5_block1_2_bn (BatchNo rmalization)</pre>	(None, 8, 8, 512)	2048	['conv5_block1_2_conv[0][0]']
<pre>conv5_block1_2_relu (Activ ation)</pre>	(None, 8, 8, 512)	Θ	['conv5_block1_2_bn[0][0]']
conv5_block1_0_conv (Conv2 D)	(None, 8, 8, 2048)	2099200	['conv4_block6_out[0][0]']
conv5_block1_3_conv (Conv2 D)	(None, 8, 8, 2048)	1050624	['conv5_block1_2_relu[0][0]']
<pre>conv5_block1_0_bn (BatchNo rmalization)</pre>	(None, 8, 8, 2048)	8192	['conv5_block1_0_conv[0][0]']
<pre>conv5_block1_3_bn (BatchNo rmalization)</pre>	(None, 8, 8, 2048)	8192	['conv5_block1_3_conv[0][0]']
conv5_block1_add (Add)	(None, 8, 8, 2048)	0	['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']
<pre>conv5_block1_out (Activati on)</pre>	(None, 8, 8, 2048)	0	['conv5_block1_add[0][0]']
conv5_block2_1_conv (Conv2 D)	(None, 8, 8, 512)	1049088	['conv5_block1_out[0][0]']
<pre>conv5_block2_1_bn (BatchNo rmalization)</pre>	(None, 8, 8, 512)	2048	['conv5_block2_1_conv[0][0]']
<pre>conv5_block2_1_relu (Activ ation)</pre>	(None, 8, 8, 512)	Θ	['conv5_block2_1_bn[0][0]']
conv5_block2_2_conv (Conv2 D)	(None, 8, 8, 512)	2359808	['conv5_block2_1_relu[0][0]']
<pre>conv5_block2_2_bn (BatchNo rmalization)</pre>	(None, 8, 8, 512)	2048	['conv5_block2_2_conv[0][0]']
<pre>conv5_block2_2_relu (Activ ation)</pre>	(None, 8, 8, 512)	Θ	['conv5_block2_2_bn[0][0]']

```
1050624
conv5_block2_3_conv (Conv2_(None, 8, 8, 2048)
                                                                     ['conv5 block2 2 relu[0][0]']
conv5 block2 3 bn (BatchNo (None, 8, 8, 2048)
                                                           8192
                                                                     ['conv5 block2 3 conv[0][0]']
rmalization)
                             (None, 8, 8, 2048)
                                                           0
                                                                     ['conv5_block1_out[0][0]'
conv5 block2 add (Add)
                                                                       conv5_block2_3_bn[0][0]']
conv5 block2 out (Activati
                             (None, 8, 8, 2048)
                                                           0
                                                                     ['conv5 block2 add[0][0]']
on)
conv5 block3 1 conv (Conv2
                             (None, 8, 8, 512)
                                                           1049088
                                                                     ['conv5 block2 out[0][0]']
conv5 block3 1 bn (BatchNo
                             (None, 8, 8, 512)
                                                           2048
                                                                     ['conv5 block3 1 conv[0][0]']
rmalization)
conv5_block3_1_relu (Activ
                             (None, 8, 8, 512)
                                                                     ['conv5_block3_1_bn[0][0]']
ation)
conv5_block3_2_conv (Conv2
                             (None, 8, 8, 512)
                                                           2359808
                                                                     ['conv5_block3_1_relu[0][0]']
conv5_block3_2_bn (BatchNo
                                                           2048
                                                                     ['conv5_block3_2_conv[0][0]']
                             (None, 8, 8, 512)
rmalization)
conv5 block3 2 relu (Activ
                             (None, 8, 8, 512)
                                                                     ['conv5 block3 2 bn[0][0]']
ation)
conv5 block3 3 conv (Conv2 (None, 8, 8, 2048)
                                                           1050624
                                                                     ['conv5_block3_2_relu[0][0]']
                                                           8192
conv5 block3 3 bn (BatchNo
                             (None, 8, 8, 2048)
                                                                     ['conv5 block3 3 conv[0][0]']
rmalization)
conv5 block3 add (Add)
                             (None, 8, 8, 2048)
                                                                     ['conv5 block2 out[0][0]'
                                                                       conv5 block3 3 bn[0][0]']
conv5_block3_out (Activati (None, 8, 8, 2048)
                                                           0
                                                                     ['conv5_block3_add[0][0]']
on)
Total params: 23587712 (89.98 MB)
```

Total params: 23587712 (89.98 MB) Trainable params: 23534592 (89.78 MB) Non-trainable params: 53120 (207.50 KB)

```
In [27]: # add a classification head to the model
```

```
headmodel = basemodel.output
headmodel = AveragePooling2D(pool_size= (4,4))(headmodel)
headmodel = Flatten(name = 'flatten')(headmodel)
headmodel = Dense(256, activation= 'relu')(headmodel)
headmodel = Dropout(0.3)(headmodel)
headmodel = Dense(256, activation= 'relu')(headmodel)
headmodel = Dropout(0.3)(headmodel)
headmodel = Dropout(0.3)(headmodel)
headmodel = Dense(2, activation= 'softmax')(headmodel)
model = Model(inputs = basemodel.input, outputs = headmodel)
```

In [28]: model.summary()

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 256, 256, 3)]	0	[]
<pre>conv1_pad (ZeroPadding2D)</pre>	(None, 262, 262, 3)	Θ	['input_1[0][0]']
conv1_conv (Conv2D)	(None, 128, 128, 64)	9472	['conv1_pad[0][0]']
<pre>conv1_bn (BatchNormalizati on)</pre>	(None, 128, 128, 64)	256	['conv1_conv[0][0]']
<pre>conv1_relu (Activation)</pre>	(None, 128, 128, 64)	0	['conv1_bn[0][0]']
<pre>pool1_pad (ZeroPadding2D)</pre>	(None, 130, 130, 64)	0	['conv1_relu[0][0]']
<pre>pool1_pool (MaxPooling2D)</pre>	(None, 64, 64, 64)	0	['pool1_pad[0][0]']
conv2_block1_1_conv (Conv2	(None, 64, 64, 64)	4160	['pool1_pool[0][0]']

D)	
----	--

D)			
<pre>conv2_block1_1_bn (BatchNo rmalization)</pre>	(None, 64, 64, 64)	256	['conv2_block1_1_conv[0][0]']
<pre>conv2_block1_1_relu (Activ ation)</pre>	(None, 64, 64, 64)	0	['conv2_block1_1_bn[0][0]']
conv2_block1_2_conv (Conv2 D)	(None, 64, 64, 64)	36928	['conv2_block1_1_relu[0][0]']
<pre>conv2_block1_2_bn (BatchNo rmalization)</pre>	(None, 64, 64, 64)	256	['conv2_block1_2_conv[0][0]']
<pre>conv2_block1_2_relu (Activ ation)</pre>	(None, 64, 64, 64)	0	['conv2_block1_2_bn[0][0]']
<pre>conv2_block1_0_conv (Conv2 D)</pre>	(None, 64, 64, 256)	16640	['pool1_pool[0][0]']
conv2_block1_3_conv (Conv2 D)	(None, 64, 64, 256)	16640	['conv2_block1_2_relu[0][0]']
<pre>conv2_block1_0_bn (BatchNo rmalization)</pre>	(None, 64, 64, 256)	1024	['conv2_block1_0_conv[0][0]']
<pre>conv2_block1_3_bn (BatchNo rmalization)</pre>	(None, 64, 64, 256)	1024	['conv2_block1_3_conv[0][0]']
conv2_block1_add (Add)	(None, 64, 64, 256)	0	['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']
<pre>conv2_block1_out (Activati on)</pre>	(None, 64, 64, 256)	0	['conv2_block1_add[0][0]']
conv2_block2_1_conv (Conv2 D)	(None, 64, 64, 64)	16448	['conv2_block1_out[0][0]']
<pre>conv2_block2_1_bn (BatchNo rmalization)</pre>	(None, 64, 64, 64)	256	['conv2_block2_1_conv[0][0]']
<pre>conv2_block2_1_relu (Activ ation)</pre>	(None, 64, 64, 64)	0	['conv2_block2_1_bn[0][0]']
conv2_block2_2_conv (Conv2 D)	(None, 64, 64, 64)	36928	['conv2_block2_1_relu[0][0]']
<pre>conv2_block2_2_bn (BatchNo rmalization)</pre>	(None, 64, 64, 64)	256	['conv2_block2_2_conv[0][0]']
<pre>conv2_block2_2_relu (Activ ation)</pre>	(None, 64, 64, 64)	0	['conv2_block2_2_bn[0][0]']
conv2_block2_3_conv (Conv2 D)	(None, 64, 64, 256)	16640	['conv2_block2_2_relu[0][0]']
<pre>conv2_block2_3_bn (BatchNo rmalization)</pre>	(None, 64, 64, 256)	1024	['conv2_block2_3_conv[0][0]']
conv2_block2_add (Add)	(None, 64, 64, 256)	0	['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']
<pre>conv2_block2_out (Activati on)</pre>	(None, 64, 64, 256)	0	['conv2_block2_add[0][0]']
conv2_block3_1_conv (Conv2 D)	(None, 64, 64, 64)	16448	['conv2_block2_out[0][0]']
<pre>conv2_block3_1_bn (BatchNo rmalization)</pre>	(None, 64, 64, 64)	256	['conv2_block3_1_conv[0][0]']
<pre>conv2_block3_1_relu (Activ ation)</pre>	(None, 64, 64, 64)	0	['conv2_block3_1_bn[0][0]']
conv2_block3_2_conv (Conv2 D)	(None, 64, 64, 64)	36928	['conv2_block3_1_relu[0][0]']
<pre>conv2_block3_2_bn (BatchNo rmalization)</pre>	(None, 64, 64, 64)	256	['conv2_block3_2_conv[0][0]']
<pre>conv2_block3_2_relu (Activ ation)</pre>	(None, 64, 64, 64)	0	['conv2_block3_2_bn[0][0]']
conv2_block3_3_conv (Conv2 D)	(None, 64, 64, 256)	16640	['conv2_block3_2_relu[0][0]']
<pre>conv2_block3_3_bn (BatchNo rmalization)</pre>	(None, 64, 64, 256)	1024	['conv2_block3_3_conv[0][0]']

conv2_block3_add (Add)	(None, 64, 64, 256)	0	['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']
<pre>conv2_block3_out (Activati on)</pre>	(None, 64, 64, 256)	0	['conv2_block3_add[0][0]']
<pre>conv3_block1_1_conv (Conv2 D)</pre>	(None, 32, 32, 128)	32896	['conv2_block3_out[0][0]']
<pre>conv3_block1_1_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block1_1_conv[0][0]']
<pre>conv3_block1_1_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block1_1_bn[0][0]']
conv3_block1_2_conv (Conv2 D)	(None, 32, 32, 128)	147584	['conv3_block1_1_relu[0][0]']
<pre>conv3_block1_2_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block1_2_conv[0][0]']
<pre>conv3_block1_2_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block1_2_bn[0][0]']
<pre>conv3_block1_0_conv (Conv2 D)</pre>	(None, 32, 32, 512)	131584	['conv2_block3_out[0][0]']
conv3_block1_3_conv (Conv2 D)	(None, 32, 32, 512)	66048	['conv3_block1_2_relu[0][0]']
<pre>conv3_block1_0_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block1_0_conv[0][0]']
<pre>conv3_block1_3_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block1_3_conv[0][0]']
conv3_block1_add (Add)	(None, 32, 32, 512)	0	['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']
<pre>conv3_block1_out (Activati on)</pre>	(None, 32, 32, 512)	0	['conv3_block1_add[0][0]']
conv3_block2_1_conv (Conv2 D)	(None, 32, 32, 128)	65664	['conv3_block1_out[0][0]']
<pre>conv3_block2_1_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block2_1_conv[0][0]']
<pre>conv3_block2_1_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block2_1_bn[0][0]']
conv3_block2_2_conv (Conv2 D)	(None, 32, 32, 128)	147584	['conv3_block2_1_relu[0][0]']
<pre>conv3_block2_2_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block2_2_conv[0][0]']
<pre>conv3_block2_2_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block2_2_bn[0][0]']
conv3_block2_3_conv (Conv2 D)	(None, 32, 32, 512)	66048	['conv3_block2_2_relu[0][0]']
<pre>conv3_block2_3_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block2_3_conv[0][0]']
conv3_block2_add (Add)	(None, 32, 32, 512)	0	['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']
<pre>conv3_block2_out (Activati on)</pre>	(None, 32, 32, 512)	0	['conv3_block2_add[0][0]']
conv3_block3_1_conv (Conv2 D)	(None, 32, 32, 128)	65664	['conv3_block2_out[0][0]']
<pre>conv3_block3_1_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block3_1_conv[0][0]']
<pre>conv3_block3_1_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block3_1_bn[0][0]']
conv3_block3_2_conv (Conv2 D)	(None, 32, 32, 128)	147584	['conv3_block3_1_relu[0][0]']
<pre>conv3_block3_2_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block3_2_conv[0][0]']
<pre>conv3_block3_2_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block3_2_bn[0][0]']

conv3_block3_3_conv (Conv2 D)	(None, 32, 32, 512)	66048	['conv3_block3_2_relu[0][0]']
<pre>conv3_block3_3_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block3_3_conv[0][0]']
conv3_block3_add (Add)	(None, 32, 32, 512)	0	['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']
<pre>conv3_block3_out (Activati on)</pre>	(None, 32, 32, 512)	0	['conv3_block3_add[0][0]']
conv3_block4_1_conv (Conv2 D)	(None, 32, 32, 128)	65664	['conv3_block3_out[0][0]']
<pre>conv3_block4_1_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block4_1_conv[0][0]']
<pre>conv3_block4_1_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block4_1_bn[0][0]']
conv3_block4_2_conv (Conv2 D)	(None, 32, 32, 128)	147584	['conv3_block4_1_relu[0][0]']
<pre>conv3_block4_2_bn (BatchNo rmalization)</pre>	(None, 32, 32, 128)	512	['conv3_block4_2_conv[0][0]']
<pre>conv3_block4_2_relu (Activ ation)</pre>	(None, 32, 32, 128)	0	['conv3_block4_2_bn[0][0]']
conv3_block4_3_conv (Conv2 D)	(None, 32, 32, 512)	66048	['conv3_block4_2_relu[0][0]']
<pre>conv3_block4_3_bn (BatchNo rmalization)</pre>	(None, 32, 32, 512)	2048	['conv3_block4_3_conv[0][0]']
conv3_block4_add (Add)	(None, 32, 32, 512)	0	['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']
<pre>conv3_block4_out (Activati on)</pre>	(None, 32, 32, 512)	0	['conv3_block4_add[0][0]']
<pre>conv4_block1_1_conv (Conv2 D)</pre>	(None, 16, 16, 256)	131328	['conv3_block4_out[0][0]']
<pre>conv4_block1_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block1_1_conv[0][0]']
<pre>conv4_block1_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block1_1_bn[0][0]']
conv4_block1_2_conv (Conv2 D)	(None, 16, 16, 256)	590080	['conv4_block1_1_relu[0][0]']
<pre>conv4_block1_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block1_2_conv[0][0]']
<pre>conv4_block1_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block1_2_bn[0][0]']
<pre>conv4_block1_0_conv (Conv2 D)</pre>	(None, 16, 16, 1024)	525312	['conv3_block4_out[0][0]']
<pre>conv4_block1_3_conv (Conv2 D)</pre>	(None, 16, 16, 1024)	263168	['conv4_block1_2_relu[0][0]']
<pre>conv4_block1_0_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block1_0_conv[0][0]']
<pre>conv4_block1_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block1_3_conv[0][0]']
conv4_block1_add (Add)	(None, 16, 16, 1024)	0	['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']
<pre>conv4_block1_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block1_add[0][0]']
<pre>conv4_block2_1_conv (Conv2 D)</pre>	(None, 16, 16, 256)	262400	['conv4_block1_out[0][0]']
<pre>conv4_block2_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block2_1_conv[0][0]']
<pre>conv4_block2_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block2_1_bn[0][0]']
conv4_block2_2_conv (Conv2	(None, 16, 16, 256)	590080	['conv4_block2_1_relu[0][0]']

D)				
<pre>conv4_block2_2_bn (BatchNo rmalization)</pre>	(None, 16, 16,	256)	1024	['conv4_block2_2_conv[0][0]']
<pre>conv4_block2_2_relu (Activ ation)</pre>	(None, 16, 16,	256)	0	['conv4_block2_2_bn[0][0]']
<pre>conv4_block2_3_conv (Conv2 D)</pre>	(None, 16, 16,	1024)	263168	['conv4_block2_2_relu[0][0]']
<pre>conv4_block2_3_bn (BatchNo rmalization)</pre>	(None, 16, 16,	1024)	4096	['conv4_block2_3_conv[0][0]']
conv4_block2_add (Add)	(None, 16, 16,	1024)	0	['conv4_block1_out[0][0]',

rmalization)	(None, 16, 16, 256)	1024	['CONV4_DLOCK2_2_CONV[0][0]']
<pre>conv4_block2_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block2_2_bn[0][0]']
conv4_block2_3_conv (Conv2 D)	(None, 16, 16, 1024)	263168	['conv4_block2_2_relu[0][0]']
<pre>conv4_block2_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block2_3_conv[0][0]']
conv4_block2_add (Add)	(None, 16, 16, 1024)	Θ	['conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]']
<pre>conv4_block2_out (Activati on)</pre>	(None, 16, 16, 1024)	Θ	['conv4_block2_add[0][0]']
conv4_block3_1_conv (Conv2 D)	(None, 16, 16, 256)	262400	['conv4_block2_out[0][0]']
<pre>conv4_block3_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block3_1_conv[0][0]']
<pre>conv4_block3_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block3_1_bn[0][0]']
<pre>conv4_block3_2_conv (Conv2 D)</pre>	(None, 16, 16, 256)	590080	['conv4_block3_1_relu[0][0]']
<pre>conv4_block3_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block3_2_conv[0][0]']
<pre>conv4_block3_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	Θ	['conv4_block3_2_bn[0][0]']
conv4_block3_3_conv (Conv2 D)	(None, 16, 16, 1024)	263168	['conv4_block3_2_relu[0][0]']
<pre>conv4_block3_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block3_3_conv[0][0]']
conv4_block3_add (Add)	(None, 16, 16, 1024)	0	['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']
<pre>conv4_block3_out (Activati on)</pre>	(None, 16, 16, 1024)	Θ	['conv4_block3_add[0][0]']
<pre>conv4_block4_1_conv (Conv2 D)</pre>	(None, 16, 16, 256)	262400	['conv4_block3_out[0][0]']
<pre>conv4_block4_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block4_1_conv[0][0]']
<pre>conv4_block4_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block4_1_bn[0][0]']
<pre>conv4_block4_2_conv (Conv2 D)</pre>	(None, 16, 16, 256)	590080	['conv4_block4_1_relu[0][0]']
<pre>conv4_block4_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block4_2_conv[0][0]']
<pre>conv4_block4_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block4_2_bn[0][0]']
<pre>conv4_block4_3_conv (Conv2 D)</pre>	(None, 16, 16, 1024)	263168	['conv4_block4_2_relu[0][0]']
<pre>conv4_block4_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block4_3_conv[0][0]']
conv4_block4_add (Add)	(None, 16, 16, 1024)	0	['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']
<pre>conv4_block4_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block4_add[0][0]']
<pre>conv4_block5_1_conv (Conv2 D)</pre>	(None, 16, 16, 256)	262400	['conv4_block4_out[0][0]']
<pre>conv4_block5_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block5_1_conv[0][0]']
<pre>conv4_block5_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	Θ	['conv4_block5_1_bn[0][0]']

conv4_block5_2_conv (Conv2 D)	(None, 16, 16, 256)	590080	['conv4_block5_1_relu[0][0]']
<pre>conv4_block5_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block5_2_conv[0][0]']
<pre>conv4_block5_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block5_2_bn[0][0]']
conv4_block5_3_conv (Conv2 D)	(None, 16, 16, 1024)	263168	['conv4_block5_2_relu[0][0]']
<pre>conv4_block5_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block5_3_conv[0][0]']
conv4_block5_add (Add)	(None, 16, 16, 1024)	0	['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']
<pre>conv4_block5_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block5_add[0][0]']
<pre>conv4_block6_1_conv (Conv2 D)</pre>	(None, 16, 16, 256)	262400	['conv4_block5_out[0][0]']
<pre>conv4_block6_1_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block6_1_conv[0][0]']
<pre>conv4_block6_1_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block6_1_bn[0][0]']
conv4_block6_2_conv (Conv2 D)	(None, 16, 16, 256)	590080	['conv4_block6_1_relu[0][0]']
<pre>conv4_block6_2_bn (BatchNo rmalization)</pre>	(None, 16, 16, 256)	1024	['conv4_block6_2_conv[0][0]']
<pre>conv4_block6_2_relu (Activ ation)</pre>	(None, 16, 16, 256)	0	['conv4_block6_2_bn[0][0]']
<pre>conv4_block6_3_conv (Conv2 D)</pre>	(None, 16, 16, 1024)	263168	['conv4_block6_2_relu[0][0]']
<pre>conv4_block6_3_bn (BatchNo rmalization)</pre>	(None, 16, 16, 1024)	4096	['conv4_block6_3_conv[0][0]']
conv4_block6_add (Add)	(None, 16, 16, 1024)	0	['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']
<pre>conv4_block6_out (Activati on)</pre>	(None, 16, 16, 1024)	0	['conv4_block6_add[0][0]']
conv5_block1_1_conv (Conv2 D)	(None, 8, 8, 512)	524800	['conv4_block6_out[0][0]']
<pre>conv5_block1_1_bn (BatchNo rmalization)</pre>	(None, 8, 8, 512)	2048	['conv5_block1_1_conv[0][0]']
<pre>conv5_block1_1_relu (Activ ation)</pre>	(None, 8, 8, 512)	0	['conv5_block1_1_bn[0][0]']
conv5_block1_2_conv (Conv2 D)	(None, 8, 8, 512)	2359808	['conv5_block1_1_relu[0][0]']
<pre>conv5_block1_2_bn (BatchNo rmalization)</pre>	(None, 8, 8, 512)	2048	['conv5_block1_2_conv[0][0]']
<pre>conv5_block1_2_relu (Activ ation)</pre>	(None, 8, 8, 512)	Θ	['conv5_block1_2_bn[0][0]']
conv5_block1_0_conv (Conv2 D)	(None, 8, 8, 2048)	2099200	['conv4_block6_out[0][0]']
conv5_block1_3_conv (Conv2 D)	(None, 8, 8, 2048)	1050624	['conv5_block1_2_relu[0][0]']
<pre>conv5_block1_0_bn (BatchNo rmalization)</pre>	(None, 8, 8, 2048)	8192	['conv5_block1_0_conv[0][0]']
<pre>conv5_block1_3_bn (BatchNo rmalization)</pre>	(None, 8, 8, 2048)	8192	['conv5_block1_3_conv[0][0]']
conv5_block1_add (Add)	(None, 8, 8, 2048)	0	['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']
<pre>conv5_block1_out (Activati on)</pre>	(None, 8, 8, 2048)	Θ	['conv5_block1_add[0][0]']
conv5_block2_1_conv (Conv2 D)	(None, 8, 8, 512)	1049088	['conv5_block1_out[0][0]']

```
conv5_block2_1_bn (BatchNo
                           (None, 8, 8, 512)
                                                           2048
                                                                      ['conv5_block2_1_conv[0][0]']
rmalization)
conv5 block2 1 relu (Activ
                             (None, 8, 8, 512)
                                                                      ['conv5 block2 1 bn[0][0]']
ation)
                             (None, 8, 8, 512)
                                                           2359808
                                                                     ['conv5 block2 1 relu[0][0]']
conv5_block2_2_conv (Conv2
conv5 block2 2 bn (BatchNo
                             (None, 8, 8, 512)
                                                           2048
                                                                      ['conv5 block2 2 conv[0][0]']
rmalization)
                                                                      ['conv5 block2 2 bn[0][0]']
conv5 block2 2 relu (Activ
                             (None, 8, 8, 512)
ation)
conv5 block2 3 conv (Conv2
                             (None, 8, 8, 2048)
                                                           1050624
                                                                      ['conv5 block2 2 relu[0][0]']
D)
conv5_block2_3_bn (BatchNo
                             (None, 8, 8, 2048)
                                                           8192
                                                                      ['conv5_block2_3_conv[0][0]']
rmalization)
                                                           0
                                                                      ['conv5_block1_out[0][0]'
conv5_block2_add (Add)
                             (None, 8, 8, 2048)
                                                                       'conv5 block2 3 bn[0][0]']
                             (None, 8, 8, 2048)
                                                           0
                                                                      ['conv5_block2_add[0][0]']
conv5_block2_out (Activati
on)
                             (None, 8, 8, 512)
                                                           1049088
                                                                     ['conv5_block2_out[0][0]']
conv5_block3_1_conv (Conv2
conv5 block3 1 bn (BatchNo
                             (None, 8, 8, 512)
                                                           2048
                                                                      ['conv5 block3 1 conv[0][0]']
rmalization)
                             (None, 8, 8, 512)
conv5 block3 1 relu (Activ
                                                                      ['conv5 block3 1 bn[0][0]']
ation)
conv5 block3 2 conv (Conv2
                             (None, 8, 8, 512)
                                                           2359808
                                                                      ['conv5_block3_1_relu[0][0]']
D)
conv5 block3 2 bn (BatchNo
                             (None, 8, 8, 512)
                                                           2048
                                                                      ['conv5 block3 2 conv[0][0]']
rmalization)
conv5 block3 2 relu (Activ
                             (None, 8, 8, 512)
                                                           0
                                                                      ['conv5 block3 2 bn[0][0]']
ation)
                             (None, 8, 8, 2048)
                                                           1050624
                                                                     ['conv5_block3_2_relu[0][0]']
conv5_block3_3_conv (Conv2
conv5 block3_3_bn (BatchNo
                             (None, 8, 8, 2048)
                                                           8192
                                                                      ['conv5 block3 3 conv[0][0]']
rmalization)
conv5 block3 add (Add)
                             (None, 8, 8, 2048)
                                                           0
                                                                      ['conv5 block2 out[0][0]'
                                                                       conv5_block3_3_bn[0][0]']
                                                                      ['conv5 block3 add[0][0]']
conv5 block3 out (Activati
                             (None, 8, 8, 2048)
                                                           0
on)
average_pooling2d (Average
                             (None, 2, 2, 2048)
                                                           0
                                                                      ['conv5_block3_out[0][0]']
Pooling2D)
flatten (Flatten)
                             (None, 8192)
                                                           0
                                                                      ['average_pooling2d[0][0]']
dense (Dense)
                             (None, 256)
                                                           2097408
                                                                      ['flatten[0][0]']
dropout (Dropout)
                             (None, 256)
                                                                      ['dense[0][0]']
                                                           0
dense 1 (Dense)
                             (None, 256)
                                                           65792
                                                                      ['dropout[0][0]']
dropout_1 (Dropout)
                                                                      ['dense_1[0][0]']
                             (None, 256)
                                                           0
dense_2 (Dense)
                             (None, 2)
                                                           514
                                                                      ['dropout_1[0][0]']
```

Total params: 25751426 (98.23 MB)

Trainable params: 25698306 (98.03 MB) Non-trainable params: 53120 (207.50 KB)

```
In [29]: # compile the model
model.compile(loss= 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])
```

WARNING:tensorflow:From C:\Users\olayi\anaconda3\Lib\site-packages\keras\src\optimizers__init__.py:309: The na me tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

```
In [30]: # use early stopping to exit training if validation loss is not decreasing even after certain epochs (patience)
earlystopping = EarlyStopping(monitor = 'val_loss', mode = 'min', verbose = 1, patience = 20)
```

```
In [31]: history = model.fit(train_generator, steps_per_epoch=train_generator.n // 16, epochs = 10, validation_data= val
           WARNING: tensorflow: From C: \Users \o ayi \o anaconda 3 \bot ib \diagdown site-packages \backprime keras \backprime src \backprime tils \backprime tf\_utils.py: 492: The name tform of the site-packages \lor the site-
           .ragged.RaggedTensorValue is deprecated. Please use tf.compat.v1.ragged.RaggedTensorValue instead.
           WARNING:tensorflow:From C:\Users\olayi\anaconda3\Lib\site-packages\keras\src\engine\base layer utils.py:384: Th
           e name tf.executing_eagerly_outside_functions is deprecated. Please use tf.compat.v1.executing_eagerly_outside_
           functions instead.
           177/177 [============] - ETA: 0s - loss: 0.9508 - accuracy: 0.6532
           Epoch 1: val_loss improved from inf to 0.65904, saving model to classifier-resnet-weights.hdf5
                               177/177 [===
           val accuracy: 0.6331
           Epoch 2/10
           Epoch 2: val loss did not improve from 0.65904
           177/177 [===
                                              ========] - 693s 4s/step - loss: 0.5868 - accuracy: 0.7535 - val loss: 0.6681 -
           val accuracy: 0.6331
           Epoch 3/10
           177/177 [==
                                    Epoch 3: val loss did not improve from 0.65904
           val_accuracy: 0.6351
           Epoch 4/10
           Epoch 4: val loss did not improve from 0.65904
           177/177 [==
                                      =========] - 705s 4s/step - loss: 0.3710 - accuracy: 0.8353 - val loss: 0.7307 -
           val_accuracy: 0.6310
           Epoch 5/10
           Epoch 5: val loss did not improve from 0.65904
           val_accuracy: 0.6694
           Epoch 6/10
           177/177 [========] - ETA: 0s - loss: 0.2807 - accuracy: 0.8874 Epoch 6: val_loss improved from 0.65904 to 0.51411, saving model to classifier-resnet-weights.hdf5
           val accuracy: 0.7722
           Epoch 7/10
           177/177 [==
                           Epoch 7: val loss did not improve from 0.51411
           13 - val accuracy: 0.6976
           Epoch 8/10
           Epoch 8: val loss improved from 0.51411 to 0.31863, saving model to classifier-resnet-weights.hdf5
           val accuracy: 0.8609
           Epoch 9/10
           177/177 [==
                              Epoch 9: val loss did not improve from 0.31863
           177/177 [=========== ] - 11596s 66s/step - loss: 0.2450 - accuracy: 0.9065 - val loss: 0.3296
            - val accuracy: 0.8589
           Epoch 10/10
           Epoch 10: val_loss did not improve from 0.31863
           177/177 [===
                                     - val accuracy: 0.8891
In [32]: # save model architecture to json file for future use
           model_json = model.to_json()
           with open('classifier resnet model.json', 'w') as json file:
                json_file.write(model_json)
           Assess Trained Model Performance
In [33]: # load json file
           with open('classifier resnet model.json', 'r') as json file:
                json_savedModel = json_file.read()
           model = tf.keras.models.model from json(json savedModel)
           model.load_weights('classifier-resnet-weights.hdf5')
           model.compile(loss = 'categorical_crossentropy', optimizer= 'adam', metrics = ['accuracy'])
In [34]: # make prediction
           test pred = model.predict(test generator, steps = test generator.n // 16, verbose = 1)
           In [35]: test pred.shape
```

checkpointer = ModelCheckpoint(filepath='classifier-resnet-weights.hdf5', verbose = 1, save_best_only = True)

save the best model with the least validation loss

```
Out[35]: (576, 2)
In [36]: test_pred
Out[36]: array([[9.9453026e-01, 5.4697674e-03],
                [9.7976160e-01, 2.0238433e-02],
                [4.4421285e-01, 5.5578715e-01],
                [9.9669302e-01, 3.3069481e-03],
                [9.2080944e-12, 1.0000000e+00],
                [9.5426548e-01, 4.5734555e-02]], dtype=float32)
In [37]: # obtain the predicted class from the model prediction
         predict = []
         for i in test_pred:
             predict.append(str(np.argmax(i)))
         predict = np.asarray(predict)
In [38]: predict
                                                                     '0',
Out[38]: array(['0', '0', '1',
                              '1', '1', '0',
                                             '0', '1', '1', '0', '0',
                                                                          '0',
                '0', '0', '1',
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In [39]: # compare the predictions to the original test data
         # since we have used test generator, it limited the images to len(predict), due to batch size
         original_test = np.asarray(test['mask'])[:len(predict)]
         len(original_test)
Out[39]: 576
In [40]:
         # obtain the accuracy of the model
         from sklearn.metrics import accuracy score
         accuracy = accuracy_score(original_test, predict)
         accuracy
Out[40]: 0.8697916666666666
         # plot the confusion matrix
In [41]:
         from sklearn.metrics import confusion_matrix
         cm = confusion_matrix(original_test, predict)
         plt.figure(figsize = (7,7))
```

```
- 300
                                                                                           - 250
            0 -
                           3.2e+02
                                                                55
                                                                                           - 200
                                                                                           - 150
                                                                                           - 100
                               20
                                                             1.8e+02
                                                                                           - 50
                                0
                                                                 1
In [42]: # Obatin the classification report
           from sklearn.metrics import classification_report
           report = classification_report(original_test, predict, labels = [0,1])
           print(report)
                                           recall f1-score
                            precision
                                                                 support
                        0
                                  0.94
                                              0.85
                                                          0.90
                                                                       378
                        1
                                  0.76
                                              0.90
                                                          0.83
                                                                       198
                                                                       576
                                              0.87
               micro avg
                                  0.87
                                                          0.87
               macro avg
                                  0.85
                                              0.88
                                                          0.86
                                                                       576
                                              0.87
                                                          0.87
           weighted avg
                                  0.88
                                                                       576
           Build A segmentation Model to Locate the Tumor
In [43]: # Get the dataframe containing MRIs which have masks associated with them
brain_df_mask = brain_df[brain_df['mask']==1]
           brain df mask shape
Out[43]: (1373, 4)
In [44]:
           # split the data into train the test set
           from sklearn.model selection import train test split
           X_train, X_val = train_test_split(brain_df_mask, test_size = 0.15)
X_test, X_val = train_test_split(X_val, test_size = 0.5)
           print('X_train ', X_train)
print('X_val ', X_val)
print('X_test ', X_test)
```

sns.heatmap(cm, annot = True)

Out[41]: <Axes: >

X train

2744 TCGA_HT_7855_19951020
3358 TCGA_HT_A5RC_19990831
1773 TCGA_DU_A5TT_19980318
3016 TCGA_HT_7879_19981009
2691 TCGA_HT_7856_19950831
...
1684 TCGA_DU_8165_19970205
2746 TCGA_HT_7855_19951020
2986 TCGA_HT_7855_19951020
2986 TCGA_HT_7882_19970125
3480 TCGA_DU_8164_19970111
3532 TCGA_DU_8168_19970503

patient id \

```
image path \
      TCGA_FG_6690_20020226/TCGA_FG_6690_20020226_27...
3358
      TCGA_DU_6401_19831001/TCGA_DU_6401_19831001_37...
      TCGA CS 6665 20010817/TCGA CS 6665 20010817 17...
      TCGA FG 6690 20020226/TCGA FG 6690 20020226 31...
3016
2691
      TCGA_HT_A61A_20000127/TCGA_HT_A61A_20000127_26...
1684
      TCGA DU 7008 19830723/TCGA DU 7008 19830723 16...
      TCGA_FG_7643_20021104/TCGA_FG_7643_20021104_27...
2746
2986
      TCGA_DU_6399_19830416/TCGA_DU_6399_19830416_31...
      TCGA DU 6405 19851005/TCGA DU 6405 19851005 42...
3480
      TCGA DU 7014 19860618/TCGA DU 7014 19860618 44...
3532
2744
      TCGA FG 6690 20020226/TCGA FG 6690 20020226 27...
3358
      TCGA_DU_6401_19831001/TCGA_DU_6401_19831001_37...
1773
      TCGA CS 6665 20010817/TCGA CS 6665 20010817 17...
      TCGA FG 6690 20020226/TCGA FG 6690 20020226 31...
3016
2691 TCGA_HT_A61A_20000127/TCGA_HT_A61A_20000127_26...
1684
      TCGA DU 7008 19830723/TCGA DU 7008 19830723 16...
      TCGA FG 7643 20021104/TCGA FG 7643 20021104 27...
TCGA_DU_6399_19830416/TCGA_DU_6399_19830416_31...
2746
2986
3480
      TCGA DU 6405 19851005/TCGA DU 6405 19851005 42...
3532
      TCGA DU 7014 19860618/TCGA DU 7014 19860618 44...
[1167 rows x 4 columns]
                          patient id \
      TCGA_HT_7616_19940813
2667
      TCGA_DU_A5TP_19970614
TCGA_FG_6691_20020405
1601
2150
      TCGA FG 6689 20020326
     TCGA_DU_7306_19930512
1519
3042
      TCGA HT 7877 19980917
      TCGA_HT_8563_19981209
TCGA_FG_7634_20000128
3205
2239
3390
      TCGA HT A61B 19991127
1835
      TCGA FG 5962 20000626
                                                  image path ∖
2667
      TCGA DU A5TY 19970709/TCGA DU A5TY 19970709 26...
      TCGA_FG_7637_20000922/TCGA_FG_7637_20000922_15...
1601
      TCGA_FG_7634_20000128/TCGA_FG_7634_20000128_20...
2150
2009
      TCGA DU 7018 19911220/TCGA DU 7018 19911220 19...
      TCGA HT 8113 19930809/TCGA HT 8113 19930809 14...
1519
      TCGA DU 6401 19831001/TCGA DU 6401 19831001 31...
3042
      TCGA HT A61A 20000127/TCGA HT A61A 20000127 34...
3205
      TCGA_DU_8166_19970322/TCGA_DU_8166_19970322_21...
TCGA_DU_7010_19860307/TCGA_DU_7010_19860307_38...
2239
3390
1835
      TCGA HT 7692 19960724/TCGA HT 7692 19960724 17...
                                                   mask path mask
      TCGA_DU_A5TY_19970709/TCGA_DU_A5TY_19970709_26...
2667
      TCGA_FG_7637_20000922/TCGA_FG_7637_20000922_15...
1601
      TCGA FG 7634 20000128/TCGA FG 7634 20000128 20...
2150
      TCGA_DU_7018_19911220/TCGA_DU_7018_19911220_19...
2009
1519
      TCGA_HT_8113_19930809/TCGA_HT_8113_19930809_14...
                                                                  1
      TCGA_DU_6401_19831001/TCGA_DU_6401_19831001_31...
TCGA_HT_A61A_20000127/TCGA_HT_A61A_20000127_34...
3042
                                                                  1
3205
                                                                  1
      TCGA DU 8166 19970322/TCGA DU 8166 19970322 21...
      TCGA_DU_7010_19860307/TCGA_DU_7010_19860307_38...
TCGA_HT_7692_19960724/TCGA_HT_7692_19960724_17...
3390
1835
[103 rows x 4 columns]
X test
                           patient id \
1816 TCGA_DU_A5TY_19970709
      TCGA_DU_7304_19930325
3783
      TCGA_DU_7306_19930512
1484
1842
      TCGA_FG_5962_20000626
995
      TCGA DU 6405 19851005
      TCGA_DU_A5TT_19980318
TCGA_DU_8166_19970322
1794
1639
2771
      TCGA_HT_7692_19960724
      TCGA_FG_6691_20020405
TCGA_DU_8163_19961119
2154
1286
                                                  image path
      TCGA FG 6692 20020606/TCGA FG 6692 20020606 17...
1816
      TCGA_FG_A60K_20040224/TCGA_FG_A60K_20040224_56...
3783
1484
      TCGA_DU_A5TU_19980312/TCGA_DU_A5TU_19980312_14...
      TCGA HT 7882 19970125/TCGA HT 7882 19970125 17...
1842
      TCGA_CS_5393_19990606/TCGA_CS_5393_19990606_10...
995
1794
      TCGA DU 7008 19830723/TCGA DU 7008 19830723 17...
      TCGA DU 8168 19970503/TCGA DU 8168 19970503 15...
1639
```

```
2771 TCGA FG 6688 20020215/TCGA FG 6688 20020215 27...
               TCGA_FG_A4MU_20030903/TCGA_FG_A4MU_20030903_20...
         2154
         1286 TCGA HT 7874 19950902/TCGA HT 7874 19950902 12...
                                                       mask path mask
         3783 TCGA FG A60K 20040224/TCGA FG A60K 20040224 56...
                                                                     1
               TCGA_DU_A5TU_19980312/TCGA_DU_A5TU_19980312_14...
TCGA_HT_7882_19970125/TCGA_HT_7882_19970125_17...
         1484
         1842
                                                                     1
               TCGA CS 5393 19990606/TCGA CS 5393 19990606 10...
         995
                                                                     1
         1794 TCGA DU 7008 19830723/TCGA DU 7008 19830723 17...
                                                                     1
               TCGA_DU_8168_19970503/TCGA_DU_8168_19970503_15...
         1639
                                                                     1
         2771
               TCGA_FG_6688_20020215/TCGA_FG_6688_20020215_27...
                                                                     1
               TCGA FG A4MU 20030903/TCGA FG A4MU 20030903 20...
         2154
                                                                     1
         1286 TCGA_HT_7874_19950902/TCGA_HT_7874_19950902_12...
         [103 rows x 4 columns]
In [45]: # Create a seperate list for the imageId, classId to pass into the generator
         train ids = list(X train.image path)
         train_mask = list(X_train.mask_path)
         val ids = list(X val.image path)
         val mask = list(X_val.mask_path)
In [46]:
         # utilities file contains the code for custom loass function and custom data generator
         from utilities import DataGenerator
         training gen = DataGenerator(train ids, train mask)
         validation_gen = DataGenerator(val_ids, val mask)
         # building ResUNet model
In [47]:
         # Using the res block
         def resblock(X, f):
             # make a copy of the input
             X copy = X
             # main path
             X = Conv2D(f, kernel_size = (1,1), strides = (1,1), kernel_initializer= 'he_normal')(X)
             X = BatchNormalization()(X)
             X = Activation('relu')(X)
             X = Conv2D(f, kernel size = (3,3), strides = (1,1), padding = 'same', kernel initializer= 'he normal')(X)
             X = BatchNormalization()(X)
             # Short path
             X copy = Conv2D(f, kernel_size = (1,1), strides = (1,1), kernel initializer= 'he normal')(X copy)
             X_copy = BatchNormalization()(X_copy)
             # Adding the output from the main path and the short path together
             X = add([X,X copy])
             X = Activation('relu')(X)
             return X
In [48]: # function to upscale and concetenate the values passed
         def upsample_concat(x, skip):
             x = UpSampling2D((2,2))(x)
             merge = Concatenate()([x, skip])
             return merge
In [49]: # build the network
         input\_shape = (256, 256, 3)
         # Input tensor shape
         X_input = Input(input_shape)
         # Stage 1
         conv1_in = Conv2D(16,3, activation ='relu', padding = 'same', kernel_initializer = 'he_normal')(X_input)
         conv1 in = BatchNormalization()(conv1 in)
         conv1 in = Conv2D(16,3, activation ='relu', padding = 'same', kernel initializer = 'he normal')(conv1 in)
         conv1 in = BatchNormalization()(conv1 in)
         pool_1 = MaxPool2D(pool_size= (2,2))(conv1_in)
         # Stage 2
         conv2_in = resblock(pool_1,32)
         pool_2 = MaxPool2D(pool_size= (2,2))(conv2_in)
         # Stage 3
         conv3_in = resblock(pool_2,64)
         pool 3 = MaxPool2D(pool size= (2,2))(conv3 in)
```

```
# Stage 4
conv4_in = resblock(pool_3,128)
pool \overline{4} = MaxPool2D(pool size= (2,2))(conv4 in)
# Stage 5 (Bottle Neck)
conv5_in = resblock(pool_4,256)
# Upsale stage 1
up_1 = upsample_concat(conv5_in, conv4_in)
up_1 = resblock(up_1, 128)
# Upsale stage 2
up_2 = upsample_concat(up_1, conv3_in)
up_2 = resblock(up_2, 64)
# Upsale stage 3
up 3 = upsample concat(up 2, conv2 in)
up_3 = resblock(up_3, 32)
# Upsale stage 4
up_4 = upsample_concat(up_3, conv1_in)
up_4 = resblock(up_4, 16)
# Final Output
output = Conv2D(1, (1,1), padding = 'same', activation = 'sigmoid')(up_4)
model_seg = Model(inputs = X_input, outputs = output)
```

In [50]: model_seg.summary()

Model: "model_1"

Layer (type)	Output Shape	Param #	Connected to
input_2 (InputLayer)	[(None, 256, 256, 3)]	0	[]
conv2d (Conv2D)	(None, 256, 256, 16)	448	['input_2[0][0]']
<pre>batch_normalization (Batch Normalization)</pre>	(None, 256, 256, 16)	64	['conv2d[0][0]']
conv2d_1 (Conv2D)	(None, 256, 256, 16)	2320	['batch_normalization[0][0]']
<pre>batch_normalization_1 (Bat chNormalization)</pre>	(None, 256, 256, 16)	64	['conv2d_1[0][0]']
<pre>max_pooling2d (MaxPooling2 D)</pre>	(None, 128, 128, 16)	0	<pre>['batch_normalization_1[0][0]']</pre>
conv2d_2 (Conv2D)	(None, 128, 128, 32)	544	['max_pooling2d[0][0]']
<pre>batch_normalization_2 (Bat chNormalization)</pre>	(None, 128, 128, 32)	128	['conv2d_2[0][0]']
activation (Activation)	(None, 128, 128, 32)	0	<pre>['batch_normalization_2[0][0]']</pre>
conv2d_3 (Conv2D)	(None, 128, 128, 32)	9248	['activation[0][0]']
conv2d_4 (Conv2D)	(None, 128, 128, 32)	544	['max_pooling2d[0][0]']
<pre>batch_normalization_3 (Bat chNormalization)</pre>	(None, 128, 128, 32)	128	['conv2d_3[0][0]']
<pre>batch_normalization_4 (Bat chNormalization)</pre>	(None, 128, 128, 32)	128	['conv2d_4[0][0]']
add (Add)	(None, 128, 128, 32)	0	<pre>['batch_normalization_3[0][0]' , 'batch_normalization_4[0][0] ']</pre>
<pre>activation_1 (Activation)</pre>	(None, 128, 128, 32)	0	['add[0][0]']
<pre>max_pooling2d_1 (MaxPoolin g2D)</pre>	(None, 64, 64, 32)	0	['activation_1[0][0]']
conv2d_5 (Conv2D)	(None, 64, 64, 64)	2112	['max_pooling2d_1[0][0]']
<pre>batch_normalization_5 (Bat chNormalization)</pre>	(None, 64, 64, 64)	256	['conv2d_5[0][0]']
activation_2 (Activation)	(None, 64, 64, 64)	0	<pre>['batch_normalization_5[0][0]']</pre>
conv2d_6 (Conv2D)	(None, 64, 64, 64)	36928	['activation_2[0][0]']

conv2d_7 (Conv2D)	(None, 64,	64,	64)	2112	['max_pooling2d_1[0][0]']
<pre>batch_normalization_6 (Bat chNormalization)</pre>	(None, 64,	64,	64)	256	['conv2d_6[0][0]']
<pre>batch_normalization_7 (Bat chNormalization)</pre>	(None, 64,	64,	64)	256	['conv2d_7[0][0]']
add_1 (Add)	(None, 64,	64,	64)	0	<pre>['batch_normalization_6[0][0]' , 'batch_normalization_7[0][0] ']</pre>
activation_3 (Activation)	(None, 64,	64,	64)	0	['add_1[0][0]']
<pre>max_pooling2d_2 (MaxPoolin g2D)</pre>	(None, 32,	32,	64)	0	['activation_3[0][0]']
conv2d_8 (Conv2D)	(None, 32,	32,	128)	8320	['max_pooling2d_2[0][0]']
<pre>batch_normalization_8 (Bat chNormalization)</pre>	(None, 32,	32,	128)	512	['conv2d_8[0][0]']
activation_4 (Activation)	(None, 32,	32,	128)	0	<pre>['batch_normalization_8[0][0]']</pre>
conv2d_9 (Conv2D)	(None, 32,	32,	128)	147584	['activation_4[0][0]']
conv2d_10 (Conv2D)	(None, 32,	32,	128)	8320	['max_pooling2d_2[0][0]']
<pre>batch_normalization_9 (Bat chNormalization)</pre>	(None, 32,	32,	128)	512	['conv2d_9[0][0]']
<pre>batch_normalization_10 (Ba tchNormalization)</pre>	(None, 32,	32,	128)	512	['conv2d_10[0][0]']
add_2 (Add)	(None, 32,	32,	128)	0	<pre>['batch_normalization_9[0][0]' , 'batch_normalization_10[0][0]']</pre>
activation_5 (Activation)	(None, 32,	32,	128)	0	['add_2[0][0]']
max_pooling2d_3 (MaxPoolin g2D)	(None, 16,	16,	128)	0	['activation_5[0][0]']
conv2d_11 (Conv2D)	(None, 16,	16,	256)	33024	['max_pooling2d_3[0][0]']
<pre>batch_normalization_11 (Ba tchNormalization)</pre>	(None, 16,	16,	256)	1024	['conv2d_11[0][0]']
activation_6 (Activation)	(None, 16,	16,	256)	0	['batch_normalization_11[0][0] ']
conv2d_12 (Conv2D)	(None, 16,	16,	256)	590080	['activation_6[0][0]']
conv2d_13 (Conv2D)	(None, 16,	16,	256)	33024	['max_pooling2d_3[0][0]']
<pre>batch_normalization_12 (Ba tchNormalization)</pre>	(None, 16,	16,	256)	1024	['conv2d_12[0][0]']
<pre>batch_normalization_13 (Ba tchNormalization)</pre>	(None, 16,	16,	256)	1024	['conv2d_13[0][0]']
add_3 (Add)	(None, 16,	16,	256)	0	['batch_normalization_12[0][0]
					batch_normalization_13[0][0]
activation_7 (Activation)	(None, 16,	16,	256)	0	['add_3[0][0]']
<pre>up_sampling2d (UpSampling2 D)</pre>	(None, 32,	32,	256)	0	['activation_7[0][0]']
concatenate (Concatenate)	(None, 32,	32,	384)	0	['up_sampling2d[0][0]', 'activation_5[0][0]']
conv2d_14 (Conv2D)	(None, 32,	32,	128)	49280	['concatenate[0][0]']
<pre>batch_normalization_14 (Ba tchNormalization)</pre>	(None, 32,	32,	128)	512	['conv2d_14[0][0]']
activation_8 (Activation)	(None, 32,	32,	128)	0	['batch_normalization_14[0][0] ']
conv2d_15 (Conv2D)	(None, 32,	32,	128)	147584	['activation_8[0][0]']
conv2d_16 (Conv2D)	(None, 32,	32,	128)	49280	['concatenate[0][0]']
<pre>batch_normalization_15 (Ba tchNormalization)</pre>	(None, 32,	32,	128)	512	['conv2d_15[0][0]']

<pre>batch_normalization_16 (Ba tchNormalization)</pre>	(None, 32, 32, 128)	512	['conv2d_16[0][0]']
add_4 (Add)	(None, 32, 32, 128)	0	<pre>['batch_normalization_15[0][0] ', 'batch_normalization_16[0][0] ']</pre>
activation 9 (Activation)	(None, 32, 32, 128)	0	['add_4[0][0]']
up_sampling2d_1 (UpSamplin	(None, 64, 64, 128)	Θ	_ ['activation_9[0][0]']
<pre>g2D) concatenate_1 (Concatenate)</pre>	(None, 64, 64, 192)	0	['up_sampling2d_1[0][0]', 'activation_3[0][0]']
conv2d_17 (Conv2D)	(None, 64, 64, 64)	12352	['concatenate_1[0][0]']
batch_normalization_17 (BatchNormalization)	(None, 64, 64, 64)	256	['conv2d_17[0][0]']
activation_10 (Activation)	(None, 64, 64, 64)	Θ	['batch_normalization_17[0][0] ']
conv2d_18 (Conv2D)	(None, 64, 64, 64)	36928	['activation_10[0][0]']
conv2d_19 (Conv2D)	(None, 64, 64, 64)	12352	['concatenate_1[0][0]']
<pre>batch_normalization_18 (Ba tchNormalization)</pre>	(None, 64, 64, 64)	256	['conv2d_18[0][0]']
<pre>batch_normalization_19 (Ba tchNormalization)</pre>	(None, 64, 64, 64)	256	['conv2d_19[0][0]']
add_5 (Add)	(None, 64, 64, 64)	0	['batch_normalization_18[0][0]
			'batch_normalization_19[0][0]
activation_11 (Activation)	(None, 64, 64, 64)	0	['add_5[0][0]']
up_sampling2d_2 (UpSamplin g2D)	(None, 128, 128, 64)	0	['activation_11[0][0]']
<pre>concatenate_2 (Concatenate)</pre>	(None, 128, 128, 96)	0	['up_sampling2d_2[0][0]', 'activation_1[0][0]']
conv2d_20 (Conv2D)	(None, 128, 128, 32)	3104	['concatenate_2[0][0]']
<pre>batch_normalization_20 (Ba tchNormalization)</pre>	(None, 128, 128, 32)	128	['conv2d_20[0][0]']
activation_12 (Activation)	(None, 128, 128, 32)	0	['batch_normalization_20[0][0] ']
conv2d_21 (Conv2D)	(None, 128, 128, 32)	9248	['activation_12[0][0]']
conv2d_22 (Conv2D)	(None, 128, 128, 32)	3104	['concatenate_2[0][0]']
<pre>batch_normalization_21 (Ba tchNormalization)</pre>	(None, 128, 128, 32)	128	['conv2d_21[0][0]']
<pre>batch_normalization_22 (Ba tchNormalization)</pre>	(None, 128, 128, 32)	128	['conv2d_22[0][0]']
add_6 (Add)	(None, 128, 128, 32)	0	['batch_normalization_21[0][0]
			'batch_normalization_22[0][0] ']
activation_13 (Activation)	(None, 128, 128, 32)	0	['add_6[0][0]']
up_sampling2d_3 (UpSamplin g2D)	(None, 256, 256, 32)	0	['activation_13[0][0]']
<pre>concatenate_3 (Concatenate)</pre>	(None, 256, 256, 48)	0	<pre>['up_sampling2d_3[0][0]', 'batch_normalization_1[0][0]']</pre>
conv2d_23 (Conv2D)	(None, 256, 256, 16)	784	['concatenate_3[0][0]']
<pre>batch_normalization_23 (Ba tchNormalization)</pre>	(None, 256, 256, 16)	64	['conv2d_23[0][0]']
activation_14 (Activation)	(None, 256, 256, 16)	0	<pre>['batch_normalization_23[0][0] ']</pre>
conv2d_24 (Conv2D)	(None, 256, 256, 16)	2320	['activation_14[0][0]']

```
784
 conv2d 25 (Conv2D)
                            (None, 256, 256, 16)
                                                                    ['concatenate_3[0][0]']
 batch normalization 24 (Ba (None, 256, 256, 16)
                                                          64
                                                                    ['conv2d 24[0][0]']
 tchNormalization)
 batch normalization 25 (Ba (None, 256, 256, 16)
                                                          64
                                                                    ['conv2d 25[0][0]']
 tchNormalization)
 add 7 (Add)
                            (None, 256, 256, 16)
                                                          Θ
                                                                    ['batch normalization 24[0][0]
                                                                     'batch normalization 25[0][0]
                                                                    ['add 7[0][0]']
 activation 15 (Activation) (None, 256, 256, 16)
conv2d 26 (Conv2D)
                            (None, 256, 256, 1)
                                                          17
                                                                    ['activation_15[0][0]']
Total params: 1210513 (4.62 MB)
Trainable params: 1206129 (4.60 MB)
Non-trainable params: 4384 (17.12 KB)
```

Train A Segmentation ResUNet Model To Localize Tumor

In [51]: # using the custom loss function Focal_tversky, tversky_loss, tversky

```
from utilities import focal tversky, tversky loss, tversky
In [52]: # code for focal_tversky
          from keras.losses import binary_crossentropy
          import keras.backend as K
          import tensorflow as tf
          ensilon = 1e-5
          smooth = 1
          def dsc(y_true, y_pred):
              smooth = 1.
              y_true_f = K.flatten(y_true)
              y_pred_f = K.flatten(y_pred)
intersection = K.sum(y_true_f * y_pred_f)
              score = (2. * intersection + smooth) / (K.sum(y_true_f) + K.sum(y_pred_f) + smooth)
              return score
          def dice_loss(y_true, y_pred):
               loss = 1 - dsc(y_true, y_pred)
               return loss
          def bce_dice_loss(y_true, y_pred):
              loss = binary crossentropy(y true, y pred) + dice loss(y true, y pred)
              return loss
          def confusion(y true, y pred):
              smooth=1
              y_pred_pos = K.clip(y_pred, 0, 1)
y_pred_neg = 1 - y_pred_pos
              y_{pos} = K.clip(y_{true}, 0, 1)
              y_neg = 1 - y_pos
              tp = K.sum(y_pos * y_pred_pos)
              fp = K.sum(y_neg * y_pred_pos)
fn = K.sum(y_pos * y_pred_neg)
              prec = (tp + smooth)/(tp+fp+smooth)
               recall = (tp+smooth)/(tp+fn+smooth)
              return prec, recall
          def tp(y_true, y_pred):
              smooth = 1
              y pred_pos = K.round(K.clip(y_pred, 0, 1))
              y_pos = K.round(K.clip(y_true, 0, 1))
              tp = (K.sum(y_pos * y_pred_pos) + smooth)/(K.sum(y_pos) + smooth)
              return tp
          def tn(y_true, y_pred):
              smooth = 1
              y pred_pos = K.round(K.clip(y_pred, 0, 1))
              y_pred_neg = 1 - y_pred_pos
              y_pos = K.round(K.clip(y_true, 0, 1))
              y_neg = 1 - y_pos
              tn = (K.sum(y_neg * y_pred_neg) + smooth) / (K.sum(y_neg) + smooth)
              return tn
          def tversky(y_true, y_pred):
    y_true_pos = K.flatten(y_true)
              y_pred_pos = K.flatten(y_pred)
```

```
true_pos = K.sum(y_true_pos * y_pred_pos)
false_neg = K.sum(y_true_pos * (1-y_pred_pos))
          false_pos = K.sum((1-y_true_pos)*y_pred_pos)
          return (true pos + smooth)/(true pos + alpha*false neg + (1-alpha)*false pos + smooth)
       def tversky loss(y true, y pred):
          return 1 - tversky(y_true,y_pred)
       def focal_tversky(y_true,y_pred):
          y_true = tf.cast(y_true, tf.float32) # Cast y_true to float32
y_pred = tf.cast(y_pred, tf.float32)
          pt_1 = tversky(y_true, y_pred)
          gamma = 0.75
          return K.pow((1-pt 1), gamma)
In [53]: # compile model
       adam = tf.keras.optimizers.legacy.Adam(lr = 0.05, epsilon = 0.1)
       model seg.compile(optimizer= adam, loss = focal tversky, metrics = [tversky])
In [54]: # use early stopping to exit training if validation loss is not decreasing even after certain epochs (patience)
       earlystopping = EarlyStopping(monitor = 'val_loss', mode = 'min', verbose = 1, patience = 20)
       # save the best model with the least validation loss
       checkpointer = ModelCheckpoint(filepath='ResUNet-weights.hdf5', verbose = 1, save best only = True)
In [55]: history = model_seg.fit(training_gen, epochs = 40, validation_data= validation_gen, callbacks= [checkpointer, e
       72/72 [=============] - ETA: 0s - loss: 0.8447 - tversky: 0.2007
       Epoch 2/40
       72/72 [==========] - ETA: 0s - loss: 0.5082 - tversky: 0.5898
       Epoch 2: val_loss improved from 0.75451 to 0.47842, saving model to ResUNet-weights.hdf5
       tversky: 0.6244
       Epoch 3/40
                 72/72 [=====
       Epoch 3: val loss did not improve from 0.47842
       72/72 [==========] - 106s 1s/step - loss: 0.3789 - tversky: 0.7236 - val loss: 0.5118 - val
        _tversky: 0.5892
       Epoch 4/40
       72/72 [=======] - ETA: 0s - loss: 0.3349 - tversky: 0.7658
       Epoch 4: val loss did not improve from 0.47842
       72/72 [==========] - 108s 2s/step - loss: 0.3349 - tversky: 0.7658 - val loss: 0.5491 - val
        tversky: 0.5476
       Epoch 5/40
       72/72 [========] - ETA: 0s - loss: 0.3042 - tversky: 0.7933
       Epoch 5: val_loss improved from 0.47842 to 0.34035, saving model to ResUNet-weights.hdf5
       72/72 [=========] - 149s 2s/step - loss: 0.3042 - tversky: 0.7933 - val loss: 0.3403 - val
       tversky: 0.7612
       Epoch 6/40
       72/72 [=========] - ETA: 0s - loss: 0.2718 - tversky: 0.8224
       Epoch 6: val_loss improved from 0.34035 to 0.28135, saving model to ResUNet-weights.hdf5
                 val tversky: 0.8152
       Epoch 7/40
       Epoch 7: val_loss did not improve from 0.28135
       72/72 [=========] - 146s 2s/step - loss: 0.2590 - tversky: 0.8336 - val loss: 0.4022 - val
        tverskv: 0.7023
       Epoch 8/40
       72/72 [==========] - ETA: 0s - loss: 0.2463 - tversky: 0.8442
       Epoch 8: val loss did not improve from 0.28135
       72/72 [==========] - 141s 2s/step - loss: 0.2463 - tversky: 0.8442 - val loss: 0.3328 - val
        tversky: 0.7686
       Epoch 9/40
       72/72 [===========] - ETA: 0s - loss: 0.2289 - tversky: 0.8588
       Epoch 9: val_loss did not improve from 0.28135
       72/72 [=========] - 138s 2s/step - loss: 0.2289 - tversky: 0.8588 - val loss: 0.3500 - val
        tversky: 0.7532
       Epoch 10/40
       72/72 [==========] - ETA: 0s - loss: 0.2254 - tversky: 0.8606
       Epoch 10: val loss improved from 0.28135 to 0.26285, saving model to ResUNet-weights.hdf5
       72/72 [=====
                          ========] - 139s 2s/step - loss: 0.2254 - tversky: 0.8606 - val loss: 0.2629 - val
        _tversky: 0.8311
       Epoch 11/40
                   Epoch 11: val\_loss improved from 0.26285 to 0.24320, saving model to ResÜNet-weights.hdf5
       tversky: 0.8473
       Epoch 12/40
       Epoch 12: val loss did not improve from 0.24320
       72/72 [============] - 134s 2s/step - loss: 0.1986 - tversky: 0.8831 - val loss: 0.2750 - val
```

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tversky: 0.8198
Epoch 13/40
72/72 [====
          Epoch 13: val loss did not improve from 0.24320
72/72 [===========] - 139s 2s/step - loss: 0.2041 - tversky: 0.8785 - val loss: 0.2935 - val
tversky: 0.8044
Epoch 14/40
72/72 [==========] - ETA: 0s - loss: 0.1905 - tversky: 0.8891
Epoch 14: val loss did not improve from 0.24320
            tversky: 0.8071
Epoch 15/40
72/72 [=========] - ETA: 0s - loss: 0.1786 - tversky: 0.8986
Epoch 15: val loss improved from 0.24320 to 0.21935, saving model to ResUNet-weights.hdf5
72/72 [=========] - 133s 2s/step - loss: 0.1786 - tversky: 0.8986 - val loss: 0.2194 - val
tversky: 0.8659
Epoch 16/40
72/72 [==========] - ETA: 0s - loss: 0.1713 - tversky: 0.9041
Epoch 16: val loss did not improve from 0.21935
72/72 [========] - 134s 2s/step - loss: 0.1713 - tversky: 0.9041 - val_loss: 0.2678 - val
tversky: 0.8269
Epoch 17/40
Epoch 17: val loss improved from 0.21935 to 0.21474, saving model to ResUNet-weights.hdf5
72/72 [========] - 130s 2s/step - loss: 0.1593 - tversky: 0.9130 - val loss: 0.2147 - val
tverskv: 0.8710
Epoch 18/40
72/72 [===========] - ETA: 0s - loss: 0.1494 - tversky: 0.9202
Epoch 18: val loss improved from 0.21474 to 0.20514, saving model to ResUNet-weights.hdf5
72/72 [==========] - 134s 2s/step - loss: 0.1494 - tversky: 0.9202 - val loss: 0.2051 - val
tversky: 0.8777
Epoch 19/40
72/72 [============= ] - ETA: 0s - loss: 0.1488 - tversky: 0.9205
Epoch 19: val loss did not improve from 0.20514
72/72 [===========] - 133s 2s/step - loss: 0.1488 - tversky: 0.9205 - val loss: 0.2670 - val
tversky: 0.8268
Epoch 20/40
72/72 [===========] - ETA: 0s - loss: 0.1505 - tversky: 0.9193
Epoch \overset{\circ}{20}: val loss improved from 0.20514 to 0.19853, saving model to ResUNet-weights.hdf5
72/72 [=========] - 133s 2s/step - loss: 0.1505 - tversky: 0.9193 - val loss: 0.1985 - val
tversky: 0.8839
Epoch 21/40
72/72 [====
           Epoch 21: val loss improved from 0.19853 to 0.18324, saving model to ResUNet-weights.hdf5
tversky: 0.8954
Epoch 22/40
Epoch 22: val loss did not improve from 0.18324
72/72 [========] - 133s 2s/step - loss: 0.1495 - tversky: 0.9200 - val loss: 0.1991 - val
tverskv: 0.8829
Epoch 23/40
          Epoch 23: val loss did not improve from 0.18324
tversky: 0.8777
Epoch 24/40
72/72 [=========] - ETA: 0s - loss: 0.1454 - tversky: 0.9231
Epoch 24: val_loss did not improve from 0.18324
          72/72 [=====
tverskv: 0.8553
Epoch 25/40
72/72 [==========] - ETA: 0s - loss: 0.1339 - tversky: 0.9312
Epoch 25: val loss did not improve from 0.18324
tversky: 0.8908
72/72 [===========] - ETA: 0s - loss: 0.1268 - tversky: 0.9361
Epoch \overset{\circ}{2}6: val_loss improved from 0.18324 to 0.17267, saving model to ResUNet-weights.hdf5
tversky: 0.9037
Epoch 27/40
72/72 [=====
          Epoch 27: val loss did not improve from 0.17267
72/72 [===========] - 532s 7s/step - loss: 0.1204 - tversky: 0.9403 - val loss: 0.1820 - val
tversky: 0.8968
Epoch 28/40
72/72 [========] - ETA: 0s - loss: 0.1201 - tversky: 0.9405
Epoch 28: val loss did not improve from 0.17267
72/72 [===
          tversky: 0.8972
Epoch 29/40
72/72 [=============] - ETA: 0s - loss: 0.1212 - tversky: 0.9397
Epoch 29: val_loss did not improve from 0.17267
72/72 [=========] - 133s 2s/step - loss: 0.1212 - tversky: 0.9397 - val loss: 0.1850 - val
_tversky: 0.8944
Epoch 30/40
```

72/72 [=========] - ETA: 0s - loss: 0.1179 - tversky: 0.9419

Epoch 30: val loss did not improve from 0.17267

```
_tversky: 0.8995
      Epoch 31/40
      72/72 [========] - ETA: 0s - loss: 0.1125 - tversky: 0.9455
      Epoch 31: val loss did not improve from 0.17267
      tversky: 0.8875
      Epoch 32/40
      Epoch 32: val_loss improved from 0.17267 to 0.17189, saving model to ResUNet-weights.hdf5
      72/72 [=========] - 132s 2s/step - loss: 0.1082 - tversky: 0.9482 - val loss: 0.1719 - val
       tversky: 0.9043
      Epoch 33/40
      Epoch 33: val loss did not improve from 0.17189
      72/72 [===========] - 132s 2s/step - loss: 0.1077 - tversky: 0.9485 - val_loss: 0.1899 - val
       tversky: 0.8907
      Epoch 34/40
      72/72 [==========] - ETA: 0s - loss: 0.1118 - tversky: 0.9459
      Epoch 34: val_loss did not improve from 0.17189
      72/72 [=========] - 131s 2s/step - loss: 0.1118 - tversky: 0.9459 - val loss: 0.1929 - val
       _tversky: 0.8874
      Epoch 35/40
      Epoch 35: val loss did not improve from 0.17189
      _tversky: 0.8973
      Epoch 36/40
      72/72 [=====
                 Epoch 36: val_loss did not improve from 0.17189
      72/72 [=========] - 131s 2s/step - loss: 0.1038 - tversky: 0.9510 - val loss: 0.2112 - val
       tversky: 0.8725
      Epoch 37/40
      72/72 [============= ] - ETA: 0s - loss: 0.1012 - tversky: 0.9527
      Epoch 37: val loss did not improve from 0.17189
      72/72 [==
                    :=========] - 131s 2s/step - loss: 0.1012 - tversky: 0.9527 - val loss: 0.1852 - val
       tversky: 0.8936
      Epoch 38/40
      72/72 [=========] - ETA: 0s - loss: 0.0993 - tversky: 0.9539
      Epoch 38: val_loss did not improve from 0.17189
      72/72 [=========] - 131s 2s/step - loss: 0.0993 - tversky: 0.9539 - val loss: 0.1868 - val
       tversky: 0.8928
      Epoch 39/40
      72/72 [===========] - ETA: 0s - loss: 0.0980 - tversky: 0.9547
      Epoch 39: val_loss improved from 0.17189 to 0.16629, saving model to ResUNet-weights.hdf5
      72/72 [=========] - 131s 2s/step - loss: 0.0980 - tversky: 0.9547 - val loss: 0.1663 - val
       tversky: 0.9081
      Epoch 40/40
      72/72 [=====
                  Epoch 40: val loss did not improve from 0.16629
      tversky: 0.9004
In [56]: # save model architecture to json file for future use
      model_json = model_seg.to_json()
      with open('ResUNet model.json', 'w') as json file:
         json_file.write(model json)
      Assess Trained Segmentation ResUNet
In [58]: from utilities import focal_tversky, tversky_loss, tversky
      with open('ResUNet model.json', 'r') as json file:
         json savedModel= json file.read()
      # load the model architecture
      model seg = tf.keras.models.model from json(json savedModel)
      model_seg.load_weights('ResUNet-weights.hdf5')
      adam = tf.keras.optimizers.Adam(lr = 0.05, epsilon = 0.1)
      model seq.compile(optimizer = adam, loss = focal tversky, metrics = [tversky])
      WARNING:absl:`lr` is deprecated in Keras optimizer, please use `learning_rate` or use the legacy optimizer, e.g
      .,tf.keras.optimizers.legacy.Adam.
In [62]: # Utilities file contains the code for custom loss function and custom data generator
      from utilities import prediction
      # making prediction
      image_id, mask, has_mask = prediction(test, model, model_seg)
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72/72 [==========] - 134s 2s/step - loss: 0.1179 - tversky: 0.9419 - val loss: 0.1781 - val

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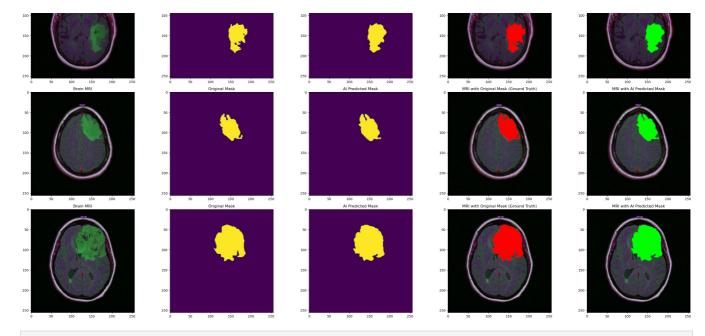
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1/1 [======] - 0s 201ms/step
1/1 [=======] - 0s 207ms/step
1/1 [=======] - 0s 115ms/step
1/1 [=======] - 0s 214ms/step
1/1 [======= ] - 0s 215ms/step
1/1 [======] - 0s 215ms/step
1/1 [=======] - 0s 205ms/step
1/1 [======] - 0s 207ms/step
1/1 [======] - 0s 201ms/step
1/1 [======] - 0s 229ms/step
1/1 [=======] - 0s 116ms/step
1/1 [=======] - 0s 233ms/step
1/1 [======] - 0s 229ms/step
1/1 [=======] - 0s 207ms/step
1/1 [======] - 0s 208ms/step
1/1 [======= ] - 0s 113ms/step
1/1 [======] - 0s 82ms/step
1/1 [=======] - 0s 220ms/step
1/1 [======] - 0s 99ms/step
1/1 [=======] - 0s 200ms/step
1/1 [======] - 0s 113ms/step
1/1 [======] - 0s 220ms/step
1/1 [======] - 0s 190ms/step
1/1 [=======] - 0s 112ms/step
```

```
1/1
                                       - 0s 106ms/step
        1/1 [======] - 0s 216ms/step
        1/1 [======] - 0s 210ms/step
        1/1 [=======] - 0s 98ms/step
        1/1 [==
                    ======= ] - 0s 209ms/step
        1/1 [======] - 0s 214ms/step
        1/1 [=======] - 0s 98ms/step
        1/1 [===
               ======| - 0s 208ms/step
        1/1 [======] - 0s 208ms/step
        [======] - 0s 190ms/step
        1/1
        # creating a dataframe for the result
In [63]:
        df pred = pd.DataFrame({'image path': image id,'predicted mask': mask,'has mask': has mask})
        df pred
Out[63]:
                                          image_path
                                                                     predicted_mask has_mask
            TCGA DU 5872 19950223/TCGA DU 5872 19950223 67...
                                                                                      0
                                                                          No mask
            TCGA_FG_A4MT_20020212/TCGA_FG_A4MT_20020212_3.tif
                                                                          No mask
                                                                                      0
         2 TCGA_DU_A5TW_19980228/TCGA_DU_A5TW_19980228_29... [[[[6.579732e-06], [5.0200333e-07], [1.8209814...
             TCGA_FG_7637_20000922/TCGA_FG_7637_20000922_17... [[[[1.3748944e-06], [5.6489725e-08], [3.797600...
         4
             TCGA_CS_6666_20011109/TCGA_CS_6666_20011109_25... [[[[4.1414055e-06], [1.4924562e-06], [7.196988...
            TCGA_HT_A61A_20000127/TCGA_HT_A61A_20000127_23... [[[[3.941212e-06], [1.2909313e-06], [7.090385e...
        585
                                                                                      1
        586
             TCGA_DU_5872_19950223/TCGA_DU_5872_19950223_69...
                                                                          No mask
                                                                                      0
             TCGA_DU_6404_19850629/TCGA_DU_6404_19850629_24...
        587
                                                                          No mask
                                                                                      0
             TCGA_DU_8165_19970205/TCGA_DU_8165_19970205_7.tif
        588
                                                                          No mask
                                                                                      0
        589
             TCGA_DU_6404_19850629/TCGA_DU_6404_19850629_53...
                                                                          No mask
                                                                                      0
       590 rows × 3 columns
In [64]: # Merge the dataframe containing predicted results with the original test data.
        df pred = test.merge(df_pred, on = 'image_path')
        df_pred.head()
                                        image path
                                                                                mask path mask predicted mask
Out[64]:
           TCGA_DU_5872_19950223/TCGA_DU_5872_19950223_67...
                                                   TCGA_DU_5872_19950223/TCGA_DU_5872_19950223_67...
                                                                                                  No mask
           TCGA_FG_A4MT_20020212/TCGA_FG_A4MT_20020212_3.iif TCGA_FG_A4MT_20020212/TCGA_FG_A4MT_20020212_3_...
                                                                                                  No mask
                                                                                               [[[6.579732e
                                                                                                     061
        2 TCGA_DU_A5TW_19980228/TCGA_DU_A5TW_19980228_29... TCGA_DU_A5TW_19980228/TCGA_DU_A5TW_19980228_29...
                                                                                             [5.0200333e-07]
                                                                                               [1.8209814...
                                                                                              [[[[1.3748944e
           TCGA_FG_7637_20000922/TCGA_FG_7637_20000922_17...
                                                   TCGA_FG_7637_20000922/TCGA_FG_7637_20000922_17...
                                                                                             [5.6489725e-08]
                                                                                                [3.797600..
                                                                                              [[[[4.1414055e
           TCGA CS 6666 20011109/TCGA CS 6666 20011109 25...
                                                   TCGA CS 6666 20011109/TCGA CS 6666 20011109 25...
                                                                                             [1.4924562e-06]
                                                                                                [7.196988...
In [65]:
        count = 0
        fig, axs = plt.subplots(10, 5, figsize=(30, 50))
        for i in range(len(df_pred)):
         if df pred['has mask'][i] == 1 and count < 10:</pre>
           # read the images and convert them to RGB format
           img = io.imread(df_pred.image_path[i])
           img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
           axs[count][0].title.set text("Brain MRI")
           axs[count][0].imshow(img)
           # Obtain the mask for the image
           mask = io.imread(df_pred.mask_path[i])
axs[count][1].title.set_text("Original Mask")
           axs[count][1].imshow(mask)
```

```
# Obtain the predicted mask for the image
    predicted_mask = np.asarray(df_pred.predicted_mask[i])[0].squeeze().round()
    axs[count][2].title.set_text("AI Predicted Mask")
    axs[count][2].imshow(predicted_mask)
    # Apply the mask to the image 'mask==255'
    img[mask == 255] = (255, 0, 0)
    axs[count][3].title.set_text("MRI with Original Mask (Ground Truth)")
    axs[count][3].imshow(img)
    img_ = io.imread(df_pred.image_path[i])
img_ = cv2.cvtColor(img_, cv2.COLOR_BGR2RGB)
img_[predicted_mask == 1] = (0, 255, 0)
    axs[count][4].title.set_text("MRI with AI Predicted Mask")
    axs[count][4].imshow(img_)
    count += 1
fig.tight_layout()
```



In []:

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