

# Brain Tumor Segmentation using Deep Learning

```
In [1]: !apt-get install -y -qq software-properties-common python-software-properties mo
!add-apt-repository -y ppa:alessandro-strada/ppa 2>&1 > /dev/null
!apt-get update -qq 2>&1 > /dev/null
!apt-get -y install -qq google-drive-ocamlfuse fuse
from google.colab import auth
auth.authenticate_user()
from oauth2client.client import GoogleCredentials
creds = GoogleCredentials.get_application_default()
import getpass
!google-drive-ocamlfuse -headless -id={creds.client_id} -secret={creds.client_se
vcode = getpass.getpass()
!echo {vcode} | google-drive-ocamlfuse -headless -id={creds.client_id} -secret={c
```

Please, open the following URL in a web browser: [https://accounts.google.com/o/oauth2/auth?client\\_id=32555940559.apps.googleusercontent.com&redirect\\_uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%3Aoob&scope=https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdrive&response\\_type=code&access\\_type=offline&approval\\_prompt=force](https://accounts.google.com/o/oauth2/auth?client_id=32555940559.apps.googleusercontent.com&redirect_uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%3Aoob&scope=https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdrive&response_type=code&access_type=offline&approval_prompt=force) (https://accounts.google.com/o/oauth2/auth?client\_id=32555940559.apps.googleusercontent.com&redirect\_uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%3Aoob&scope=https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdrive&response\_type=code&access\_type=offline&approval\_prompt=force)

.....

Please, open the following URL in a web browser: [https://accounts.google.com/o/oauth2/auth?client\\_id=32555940559.apps.googleusercontent.com&redirect\\_uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%3Aoob&scope=https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdrive&response\\_type=code&access\\_type=offline&approval\\_prompt=force](https://accounts.google.com/o/oauth2/auth?client_id=32555940559.apps.googleusercontent.com&redirect_uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%3Aoob&scope=https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdrive&response_type=code&access_type=offline&approval_prompt=force) (https://accounts.google.com/o/oauth2/auth?client\_id=32555940559.apps.googleusercontent.com&redirect\_uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%3Aoob&scope=https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdrive&response\_type=code&access\_type=offline&approval\_prompt=force)

Please enter the verification code: Access token retrieved correctly.

#Dataset link <https://www.kaggle.com/awsaf49/brats2020-training-data>  
(<https://www.kaggle.com/awsaf49/brats2020-training-data>)

```
In [2]: !mkdir -p drive
!google-drive-ocamlfuse drive
```

```
In [0]: import os
os.chdir('drive/brat')
```

In [7]: `!ls`

```
Brain Tumor Segmentation.odt  slice_126_0
data.ipynb                   slice_126_1
data_trial_81.h5              slice_126_2
data_trial_dim2_128.h5        slice_126_3
data_Y_0001.pickle            slice_126_GT
FinalCode.ipynb               training.ipynb
HG                             trial_0001_2path_acc.h5
info1_input.h5                trial_0001_accuracy.h5
info_mfc.h5                   trial_0001_input_cascade_acc.h5
LG                             trial_0001_MFCcas_dim2_128_acc.h5
MFC_cascade_siam.h5           trial_MFCcascade_acc.h5
model.ipynb                   untitled.ipynb
siamese_training.ipynb
```

In [2]: `!pip3 install SimpleITK`

Requirement already satisfied: SimpleITK in /usr/local/lib/python3.6/dist-packages (1.1.0)

In [0]: `import SimpleITK as sitk`  
`import numpy as np`

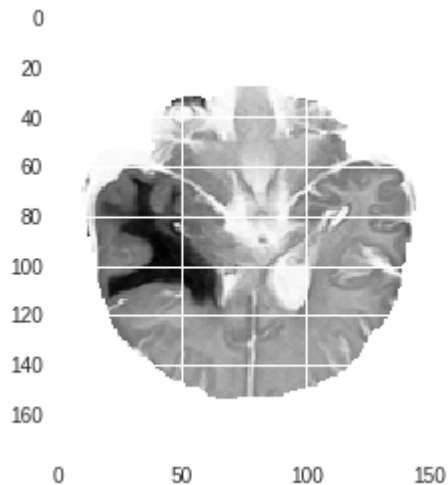
### Data Visualisations

In [7]: `import os`  
`path = 'HG/0001'`  
`p = os.listdir(path)`  
`p.sort(key=str.lower)`  
`arr = []`  
`for i in range(len(p)):`  
 `if(i != 4):`  
 `p1 = os.listdir(path+'/'+p[i])`  
 `p1.sort()`  
 `img = sitk.ReadImage(path+'/'+p[i]+'/' +p1[-1])`  
 `arr.append(sitk.GetArrayFromImage(img))`  
 `else:`  
 `p1 = os.listdir(path+'/'+p[i])`  
 `img = sitk.ReadImage(path+'/'+p[i]+'/' +p1[0])`  
 `Y_labels = sitk.GetArrayFromImage(img)`  
 `print(Y_labels.shape)`  
`data = np.zeros((Y_labels.shape[1],Y_labels.shape[0],Y_labels.shape[2],4))`  
`for i in range(Y_labels.shape[1]):`  
 `data[i,:,:0] = arr[0][:,i,:]`  
 `data[i,:,:1] = arr[1][:,i,:]`  
 `data[i,:,:2] = arr[2][:,i,:]`  
 `data[i,:,:3] = arr[3][:,i,:]`

(176, 216, 160)

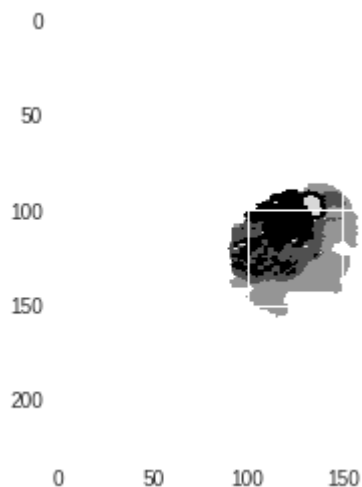
```
In [14]: %pylab inline
import matplotlib.pyplot as plt
img = data[126,:,:0]
imgplot = plt.imshow(img)
plt.show()
plt.imsave('slice_126_4',img,cmap='gray')
```

Populating the interactive namespace from numpy and matplotlib



```
In [28]: %pylab inline
import matplotlib.pyplot as plt
img = Y_labels[:,126,:]
imgplot = plt.imshow(img)
plt.show()
#plt.imsave('slice_126_GT',img,cmap='gray')
```

Populating the interactive namespace from numpy and matplotlib



```
In [0]: def model_gen(input_dim,x,y,slice_no):
    X1 = []
    X2 = []
    Y = []

    for i in range(int((input_dim)/2),y.shape[0]-int((input_dim)/2)):
        for j in range(int((input_dim)/2),y.shape[2]-int((input_dim)/2)):
            #Filtering all 0 patches
            if(x[i-16:i+17,j-16:j+17,:].any != 0):
                X2.append(x[i-16:i+17,j-16:j+17,:])
                X1.append(x[i-int((input_dim)/2):i+int((input_dim)/2)+1,j-int((input_dim)/2):j+int((input_dim)/2)+1,:])
                Y.append(y[i,slice_no,j])

    X1 = np.asarray(X1)
    X2 = np.asarray(X2)
    Y = np.asarray(Y)
    d = [X1,X2,Y]
    return d
```

```
In [0]: def data_gen(data,y,slice_no,model_no):
    d = []
    x = data[slice_no]
    #filtering all 0 slices and non-tumor slices
    if(x.any() != 0 and y.any() != 0):
        if(model_no == 0):
            X1 = []
            for i in range(16,159):
                for j in range(16,199):
                    if(x[i-16:i+17,j-16:j+17,:].all != 0):
                        X1.append(x[i-16:i+17,j-16:j+17,:])
            Y1 = []
            for i in range(16,159):
                for j in range(16,199):
                    if(x[i-16:i+17,j-16:j+17,:].all != 0):
                        Y1.append(y[i,slice_no,j])
            X1 = np.asarray(X1)
            Y1 = np.asarray(Y1)
            d = [X1,Y1]
        elif(model_no == 1):
            d = model_gen(65,x,y,slice_no)
        elif(model_no == 2):
            d = model_gen(56,x,y,slice_no)
        elif(model_no == 3):
            d = model_gen(53,x,y,slice_no)

    return d
```

Model Definations

```
In [8]: import keras
```

Using TensorFlow backend.

```
In [0]: import keras
from keras import layers
from keras.layers import Input, Dense, Activation, ZeroPadding2D, BatchNormalizat
from keras.layers import AveragePooling2D, MaxPooling2D, Dropout, GlobalMaxPoolin
from keras.models import Model
from keras import regularizers
from keras.preprocessing import image
from keras.utils import layer_utils
from keras.utils.data_utils import get_file
from keras.applications.imagenet_utils import preprocess_input
from keras.initializers import glorot_normal
#import pydot
from IPython.display import SVG
from keras.utils.vis_utils import model_to_dot
from keras.utils import plot_model
```

```
In [37]: def two_path(X_input):
    # Local path Conv1
    X = Conv2D(64,(7,7),strides=(1,1),padding='valid')(X_input)
    # Batch-norm
    X = BatchNormalization()(X)
    X1 = Conv2D(64,(7,7),strides=(1,1),padding='valid')(X_input)
    X1 = BatchNormalization()(X1)
    # Max-out
    X = layers.Maximum()([X,X1])
    X = Conv2D(64,(4,4),strides=(1,1),padding='valid',activation='relu')(X)

    # Global path
    X2 = Conv2D(160,(13,13),strides=(1,1),padding='valid')(X_input)
    X2 = BatchNormalization()(X2)
    X21 = Conv2D(160,(13,13),strides=(1,1),padding='valid')(X_input)
    X21 = BatchNormalization()(X21)
    # Max-out
    X2 = layers.Maximum()([X2,X21])

    # Local path Conv2
    X3 = Conv2D(64,(3,3),strides=(1,1),padding='valid')(X)
    X3 = BatchNormalization()(X3)
    X31 = Conv2D(64,(3,3),strides=(1,1),padding='valid')(X)
    X31 = BatchNormalization()(X31)
    X = layers.Maximum()([X3,X31])
    X = Conv2D(64,(2,2),strides=(1,1),padding='valid',activation='relu')(X)

    # Merging the two paths
    X = Concatenate()([X2,X])
    #X = Conv2D(5,(21,21),strides=(1,1))(X)
    #X = Activation('softmax')(X)

    #model = Model(inputs = X_input, outputs = X)
    return X
```

```
In [38]: def input_cascade(input_shape1,input_shape2):

    X1_input = Input(input_shape1)
    # 1st two-path of cascade
    X1 = two_path(X1_input)
    X1 = Conv2D(5,(21,21),strides=(1,1),padding='valid',activation='relu')(X1)
    X1 = BatchNormalization()(X1)

    X2_input = Input(input_shape2)
    # Concatenating the output of 1st to input of 2nd
    X2_input1 = Concatenate()([X1,X2_input])
    #X2_input1 = Input(tensor = X2_input1)
    X2 = two_path(X2_input1)
    # Fully convolutional softmax classification
    X2 = Conv2D(5,(21,21),strides=(1,1),padding='valid')(X2)
    X2 = BatchNormalization()(X2)
    X2 = Activation('softmax')(X2)

    model = Model(inputs=[X1_input,X2_input],outputs=X2)
    return model
```

```
In [39]: def MFCcascade(input_shape1,input_shape2):

    # 1st two-path
    X1_input = Input(input_shape1)
    X1 = two_path(X1_input)
    X1 = Conv2D(5,(21,21),strides=(1,1),padding='valid',activation='relu')(X1)
    X1 = BatchNormalization()(X1)
    #X1 = MaxPooling2D((2,2))(X1)

    #2nd two-path
    X2_input = Input(input_shape2)
    X2 = two_path(X2_input)

    # Concatenate before classification
    X2 = Concatenate()([X1,X2])
    X2 = Conv2D(5,(21,21),strides=(1,1),padding='valid',activation='relu')(X2)
    X2 = BatchNormalization()(X2)
    X2 = Activation('softmax')(X2)

    model = Model(inputs=[X1_input,X2_input],outputs=X2)
    return model
```

```
In [12]: def two_pathcnn(input_shape):

    X_input = Input(input_shape)

    X = Conv2D(64,(7,7),strides=(1,1),padding='valid')(X_input)
    X = BatchNormalization()(X)
    X1 = Conv2D(64,(7,7),strides=(1,1),padding='valid')(X_input)
    X1 = BatchNormalization()(X1)
    X = layers.Maximum()([X,X1])
    X = Conv2D(64,(4,4),strides=(1,1),padding='valid',activation='relu')(X)

    X2 = Conv2D(160,(13,13),strides=(1,1),padding='valid')(X_input)
    X2 = BatchNormalization()(X2)
    X21 = Conv2D(160,(13,13),strides=(1,1),padding='valid')(X_input)
    X21 = BatchNormalization()(X21)
    X2 = layers.Maximum()([X2,X21])

    X3 = Conv2D(64,(3,3),strides=(1,1),padding='valid')(X)
    X3 = BatchNormalization()(X3)
    X31 = Conv2D(64,(3,3),strides=(1,1),padding='valid')(X)
    X31 = BatchNormalization()(X31)
    X = layers.Maximum()([X3,X31])
    X = Conv2D(64,(2,2),strides=(1,1),padding='valid',activation='relu')(X)

    X = Concatenate()([X2,X])
    X = Conv2D(5,(21,21),strides=(1,1),padding='valid')(X)
    X = Activation('softmax')(X)

    model = Model(inputs = X_input, outputs = X)
    return model
```

```
In [13]: m0 = two_pathcnn((33,33,4))
m0.summary()
```

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	(None, 33, 33, 4)	0	
conv2d_1 (Conv2D)	(None, 27, 27, 64)	12608	input_1[0][0]
conv2d_2 (Conv2D)	(None, 27, 27, 64)	12608	input_1[0][0]
batch_normalization_1 (BatchNor	(None, 27, 27, 64)	256	conv2d_1[0][0]
batch_normalization_2 (BatchNor	(None, 27, 27, 64)	256	conv2d_2[0][0]
maximum_1 (Maximum)	(None, 27, 27, 64)	0	batch_normaliz
ation_1[0][0]			ation_2[0][0]
conv2d_3 (Conv2D)	(None, 24, 24, 64)	65600	maximum_1[0]
conv2d_6 (Conv2D)	(None, 22, 22, 64)	36928	conv2d_3[0][0]
conv2d_7 (Conv2D)	(None, 22, 22, 64)	36928	conv2d_3[0][0]
conv2d_4 (Conv2D)	(None, 21, 21, 160)	108320	input_1[0][0]
conv2d_5 (Conv2D)	(None, 21, 21, 160)	108320	input_1[0][0]
batch_normalization_5 (BatchNor	(None, 22, 22, 64)	256	conv2d_6[0][0]
batch_normalization_6 (BatchNor	(None, 22, 22, 64)	256	conv2d_7[0][0]
batch_normalization_3 (BatchNor	(None, 21, 21, 160)	640	conv2d_4[0][0]
batch_normalization_4 (BatchNor	(None, 21, 21, 160)	640	conv2d_5[0][0]



maximum_3 (Maximum) ation_5[0][0]	(None, 22, 22, 64)	0	batch_normaliz batch_normaliz ation_6[0][0]
maximum_2 (Maximum) ation_3[0][0]	(None, 21, 21, 160)	0	batch_normaliz batch_normaliz ation_4[0][0]
conv2d_8 (Conv2D) [0]	(None, 21, 21, 64)	16448	maximum_3[0]
concatenate_1 (Concatenate) [0]	(None, 21, 21, 224)	0	maximum_2[0] conv2d_8[0][0]
conv2d_9 (Conv2D) [0][0]	(None, 1, 1, 5)	493925	concatenate_1
activation_1 (Activation)	(None, 1, 1, 5)	0	conv2d_9[0][0]
=====			
Total params: 893,989			
Trainable params: 892,837			
Non-trainable params: 1,152			



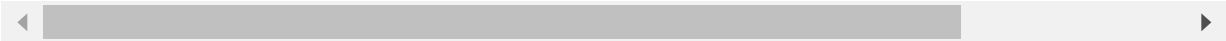
```
In [40]: m1 = MFCascade((53,53,4),(33,33,4))
m1.summary()
```

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	(None, 53, 53, 4)	0	
conv2d_1 (Conv2D)	(None, 47, 47, 64)	12608	input_1[0][0]
conv2d_2 (Conv2D)	(None, 47, 47, 64)	12608	input_1[0][0]
batch_normalization_1 (BatchNor	(None, 47, 47, 64)	256	conv2d_1[0][0]
batch_normalization_2 (BatchNor	(None, 47, 47, 64)	256	conv2d_2[0][0]
input_2 (InputLayer)	(None, 33, 33, 4)	0	
maximum_1 (Maximum)	(None, 47, 47, 64)	0	batch_normaliz
ation_1[0][0]			batch_normaliz
ation_2[0][0]			
conv2d_10 (Conv2D)	(None, 27, 27, 64)	12608	input_2[0][0]
conv2d_11 (Conv2D)	(None, 27, 27, 64)	12608	input_2[0][0]
conv2d_3 (Conv2D)	(None, 44, 44, 64)	65600	maximum_1[0]
[0]			
batch_normalization_8 (BatchNor	(None, 27, 27, 64)	256	conv2d_10[0]
[0]			
batch_normalization_9 (BatchNor	(None, 27, 27, 64)	256	conv2d_11[0]
[0]			
conv2d_6 (Conv2D)	(None, 42, 42, 64)	36928	conv2d_3[0][0]
conv2d_7 (Conv2D)	(None, 42, 42, 64)	36928	conv2d_3[0][0]

maximum_4 (Maximum) ation_8[0][0]	(None, 27, 27, 64)	0	batch_normaliz ation_9[0][0]
conv2d_4 (Conv2D)	(None, 41, 41, 160)	108320	input_1[0][0]
conv2d_5 (Conv2D)	(None, 41, 41, 160)	108320	input_1[0][0]
batch_normalization_5 (BatchNor	(None, 42, 42, 64)	256	conv2d_6[0][0]
batch_normalization_6 (BatchNor	(None, 42, 42, 64)	256	conv2d_7[0][0]
conv2d_12 (Conv2D) [0]	(None, 24, 24, 64)	65600	maximum_4[0]
batch_normalization_3 (BatchNor	(None, 41, 41, 160)	640	conv2d_4[0][0]
batch_normalization_4 (BatchNor	(None, 41, 41, 160)	640	conv2d_5[0][0]
maximum_3 (Maximum) ation_5[0][0]	(None, 42, 42, 64)	0	batch_normaliz ation_6[0][0]
conv2d_15 (Conv2D) [0]	(None, 22, 22, 64)	36928	conv2d_12[0]
conv2d_16 (Conv2D) [0]	(None, 22, 22, 64)	36928	conv2d_12[0]
maximum_2 (Maximum) ation_3[0][0]	(None, 41, 41, 160)	0	batch_normaliz ation_4[0][0]
conv2d_8 (Conv2D) [0]	(None, 41, 41, 64)	16448	maximum_3[0]
conv2d_13 (Conv2D)	(None, 21, 21, 160)	108320	input_2[0][0]
conv2d_14 (Conv2D)	(None, 21, 21, 160)	108320	input_2[0][0]

batch_normalization_12 (BatchNo	(None, 22, 22, 64)	256	conv2d_15[0]
[0]			
batch_normalization_13 (BatchNo	(None, 22, 22, 64)	256	conv2d_16[0]
[0]			
concatenate_1 (Concatenate)	(None, 41, 41, 224)	0	maximum_2[0]
[0]			conv2d_8[0][0]
batch_normalization_10 (BatchNo	(None, 21, 21, 160)	640	conv2d_13[0]
[0]			
batch_normalization_11 (BatchNo	(None, 21, 21, 160)	640	conv2d_14[0]
[0]			
maximum_6 (Maximum)	(None, 22, 22, 64)	0	batch_normaliz
ation_12[0][0]			batch_normaliz
ation_13[0][0]			
conv2d_9 (Conv2D)	(None, 21, 21, 5)	493925	concatenate_1
[0][0]			
maximum_5 (Maximum)	(None, 21, 21, 160)	0	batch_normaliz
ation_10[0][0]			batch_normaliz
ation_11[0][0]			
conv2d_17 (Conv2D)	(None, 21, 21, 64)	16448	maximum_6[0]
[0]			
batch_normalization_7 (BatchNor	(None, 21, 21, 5)	20	conv2d_9[0][0]
concatenate_2 (Concatenate)	(None, 21, 21, 224)	0	maximum_5[0]
[0]			conv2d_17[0]
[0]			
concatenate_3 (Concatenate)	(None, 21, 21, 229)	0	batch_normaliz
ation_7[0][0]			concatenate_2
[0][0]			

conv2d_18 (Conv2D)	(None, 1, 1, 5)	504950	concatenate_3 [0][0]
<hr/>			
batch_normalization_14 (Batch Normalization)	(None, 1, 1, 5)	20	conv2d_18[0] [0]
<hr/>			
activation_1 (Activation)	(None, 1, 1, 5)	0	batch_normalization_14[0][0]
<hr/>			
=====			
=====			
Total params: 1,799,043			
Trainable params: 1,796,719			
Non-trainable params: 2,324			
<hr/>			



```
In [41]: m1 = input_cascade((65,65,4),(33,33,4))
m1.summary()
```

Layer (type)	Output Shape	Param #	Connected to
=====			
input_3 (InputLayer)	(None, 65, 65, 4)	0	
conv2d_19 (Conv2D)	(None, 59, 59, 64)	12608	input_3[0][0]
conv2d_20 (Conv2D)	(None, 59, 59, 64)	12608	input_3[0][0]
batch_normalization_15 (Batch Normalization)	(None, 59, 59, 64)	256	conv2d_19[0]
batch_normalization_16 (Batch Normalization)	(None, 59, 59, 64)	256	conv2d_20[0]
maximum_7 (Maximum)	(None, 59, 59, 64)	0	batch_normalization_15[0][0] batch_normalization_16[0][0]
conv2d_21 (Conv2D)	(None, 56, 56, 64)	65600	maximum_7[0]
conv2d_24 (Conv2D)	(None, 54, 54, 64)	36928	conv2d_21[0]
conv2d_25 (Conv2D)	(None, 54, 54, 64)	36928	conv2d_21[0]
conv2d_22 (Conv2D)	(None, 53, 53, 160)	108320	input_3[0][0]
conv2d_23 (Conv2D)	(None, 53, 53, 160)	108320	input_3[0][0]
batch_normalization_19 (Batch Normalization)	(None, 54, 54, 64)	256	conv2d_24[0]
batch_normalization_20 (Batch Normalization)	(None, 54, 54, 64)	256	conv2d_25[0]

batch_normalization_17 (BatchNo	(None, 53, 53, 160)	640	conv2d_22[0]
[0]			
batch_normalization_18 (BatchNo	(None, 53, 53, 160)	640	conv2d_23[0]
[0]			
maximum_9 (Maximum)	(None, 54, 54, 64)	0	batch_normaliz
ation_19[0][0]			batch_normaliz
ation_20[0][0]			
maximum_8 (Maximum)	(None, 53, 53, 160)	0	batch_normaliz
ation_17[0][0]			batch_normaliz
ation_18[0][0]			
conv2d_26 (Conv2D)	(None, 53, 53, 64)	16448	maximum_9[0]
[0]			
concatenate_4 (Concatenate)	(None, 53, 53, 224)	0	maximum_8[0]
[0]			conv2d_26[0]
[0]			
conv2d_27 (Conv2D)	(None, 33, 33, 5)	493925	concatenate_4
[0][0]			
batch_normalization_21 (BatchNo	(None, 33, 33, 5)	20	conv2d_27[0]
[0]			
input_4 (InputLayer)	(None, 33, 33, 4)	0	
concatenate_5 (Concatenate)	(None, 33, 33, 9)	0	batch_normaliz
ation_21[0][0]			input_4[0][0]
conv2d_28 (Conv2D)	(None, 27, 27, 64)	28288	concatenate_5
[0][0]			
conv2d_29 (Conv2D)	(None, 27, 27, 64)	28288	concatenate_5
[0][0]			
batch_normalization_22 (BatchNo	(None, 27, 27, 64)	256	conv2d_28[0]
[0]			

<u>batch_normalization_23</u> (BatchNo	(None, 27, 27, 64)	256	conv2d_29[0]
[0]			
<u>maximum_10</u> (Maximum)	(None, 27, 27, 64)	0	batch_normaliz
ation_22[0][0]			batch_normaliz
ation_23[0][0]			
<u>conv2d_30</u> (Conv2D)	(None, 24, 24, 64)	65600	maximum_10[0]
[0]			
<u>conv2d_33</u> (Conv2D)	(None, 22, 22, 64)	36928	conv2d_30[0]
[0]			
<u>conv2d_34</u> (Conv2D)	(None, 22, 22, 64)	36928	conv2d_30[0]
[0]			
<u>conv2d_31</u> (Conv2D)	(None, 21, 21, 160)	243520	concatenate_5
[0][0]			
<u>conv2d_32</u> (Conv2D)	(None, 21, 21, 160)	243520	concatenate_5
[0][0]			
<u>batch_normalization_26</u> (BatchNo	(None, 22, 22, 64)	256	conv2d_33[0]
[0]			
<u>batch_normalization_27</u> (BatchNo	(None, 22, 22, 64)	256	conv2d_34[0]
[0]			
<u>batch_normalization_24</u> (BatchNo	(None, 21, 21, 160)	640	conv2d_31[0]
[0]			
<u>batch_normalization_25</u> (BatchNo	(None, 21, 21, 160)	640	conv2d_32[0]
[0]			
<u>maximum_12</u> (Maximum)	(None, 22, 22, 64)	0	batch_normaliz
ation_26[0][0]			batch_normaliz
ation_27[0][0]			
<u>maximum_11</u> (Maximum)	(None, 21, 21, 160)	0	batch_normaliz
ation_24[0][0]			batch_normaliz



ation\_25[0][0]

conv2d_35 (Conv2D)	(None, 21, 21, 64)	16448	maximum_12[0]
concatenate_6 (Concatenate)	(None, 21, 21, 224)	0	maximum_11[0]
			conv2d_35[0]
conv2d_36 (Conv2D)	(None, 1, 1, 5)	493925	concatenate_6
			[0]
batch_normalization_28 (Batch Normalization)	(None, 1, 1, 5)	20	conv2d_36[0]
activation_2 (Activation)	(None, 1, 1, 5)	0	batch_normalization_28[0][0]
=====			
Total params: 2,089,778			
Trainable params: 2,087,454			
Non-trainable params: 2,324			



Training the architectures

```
In [10]: from sklearn.utils import class_weight
```

```
In [0]: m1 = keras.models.load_model('trial_0001_input_cascade_acc.h5')
```

Training for the InputCascadeCNN model

```

In [0]: fold = os.listdir('HG/')
        fold.sort(key=str.lower)

        for path in fold:
            print(path)
            path = 'HG/'+path
            p = os.listdir(path)
            p.sort(key=str.lower)
            arr = []

            # Reading from 4 images and creating 4 channel slice-wise
            for i in range(len(p)):
                if(i != 4):
                    p1 = os.listdir(path+'/'+p[i])
                    p1.sort()
                    img = sitk.ReadImage(path+'/'+p[i]+'/' +p1[-1])
                    arr.append(sitk.GetArrayFromImage(img))
                else:
                    p1 = os.listdir(path+'/'+p[i])
                    img = sitk.ReadImage(path+'/'+p[i]+'/' +p1[0])
                    Y_labels = sitk.GetArrayFromImage(img)
                    data = np.zeros((Y_labels.shape[1],Y_labels.shape[0],Y_labels.shape[2],4))
                    for i in range(Y_labels.shape[1]):
                        data[i,:,:0] = arr[0][:,i,:]
                        data[i,:,:1] = arr[1][:,i,:]
                        data[i,:,:2] = arr[2][:,i,:]
                        data[i,:,:3] = arr[3][:,i,:]
                    print(data.shape)
                    info = []

            # Creating patches for each slice and training(slice-wise)
            for i in range(data.shape[0]):
                d = data_gen(data,Y_labels,i,1)
                if(len(d) != 0):
                    y = np.zeros((d[2].shape[0],1,1,5))
                    for j in range(y.shape[0]):
                        y[j,:,:d[2][j]] = 1
                    X1 = d[0]
                    X2 = d[1]
                    class_weights = class_weight.compute_class_weight('balanced',
                                                                        np.unique(d[2]),
                                                                        d[2])

                    print('slice no: '+str(i))
                    info.append(m1.fit([X1,X2],y,epochs=5,batch_size=128,class_weight= class_
                    m1.save('trial_0001_input_cascade_acc.h5')

```

### Testing

```

In [0]: model = keras.models.load_model('trial_0001_input_cascade_acc.h5')

```

```

In [16]: path = 'HG/0027'
p = os.listdir(path)
p.sort(key=str.lower)
arr = []
for i in range(len(p)):
    if(i != 4):
        p1 = os.listdir(path+'/'+p[i])
        p1.sort()
        img = sitk.ReadImage(path+'/'+p[i]+'/' + p1[-1])
        arr.append(sitk.GetArrayFromImage(img))
    else:
        p1 = os.listdir(path+'/'+p[i])
        img = sitk.ReadImage(path+'/'+p[i]+'/' + p1[0])
        Y_labels = sitk.GetArrayFromImage(img)
data = np.zeros((Y_labels.shape[1],Y_labels.shape[0],Y_labels.shape[2],4))
for i in range(Y_labels.shape[1]):
    data[i,:,:,:0] = arr[0][:,i,:]
    data[i,:,:,:1] = arr[1][:,i,:]
    data[i,:,:,:2] = arr[2][:,i,:]
    data[i,:,:,:3] = arr[3][:,i,:]
info = []

d = data_gen(data,Y_labels,113,3)
if(len(d) != 0):
    y = np.zeros((d[2].shape[0],1,1,5))
    for j in range(y.shape[0]):
        y[j,:,:,:d[2][j]] = 1
    X1 = d[0]
    X2 = d[1]
    pred = model1.predict([X1,X2],batch_size = 256)
    pred = np.around(pred)
    #print(pred.shape)
    pred1 = np.argmax(pred.reshape(y.shape[0],5)[:,:1:4],axis = 1)
    y2 = np.argmax(y.reshape(y.shape[0],5)[:,:1:4],axis = 1)
    f1 = metrics.f1_score(y2,pred1,average='micro')
    print(f1)

```

0.792288049029622

```

In [0]: pred = model.predict([X1,X2],batch_size = 256)

```

```

In [55]: pred = np.around(pred)
print(pred.shape)
pred1 = np.argmax(pred.reshape(y.shape[0],5)[:,:1:4],axis = 1)
y2 = np.argmax(y.reshape(y.shape[0],5)[:,:1:4],axis = 1)

```

(16268, 1, 1, 5)

```

In [0]: from sklearn import metrics

```

```
In [56]: f1 = metrics.f1_score(y2,pred1,average='micro')  
f1
```

Out[56]: 0.927710843373494

```
In [0]: model1 = keras.models.load_model('trial_MFCcascade_acc.h5')
```

```
In [0]: pred2 = model1.predict([X1,X2],batch_size = 256)
```

```
In [0]: pred2 = np.around(pred2)  
pred3 = np.argmax(pred2.reshape(y.shape[0],5)[:,:1:4],axis = 1)  
y2 = np.argmax(y.reshape(y.shape[0],5)[:,:1:4],axis = 1)
```

```
In [62]: f1 = metrics.f1_score(y2,pred3,average='micro')  
f1
```

Out[62]: 0.7986210418794688