Assignment1_EE782

November 4, 2020

1 HyperParameter Tuning: Experiments and Reports

1.1 Advanced Machine Learning

1.1.1 T Sanjev Vishnu

```
[257]: import tensorflow as tf
       import numpy as np
       import pandas as pd
       import numpy as np
       import os
       import pydicom
       import matplotlib.pyplot as plt
       import cv2 as cv
       from PIL import Image, ImageDraw
       import random
       from tensorflow import keras
       from sklearn.model_selection import train_test_split
       from tensorflow.keras.layers import Activation, Conv2D, MaxPooling2D, Dropout,
       \rightarrowReshape, Conv2DTranspose, concatenate, Input, BatchNormalization
       from tensorflow.keras import models, layers
       from keras.optimizers import Adam
       from tensorflow.keras.losses import binary_crossentropy
```

```
[85]: def pad_up_to(t, max_in_dims, constant_values):
    s = tf.shape(t)
    paddings = [[0, m-s[i]] for (i,m) in enumerate(max_in_dims)]
    return tf.pad(t, paddings, 'CONSTANT', constant_values=constant_values)
```

```
if len(i) > 1:
                   y.append(i[-28:])
  [5]: images = [line[:8] for line in y]
  [7]: img_ext = ".dcm"
       i = 0
       x_train = []
       for i in range(len(images)):
           if i\%2 == 0:
               ds_img = pydicom.dcmread("TrainingSet/"+images[i]+img_ext)
               numpy_array = ds_img.pixel_array
               #cv.imwrite(images[i]+".jpg",numpy_array)
               x_train.append(numpy_array)
               #plt.figure(frameon=False)
               #plt.title("Patient ID "+images[i])
               #plt.axis("off")
               plt.imsave(images[i]+".png",ds_img.pixel_array)
               #plt.imshow(ds_imq.pixel_array)
[107]: y_train = []
       for h in range(243):
           cont_f = open("TrainingSet/"+y[(h*2)])
           cont_f = cont_f.read()
           coordinates = cont_f.split("\n")
           c = \prod
           for i in range(len(coordinates) - 1):
               cc = coordinates[i].split()
               x1, y1 = float(cc[0]), float(cc[1])
               ccc = (x1, y1)
               c.append(ccc)
           ocont_f = open("TrainingSet/"+y[(h*2)+1])
           ocont_f = ocont_f.read()
           ocoordinates = ocont_f.split("\n")
           o = []
           for i in range(len(ocoordinates) - 1):
               oo = ocoordinates[i].split()
               x1, y1 = float(oo[0]), float(oo[1])
               000 = (x1, y1)
               o.append(ooo)
           #print(images[2*h])
           w, height = x_train[h].shape
           oim = Image.new("L",size=(height, w))
           odraw = ImageDraw.Draw(oim)
           iim = Image.new("L",size=(height, w))
           idraw = ImageDraw.Draw(iim)
           idraw.polygon(c, fill="white")
```

```
odraw.polygon(o, fill="white")
oim.save("OuterContour/"+images[h*2]+"o"+".png","PNG")
iim.save("InnerContour/"+images[h*2]+"i"+".png","PNG")
yy = cv.imread("InnerContour/"+images[h*2]+"i"+".png",0)
#print(yy.shape)
y_train.append(yy)
if h%10 == 2:
    print("*", end="")
```

```
[254]: X, Y = [], []
for i in range(len(x_train)):
    #if x_train[i].shape == (256,216):
    wid, hei = x_train[i].shape
    temp = tf.convert_to_tensor((x_train[i].reshape(1,wid,hei,1))/255.000,
    dtype=tf.float32)
    X.append(pad_up_to(temp, [1,256,256,1], 0))
    wid, hei = y_train[i].shape
    temp = tf.convert_to_tensor((y_train[i].reshape(1,wid,hei,1))/255.000,
    dtype=tf.float32)
    Y.append(pad_up_to(temp, [1,256,256,1], 0))
    #print(x_train[i].shape, y_train[i].shape)
```

```
[304]: \#print(type(Y))
       def split dataset(split):
           if split <= 1 and split > 0:
               split point = int(len(X) - split*len(X))
               train_split_X = X[:split_point]
               train_split_Y = Y[:split_point]
               test_split_X = X[split_point:]
               test_split_Y = Y[split_point:]
               tr_lsdx = tf.data.Dataset.from_tensor_slices(train_split_X)
               tr_lsdy = tf.data.Dataset.from_tensor_slices(train_split_Y)
               tr_lsdd = tf.data.Dataset.zip((tr_lsdx, tr_lsdy))
               ts_lsdx = tf.data.Dataset.from_tensor_slices(test_split_X)
               ts_lsdy = tf.data.Dataset.from_tensor_slices(test_split_Y)
               ts_lsdd = tf.data.Dataset.zip((ts_lsdx, ts_lsdy))
               return tr_lsdd, ts_lsdd
           else :
               return("Split cannot be negative")
```

```
[149]: def display(display_list):
    plt.figure(figsize=(15, 15))

    title = ['Input Image', 'True Mask', 'Predicted Mask']
```

```
for i in range(len(display_list)):
           plt.subplot(1, len(display_list), i+1)
           plt.title(title[i])
           plt.imshow(tf.keras.preprocessing.image.array_to_img(display_list[i]))
           plt.axis('off')
         plt.show()
[474]: def conv2d_block(input_tensor,11, 12, n_filters, kernel_size = 3, batchnorm = ____
           """Function to add 2 convolutional layers with the parameters passed to \Box
        \hookrightarrow i.t."""
           # first layer
           x = Conv2D(filters = n_filters, kernel_size = (kernel_size, kernel_size),\
                     kernel_initializer = 'he_normal', padding = 'same',
        wkernel_regularizer=tf.keras.regularizers.11_12(11, 12))(input_tensor)
           if batchnorm:
               x = BatchNormalization()(x)
           x = Activation('relu')(x)
           # second layer
           x = Conv2D(filters = n_filters, kernel_size = (kernel_size, kernel_size),\
                     kernel_initializer = 'he_normal', padding = 'same',
        →kernel_regularizer=tf.keras.regularizers.11_12(11, 12))(input_tensor)
           if batchnorm:
               x = BatchNormalization()(x)
           x = Activation('relu')(x)
```

return x

```
c4 = conv2d_block(p3,11, 12, n_filters * 8, kernel_size = 3, batchnorm = __
        →batchnorm)
           p4 = MaxPooling2D((2, 2))(c4)
           p4 = Dropout(dropout)(p4)
           c5 = conv2d_block(p4,11, 12, n_filters = n_filters * 16, kernel_size = 3,_
        →batchnorm = batchnorm)
           # Expansive Path
           u6 = Conv2DTranspose(n_filters * 8, (3, 3), strides = (2, 2), padding = __
        → 'same', kernel regularizer=tf.keras.regularizers.l1 12(11, 12))(c5)
           u6 = concatenate([u6, c4])
           u6 = Dropout(dropout)(u6)
           c6 = conv2d_block(u6,11, 12, n_filters * 8, kernel_size = 3, batchnorm = u
        →batchnorm)
           u7 = Conv2DTranspose(n_filters * 4, (3, 3), strides = (2, 2), padding = __
        → 'same', kernel_regularizer=tf.keras.regularizers.11_12(11, 12))(c6)
           u7 = concatenate([u7, c3])
           u7 = Dropout(dropout)(u7)
           c7 = conv2d_block(u7,11, 12, n_filters * 4, kernel_size = 3, batchnorm = u
        →batchnorm)
           u8 = Conv2DTranspose(n_filters * 2, (3, 3), strides = (2, 2), padding = (2, 2)
        → 'same', kernel_regularizer=tf.keras.regularizers.l1_l2(l1, l2))(c7)
           u8 = concatenate([u8, c2])
           u8 = Dropout(dropout)(u8)
           c8 = conv2d_block(u8,11, 12, n_filters * 2, kernel_size = 3, batchnorm =__
        ⇒batchnorm)
           u9 = Conv2DTranspose(n_filters * 1, (3, 3), strides = (2, 2), padding = __
        → 'same', kernel_regularizer=tf.keras.regularizers.11_12(11, 12))(c8)
           u9 = concatenate([u9, c1])
           u9 = Dropout(dropout)(u9)
           c9 = conv2d_block(u9,11, 12, n_filters * 1, kernel_size = 3, batchnorm = u
        →batchnorm)
           outputs = Conv2D(1, (1, 1), activation='sigmoid')(c9)
           model1 = models.Model(inputs=[input_img], outputs=[outputs])
           return model1
[466]: def dice_coefficient(y_true, y_pred):
           numerator = 2 * tf.reduce_sum(y_true * y_pred)
           denominator = tf.reduce_sum(y_true + y_pred)
           return numerator / (denominator + tf.keras.backend.epsilon())
```

```
[487]: def loss(y_true, y_pred):
        return (binary_crossentropy(y_true, y_pred))
[476]: model1.summary()
    Model: "model 6"
    ______
                            Output Shape Param # Connected to
    Layer (type)
    ______
                            [(None, 256, 256, 1) 0
    img (InputLayer)
                           (None, 256, 256, 16) 160
    conv2d_241 (Conv2D)
                                                   img[0][0]
    batch_normalization_138 (BatchN (None, 256, 256, 16) 64
    conv2d_241[0][0]
    activation_137 (Activation) (None, 256, 256, 16) 0
    batch_normalization_138[0][0]
    ______
    max_pooling2d_52 (MaxPooling2D) (None, 128, 128, 16) 0
    activation_137[0][0]
                           (None, 128, 128, 16) 0
    dropout_100 (Dropout)
    max_pooling2d_52[0][0]
    conv2d_243 (Conv2D)
                           (None, 128, 128, 32) 4640
    dropout_100[0][0]
    batch_normalization_140 (BatchN (None, 128, 128, 32) 128
    conv2d_243[0][0]
    ______
    activation_139 (Activation) (None, 128, 128, 32) 0
    batch_normalization_140[0][0]
    max_pooling2d_53 (MaxPooling2D) (None, 64, 64, 32) 0
    activation_139[0][0]
```

dropout_101 (Dropout) max_pooling2d_53[0][0]	(None,	64,	64,	32)	0
 conv2d_245 (Conv2D) dropout_101[0][0]	(None,	64,	64,	64)	18496
batch_normalization_142 (BatchN conv2d_245[0][0]	(None,	64,	64,	64)	256
activation_141 (Activation) batch_normalization_142[0][0]	(None,	64,	64,	64)	0
max_pooling2d_54 (MaxPooling2D) activation_141[0][0]				64)	0
dropout_102 (Dropout) max_pooling2d_54[0][0]	(None,			64)	0
conv2d_247 (Conv2D) dropout_102[0][0]	(None,	32,	32,	128)	73856
batch_normalization_144 (BatchN conv2d_247[0][0]	(None,	32,	32,	128)	512
activation_143 (Activation) batch_normalization_144[0][0]	(None,				
max_pooling2d_55 (MaxPooling2D) activation_143[0][0]	(None,	16,	16,	128)	0
dropout_103 (Dropout) max_pooling2d_55[0][0]	(None,	16,	16,	128)	
conv2d_249 (Conv2D) dropout_103[0][0]					295168

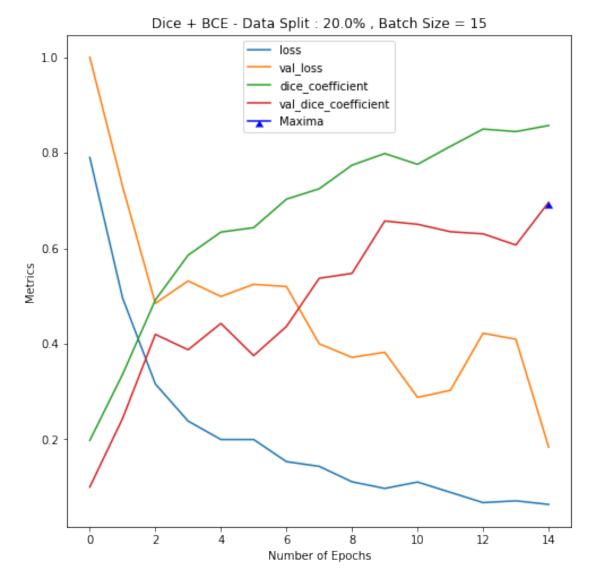
batch_normalization_146 (BatchN conv2d_249[0][0]	(None,	16,	16,	256)	1024
activation_145 (Activation) batch_normalization_146[0][0]	(None,	16,	16,	256)	0
conv2d_transpose_49 (Conv2DTran activation_145[0][0]	(None,	32,	32,	128)	295040
concatenate_29 (Concatenate) conv2d_transpose_49[0][0] activation_143[0][0]	(None,	32,	32,	256)	0
dropout_104 (Dropout) concatenate_29[0][0]	(None,	32,	32,	256)	0
conv2d_251 (Conv2D) dropout_104[0][0]	(None,	32,	32,	128)	295040
batch_normalization_148 (BatchN conv2d_251[0][0]	(None,	32,	32,	128)	512
activation_147 (Activation) batch_normalization_148[0][0]	(None,	32,	32,	128)	0
conv2d_transpose_50 (Conv2DTran activation_147[0][0]					73792
concatenate_30 (Concatenate) conv2d_transpose_50[0][0] activation_141[0][0]	(None,	64,	64,	128)	
dropout_105 (Dropout) concatenate_30[0][0]	(None,	64,	64,	128)	0
					

conv2d_253 (Conv2D) dropout_105[0][0]	(None,	64, 64, 64)	73792
batch_normalization_150 (BatchN conv2d_253[0][0]	(None,	64, 64, 64)	256
activation_149 (Activation) batch_normalization_150[0][0]	(None,	64, 64, 64)	0
conv2d_transpose_51 (Conv2DTran activation_149[0][0]	(None,	128, 128, 32) 18464
concatenate_31 (Concatenate) conv2d_transpose_51[0][0] activation_139[0][0]	(None,	128, 128, 64) 0
dropout_106 (Dropout) concatenate_31[0][0]	(None,	128, 128, 64) 0
conv2d_255 (Conv2D) dropout_106[0][0]	(None,	128, 128, 32) 18464
batch_normalization_152 (BatchN conv2d_255[0][0]	(None,	128, 128, 32) 128
activation_151 (Activation) batch_normalization_152[0][0]		128, 128, 32	
conv2d_transpose_52 (Conv2DTran activation_151[0][0]	(None,	256, 256, 16	
concatenate_32 (Concatenate) conv2d_transpose_52[0][0] activation_137[0][0]		256, 256, 32	
dropout_107 (Dropout) concatenate_32[0][0]	(None,	256, 256, 32) 0

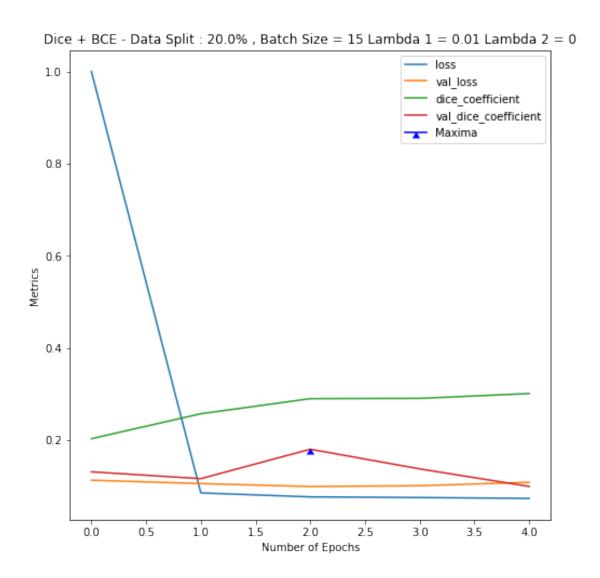
```
(None, 256, 256, 16) 4624
     conv2d_257 (Conv2D)
     dropout_107[0][0]
     batch normalization 154 (BatchN (None, 256, 256, 16) 64
     conv2d_257[0][0]
     ______
     activation_153 (Activation) (None, 256, 256, 16) 0
     batch_normalization_154[0][0]
     conv2d_258 (Conv2D)
                                 (None, 256, 256, 1) 17
     activation_153[0][0]
     ______
     Total params: 1,179,121
     Trainable params: 1,177,649
     Non-trainable params: 1,472
     ______
     _____
[477]: def dice_metric_plot(m,titl):
         plt.figure(figsize = (8,8))
         plt.title(titl)
         loss_normal = max(np.max(m.history["loss"]), np.max(m.history["val_loss"]))
         plt.plot(m.history["loss"]/loss_normal, label="loss")
         plt.plot(m.history["val_loss"]/loss_normal, label="val_loss")
         plt.plot(m.history["dice_coefficient"], label="dice_coefficient")
         plt.plot(m.history["val_dice_coefficient"], label="val_dice_coefficient")
         plt.plot( np.argmax(m.history["val_dice_coefficient"]), np.max(m.
      →history["val_dice_coefficient"]), marker=6, color="b", label="Maxima")
         plt.xlabel("Number of Epochs")
         plt.ylabel("Metrics")
         plt.legend();
         plt.savefig(str(m)+".png")
 []: for image, mask in lsdd.take(100):
         pred_mask = dicelossmodel.predict(image)
         #print(tf.shape(pred_mask))
         sample_image, sample_mask, sample_pred_mask = tf.reshape(image, [ 256, 256, u
      \rightarrow1]), tf.reshape(mask, [ 256, 256, 1]), tf.reshape(pred_mask, [ 256, 256, 1])
         display([sample_image, sample_mask, sample_pred_mask])
```

```
[488]: def experiment1(split_percent, batch_size, epochs, lamb1, lamb2):
       im_width = 256
       im_height = 256
       input_img = Input((im_height, im_width, 1), name='img')
       dicelossmodel = get_unet(input_img,lamb1, lamb2, n_filters=16, dropout=0.
     →05, batchnorm=True)
       dicelossmodel.compile(optimizer=Adam(), loss = loss ,__
     →metrics=[dice_coefficient])
       tra, tes = split_dataset(split_percent)
       diceresults = dicelossmodel.fit(tra, batch_size = batch_size, epochs = ___
     →epochs, validation_data = tes)
       dice_metric_plot(diceresults, "BCE - Data Split : __
     →"+str(split_percent*100)+"%, Batch Size = "+ str(batch_size) + " Lambda 1 = "
     \rightarrow"+str(lamb1)+" Lambda 2 = "+str(lamb2))
[461]: experiment1(0.2, 15, 15)
    Epoch 1/15
    dice_coefficient: 0.1974 - val_loss: 3.0428 - val_dice_coefficient: 0.0999
    Epoch 2/15
    dice_coefficient: 0.3358 - val_loss: 2.2203 - val_dice_coefficient: 0.2434
    Epoch 3/15
    dice_coefficient: 0.4916 - val_loss: 1.4743 - val_dice_coefficient: 0.4197
    Epoch 4/15
    dice_coefficient: 0.5855 - val_loss: 1.6175 - val_dice_coefficient: 0.3875
    dice_coefficient: 0.6339 - val_loss: 1.5183 - val_dice_coefficient: 0.4426
    Epoch 6/15
    dice_coefficient: 0.6432 - val_loss: 1.5960 - val_dice_coefficient: 0.3751
    Epoch 7/15
    dice_coefficient: 0.7028 - val_loss: 1.5816 - val_dice_coefficient: 0.4362
    Epoch 8/15
    dice_coefficient: 0.7246 - val_loss: 1.2168 - val_dice_coefficient: 0.5371
    Epoch 9/15
    dice_coefficient: 0.7741 - val_loss: 1.1303 - val_dice_coefficient: 0.5475
    Epoch 10/15
```

dice_coefficient: 0.7984 - val_loss: 1.1629 - val_dice_coefficient: 0.6571



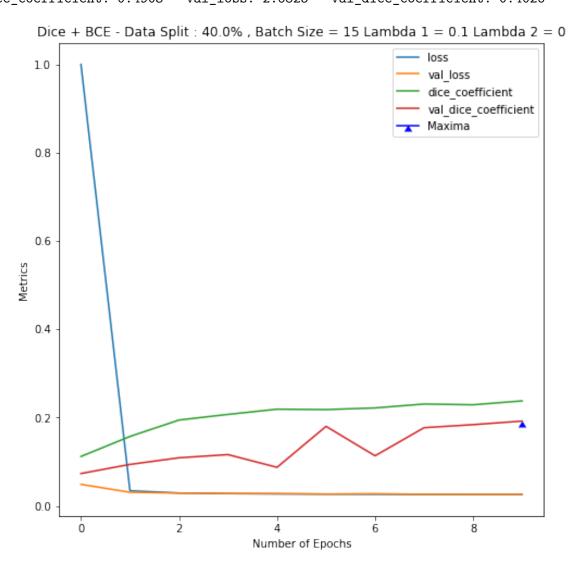
[479]: experiment1(0.2,15,5, 0.01, 0)

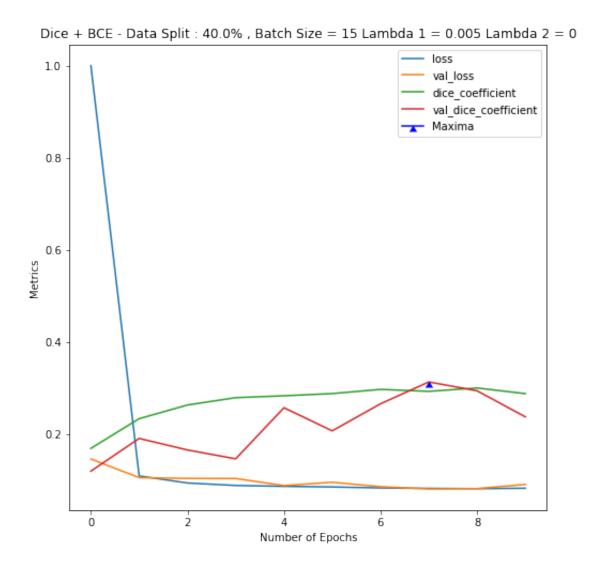


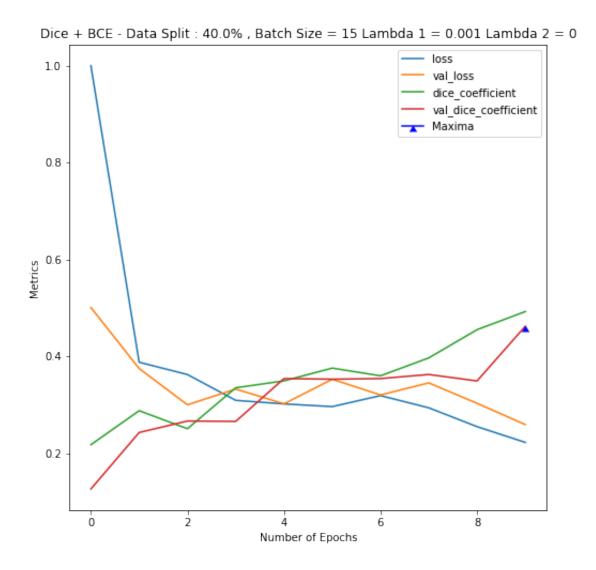
```
[481]: experiment1(0.4, 15, 10, 0.1, 0) experiment1(0.4, 15, 10, 0.005, 0) experiment1(0.4, 15, 10, 0.001, 0) experiment1(0.4, 15, 10, 0.0005, 0)
```

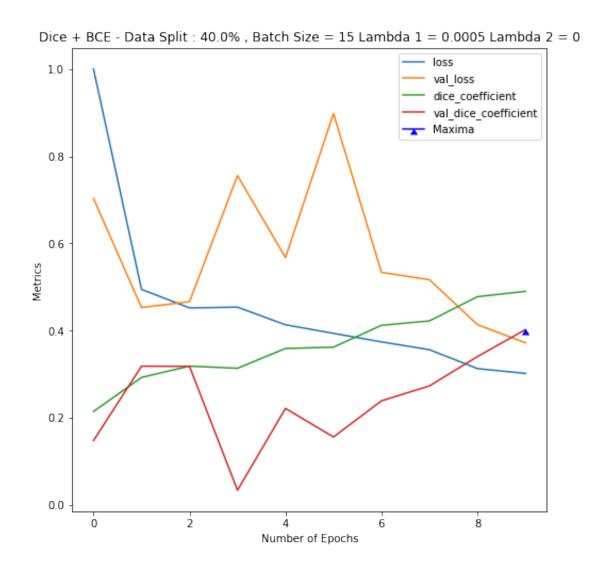
```
dice_coefficient: 0.2072 - val_loss: 18.5358 - val_dice_coefficient: 0.1161
Epoch 5/10
145/145 [============= ] - 46s 314ms/step - loss: 17.9862 -
dice_coefficient: 0.2190 - val_loss: 18.5002 - val_dice_coefficient: 0.0873
Epoch 6/10
145/145 [============= ] - 44s 304ms/step - loss: 17.6625 -
dice_coefficient: 0.2181 - val_loss: 17.8203 - val_dice_coefficient: 0.1801
Epoch 7/10
dice_coefficient: 0.2219 - val_loss: 18.2768 - val_dice_coefficient: 0.1135
Epoch 8/10
145/145 [============= ] - 45s 308ms/step - loss: 17.2816 -
dice_coefficient: 0.2307 - val_loss: 17.4000 - val_dice_coefficient: 0.1771
dice_coefficient: 0.2290 - val_loss: 17.1992 - val_dice_coefficient: 0.1840
Epoch 10/10
dice_coefficient: 0.2378 - val_loss: 17.2359 - val_dice_coefficient: 0.1921
dice_coefficient: 0.1697 - val_loss: 5.3695 - val_dice_coefficient: 0.1203
Epoch 2/10
dice_coefficient: 0.2342 - val_loss: 3.8886 - val_dice_coefficient: 0.1913
Epoch 3/10
dice_coefficient: 0.2638 - val_loss: 3.8281 - val_dice_coefficient: 0.1663
Epoch 4/10
dice_coefficient: 0.2797 - val_loss: 3.8201 - val_dice_coefficient: 0.1471
Epoch 5/10
dice coefficient: 0.2838 - val loss: 3.2595 - val dice coefficient: 0.2579
Epoch 6/10
dice_coefficient: 0.2886 - val_loss: 3.5231 - val_dice_coefficient: 0.2077
Epoch 7/10
dice_coefficient: 0.2977 - val_loss: 3.1718 - val_dice_coefficient: 0.2665
Epoch 8/10
dice_coefficient: 0.2935 - val_loss: 2.9914 - val_dice_coefficient: 0.3138
Epoch 9/10
dice_coefficient: 0.3007 - val_loss: 3.0210 - val_dice_coefficient: 0.2947
Epoch 10/10
```

```
dice_coefficient: 0.2885 - val_loss: 3.3508 - val_dice_coefficient: 0.2381
Epoch 1/10
145/145 [============= ] - 47s 327ms/step - loss: 10.7543 -
dice_coefficient: 0.2173 - val_loss: 5.3841 - val_dice_coefficient: 0.1260
Epoch 2/10
dice_coefficient: 0.2875 - val_loss: 4.0290 - val_dice_coefficient: 0.2426
Epoch 3/10
dice_coefficient: 0.2503 - val_loss: 3.2233 - val_dice_coefficient: 0.2662
Epoch 4/10
dice_coefficient: 0.3349 - val_loss: 3.5740 - val_dice_coefficient: 0.2655
dice_coefficient: 0.3492 - val_loss: 3.2451 - val_dice_coefficient: 0.3539
dice_coefficient: 0.3755 - val_loss: 3.7905 - val_dice_coefficient: 0.3526
dice_coefficient: 0.3596 - val_loss: 3.4421 - val_dice_coefficient: 0.3539
Epoch 8/10
dice_coefficient: 0.3966 - val_loss: 3.7102 - val_dice_coefficient: 0.3623
Epoch 9/10
dice_coefficient: 0.4548 - val_loss: 3.2558 - val_dice_coefficient: 0.3490
Epoch 10/10
dice_coefficient: 0.4923 - val_loss: 2.7838 - val_dice_coefficient: 0.4620
Epoch 1/10
dice coefficient: 0.2148 - val loss: 5.0665 - val dice coefficient: 0.1481
Epoch 2/10
dice_coefficient: 0.2930 - val_loss: 3.2653 - val_dice_coefficient: 0.3187
Epoch 3/10
dice_coefficient: 0.3189 - val_loss: 3.3608 - val_dice_coefficient: 0.3184
Epoch 4/10
dice_coefficient: 0.3139 - val_loss: 5.4440 - val_dice_coefficient: 0.0343
Epoch 5/10
dice_coefficient: 0.3594 - val_loss: 4.0886 - val_dice_coefficient: 0.2219
Epoch 6/10
```



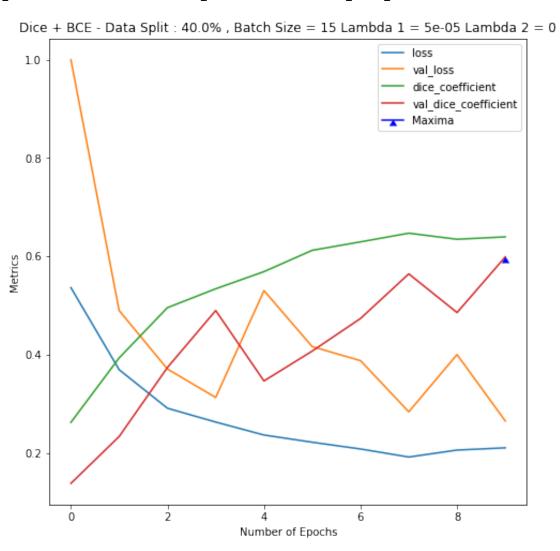






[482]: experiment1(0.4, 15, 10, 0.00005, 0)

```
dice_coefficient: 0.5690 - val_loss: 3.0626 - val_dice_coefficient: 0.3466
Epoch 6/10
dice_coefficient: 0.6123 - val_loss: 2.4044 - val_dice_coefficient: 0.4074
Epoch 7/10
dice_coefficient: 0.6295 - val_loss: 2.2396 - val_dice_coefficient: 0.4736
Epoch 8/10
dice_coefficient: 0.6472 - val_loss: 1.6387 - val_dice_coefficient: 0.5644
Epoch 9/10
dice_coefficient: 0.6348 - val_loss: 2.3111 - val_dice_coefficient: 0.4857
Epoch 10/10
dice_coefficient: 0.6397 - val_loss: 1.5308 - val_dice_coefficient: 0.5990
```

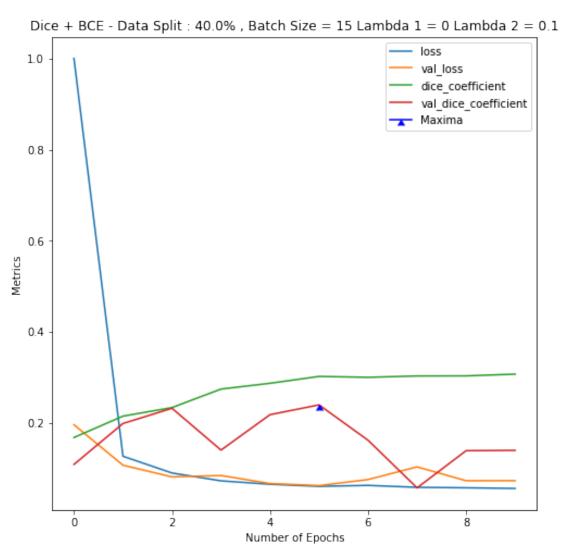


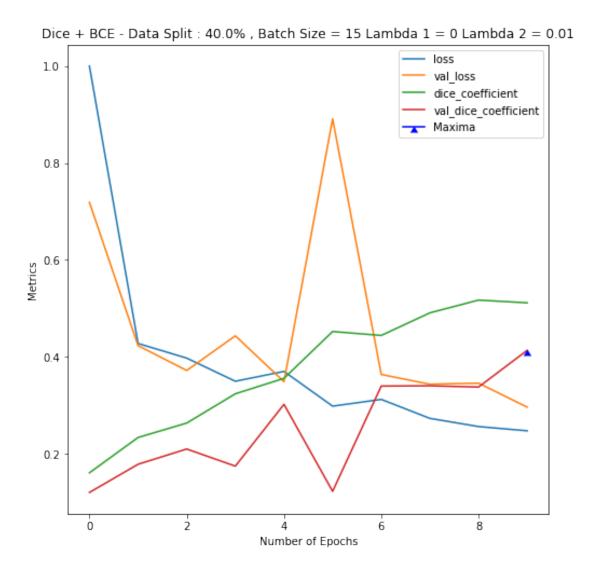
```
[483]: | # Experimenting with values of Lambda 2 hyperparameter of L2 Penalty
    experiment1(0.4, 15, 10, 0, 0.1)
    experiment1(0.4, 15, 10, 0, 0.01)
    experiment1(0.4, 15, 10, 0, 0.001)
    experiment1(0.4, 15, 10, 0, 0.0005)
    experiment1(0.4, 15, 10, 0, 0.00005)
   Epoch 1/10
   dice_coefficient: 0.1686 - val_loss: 7.3513 - val_dice_coefficient: 0.1096
   Epoch 2/10
   dice_coefficient: 0.2155 - val_loss: 4.0326 - val_dice_coefficient: 0.1994
   dice_coefficient: 0.2342 - val_loss: 3.0740 - val_dice_coefficient: 0.2328
   Epoch 4/10
   145/145 [============ ] - 48s 333ms/step - loss: 2.7500 -
   dice_coefficient: 0.2749 - val_loss: 3.1865 - val_dice_coefficient: 0.1410
   Epoch 5/10
   dice_coefficient: 0.2875 - val_loss: 2.5310 - val_dice_coefficient: 0.2187
   Epoch 6/10
   dice_coefficient: 0.3026 - val_loss: 2.3597 - val_dice_coefficient: 0.2402
   Epoch 7/10
   dice_coefficient: 0.3006 - val_loss: 2.8559 - val_dice_coefficient: 0.1626
   Epoch 8/10
   dice_coefficient: 0.3036 - val_loss: 3.8952 - val_dice_coefficient: 0.0582
   Epoch 9/10
   dice_coefficient: 0.3038 - val_loss: 2.7594 - val_dice_coefficient: 0.1397
   Epoch 10/10
   dice_coefficient: 0.3076 - val_loss: 2.7632 - val_dice_coefficient: 0.1404
   Epoch 1/10
   dice_coefficient: 0.1610 - val_loss: 4.9436 - val_dice_coefficient: 0.1205
   Epoch 2/10
   dice_coefficient: 0.2339 - val_loss: 2.9112 - val_dice_coefficient: 0.1787
   Epoch 3/10
```

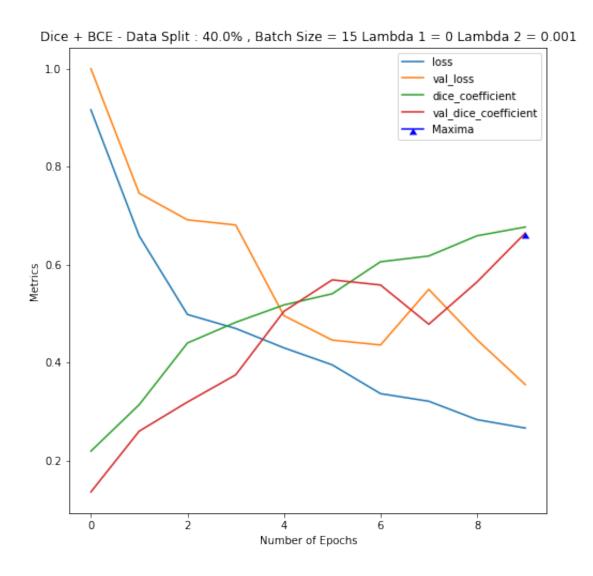
```
dice_coefficient: 0.2636 - val_loss: 2.5588 - val_dice_coefficient: 0.2102
Epoch 4/10
dice_coefficient: 0.3240 - val_loss: 3.0502 - val_dice_coefficient: 0.1748
Epoch 5/10
dice_coefficient: 0.3559 - val_loss: 2.3996 - val_dice_coefficient: 0.3023
Epoch 6/10
dice_coefficient: 0.4525 - val_loss: 6.1272 - val_dice_coefficient: 0.1228
Epoch 7/10
dice_coefficient: 0.4444 - val_loss: 2.5020 - val_dice_coefficient: 0.3400
Epoch 8/10
dice_coefficient: 0.4910 - val_loss: 2.3638 - val_dice_coefficient: 0.3405
Epoch 9/10
dice_coefficient: 0.5173 - val_loss: 2.3787 - val_dice_coefficient: 0.3379
Epoch 10/10
dice_coefficient: 0.5117 - val_loss: 2.0391 - val_dice_coefficient: 0.4136
Epoch 1/10
dice_coefficient: 0.2189 - val_loss: 3.5211 - val_dice_coefficient: 0.1355
Epoch 2/10
dice_coefficient: 0.3137 - val_loss: 2.6253 - val_dice_coefficient: 0.2596
dice_coefficient: 0.4398 - val_loss: 2.4342 - val_dice_coefficient: 0.3192
Epoch 4/10
dice_coefficient: 0.4820 - val_loss: 2.3969 - val_dice_coefficient: 0.3747
Epoch 5/10
dice_coefficient: 0.5175 - val_loss: 1.7447 - val_dice_coefficient: 0.5045
Epoch 6/10
dice_coefficient: 0.5403 - val_loss: 1.5691 - val_dice_coefficient: 0.5687
Epoch 7/10
dice_coefficient: 0.6056 - val_loss: 1.5340 - val_dice_coefficient: 0.5582
Epoch 8/10
dice_coefficient: 0.6174 - val_loss: 1.9343 - val_dice_coefficient: 0.4781
Epoch 9/10
```

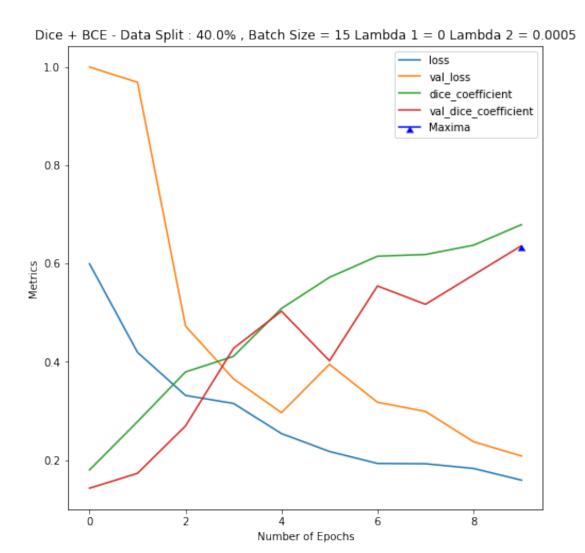
```
dice_coefficient: 0.6587 - val_loss: 1.5714 - val_dice_coefficient: 0.5644
Epoch 10/10
dice_coefficient: 0.6767 - val_loss: 1.2492 - val_dice_coefficient: 0.6644
Epoch 1/10
dice_coefficient: 0.1798 - val_loss: 5.2701 - val_dice_coefficient: 0.1425
Epoch 2/10
dice_coefficient: 0.2780 - val_loss: 5.1061 - val_dice_coefficient: 0.1729
Epoch 3/10
dice_coefficient: 0.3792 - val_loss: 2.4911 - val_dice_coefficient: 0.2693
Epoch 4/10
dice_coefficient: 0.4109 - val_loss: 1.9242 - val_dice_coefficient: 0.4272
Epoch 5/10
dice_coefficient: 0.5085 - val_loss: 1.5616 - val_dice_coefficient: 0.5029
Epoch 6/10
dice_coefficient: 0.5718 - val_loss: 2.0806 - val_dice_coefficient: 0.4020
Epoch 7/10
dice_coefficient: 0.6148 - val_loss: 1.6736 - val_dice_coefficient: 0.5543
Epoch 8/10
dice_coefficient: 0.6183 - val_loss: 1.5746 - val_dice_coefficient: 0.5169
dice_coefficient: 0.6372 - val_loss: 1.2507 - val_dice_coefficient: 0.5766
Epoch 10/10
dice_coefficient: 0.6791 - val_loss: 1.0966 - val_dice_coefficient: 0.6361
dice_coefficient: 0.1765 - val_loss: 6.2330 - val_dice_coefficient: 0.0597
Epoch 2/10
dice_coefficient: 0.2688 - val_loss: 3.9465 - val_dice_coefficient: 0.0857
Epoch 3/10
dice_coefficient: 0.3636 - val_loss: 1.9575 - val_dice_coefficient: 0.3705
Epoch 4/10
dice_coefficient: 0.4534 - val_loss: 1.4831 - val_dice_coefficient: 0.4578
Epoch 5/10
```

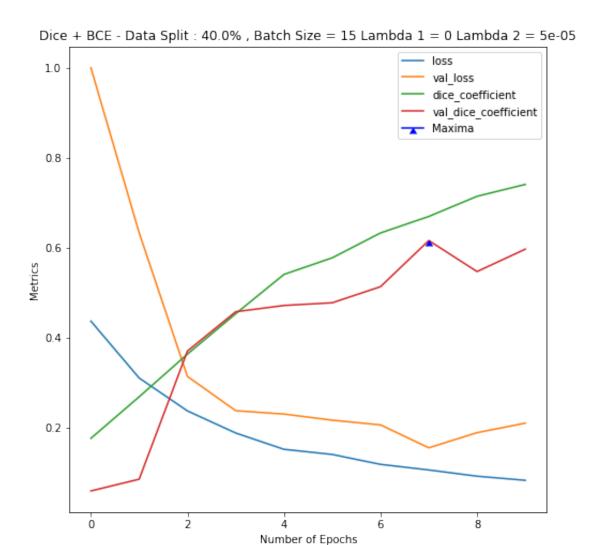
```
dice_coefficient: 0.5407 - val_loss: 1.4365 - val_dice_coefficient: 0.4717
Epoch 6/10
dice_coefficient: 0.5776 - val_loss: 1.3517 - val_dice_coefficient: 0.4777
Epoch 7/10
dice_coefficient: 0.6328 - val_loss: 1.2863 - val_dice_coefficient: 0.5135
Epoch 8/10
dice_coefficient: 0.6695 - val_loss: 0.9694 - val_dice_coefficient: 0.6160
Epoch 9/10
dice_coefficient: 0.7143 - val_loss: 1.1787 - val_dice_coefficient: 0.5471
Epoch 10/10
dice_coefficient: 0.7407 - val_loss: 1.3109 - val_dice_coefficient: 0.5969
```





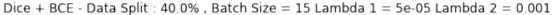


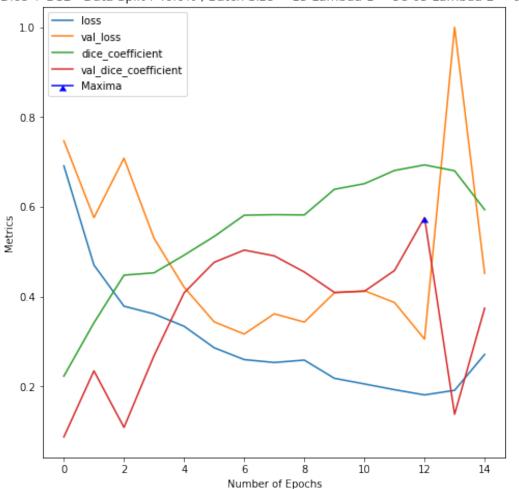




[485]: experiment1(0.4, 15, 15, 0.00005, 0.001)

```
dice_coefficient: 0.4925 - val_loss: 2.3462 - val_dice_coefficient: 0.4088
Epoch 6/15
dice_coefficient: 0.5342 - val_loss: 1.9164 - val_dice_coefficient: 0.4767
Epoch 7/15
dice_coefficient: 0.5815 - val_loss: 1.7646 - val_dice_coefficient: 0.5039
Epoch 8/15
dice_coefficient: 0.5827 - val_loss: 2.0164 - val_dice_coefficient: 0.4910
dice_coefficient: 0.5822 - val_loss: 1.9133 - val_dice_coefficient: 0.4551
dice_coefficient: 0.6391 - val_loss: 2.2761 - val_dice_coefficient: 0.4097
Epoch 11/15
dice_coefficient: 0.6517 - val_loss: 2.3021 - val_dice_coefficient: 0.4122
Epoch 12/15
dice_coefficient: 0.6811 - val_loss: 2.1558 - val_dice_coefficient: 0.4583
Epoch 13/15
dice_coefficient: 0.6935 - val_loss: 1.7021 - val_dice_coefficient: 0.5763
Epoch 14/15
dice_coefficient: 0.6805 - val_loss: 5.5685 - val_dice_coefficient: 0.1384
Epoch 15/15
dice_coefficient: 0.5939 - val_loss: 2.5172 - val_dice_coefficient: 0.3746
```



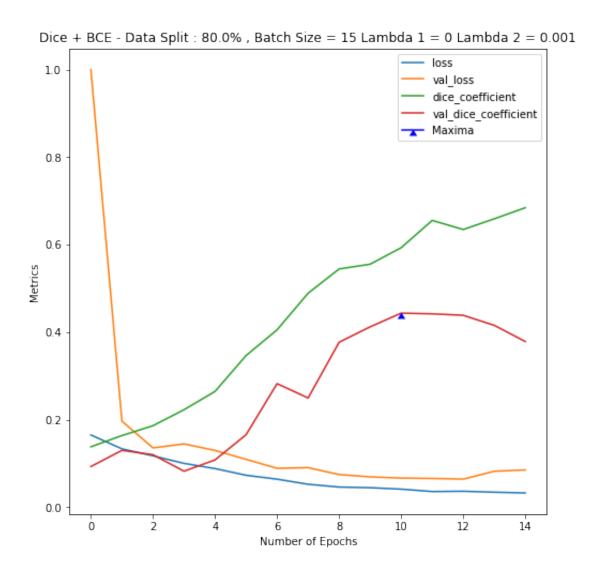


```
[486]: experiment1(0.8, 15, 15, 0, 0.001) experiment1(0.6, 15, 15, 0, 0.001) experiment1(0.2, 15, 15, 0, 0.001)
```

```
Epoch 5/15
dice_coefficient: 0.2647 - val_loss: 3.4465 - val_dice_coefficient: 0.1081
dice_coefficient: 0.3466 - val_loss: 2.9075 - val_dice_coefficient: 0.1656
Epoch 7/15
dice_coefficient: 0.4055 - val_loss: 2.3615 - val_dice_coefficient: 0.2824
Epoch 8/15
dice_coefficient: 0.4891 - val_loss: 2.4018 - val_dice_coefficient: 0.2497
Epoch 9/15
48/48 [============ ] - 23s 473ms/step - loss: 1.2249 -
dice_coefficient: 0.5446 - val_loss: 1.9836 - val_dice_coefficient: 0.3772
Epoch 10/15
48/48 [============= ] - 23s 476ms/step - loss: 1.1855 -
dice_coefficient: 0.5555 - val_loss: 1.8436 - val_dice_coefficient: 0.4123
Epoch 11/15
dice_coefficient: 0.5932 - val_loss: 1.7695 - val_dice_coefficient: 0.4436
Epoch 12/15
dice_coefficient: 0.6555 - val_loss: 1.7466 - val_dice_coefficient: 0.4420
Epoch 13/15
dice_coefficient: 0.6347 - val_loss: 1.7073 - val_dice_coefficient: 0.4387
Epoch 14/15
dice_coefficient: 0.6592 - val_loss: 2.1839 - val_dice_coefficient: 0.4157
Epoch 15/15
dice_coefficient: 0.6845 - val_loss: 2.2626 - val_dice_coefficient: 0.3787
Epoch 1/15
dice_coefficient: 0.1736 - val_loss: 5.3943 - val_dice_coefficient: 0.1102
Epoch 2/15
97/97 [=========== - 39s 397ms/step - loss: 2.5087 -
dice_coefficient: 0.2591 - val_loss: 2.6521 - val_dice_coefficient: 0.2469
Epoch 3/15
dice_coefficient: 0.3602 - val_loss: 2.5421 - val_dice_coefficient: 0.3027
dice_coefficient: 0.4767 - val_loss: 2.7085 - val_dice_coefficient: 0.2802
Epoch 5/15
dice_coefficient: 0.5400 - val_loss: 1.9545 - val_dice_coefficient: 0.4511
```

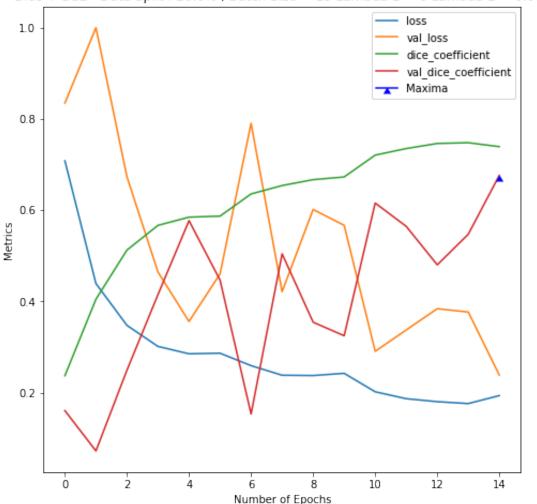
```
Epoch 6/15
dice_coefficient: 0.5834 - val_loss: 1.9717 - val_dice_coefficient: 0.4153
Epoch 7/15
97/97 [============ ] - 33s 345ms/step - loss: 1.3088 -
dice_coefficient: 0.5357 - val_loss: 3.8006 - val_dice_coefficient: 0.2685
Epoch 8/15
dice_coefficient: 0.5970 - val_loss: 2.2720 - val_dice_coefficient: 0.3832
Epoch 9/15
dice_coefficient: 0.6437 - val_loss: 4.1085 - val_dice_coefficient: 0.1238
Epoch 10/15
97/97 [========== ] - 35s 357ms/step - loss: 0.9129 -
dice_coefficient: 0.6932 - val_loss: 3.4194 - val_dice_coefficient: 0.2391
Epoch 11/15
97/97 [========== ] - 35s 361ms/step - loss: 0.8739 -
dice_coefficient: 0.7027 - val_loss: 3.0801 - val_dice_coefficient: 0.3176
Epoch 12/15
dice_coefficient: 0.7403 - val_loss: 2.4698 - val_dice_coefficient: 0.4232
Epoch 13/15
97/97 [============= ] - 35s 366ms/step - loss: 0.7812 -
dice_coefficient: 0.7380 - val_loss: 3.8237 - val_dice_coefficient: 0.2501
Epoch 14/15
97/97 [========== ] - 34s 355ms/step - loss: 0.8731 -
dice_coefficient: 0.7083 - val_loss: 3.1624 - val_dice_coefficient: 0.3475
Epoch 15/15
dice_coefficient: 0.7561 - val_loss: 1.6883 - val_dice_coefficient: 0.5711
Epoch 1/15
dice_coefficient: 0.2370 - val_loss: 3.5258 - val_dice_coefficient: 0.1604
Epoch 2/15
dice_coefficient: 0.4046 - val_loss: 4.2236 - val_dice_coefficient: 0.0720
Epoch 3/15
dice_coefficient: 0.5125 - val_loss: 2.8424 - val_dice_coefficient: 0.2491
Epoch 4/15
dice_coefficient: 0.5666 - val_loss: 1.9611 - val_dice_coefficient: 0.4149
dice_coefficient: 0.5846 - val_loss: 1.5030 - val_dice_coefficient: 0.5769
Epoch 6/15
dice_coefficient: 0.5869 - val_loss: 1.9437 - val_dice_coefficient: 0.4466
```

```
Epoch 7/15
dice_coefficient: 0.6355 - val_loss: 3.3381 - val_dice_coefficient: 0.1531
dice_coefficient: 0.6539 - val_loss: 1.7788 - val_dice_coefficient: 0.5042
dice_coefficient: 0.6668 - val_loss: 2.5402 - val_dice_coefficient: 0.3543
Epoch 10/15
dice_coefficient: 0.6727 - val_loss: 2.3936 - val_dice_coefficient: 0.3246
Epoch 11/15
dice_coefficient: 0.7206 - val_loss: 1.2267 - val_dice_coefficient: 0.6155
Epoch 12/15
dice_coefficient: 0.7350 - val_loss: 1.4243 - val_dice_coefficient: 0.5644
Epoch 13/15
dice_coefficient: 0.7462 - val_loss: 1.6213 - val_dice_coefficient: 0.4800
Epoch 14/15
dice_coefficient: 0.7478 - val_loss: 1.5904 - val_dice_coefficient: 0.5472
Epoch 15/15
dice_coefficient: 0.7392 - val_loss: 1.0066 - val_dice_coefficient: 0.6763
```



Dice + BCE - Data Split : 60.0% , Batch Size = 15 Lambda 1 = 0 Lambda 2 = 0.001loss 1.0 val_loss dice_coefficient val_dice_coefficient Maxima 0.8 Metrics 9.0 0.4 0.2 2 6 8 Number of Epochs ò 10 12 14 4





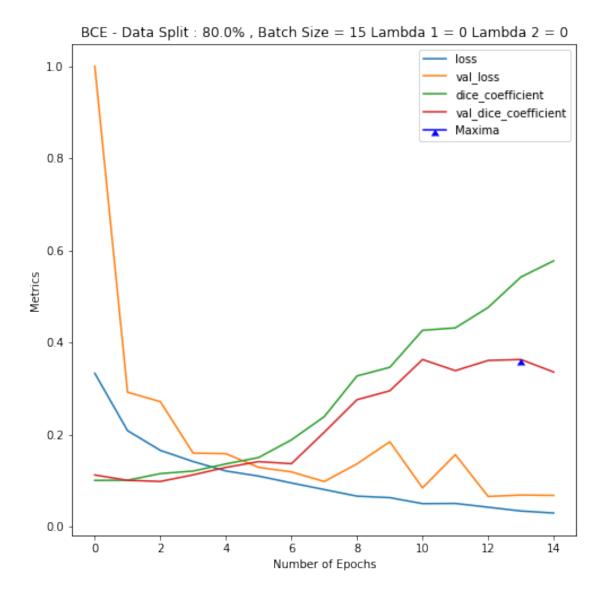
```
[489]: experiment1(0.8, 15, 15, 0, 0)
experiment1(0.6, 15, 15, 0, 0)
experiment1(0.4, 15, 15, 0, 0)
experiment1(0.2, 15, 15, 0, 0)
```

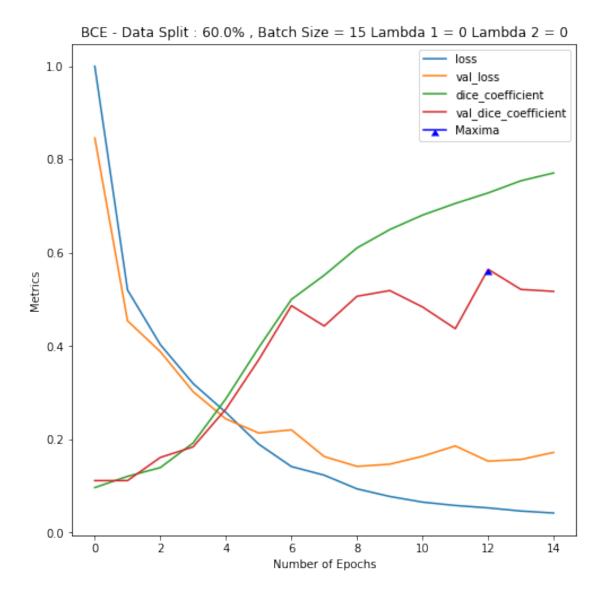
```
dice_coefficient: 0.1207 - val_loss: 0.2437 - val_dice_coefficient: 0.1125
Epoch 5/15
dice_coefficient: 0.1363 - val_loss: 0.2418 - val_dice_coefficient: 0.1286
Epoch 6/15
dice_coefficient: 0.1501 - val_loss: 0.1965 - val_dice_coefficient: 0.1412
Epoch 7/15
dice_coefficient: 0.1885 - val_loss: 0.1814 - val_dice_coefficient: 0.1367
Epoch 8/15
dice_coefficient: 0.2389 - val_loss: 0.1495 - val_dice_coefficient: 0.2050
dice_coefficient: 0.3273 - val_loss: 0.2080 - val_dice_coefficient: 0.2757
Epoch 10/15
dice_coefficient: 0.3460 - val_loss: 0.2812 - val_dice_coefficient: 0.2950
48/48 [============= ] - 22s 463ms/step - loss: 0.0763 -
dice_coefficient: 0.4264 - val_loss: 0.1285 - val_dice_coefficient: 0.3630
Epoch 12/15
dice_coefficient: 0.4317 - val_loss: 0.2386 - val_dice_coefficient: 0.3388
Epoch 13/15
dice_coefficient: 0.4760 - val_loss: 0.1001 - val_dice_coefficient: 0.3610
Epoch 14/15
dice_coefficient: 0.5422 - val_loss: 0.1047 - val_dice_coefficient: 0.3631
Epoch 15/15
dice coefficient: 0.5773 - val loss: 0.1035 - val dice coefficient: 0.3358
Epoch 1/15
97/97 [=========== ] - 37s 382ms/step - loss: 0.4102 -
dice_coefficient: 0.0963 - val_loss: 0.3471 - val_dice_coefficient: 0.1112
Epoch 2/15
dice_coefficient: 0.1204 - val_loss: 0.1862 - val_dice_coefficient: 0.1113
Epoch 3/15
dice_coefficient: 0.1392 - val_loss: 0.1591 - val_dice_coefficient: 0.1610
Epoch 4/15
dice_coefficient: 0.1920 - val_loss: 0.1239 - val_dice_coefficient: 0.1836
Epoch 5/15
```

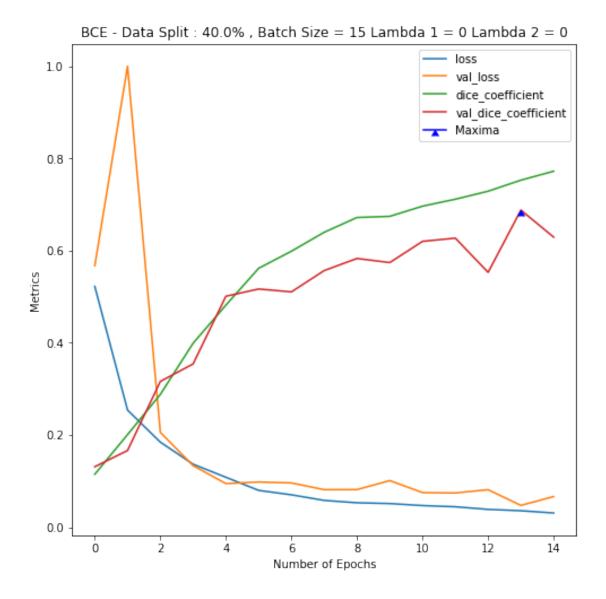
```
dice_coefficient: 0.2868 - val_loss: 0.0999 - val_dice_coefficient: 0.2647
Epoch 6/15
dice_coefficient: 0.3963 - val_loss: 0.0875 - val_dice_coefficient: 0.3702
Epoch 7/15
dice_coefficient: 0.4999 - val_loss: 0.0903 - val_dice_coefficient: 0.4864
Epoch 8/15
dice_coefficient: 0.5517 - val_loss: 0.0668 - val_dice_coefficient: 0.4429
dice_coefficient: 0.6103 - val_loss: 0.0582 - val_dice_coefficient: 0.5066
dice_coefficient: 0.6497 - val_loss: 0.0601 - val_dice_coefficient: 0.5189
Epoch 11/15
dice_coefficient: 0.6808 - val_loss: 0.0671 - val_dice_coefficient: 0.4837
97/97 [========== - 36s 371ms/step - loss: 0.0238 -
dice_coefficient: 0.7058 - val_loss: 0.0762 - val_dice_coefficient: 0.4371
Epoch 13/15
dice_coefficient: 0.7283 - val_loss: 0.0627 - val_dice_coefficient: 0.5645
Epoch 14/15
dice_coefficient: 0.7543 - val_loss: 0.0642 - val_dice_coefficient: 0.5213
Epoch 15/15
97/97 [=========== ] - 36s 373ms/step - loss: 0.0172 -
dice_coefficient: 0.7712 - val_loss: 0.0704 - val_dice_coefficient: 0.5171
Epoch 1/15
dice coefficient: 0.1150 - val loss: 0.3504 - val dice coefficient: 0.1313
Epoch 2/15
dice_coefficient: 0.2011 - val_loss: 0.6175 - val_dice_coefficient: 0.1666
Epoch 3/15
dice_coefficient: 0.2882 - val_loss: 0.1272 - val_dice_coefficient: 0.3164
Epoch 4/15
dice_coefficient: 0.3993 - val_loss: 0.0825 - val_dice_coefficient: 0.3543
Epoch 5/15
dice_coefficient: 0.4819 - val_loss: 0.0585 - val_dice_coefficient: 0.5011
Epoch 6/15
```

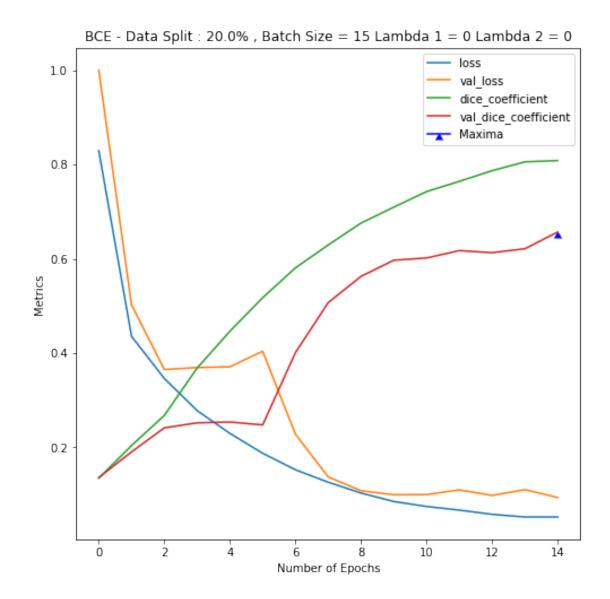
```
dice_coefficient: 0.5618 - val_loss: 0.0607 - val_dice_coefficient: 0.5168
Epoch 7/15
dice_coefficient: 0.5988 - val_loss: 0.0594 - val_dice_coefficient: 0.5106
Epoch 8/15
dice_coefficient: 0.6399 - val_loss: 0.0504 - val_dice_coefficient: 0.5568
Epoch 9/15
dice_coefficient: 0.6719 - val_loss: 0.0505 - val_dice_coefficient: 0.5831
Epoch 10/15
dice_coefficient: 0.6745 - val_loss: 0.0625 - val_dice_coefficient: 0.5742
dice_coefficient: 0.6965 - val_loss: 0.0465 - val_dice_coefficient: 0.6202
Epoch 12/15
dice_coefficient: 0.7115 - val_loss: 0.0460 - val_dice_coefficient: 0.6271
dice_coefficient: 0.7289 - val_loss: 0.0503 - val_dice_coefficient: 0.5529
Epoch 14/15
dice_coefficient: 0.7529 - val_loss: 0.0293 - val_dice_coefficient: 0.6880
Epoch 15/15
dice_coefficient: 0.7723 - val_loss: 0.0411 - val_dice_coefficient: 0.6293
Epoch 1/15
dice_coefficient: 0.1347 - val_loss: 0.3118 - val_dice_coefficient: 0.1361
Epoch 2/15
dice coefficient: 0.2040 - val loss: 0.1566 - val dice coefficient: 0.1906
Epoch 3/15
dice_coefficient: 0.2679 - val_loss: 0.1139 - val_dice_coefficient: 0.2416
Epoch 4/15
dice_coefficient: 0.3683 - val_loss: 0.1152 - val_dice_coefficient: 0.2521
Epoch 5/15
dice_coefficient: 0.4464 - val_loss: 0.1157 - val_dice_coefficient: 0.2539
Epoch 6/15
dice_coefficient: 0.5175 - val_loss: 0.1260 - val_dice_coefficient: 0.2477
Epoch 7/15
```

```
dice_coefficient: 0.5808 - val_loss: 0.0711 - val_dice_coefficient: 0.4011
Epoch 8/15
dice_coefficient: 0.6294 - val_loss: 0.0429 - val_dice_coefficient: 0.5072
Epoch 9/15
dice_coefficient: 0.6758 - val_loss: 0.0338 - val_dice_coefficient: 0.5628
Epoch 10/15
dice_coefficient: 0.7096 - val_loss: 0.0311 - val_dice_coefficient: 0.5970
Epoch 11/15
dice_coefficient: 0.7426 - val_loss: 0.0312 - val_dice_coefficient: 0.6018
dice_coefficient: 0.7643 - val_loss: 0.0343 - val_dice_coefficient: 0.6175
Epoch 13/15
dice_coefficient: 0.7871 - val_loss: 0.0306 - val_dice_coefficient: 0.6130
Epoch 14/15
dice_coefficient: 0.8056 - val_loss: 0.0345 - val_dice_coefficient: 0.6214
Epoch 15/15
dice_coefficient: 0.8085 - val_loss: 0.0292 - val_dice_coefficient: 0.6569
```









[]: