

```

File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:1411 in
fbeta_score
    _, _, f, _ = precision_recall_fscore_support(

File ~\anaconda3\Lib\site-packages\sklearn\utils\_param_validation.py:184 in wrapper
    return func(*args, **kwargs)

File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:1721 in
precision_recall_fscore_support
    labels = _check_set_wise_labels(y_true, y_pred, average, labels, pos_label)

File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:1499 in
_check_set_wise_labels
    y_type, y_true, y_pred = _check_targets(y_true, y_pred)

File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:93 in
_check_targets
    raise ValueError(

ValueError: Classification metrics can't handle a mix of continuous and binary targets

```

```

In [99]:
...:     = .
...:     = > 0.5
...:
...:     = .
...:     = .
...:     = . / .
...: print "IoU socre is: "
3/3 [=====] - 0s 76ms/step
Traceback (most recent call last):

```

```

Cell In[99], line 4
    intersection = np.logical_and(y_valid, y_pred_thresholded)

```

MemoryError: Unable to allocate 1.27 TiB for an array with shape (72, 128, 128, 1179648) and data type bool

In [100]:

Removing all variables...

```

In [100]:
...: """
...: Created on Sat Jan 6 17:25:02 2024
...:
...: @author: Sabbir Ahmed SibLi
...: source: https://github.com/hLamba28/UNET-TGS/blob/master/TGS%20UNET.ipynb
...: """
...:
...: from          import
...: import
...: import        as

```

```

...: import                                as
...: import
...: from                                import
...:
...: from                                import
...: from                                import
...:
...: from                                import
...: from                                import
...: from                                import
...:
...: # Set some params
...:     = 128
...:     = 128
...:     = 5
...:
...: # Loading the dataset and the masks
...: # Loading the dataset (Original and Masked)
...:     = 'D:/Course Materials [Erasmus MSc]/University of Kragujevac/
Biomedical Image Processing/Assignments/Datasets/Spine_DICOM'
...:     = 'D:/Course Materials [Erasmus MSc]/University of Kragujevac/
Biomedical Image Processing/Assignments/Datasets/masked_spines'
...:
...: # Reading original dicom slices
...:     = .                                format='dcm'
...:
...: # Reading masked images
...:     =
...: for         in
...:     if not
...:         continue
...:     .
...:
...:     = .                                # converting list into numpy array
...:     = -1
...:
...: # Resize Image to 128x128
...:     = 128
...:     = 128
...:
Reading DICOM (examining files): 717/717 files (100.0%)
Found 1 correct series.
Reading DICOM (loading data): 717/717 (100.0%)
C:\Users\Sabbir Ahmed Sibli\AppData\Local\Temp\ipykernel_18272\312730056.py:41:
DeprecationWarning: Starting with ImageIO v3 the behavior of this function will switch
to that of iio.v3.imread. To keep the current behavior (and make this warning disappear)
use `import imageio.v2 as imageio` or call `imageio.v2.imread` directly.
    masks.append(imageio.imread(file))

In [101]: 'D:/Course Materials [Erasmus MSc]/University of Kragujevac/Biomedical
Image Processing/Assignments/Seminar Paper/Codes/UNET_Segmentation/UNET_Example_02/
Unet_Portion.py' = 'D:/Course Materials [Erasmus MSc]/University of Kragujevac/
Biomedical Image Processing/Assignments/Seminar Paper/Codes/UNET_Segmentation/
UNET_Example_02'
Reloaded modules: Unet_Portion

In [102]:
...:     = .                                len                                1

```

```

# Create array of zeros for data
...: = . len 1 = .
#Create array of zeros for masks
...: for in range 0 len
...: # Load original images
...: =
...: = 1 = 'constant'
...: = True
...: # Load masks
...: =
...: = 1 = 'constant'
...: = True
...: # Creating Normalized image (converting all pixel values between 0 and 1)
...: = /255.0
...: = /255.0
...:
...: # Split train and valid
...: = =0.1
...: =42

```

```

In [103]:
...: = . 0 len
...: = 12 6
...: 121
...: 'Dicom Slice'
...: 128 128 = 'gray'
...: 122
...: 'Corresponding Mask'
...: 128 128 = 'gray'
...:

```

```

In [104]:
...: =
...: 1 = 'img'
...: =16 =0.05 =True
...: = "binary_crossentropy"
...: = "accuracy"

```

```

In [105]:
...:
...:
...: = 'best_model.h5' = 'val_loss'
...: =True = 'min' =1
...: =32 =20
...: =

```

Model: "model_4"

Layer (type)	Output Shape	Param #	Connected to
=====			
img (InputLayer)	[(None, 128, 128, 1)]	0	[]
conv2d_115 (Conv2D)	(None, 128, 128, 16)	160	['img[0][0]']

batch_normalization_109 (BatchNormalization)	(None, 128, 128, 16)	64	['conv2d_115[0][0]']
activation_109 (Activation)	(None, 128, 128, 16)	0	['batch_normalization_109[0][0]']
max_pooling2d_24 (MaxPooling2D)	(None, 64, 64, 16)	0	['activation_109[0][0]']
dropout_32 (Dropout)	(None, 64, 64, 16)	0	['max_pooling2d_24[0][0]']
conv2d_117 (Conv2D)	(None, 64, 64, 32)	4640	['dropout_32[0][0]']
batch_normalization_111 (BatchNormalization)	(None, 64, 64, 32)	128	['conv2d_117[0][0]']
activation_111 (Activation)	(None, 64, 64, 32)	0	['batch_normalization_111[0][0]']
max_pooling2d_25 (MaxPooling2D)	(None, 32, 32, 32)	0	['activation_111[0][0]']
dropout_33 (Dropout)	(None, 32, 32, 32)	0	['max_pooling2d_25[0][0]']
conv2d_119 (Conv2D)	(None, 32, 32, 64)	18496	['dropout_33[0][0]']
batch_normalization_113 (BatchNormalization)	(None, 32, 32, 64)	256	['conv2d_119[0][0]']
activation_113 (Activation)	(None, 32, 32, 64)	0	['batch_normalization_113[0][0]']
max_pooling2d_26 (MaxPooling2D)	(None, 16, 16, 64)	0	['activation_113[0][0]']
dropout_34 (Dropout)	(None, 16, 16, 64)	0	['max_pooling2d_26[0][0]']
conv2d_121 (Conv2D)	(None, 16, 16, 128)	73856	['dropout_34[0][0]']
batch_normalization_115 (BatchNormalization)	(None, 16, 16, 128)	512	['conv2d_121[0][0]']
activation_115 (Activation)	(None, 16, 16, 128)	0	['batch_normalization_115[0][0]']
max_pooling2d_27 (MaxPooling2D)	(None, 8, 8, 128)	0	['activation_115[0][0]']

dropout_35 (Dropout)	(None, 8, 8, 128)	0	
['max_pooling2d_27[0][0]']			
conv2d_123 (Conv2D)	(None, 8, 8, 256)	295168	['dropout_35[0][0]']
batch_normalization_117 (Batch Normalization)	(None, 8, 8, 256)	1024	['conv2d_123[0][0]']
activation_117 (Activation)	(None, 8, 8, 256)	0	
['batch_normalization_117[0][0]']			
conv2d_transpose_24 (Conv2DTranspose)	(None, 16, 16, 128)	295040	['activation_117[0]']
concatenate_24 (Concatenate)	(None, 16, 16, 256)	0	
['conv2d_transpose_24[0][0]', 'activation_115[0]']			
dropout_36 (Dropout)	(None, 16, 16, 256)	0	['concatenate_24[0]']
conv2d_125 (Conv2D)	(None, 16, 16, 128)	295040	['dropout_36[0][0]']
batch_normalization_119 (Batch Normalization)	(None, 16, 16, 128)	512	['conv2d_125[0][0]']
activation_119 (Activation)	(None, 16, 16, 128)	0	
['batch_normalization_119[0][0]']			
conv2d_transpose_25 (Conv2DTranspose)	(None, 32, 32, 64)	73792	['activation_119[0]']
concatenate_25 (Concatenate)	(None, 32, 32, 128)	0	
['conv2d_transpose_25[0][0]', 'activation_113[0]']			
dropout_37 (Dropout)	(None, 32, 32, 128)	0	['concatenate_25[0]']
conv2d_127 (Conv2D)	(None, 32, 32, 64)	73792	['dropout_37[0][0]']
batch_normalization_121 (Batch Normalization)	(None, 32, 32, 64)	256	['conv2d_127[0][0]']
activation_121 (Activation)	(None, 32, 32, 64)	0	
['batch_normalization_121[0][0]']			
conv2d_transpose_26 (Conv2DTranspose)	(None, 64, 64, 32)	18464	['activation_121[0]']

concatenate_26 (Concatenat ['conv2d_transpose_26[0][0]', e) [0]']	(None, 64, 64, 64)	0	'activation_111[0]
dropout_38 (Dropout) [0]']	(None, 64, 64, 64)	0	['concatenate_26[0]
conv2d_129 (Conv2D)	(None, 64, 64, 32)	18464	['dropout_38[0][0]']
batch_normalization_123 (B atchNormalization)	(None, 64, 64, 32)	128	['conv2d_129[0][0]']
activation_123 (Activation ['batch_normalization_123[0][0)	(None, 64, 64, 32)	0	']']
conv2d_transpose_27 (Conv2 DTranspose)	(None, 128, 128, 16)	4624	['activation_123[0]
concatenate_27 (Concatenat ['conv2d_transpose_27[0][0]', e) [0]']	(None, 128, 128, 32)	0	'activation_109[0]
dropout_39 (Dropout) [0]']	(None, 128, 128, 32)	0	['concatenate_27[0]
conv2d_131 (Conv2D)	(None, 128, 128, 16)	4624	['dropout_39[0][0]']
batch_normalization_125 (B atchNormalization)	(None, 128, 128, 16)	64	['conv2d_131[0][0]']
activation_125 (Activation ['batch_normalization_125[0][0)	(None, 128, 128, 16)	0	']']
conv2d_132 (Conv2D) [0]']	(None, 128, 128, 1)	17	['activation_125[0]

=====

=====

Total params: 1179121 (4.50 MB)
Trainable params: 1177649 (4.49 MB)
Non-trainable params: 1472 (5.75 KB)

Epoch 1/20

21/21 [=====] - ETA: 0s - loss: 0.4104 - accuracy: 0.8572

Epoch 1: val_loss improved from inf to 0.23607, saving model to best_model.h5

C:\Users\Sabbir Ahmed Sibli\anaconda3\Lib\site-packages\keras\src\engine\training.py:
3103: UserWarning: You are saving your model as an HDF5 file via `model.save()`. This
file format is considered legacy. We recommend using instead the native Keras format,
e.g. `model.save('my_model.keras')`.

saving_api.save_model(

21/21 [=====] - 15s 567ms/step - loss: 0.4104 - accuracy:

```

0.8572 - val_loss: 0.2361 - val_accuracy: 0.8944
Epoch 2/20
21/21 [=====] - ETA: 0s - loss: 0.2596 - accuracy: 0.9216
Epoch 2: val_loss improved from 0.23607 to 0.21248, saving model to best_model.h5
21/21 [=====] - 24s 1s/step - loss: 0.2596 - accuracy: 0.9216 -
val_loss: 0.2125 - val_accuracy: 0.9039
Epoch 3/20
21/21 [=====] - ETA: 0s - loss: 0.1964 - accuracy: 0.9246
Epoch 3: val_loss did not improve from 0.21248
21/21 [=====] - 25s 1s/step - loss: 0.1964 - accuracy: 0.9246 -
val_loss: 0.5947 - val_accuracy: 0.7334
Epoch 4/20
21/21 [=====] - ETA: 0s - loss: 0.1609 - accuracy: 0.9248
Epoch 4: val_loss did not improve from 0.21248
21/21 [=====] - 24s 1s/step - loss: 0.1609 - accuracy: 0.9248 -
val_loss: 0.4217 - val_accuracy: 0.7762
Epoch 5/20
21/21 [=====] - ETA: 0s - loss: 0.1358 - accuracy: 0.9249
Epoch 5: val_loss did not improve from 0.21248
21/21 [=====] - 24s 1s/step - loss: 0.1358 - accuracy: 0.9249 -
val_loss: 0.3346 - val_accuracy: 0.8301
Epoch 6/20
21/21 [=====] - ETA: 0s - loss: 0.1164 - accuracy: 0.9249
Epoch 6: val_loss improved from 0.21248 to 0.14507, saving model to best_model.h5
21/21 [=====] - 18s 861ms/step - loss: 0.1164 - accuracy:
0.9249 - val_loss: 0.1451 - val_accuracy: 0.9178
Epoch 7/20
21/21 [=====] - ETA: 0s - loss: 0.1005 - accuracy: 0.9249
Epoch 7: val_loss improved from 0.14507 to 0.10951, saving model to best_model.h5
21/21 [=====] - 15s 736ms/step - loss: 0.1005 - accuracy:
0.9249 - val_loss: 0.1095 - val_accuracy: 0.9236
Epoch 8/20
21/21 [=====] - ETA: 0s - loss: 0.0874 - accuracy: 0.9249
Epoch 8: val_loss improved from 0.10951 to 0.08731, saving model to best_model.h5
21/21 [=====] - 15s 716ms/step - loss: 0.0874 - accuracy:
0.9249 - val_loss: 0.0873 - val_accuracy: 0.9248
Epoch 9/20
21/21 [=====] - ETA: 0s - loss: 0.0765 - accuracy: 0.9249
Epoch 9: val_loss improved from 0.08731 to 0.07442, saving model to best_model.h5
21/21 [=====] - 20s 971ms/step - loss: 0.0765 - accuracy:
0.9249 - val_loss: 0.0744 - val_accuracy: 0.9249
Epoch 10/20
21/21 [=====] - ETA: 0s - loss: 0.0676 - accuracy: 0.9249
Epoch 10: val_loss improved from 0.07442 to 0.06252, saving model to best_model.h5
21/21 [=====] - 25s 1s/step - loss: 0.0676 - accuracy: 0.9249 -
val_loss: 0.0625 - val_accuracy: 0.9249
Epoch 11/20
21/21 [=====] - ETA: 0s - loss: 0.0601 - accuracy: 0.9249
Epoch 11: val_loss improved from 0.06252 to 0.05419, saving model to best_model.h5
21/21 [=====] - 25s 1s/step - loss: 0.0601 - accuracy: 0.9249 -
val_loss: 0.0542 - val_accuracy: 0.9249
Epoch 12/20
21/21 [=====] - ETA: 0s - loss: 0.0540 - accuracy: 0.9249
Epoch 12: val_loss improved from 0.05419 to 0.04805, saving model to best_model.h5
21/21 [=====] - 24s 1s/step - loss: 0.0540 - accuracy: 0.9249 -
val_loss: 0.0481 - val_accuracy: 0.9249
Epoch 13/20
21/21 [=====] - ETA: 0s - loss: 0.0490 - accuracy: 0.9249

```

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Epoch 13: val_loss improved from 0.04805 to 0.04250, saving model to best_model.h5
21/21 [=====] - 24s 1s/step - loss: 0.0490 - accuracy: 0.9249 -
val_loss: 0.0425 - val_accuracy: 0.9249
Epoch 14/20
21/21 [=====] - ETA: 0s - loss: 0.0444 - accuracy: 0.9249
Epoch 14: val_loss improved from 0.04250 to 0.03971, saving model to best_model.h5
21/21 [=====] - 16s 752ms/step - loss: 0.0444 - accuracy:
0.9249 - val_loss: 0.0397 - val_accuracy: 0.9249
Epoch 15/20
21/21 [=====] - ETA: 0s - loss: 0.0408 - accuracy: 0.9249
Epoch 15: val_loss improved from 0.03971 to 0.03616, saving model to best_model.h5
21/21 [=====] - 24s 1s/step - loss: 0.0408 - accuracy: 0.9249 -
val_loss: 0.0362 - val_accuracy: 0.9249
Epoch 16/20
21/21 [=====] - ETA: 0s - loss: 0.0375 - accuracy: 0.9249
Epoch 16: val_loss improved from 0.03616 to 0.03385, saving model to best_model.h5
21/21 [=====] - 25s 1s/step - loss: 0.0375 - accuracy: 0.9249 -
val_loss: 0.0338 - val_accuracy: 0.9249
Epoch 17/20
21/21 [=====] - ETA: 0s - loss: 0.0347 - accuracy: 0.9249
Epoch 17: val_loss improved from 0.03385 to 0.03341, saving model to best_model.h5
21/21 [=====] - 16s 768ms/step - loss: 0.0347 - accuracy:
0.9249 - val_loss: 0.0334 - val_accuracy: 0.9249
Epoch 18/20
21/21 [=====] - ETA: 0s - loss: 0.0325 - accuracy: 0.9249
Epoch 18: val_loss improved from 0.03341 to 0.03024, saving model to best_model.h5
21/21 [=====] - 19s 894ms/step - loss: 0.0325 - accuracy:
0.9249 - val_loss: 0.0302 - val_accuracy: 0.9249
Epoch 19/20
21/21 [=====] - ETA: 0s - loss: 0.0304 - accuracy: 0.9249
Epoch 19: val_loss improved from 0.03024 to 0.02858, saving model to best_model.h5
21/21 [=====] - 16s 779ms/step - loss: 0.0304 - accuracy:
0.9249 - val_loss: 0.0286 - val_accuracy: 0.9249
Epoch 20/20
21/21 [=====] - ETA: 0s - loss: 0.0285 - accuracy: 0.9249
Epoch 20: val_loss improved from 0.02858 to 0.02718, saving model to best_model.h5
21/21 [=====] - 16s 757ms/step - loss: 0.0285 - accuracy:
0.9249 - val_loss: 0.0272 - val_accuracy: 0.9249

```

```

In [106]:
...:
...:
...:      =      .      'loss'
...:      =      .      'val_loss'
...:      = range 1 len      + 1
...:      .      'y'      = 'Training loss'
...:      .      'r'      = 'Validation loss'
...:      .      'Training and validation loss'
...:      .      'Epochs'
...:      .      'Loss'
...:      .
...:

```

```

In [107]:
...:
...:      =      .      'accuracy'
...:      =      .      'val_accuracy'
...:      = range 1 len      + 1

```



```

...:         'y'         ='Training acc'
...:         'r'         ='Validation acc'
...:         'Training and validation accuracy'
...:         'Epochs'
...:         'Accuracy'
...:
In [108]:
...:
...:         =         .         0 len
...:         =
...:         =         0         None
...:         =         .
...:         =         .         0         0         > 0.2 .
1/1 [=====] - 0s 301ms/step

In [109]:
...:
...:         = 16 8
...:         231
...:         'Testing Image'
...:         0         ='gray'
...:         232
...:         'Testing Label'
...:         0         ='gray'
...:         233
...:         'Prediction on test image'
...:         ='gray'
...:
In [110]:
...:
...:         =         .
...:         =         > 0.5
...:
...:         =         .
...:         =         .
...:         =         .         /
...: print "IoU socre is: "
3/3 [=====] - 0s 74ms/step
IoU socre is: 0.5673644703919933

In [111]:
...:
...: from         import
...:
...: # Reshape y_valid and y_pred_thresholded if needed (e.g., if they are 3D)
...:         =         .         -1
...:         =         .         -1
...:
...: # Calculate Dice Coefficient using the f1_score function
...:         =
...:
...: print "Dice Coefficient Score:"
Traceback (most recent call last):

```

```

Cell In[111], line 8
    dice_coefficient = f1_score(y_valid_reshape, y_pred_thresholded_reshape)

File ~\anaconda3\Lib\site-packages\sklearn\utils\_param_validation.py:211 in wrapper
    return func(*args, **kwargs)

File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:1238 in f1_score
    return fbeta_score(

File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:1411 in
fbeta_score
    _, _, f, _ = precision_recall_fscore_support(

File ~\anaconda3\Lib\site-packages\sklearn\utils\_param_validation.py:184 in wrapper
    return func(*args, **kwargs)

File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:1721 in
precision_recall_fscore_support
    labels = _check_set_wise_labels(y_true, y_pred, average, labels, pos_label)

File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:1499 in
_check_set_wise_labels
    y_type, y_true, y_pred = _check_targets(y_true, y_pred)

File ~\anaconda3\Lib\site-packages\sklearn\metrics\_classification.py:93 in
_check_targets
    raise ValueError(

```

ValueError: Classification metrics can't handle a mix of continuous and binary targets

```

In [112]:
    ...:
    ...:         =
    ...:         = > 0.5
    ...:
    ...:         =
    ...:         =
    ...:         =
    ...:         =
    ...: print "IoU socre is: "
3/3 [=====] - 0s 78ms/step
IoU socre is: 0.5673644703919933

```

```

In [113]:
    ...:         = 2.0 *
    ...:         + 1e-8
    ...: print "Dice Coefficient:"
Dice Coefficient: 0.9992650215136403

```

```

In [114]:
    ...:         =
    ...:         = > 0.5
    ...:
    ...:         =

```

```

...:         = .
...:         = . / .
...: print "IoU socre is: "
3/3 [=====] - 0s 76ms/step
IoU socre is: 0.5673644703919933

In [115]:         = 2.0 * . / . +
.             + 1e-8
...: print "Dice Coefficient:"
Dice Coefficient: 0.9992650215136403

In [116]:

```