

```
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
import os
import numpy as np
from sklearn.model_selection import train_test_split
```

```
# Define your directories
base_dir = '/content/drive/MyDrive/Data/Data'
```

```
# Create lists to hold image and mask paths
image_paths = []
mask_paths = []
```

```
# Iterate through all folders in the base directory
for folder in os.listdir(base_dir):
    folder_path = os.path.join(base_dir, folder)
    if os.path.isdir(folder_path): # Check if it is a directory
        # List all image and mask files
        images = sorted([os.path.join(folder_path, img) for img in os.listdir(folder_path) if img.endswith('.tif') and not img.endswith('_mask.tif')])
        masks = sorted([os.path.join(folder_path, msk) for msk in os.listdir(folder_path) if msk.endswith('_mask.tif')])
```

```
    # Ensure we only keep image and mask pairs
    for img_path in images:
        # Create the corresponding mask path
        mask_name = os.path.basename(img_path).replace('.tif', '_mask.tif')
        mask_path = os.path.join(folder_path, mask_name)
```

```
        if mask_path in masks: # Check if the corresponding mask exists
            image_paths.append(img_path)
            mask_paths.append(mask_path)
```

```
# Ensure we have images and masks
print(f"Total images: {len(image_paths)}")
print(f"Total masks: {len(mask_paths)}")
```

```
# Check if we have matching lengths
if len(image_paths) != len(mask_paths):
    print("Warning: Number of images does not match number of masks!")
```

```
# Split into 80% training and 20% validation
train_image_paths, val_image_paths, train_mask_paths, val_mask_paths = train_test_split(image_paths, mask_paths, test_size=0.2, random_state=42)
```

```
print(f"Total training images: {len(train_image_paths)}")
print(f"Total validation images: {len(val_image_paths)}")
```

Total images: 116
Total masks: 116
Total training images: 92
Total validation images: 24

```

import numpy as np
import cv2
from tensorflow.keras.utils import Sequence
import os

class DataGenerator(Sequence):
    def __init__(self, image_paths, mask_paths, batch_size=16, img_size=(256, 256), shuffle=True):
        self.image_paths = image_paths
        self.mask_paths = mask_paths
        self.batch_size = batch_size
        self.img_size = img_size
        self.shuffle = shuffle
        self.indices = np.arange(len(self.image_paths))
        self.on_epoch_end()

    def __len__(self):
        return int(np.floor(len(self.image_paths) / self.batch_size))

    def __getitem__(self, index):
        indices = self.indices[index * self.batch_size:(index + 1) * self.batch_size]
        batch_image_paths = [self.image_paths[i] for i in indices]
        batch_mask_paths = [self.mask_paths[i] for i in indices]
        return self.__data_generation(batch_image_paths, batch_mask_paths)

    def on_epoch_end(self):
        if self.shuffle:
            np.random.shuffle(self.indices)

    def __data_generation(self, batch_image_paths, batch_mask_paths):
        images = np.empty((self.batch_size, *self.img_size, 3))
        masks = np.empty((self.batch_size, *self.img_size, 1))

        for i, (img_path, mask_path) in enumerate(zip(batch_image_paths, batch_mask_paths)):
            # Load and preprocess the image
            img = cv2.imread(img_path)
            img = cv2.resize(img, self.img_size)
            img = img / 255.0 # Normalize the image
            images[i,] = img

            # Load and preprocess the mask
            mask = cv2.imread(mask_path, cv2.IMREAD_GRAYSCALE)
            mask = cv2.resize(mask, self.img_size)
            mask = np.expand_dims(mask / 255.0, axis=-1) # Normalize and expand dims
            masks[i,] = mask

        return images, masks

import tensorflow as tf
from tensorflow.keras import layers, models

def nested_unet(input_shape=(256, 256, 3), num_classes=1):
    inputs = layers.Input(shape=input_shape)

```

```
# Encoder
conv1 = layers.Conv2D(64, (3, 3), padding='same')(inputs)
conv1 = layers.BatchNormalization()(conv1)
conv1 = layers.Activation('relu')(conv1)
conv1 = layers.Conv2D(64, (3, 3), padding='same')(conv1)
conv1 = layers.BatchNormalization()(conv1)
conv1 = layers.Activation('relu')(conv1)
pool1 = layers.MaxPooling2D((2, 2))(conv1)

conv2 = layers.Conv2D(128, (3, 3), padding='same')(pool1)
conv2 = layers.BatchNormalization()(conv2)
conv2 = layers.Activation('relu')(conv2)
conv2 = layers.Conv2D(128, (3, 3), padding='same')(conv2)
conv2 = layers.BatchNormalization()(conv2)
conv2 = layers.Activation('relu')(conv2)
pool2 = layers.MaxPooling2D((2, 2))(conv2)

conv3 = layers.Conv2D(256, (3, 3), padding='same')(pool2)
conv3 = layers.BatchNormalization()(conv3)
conv3 = layers.Activation('relu')(conv3)
conv3 = layers.Conv2D(256, (3, 3), padding='same')(conv3)
conv3 = layers.BatchNormalization()(conv3)
conv3 = layers.Activation('relu')(conv3)
pool3 = layers.MaxPooling2D((2, 2))(conv3)

conv4 = layers.Conv2D(512, (3, 3), padding='same')(pool3)
conv4 = layers.BatchNormalization()(conv4)
conv4 = layers.Activation('relu')(conv4)
conv4 = layers.Conv2D(512, (3, 3), padding='same')(conv4)
conv4 = layers.BatchNormalization()(conv4)
conv4 = layers.Activation('relu')(conv4)
pool4 = layers.MaxPooling2D((2, 2))(conv4)

conv5 = layers.Conv2D(1024, (3, 3), padding='same')(pool4)
conv5 = layers.BatchNormalization()(conv5)
conv5 = layers.Activation('relu')(conv5)
conv5 = layers.Conv2D(1024, (3, 3), padding='same')(conv5)
conv5 = layers.BatchNormalization()(conv5)
conv5 = layers.Activation('relu')(conv5)

# Decoder
up6 = layers.Conv2DTranspose(512, (2, 2), strides=(2, 2), padding='same')(conv5)
merge6 = layers.concatenate([up6, conv4], axis=3)
conv6 = layers.Conv2D(512, (3, 3), padding='same')(merge6)
conv6 = layers.BatchNormalization()(conv6)
conv6 = layers.Activation('relu')(conv6)
conv6 = layers.Conv2D(512, (3, 3), padding='same')(conv6)
conv6 = layers.BatchNormalization()(conv6)
conv6 = layers.Activation('relu')(conv6)

up7 = layers.Conv2DTranspose(256, (2, 2), strides=(2, 2), padding='same')(conv6)
merge7 = layers.concatenate([up7, conv3], axis=3)
conv7 = layers.Conv2D(256, (3, 3), padding='same')(merge7)
conv7 = layers.BatchNormalization()(conv7)
conv7 = layers.Activation('relu')(conv7)
conv7 = layers.Conv2D(256, (3, 3), padding='same')(conv7)
```

```
conv7 = layers.BatchNormalization()(conv7)
conv7 = layers.Activation('relu')(conv7)


up8 = layers.Conv2DTranspose(128, (2, 2), strides=(2, 2), padding='same')(conv7)
merge8 = layers.concatenate([up8, conv2], axis=3)
conv8 = layers.Conv2D(128, (3, 3), padding='same')(merge8)
conv8 = layers.BatchNormalization()(conv8)
conv8 = layers.Activation('relu')(conv8)
conv8 = layers.Conv2D(128, (3, 3), padding='same')(conv8)
conv8 = layers.BatchNormalization()(conv8)
conv8 = layers.Activation('relu')(conv8)

up9 = layers.Conv2DTranspose(64, (2, 2), strides=(2, 2), padding='same')(conv8)
merge9 = layers.concatenate([up9, conv1], axis=3)
conv9 = layers.Conv2D(64, (3, 3), padding='same')(merge9)
conv9 = layers.BatchNormalization()(conv9)
conv9 = layers.Activation('relu')(conv9)
conv9 = layers.Conv2D(64, (3, 3), padding='same')(conv9)
conv9 = layers.BatchNormalization()(conv9)
conv9 = layers.Activation('relu')(conv9)

outputs = layers.Conv2D(num_classes, (1, 1), activation='sigmoid')(conv9)

model = models.Model(inputs=inputs, outputs=outputs)
return model

# Create the Nested U-Net model
nested_unet_model = nested_unet()
nested_unet_model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
nested_unet_model.summary()
```

 Model: "functional"

Layer (type)	Output Shape	Param #	Connected to
input_layer (InputLayer)	(None, 256, 256, 3)	0	-
conv2d (Conv2D)	(None, 256, 256, 64)	1,792	input_layer[0][0]
batch_normalization (BatchNormalization)	(None, 256, 256, 64)	256	conv2d[0][0]
activation (Activation)	(None, 256, 256, 64)	0	batch_normalization[0...
conv2d_1 (Conv2D)	(None, 256, 256, 64)	36,928	activation[0][0]
batch_normalization_1 (BatchNormalization)	(None, 256, 256, 64)	256	conv2d_1[0][0]
activation_1 (Activation)	(None, 256, 256, 64)	0	batch_normalization_1...
max_pooling2d (MaxPooling2D)	(None, 128, 128, 64)	0	activation_1[0][0]
conv2d_2 (Conv2D)	(None, 128, 128, 128)	73,856	max_pooling2d[0][0]
batch_normalization_2 (BatchNormalization)	(None, 128, 128, 128)	512	conv2d_2[0][0]
activation_2 (Activation)	(None, 128, 128, 128)	0	batch_normalization_2...
conv2d_3 (Conv2D)	(None, 128, 128, 128)	147,584	activation_2[0][0]
batch_normalization_3 (BatchNormalization)	(None, 128, 128, 128)	512	conv2d_3[0][0]
activation_3 (Activation)	(None, 128, 128, 128)	0	batch_normalization_3...
max_pooling2d_1 (MaxPooling2D)	(None, 64, 64, 128)	0	activation_3[0][0]
conv2d_4 (Conv2D)	(None, 64, 64, 256)	295,168	max_pooling2d_1[0][0]
batch_normalization_4 (BatchNormalization)	(None, 64, 64, 256)	1,024	conv2d_4[0][0]
activation_4 (Activation)	(None, 64, 64, 256)	0	batch_normalization_4...
conv2d_5 (Conv2D)	(None, 64, 64, 256)	590,080	activation_4[0][0]
batch_normalization_5 (BatchNormalization)	(None, 64, 64, 256)	1,024	conv2d_5[0][0]
activation_5 (Activation)	(None, 64, 64, 256)	0	batch_normalization_5...
max_pooling2d_2 (MaxPooling2D)	(None, 32, 32, 256)	0	activation_5[0][0]
conv2d_6 (Conv2D)	(None, 32, 32, 512)	1,180,160	max_pooling2d_2[0][0]
batch_normalization_6 (BatchNormalization)	(None, 32, 32, 512)	2,048	conv2d_6[0][0]
activation_6 (Activation)	(None, 32, 32, 512)	0	batch_normalization_6...

conv2d_7 (Conv2D)	(None, 32, 32, 512)	2,359,808	activation_6[0][0]
batch_normalization_7 (BatchNormalization)	(None, 32, 32, 512)	2,048	conv2d_7[0][0]
activation_7 (Activation)	(None, 32, 32, 512)	0	batch_normalization_7...
max_pooling2d_3 (MaxPooling2D)	(None, 16, 16, 512)	0	activation_7[0][0]
conv2d_8 (Conv2D)	(None, 16, 16, 1024)	4,719,616	max_pooling2d_3[0][0]
batch_normalization_8 (BatchNormalization)	(None, 16, 16, 1024)	4,096	conv2d_8[0][0]
activation_8 (Activation)	(None, 16, 16, 1024)	0	batch_normalization_8...
conv2d_9 (Conv2D)	(None, 16, 16, 1024)	9,438,208	activation_8[0][0]
batch_normalization_9 (BatchNormalization)	(None, 16, 16, 1024)	4,096	conv2d_9[0][0]
activation_9 (Activation)	(None, 16, 16, 1024)	0	batch_normalization_9...
conv2d_transpose (Conv2DTranspose)	(None, 32, 32, 512)	2,097,664	activation_9[0][0]
concatenate (Concatenate)	(None, 32, 32, 1024)	0	conv2d_transpose[0][0]... activation_7[0][0]
conv2d_10 (Conv2D)	(None, 32, 32, 512)	4,719,104	concatenate[0][0]
batch_normalization_10 (BatchNormalization)	(None, 32, 32, 512)	2,048	conv2d_10[0][0]
activation_10 (Activation)	(None, 32, 32, 512)	0	batch_normalization_1...
conv2d_11 (Conv2D)	(None, 32, 32, 512)	2,359,808	activation_10[0][0]
batch_normalization_11 (BatchNormalization)	(None, 32, 32, 512)	2,048	conv2d_11[0][0]
activation_11 (Activation)	(None, 32, 32, 512)	0	batch_normalization_1...
conv2d_transpose_1 (Conv2DTranspose)	(None, 64, 64, 256)	524,544	activation_11[0][0]
concatenate_1 (Concatenate)	(None, 64, 64, 512)	0	conv2d_transpose_1[0]... activation_5[0][0]
conv2d_12 (Conv2D)	(None, 64, 64, 256)	1,179,904	concatenate_1[0][0]
batch_normalization_12 (BatchNormalization)	(None, 64, 64, 256)	1,024	conv2d_12[0][0]
activation_12 (Activation)	(None, 64, 64, 256)	0	batch_normalization_1...
conv2d_13 (Conv2D)	(None, 64, 64, 256)	590,080	activation_12[0][0]

batch_normalization_13 (BatchNormalization)	(None, 64, 64, 256)	1,024	conv2d_13[0][0]
activation_13 (Activation)	(None, 64, 64, 256)	0	batch_normalization_1...
conv2d_transpose_2 (Conv2DTranspose)	(None, 128, 128, 128)	131,200	activation_13[0][0]
concatenate_2 (Concatenate)	(None, 128, 128, 256)	0	conv2d_transpose_2[0]... activation_3[0][0]
conv2d_14 (Conv2D)	(None, 128, 128, 128)	295,040	concatenate_2[0][0]
batch_normalization_14 (BatchNormalization)	(None, 128, 128, 128)	512	conv2d_14[0][0]
activation_14 (Activation)	(None, 128, 128, 128)	0	batch_normalization_1...
conv2d_15 (Conv2D)	(None, 128, 128, 128)	147,584	activation_14[0][0]
batch_normalization_15 (BatchNormalization)	(None, 128, 128, 128)	512	conv2d_15[0][0]
activation_15 (Activation)	(None, 128, 128, 128)	0	batch_normalization_1...
conv2d_transpose_3 (Conv2DTranspose)	(None, 256, 256, 64)	32,832	activation_15[0][0]
concatenate_3 (Concatenate)	(None, 256, 256, 128)	0	conv2d_transpose_3[0]... activation_1[0][0]
conv2d_16 (Conv2D)	(None, 256, 256, 64)	73,792	concatenate_3[0][0]
batch_normalization_16 (BatchNormalization)	(None, 256, 256, 64)	256	conv2d_16[0][0]
activation_16 (Activation)	(None, 256, 256, 64)	0	batch_normalization_1...
conv2d_17 (Conv2D)	(None, 256, 256, 64)	36,928	activation_16[0][0]
batch_normalization_17 (BatchNormalization)	(None, 256, 256, 64)	256	conv2d_17[0][0]
activation_17 (Activation)	(None, 256, 256, 64)	0	batch_normalization_1...
conv2d_18 (Conv2D)	(None, 256, 256, 1)	65	activation_17[0][0]

Total params: 31,055,297 (118.47 MB)

Trainable params: 31,043,521 (118.42 MB)

Non-trainable params: 11,776 (46.05 KB)

```

def attention_gate(x, g, inter_channel):
    theta_x = layers.Conv2D(inter_channel, (1, 1), strides=(1, 1), padding='same')(x)
    theta_g = layers.Conv2D(inter_channel, (1, 1), strides=(1, 1), padding='same')(g)
    phi_g = layers.Conv2D(inter_channel, (1, 1), strides=(1, 1), padding='same')(g)

    f = layers.add([theta_x, phi_g])
    f = layers.Activation('relu')(f)

    psi_f = layers.Conv2D(1, (1, 1), padding='same')(f)
    psi_f = layers.Activation('sigmoid')(psi_f)

    return layers.multiply([x, psi_f])

def attention_unet(input_shape=(256, 256, 3), num_classes=1):
    inputs = layers.Input(shape=input_shape)

    # Encoder
    conv1 = layers.Conv2D(64, (3, 3), padding='same')(inputs)
    conv1 = layers.BatchNormalization()(conv1)
    conv1 = layers.Activation('relu')(conv1)
    conv1 = layers.Conv2D(64, (3, 3), padding='same')(conv1)
    conv1 = layers.BatchNormalization()(conv1)
    conv1 = layers.Activation('relu')(conv1)
    pool1 = layers.MaxPooling2D((2, 2))(conv1)

    conv2 = layers.Conv2D(128, (3, 3), padding='same')(pool1)
    conv2 = layers.BatchNormalization()(conv2)
    conv2 = layers.Activation('relu')(conv2)
    conv2 = layers.Conv2D(128, (3, 3), padding='same')(conv2)
    conv2 = layers.BatchNormalization()(conv2)
    conv2 = layers.Activation('relu')(conv2)
    pool2 = layers.MaxPooling2D((2, 2))(conv2)

    conv3 = layers.Conv2D(256, (3, 3), padding='same')(pool2)
    conv3 = layers.BatchNormalization()(conv3)
    conv3 = layers.Activation('relu')(conv3)
    conv3 = layers.Conv2D(256, (3, 3), padding='same')(conv3)
    conv3 = layers.BatchNormalization()(conv3)
    conv3 = layers.Activation('relu')(conv3)
    pool3 = layers.MaxPooling2D((2, 2))(conv3)

    conv4 = layers.Conv2D(512, (3, 3), padding='same')(pool3)
    conv4 = layers.BatchNormalization()(conv4)
    conv4 = layers.Activation('relu')(conv4)
    conv4 = layers.Conv2D(512, (3, 3), padding='same')(conv4)
    conv4 = layers.BatchNormalization()(conv4)
    conv4 = layers.Activation('relu')(conv4)
    pool4 = layers.MaxPooling2D((2, 2))(conv4)

    conv5 = layers.Conv2D(1024, (3, 3), padding='same')(pool4)
    conv5 = layers.BatchNormalization()(conv5)
    conv5 = layers.Activation('relu')(conv5)
    conv5 = layers.Conv2D(1024, (3, 3), padding='same')(conv5)
    conv5 = layers.BatchNormalization()(conv5)
    conv5 = layers.Activation('relu')(conv5)

```



```

# Decoder with Attention Gates
up6 = layers.Conv2DTranspose(512, (2, 2), strides=(2, 2), padding='same')(conv5)
att6 = attention_gate(conv4, up6, 512)
merge6 = layers.concatenate([up6, att6], axis=3)
conv6 = layers.Conv2D(512, (3, 3), padding='same')(merge6)
conv6 = layers.BatchNormalization()(conv6)
conv6 = layers.Activation('relu')(conv6)
conv6 = layers.Conv2D(512, (3, 3), padding='same')(conv6)
conv6 = layers.BatchNormalization()(conv6)
conv6 = layers.Activation('relu')(conv6)

up7 = layers.Conv2DTranspose(256, (2, 2), strides=(2, 2), padding='same')(conv6)
att7 = attention_gate(conv3, up7, 256)
merge7 = layers.concatenate([up7, att7], axis=3)
conv7 = layers.Conv2D(256, (3, 3), padding='same')(merge7)
conv7 = layers.BatchNormalization()(conv7)
conv7 = layers.Activation('relu')(conv7)
conv7 = layers.Conv2D(256, (3, 3), padding='same')(conv7)
conv7 = layers.BatchNormalization()(conv7)
conv7 = layers.Activation('relu')(conv7)

up8 = layers.Conv2DTranspose(128, (2, 2), strides=(2, 2), padding='same')(conv7)
att8 = attention_gate(conv2, up8, 128)
merge8 = layers.concatenate([up8, att8], axis=3)
conv8 = layers.Conv2D(128, (3, 3), padding='same')(merge8)
conv8 = layers.BatchNormalization()(conv8)
conv8 = layers.Activation('relu')(conv8)
conv8 = layers.Conv2D(128, (3, 3), padding='same')(conv8)
conv8 = layers.BatchNormalization()(conv8)
conv8 = layers.Activation('relu')(conv8)


up9 = layers.Conv2DTranspose(64, (2, 2), strides=(2, 2), padding='same')(conv8)
att9 = attention_gate(conv1, up9, 64)
merge9 = layers.concatenate([up9, att9], axis=3)
conv9 = layers.Conv2D(64, (3, 3), padding='same')(merge9)
conv9 = layers.BatchNormalization()(conv9)
conv9 = layers.Activation('relu')(conv9)
conv9 = layers.Conv2D(64, (3, 3), padding='same')(conv9)
conv9 = layers.BatchNormalization()(conv9)
conv9 = layers.Activation('relu')(conv9)

outputs = layers.Conv2D(num_classes, (1, 1), activation='sigmoid')(conv9)

model = models.Model(inputs=inputs, outputs=outputs)
return model

# Create the Attention U-Net model
attention_unet_model = attention_unet()
attention_unet_model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
attention_unet_model.summary()

```

 Model: "functional_1"

Layer (type)	Output Shape	Param #	Connected to
input_layer_1 (InputLayer)	(None, 256, 256, 3)	0	-
conv2d_19 (Conv2D)	(None, 256, 256, 64)	1,792	input_layer_1[0][0]
batch_normalization_18 (BatchNormalization)	(None, 256, 256, 64)	256	conv2d_19[0][0]
activation_18 (Activation)	(None, 256, 256, 64)	0	batch_normalization_1...
conv2d_20 (Conv2D)	(None, 256, 256, 64)	36,928	activation_18[0][0]
batch_normalization_19 (BatchNormalization)	(None, 256, 256, 64)	256	conv2d_20[0][0]
activation_19 (Activation)	(None, 256, 256, 64)	0	batch_normalization_1...
max_pooling2d_4 (MaxPooling2D)	(None, 128, 128, 64)	0	activation_19[0][0]
conv2d_21 (Conv2D)	(None, 128, 128, 128)	73,856	max_pooling2d_4[0][0]
batch_normalization_20 (BatchNormalization)	(None, 128, 128, 128)	512	conv2d_21[0][0]
activation_20 (Activation)	(None, 128, 128, 128)	0	batch_normalization_2...
conv2d_22 (Conv2D)	(None, 128, 128, 128)	147,584	activation_20[0][0]
batch_normalization_21 (BatchNormalization)	(None, 128, 128, 128)	512	conv2d_22[0][0]
activation_21 (Activation)	(None, 128, 128, 128)	0	batch_normalization_2...
max_pooling2d_5 (MaxPooling2D)	(None, 64, 64, 128)	0	activation_21[0][0]
conv2d_23 (Conv2D)	(None, 64, 64, 256)	295,168	max_pooling2d_5[0][0]
batch_normalization_22 (BatchNormalization)	(None, 64, 64, 256)	1,024	conv2d_23[0][0]
activation_22 (Activation)	(None, 64, 64, 256)	0	batch_normalization_2...
conv2d_24 (Conv2D)	(None, 64, 64, 256)	590,080	activation_22[0][0]
batch_normalization_23 (BatchNormalization)	(None, 64, 64, 256)	1,024	conv2d_24[0][0]
activation_23 (Activation)	(None, 64, 64, 256)	0	batch_normalization_2...
max_pooling2d_6 (MaxPooling2D)	(None, 32, 32, 256)	0	activation_23[0][0]

conv2d_25 (Conv2D)	(None, 32, 32, 512)	1,180,160	max_pooling2d_6[0][0]
batch_normalization_24 (BatchNormalization)	(None, 32, 32, 512)	2,048	conv2d_25[0][0]
activation_24 (Activation)	(None, 32, 32, 512)	0	batch_normalization_2...
conv2d_26 (Conv2D)	(None, 32, 32, 512)	2,359,808	activation_24[0][0]
batch_normalization_25 (BatchNormalization)	(None, 32, 32, 512)	2,048	conv2d_26[0][0]
activation_25 (Activation)	(None, 32, 32, 512)	0	batch_normalization_2...
max_pooling2d_7 (MaxPooling2D)	(None, 16, 16, 512)	0	activation_25[0][0]
conv2d_27 (Conv2D)	(None, 16, 16, 1024)	4,719,616	max_pooling2d_7[0][0]
batch_normalization_26 (BatchNormalization)	(None, 16, 16, 1024)	4,096	conv2d_27[0][0]
activation_26 (Activation)	(None, 16, 16, 1024)	0	batch_normalization_2...
conv2d_28 (Conv2D)	(None, 16, 16, 1024)	9,438,208	activation_26[0][0]
batch_normalization_27 (BatchNormalization)	(None, 16, 16, 1024)	4,096	conv2d_28[0][0]
activation_27 (Activation)	(None, 16, 16, 1024)	0	batch_normalization_2...
conv2d_transpose_4 (Conv2DTranspose)	(None, 32, 32, 512)	2,097,664	activation_27[0][0]
conv2d_29 (Conv2D)	(None, 32, 32, 512)	262,656	activation_25[0][0]
conv2d_31 (Conv2D)	(None, 32, 32, 512)	262,656	conv2d_transpose_4[0]...
add (Add)	(None, 32, 32, 512)	0	conv2d_29[0][0], conv2d_31[0][0]
activation_28 (Activation)	(None, 32, 32, 512)	0	add[0][0]
conv2d_32 (Conv2D)	(None, 32, 32, 1)	513	activation_28[0][0]
activation_29 (Activation)	(None, 32, 32, 1)	0	conv2d_32[0][0]
multiply (Multiply)	(None, 32, 32, 512)	0	activation_25[0][0], activation_29[0][0]
concatenate_4 (Concatenate)	(None, 32, 32, 1024)	0	conv2d_transpose_4[0]... multiply[0][0]
conv2d_33 (Conv2D)	(None, 32, 32, 512)	4,719,104	concatenate_4[0][0]
batch_normalization_28 (BatchNormalization)	(None, 32, 32, 512)	2,048	conv2d_33[0][0]

(BatchNormalization)			
activation_30 (Activation)	(None, 32, 32, 512)	0	batch_normalization_2...
conv2d_34 (Conv2D)	(None, 32, 32, 512)	2,359,808	activation_30[0][0]
batch_normalization_29 (BatchNormalization)	(None, 32, 32, 512)	2,048	conv2d_34[0][0]
activation_31 (Activation)	(None, 32, 32, 512)	0	batch_normalization_2...
conv2d_transpose_5 (Conv2DTranspose)	(None, 64, 64, 256)	524,544	activation_31[0][0]
conv2d_35 (Conv2D)	(None, 64, 64, 256)	65,792	activation_23[0][0]
conv2d_37 (Conv2D)	(None, 64, 64, 256)	65,792	conv2d_transpose_5[0]...
add_1 (Add)	(None, 64, 64, 256)	0	conv2d_35[0][0], conv2d_37[0][0]
activation_32 (Activation)	(None, 64, 64, 256)	0	add_1[0][0]
conv2d_38 (Conv2D)	(None, 64, 64, 1)	257	activation_32[0][0]
activation_33 (Activation)	(None, 64, 64, 1)	0	conv2d_38[0][0]
multiply_1 (Multiply)	(None, 64, 64, 256)	0	activation_23[0][0], activation_33[0][0]
concatenate_5 (Concatenate)	(None, 64, 64, 512)	0	conv2d_transpose_5[0]... multiply_1[0][0]
conv2d_39 (Conv2D)	(None, 64, 64, 256)	1,179,904	concatenate_5[0][0]
batch_normalization_30 (BatchNormalization)	(None, 64, 64, 256)	1,024	conv2d_39[0][0]
activation_34 (Activation)	(None, 64, 64, 256)	0	batch_normalization_3...
conv2d_40 (Conv2D)	(None, 64, 64, 256)	590,080	activation_34[0][0]
batch_normalization_31 (BatchNormalization)	(None, 64, 64, 256)	1,024	conv2d_40[0][0]
activation_35 (Activation)	(None, 64, 64, 256)	0	batch_normalization_3...
conv2d_transpose_6 (Conv2DTranspose)	(None, 128, 128, 128)	131,200	activation_35[0][0]
conv2d_41 (Conv2D)	(None, 128, 128, 128)	16,512	activation_21[0][0]
conv2d_43 (Conv2D)	(None, 128, 128, 128)	16,512	conv2d_transpose_6[0]...
add_2 (Add)	(None, 128, 128, 128)	0	conv2d_41[0][0], conv2d_43[0][0]
activation_36	(None, 128, 128, 128)	0	add_2[0][0]

(Activation)			
conv2d_44 (Conv2D)	(None, 128, 128, 1)	129	activation_36[0][0]
activation_37 (Activation)	(None, 128, 128, 1)	0	conv2d_44[0][0]
multiply_2 (Multiply)	(None, 128, 128, 128)	0	activation_21[0][0], activation_37[0][0]
concatenate_6 (Concatenate)	(None, 128, 128, 256)	0	conv2d_transpose_6[0]... multiply_2[0][0]
conv2d_45 (Conv2D)	(None, 128, 128, 128)	295,040	concatenate_6[0][0]
batch_normalization_32 (BatchNormalization)	(None, 128, 128, 128)	512	conv2d_45[0][0]
activation_38 (Activation)	(None, 128, 128, 128)	0	batch_normalization_3...
conv2d_46 (Conv2D)	(None, 128, 128, 128)	147,584	activation_38[0][0]
batch_normalization_33 (BatchNormalization)	(None, 128, 128, 128)	512	conv2d_46[0][0]
activation_39 (Activation)	(None, 128, 128, 128)	0	batch_normalization_3...
conv2d_transpose_7 (Conv2DTranspose)	(None, 256, 256, 64)	32,832	activation_39[0][0]
conv2d_47 (Conv2D)	(None, 256, 256, 64)	4,160	activation_19[0][0]
conv2d_49 (Conv2D)	(None, 256, 256, 64)	4,160	conv2d_transpose_7[0]...
add_3 (Add)	(None, 256, 256, 64)	0	conv2d_47[0][0], conv2d_49[0][0]
activation_40 (Activation)	(None, 256, 256, 64)	0	add_3[0][0]
conv2d_50 (Conv2D)	(None, 256, 256, 1)	65	activation_40[0][0]
activation_41 (Activation)	(None, 256, 256, 1)	0	conv2d_50[0][0]
multiply_3 (Multiply)	(None, 256, 256, 64)	0	activation_19[0][0], activation_41[0][0]
concatenate_7 (Concatenate)	(None, 256, 256, 128)	0	conv2d_transpose_7[0]... multiply_3[0][0]
conv2d_51 (Conv2D)	(None, 256, 256, 64)	73,792	concatenate_7[0][0]
batch_normalization_34 (BatchNormalization)	(None, 256, 256, 64)	256	conv2d_51[0][0]
activation_42 (Activation)	(None, 256, 256, 64)	0	batch_normalization_3...
conv2d_52 (Conv2D)	(None, 256, 256, 64)	36,928	activation_42[0][0]

batch_normalization_35 (BatchNormalization)	(None, 256, 256, 64)	256	conv2d_52[0][0]
activation_43 (Activation)	(None, 256, 256, 64)	0	batch_normalization_3...
conv2d_53 (Conv2D)	(None, 256, 256, 1)	65	activation_43[0][0]

Total params: 31,754,501 (121.13 MB)

Trainable params: 31,742,725 (121.09 MB)

Assuming you have already split your dataset into train and validation sets

```
train_generator = DataGenerator(train_image_paths, train_mask_paths, batch_size=16, img_size=(256, 256), shuffle=True)
```

```
val_generator = DataGenerator(val_image_paths, val_mask_paths, batch_size=16, img_size=(256, 256), shuffle=False)
```

Set the number of epochs

```
EPOCHS = 50 # Adjust this as needed
```

Train Nested U-Net

```
history_nested_unet = nested_unet_model.fit(train_generator, epochs=EPOCHS, validation_data=val_generator)
```

Train Attention U-Net

```
history_attention_unet = attention_unet_model.fit(train_generator, epochs=EPOCHS, validation_data=val_generator)
```

*** Epoch 1/50

```
/usr/local/lib/python3.10/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset` class should call `super().__init__(**kwargs)` in its constructor
self._warn_if_super_not_called()
```