```
In []: | wget --header 'Host: storage.googleapis.com' --user-agent 'Mozilla/5.0 (X11; Ubuntu; Linux x86 64; rv:87.0) Gecko/20100
          # !pip install q keras==2.4.1
          # !pip install segmentation_models
          # !pip install tensorflow io
          !unzip '/content/ultrasound-nerve-segmentation.zip'
In [6]: import os
         import re
         import numpy as np
         import pandas as pd
          import matplotlib.pyplot as plt
          import re
          import cv2
         from PIL import Image
         from skimage.transform import resize
         from sklearn.model_selection import train_test_split, KFold
         import tensorflow io as tfio
          import keras
          import tensorflow as tf
          from keras import backend as K
         \textbf{from} \text{ keras.preprocessing.image} \text{ } \textbf{import} \text{ } \textbf{ImageDataGenerator}
         from keras.callbacks import EarlyStopping, ModelCheckpoint
         plt.style.use("ggplot")
          %matplotlib inline
          from skimage.measure import compare_ssim as ssim
          from tqdm import tqdm_notebook, tnrange
         \textbf{from} \text{ itertools } \textbf{import} \text{ chain}
         from skimage.io import imread, imshow, concatenate_images
         from skimage.transform import resize
         from skimage.morphology import label
         from sklearn.model selection import train test split
          from keras.layers.core import Lambda, RepeatVector, Reshape
         from keras.preprocessing.image import ImageDataGenerator, array to img, img to array, load img
         from PIL import Image,ImageFilter
         K.set_image_data_format('channels_last')
```

Data preprocessing

7704 train/13_1.tif

```
In [7]: #creating train and test dataframes of images
    file_path=pd.DataFrame({'image_path':os.listdir('/content/train')})
    file_path_test=pd.DataFrame({'image_path':os.listdir('/content/test')})
           #getting subject and image name from train data
 In [8]:
            subject=file_path.image_path.apply(lambda x:re.split('_',x)[0])
           img_name=file_path_test.image_path.apply(lambda x:x.split('.')[0])
 In [9]:
           \label{lem:grb=file_path.image_path.apply(lambda x:re.search('_[0-9]+',x).group())} \\
           image_name=grb.apply(lambda x:re.split('_',x)[1])
           #adding two columns image name and subject name
In [10]:
           file_path['image_name']=image_name
file_path['subject_name']=subject
            file path test['image name']=img name
In [11]: #adding file location to file name.
            #separating mask images from train images
           msk=[i for i in range(len(file_path.image_path)) if 'mask' in file_path.image_path[i]]
img=[i for i in range(len(file_path.image_path)) if 'mask' not in file_path.image_path[i]]
           msk=file_path.loc[msk]
           img=file_path.loc[img]
In [13]: #aligning both to be concatenated later
           msk=msk.sort_values(by=['image_name','subject_name'])
img=img.sort_values(by=['image_name','subject_name'])
In [14]: #adding mask image path to the df
           img['mask path']=msk.image path.values
In [15]: print('Number of unique subjects',len(img.subject_name.value_counts()))
           print('Number of images per subject',len(img.image name.value counts()))
          Number of unique subjects 47
Number of images per subject 120
In [16]: img.head()
                 image_path image_name subject_name
                                                             mask_path
          10088 train/1 1.tif
                                       1
                                                    1 train/1 1 mask.tif
           6998 train/10_1.tif
                                                   10 train/10_1_mask.tif
            2298 train/11_1.tif
                                       1
                                                    11 train/11_1_mask.tif
           10602 train/12_1.tif
                                       1
                                                  12 train/12_1_mask.tif
```

13 train/13_1_mask.tif

```
In [17]: #visualising a single ultrasound image and its mask
           img_arr=np.array(Image.open(img.iloc[5].image_path))
           image1 mask = np.array(Image.open(img.iloc[5].mask path))
           fig, ax = plt.subplots(1, 3, figsize = (16, 12))
          ax[0].imshow(img_arr, cmap = 'gray')
          ax[0].set_title('Original')
          ax[1].imshow(image1_mask, cmap = 'gray')
          ax[1].set title('Mask')
          ax[2].imshow(img_arr, cmap = 'gray', interpolation = 'none')
          ax[2].imshow(image1 mask, interpolation = 'none', alpha = 0.7)
          ax[2].set_title('Mask overlay')
          plt.show()
                           Original
                                                                       Mask
                                                                                                             Mask overlay
           50
                                                     50
                                                                                                50
          100
                                                    100
                                                                                               100
          150
                                                    150
                                                                                               150
          200
                                                    200
                                                                                               200
          250
                                                    250
                                                                                               250
          300
                                                    300
                                                                                               300
          350
                                                    350
                                                                                               350
          400
                                                    400
                                                                                               400
                                                                                                       100
                                                                                                             200
                                                                                                                    300
                                                                                                                          400
In [19]: #creating an array that will store all images in 12*12 resized format
           X1 = np.zeros((len(img), 12, 12), dtype=np.float32)
           for i in range(len(img)):
             img1=cv2.imread(img.image_path.iloc[i],cv2.IMREAD_GRAYSCALE)
            img1=resize(img1, (12, 12), mode = 'constant', preserve_range = True)
            X1[i]=imq1
 In [ ]: #finding similarity between resized images and storing it in a dictionary
          from tgdm import tgdm
          sim=dict()
          for i in tqdm(range(len(X1))):
            for j in range(len(X1)):
                if i!=j:
                  if tuple(sorted((i,j))) not in sim.keys():
    simil=ssim(X1[i],X1[j])
                     if abs(simil)>0.95:
                       sim[tuple(sorted((i,j)))]=simil
In [29]: from tqdm import tqdm
          #storing conflicting images' keys in a list if their similarity is >0.99 in their resized form
           conf=[]
          for i in tqdm(sim.keys()):
   if sim[i]>0.999:
              msk1=cv2.imread(img.mask path.iloc[i[0]],cv2.IMREAD GRAYSCALE)
              msk2=cv2.imread(img.mask_path.iloc[i[1]],cv2.IMREAD_GRAYSCALE)
               if msk1.any()!=msk2.any():
                conf.append(i)
         100%| 18788/18788 [00:03<00:00, 5124.27it/s]
In [31]: #plotting few of the conflicting image and mask pairs
          cnt=0
           for i in conf:
            fig,ax=plt.subplots(1,4,figsize=(14,7))
            cnt+=1
            img1=cv2.imread(img.image path.iloc[i[0]],cv2.IMREAD GRAYSCALE)
            ax[0].imshow(img1,cmap='gray')
            ax[0].set_title('Image 1')
            img2=cv2.imread(img.image_path.iloc[i[1]],cv2.IMREAD_GRAYSCALE)
            ax[1].imshow(img2,cmap='gray')
            ax[1].set_title('Image 2')
msk1=cv2.imread(img.mask_path.iloc[i[0]],cv2.IMREAD_GRAYSCALE)
            ax[2].imshow(msk1,cmap='gray')
            ax[2].set title('Mask 1')
            msk2=cv2.imread(img.mask_path.iloc[i[1]],cv2.IMREAD_GRAYSCALE)
             ax[3].imshow(msk2,cmap='gray')
            ax[3].set_title('Mask 2')
            if cnt==5:
              break
                     Image 1
                                                 Image 2
                                                                             Mask 1
                                                                                                         Mask 2
                                                                   0
                                                                                             100
          100
                                                                  100
          200
                                                                  200
                                                                                             200
          300
                                                                  300
                                                                                             300
```

400

400

200

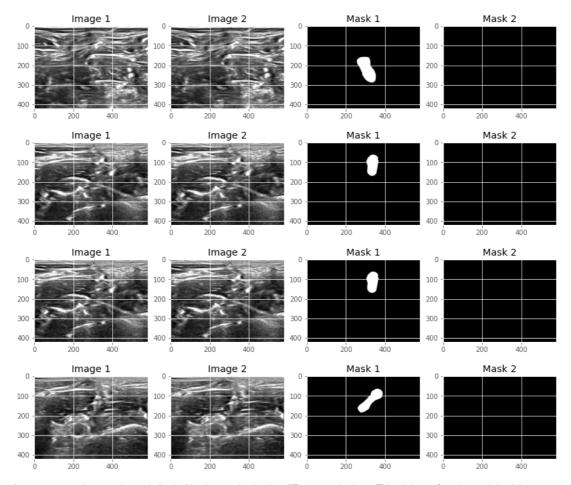
400

400

200

400

400



As you can see above we have similar looking images but having different masks them. This might confuse the model training.