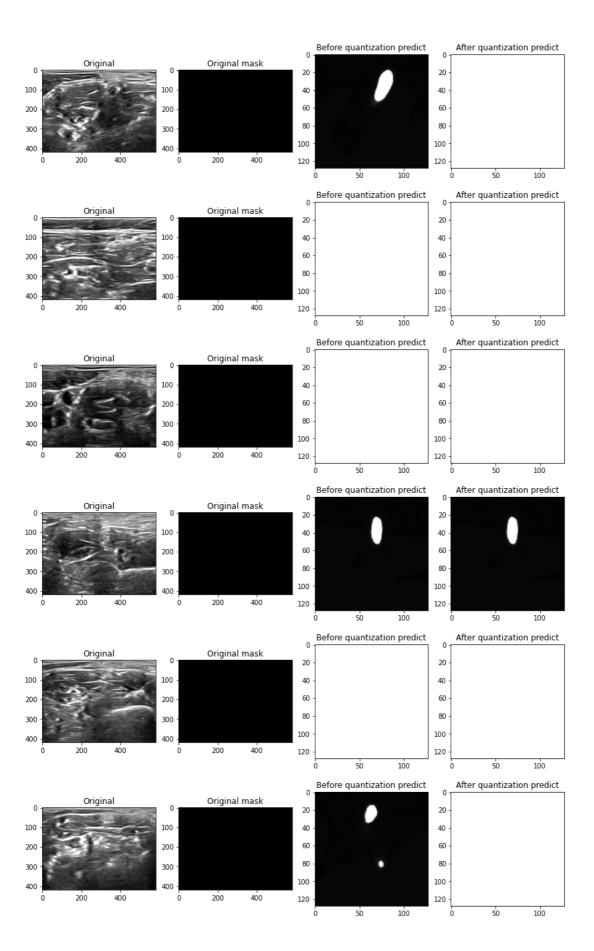
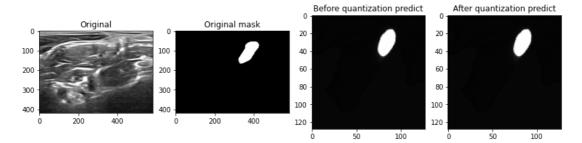
```
In []: !wget --header 'Host: storage.googleapis.com' --user-agent 'Mozilla/5.0 (X11; Ubuntu; Linux x86_64; rv:87.0) Gecko/20100
           !pip install q keras==2.4.1
           !pip install segmentation_models
           !pip install tensorflow io
           !pip install tensorflow_model_optimization
           !unzip '/content/ultrasound-nerve-segmentation.zip'
In [13]: #Loading dependancies
           import os
           import re
           import random
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as plt
           import re
           import seaborn as sns
           import cv2
           from PIL import Image
           from sklearn.model_selection import train_test_split, KFold
           import tensorflow_io as tfio
           import keras
           import tensorflow as tf
            # tf.compat.v1.enable eager execution()
           from tensorflow import keras
           from tensorflow.keras.layers import *
           \textbf{from} \ \texttt{tensorflow.keras.preprocessing} \ \textbf{import} \ \texttt{image}
           from tensorflow.keras.models import Model, load_model
           \textbf{from} \ \texttt{tensorflow.keras.layers} \ \textbf{import} \ \texttt{UpSampling2D}
           from tensorflow.keras.layers import MaxPooling2D, GlobalAveragePooling2D
from tensorflow.keras.layers import concatenate,Dropout
           from tensorflow.keras.layers import Multiply, MaxPooling2D, GlobalMaxPooling2D
           from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint
           from tensorflow.keras import backend as K
           from tensorflow.keras.layers import Input, Add, Dense, Activation, ZeroPadding2D
           from tensorflow.keras.layers import BatchNormalization, Flatten, Conv2D, AveragePooling2D
           from tensorflow.keras.models import Model, load_model,save_model from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint, ReduceLROnPlateau
           from tensorflow.keras.utils import plot_model
           from tensorflow.keras.initializers import glorot_uniform
           from tensorflow.keras.optimizers import Adam
           \textbf{from} \ \texttt{tensorflow.keras.utils} \ \textbf{import} \ \texttt{plot}\_\texttt{model}
           from keras.callbacks import ModelCheckpoint
           import tensorflow
           import keras
           from sklearn.metrics import confusion_matrix
           import cv2
           import imgaug.augmenters as iaa
           os.environ['TF_FORCE_GPU_ALLOW_GROWTH'] = 'true'
           import segmentation models as sm
           \begin{tabular}{ll} \hline \textbf{from} & \texttt{segmentation\_models.metrics} & \textbf{import} \\ \hline \end{tabular} & \texttt{iou} & \texttt{score} \\ \hline \end{tabular}
           from segmentation models import Unet
           focal_loss = sm.losses.cce_dice_loss
           import random
           import segmentation_models as sm
           from segmentation_models import Unet
           import tempfile
              sm.set framework('tf.keras')
           tf.keras.backend.set image data format('channels last')
 In [ ]: def classifier_generator(images):
                '''Construct a data generator using tf.Dataset to load only images'''
               image_string=tf.io.read_file(images)
               image = tfio.experimental.image.decode_tiff(image_string)
               image = tf.image.convert_image_dtype(image, tf.float32)
                image = tf.image.resize(image, [128, 128])
               return image
 In [ ]: def train_generator(images, masks):
                 ''Construct a data generator using tf.Dataset to load image+masks'''
               image string=tf.io.read file(images)
               image = tfio.experimental.image.decode_tiff(image string)
               image = tf.image.convert_image_dtype(image, tf.float32)
               image = tf.image.resize(image, [128, 128])
               mask=tf.io.read file(masks)
               mask = tfio.experimental.image.decode_tiff(mask)
               mask = tf.image.convert_image_dtype(mask, tf.float32)
               mask = tf.image.resize(mask, [128, 128],method='nearest')
               return image, mask
 In [ ]: def iou_coef(y_true, y_pred, smooth=1):
                 This function calculates iou
             intersection = K.sum(K.abs(y_true * y_pred), axis=[1,2])
             union = K.sum(y_true,[1,2])+K.sum(y_pred,[1,2])-intersection
             iou = K.mean((intersection + smooth) / (union + smooth))
             return iou.numpy()
In [14]: #loading image df
           img=pd.read csv('/content/drive/MyDrive/dup_rem.csv')
           df=pd.read csv('/content/train masks.csv'
           img.drop('Unnamed: 0',axis=1,inplace=True)
           img.image_name=img.image_name.astype(int)
           img.subject_name=img.subject_name.astype(int)
```

```
img.columns=['image_path','img','subject','mask_path']
           new_df=pd.merge(img,df,on=['img','subject'])
           new_df.pixels.fillna(0,inplace=True)
          val=[0 if i==0 else 1 for i in new_df.pixels]
new_df['mask_pres']=val
          pos_df=new_df[new_df.mask_pres==1]
In [ ]: from google.colab import drive
          drive.mount('/content/drive')
          Mounted at /content/drive
In [15]: X_train, X_valid, y_train, y_valid = train_test_split(new_df.image_path, new_df.mask_path, test_size=0.2, random_state=4
         Loading model weights
In [16]: #loading the classifier model
          model_clf = keras.models.load_model('/content/drive/MyDrive/classfier_nerve')
In [17]: #loading the segmentor model
          model=keras.models.load model('/content/drive/MyDrive/segmetnor',custom objects={'categorical crossentropy plus dice los
         Quantizing models
In [18]: #quantizing classifier model
          converter = tf.lite.TFLiteConverter.from_saved_model('/content/drive/MyDrive/classfier_nerve')
           converter.optimizations = [tf.lite.Optimize.DEFAULT]
           quant model clf = converter.convert()
           #quantizing segmentor model
          converter = tf.lite.TFLiteConverter.from saved model('/content/drive/MyDrive/segmetnor')
          converter.optimizations = [tf.lite.Optimize.DEFAULT]
          quant model = converter.convert()
         Saving quantized models as tflite file
In [19]: #saving the tflite file of classifier quantized model
           f = open("quant classifier.tflite", "wb")
           f.write(quant_model_clf)
           f.close()
          #saving the tflite file of segmentor quantized model
f = open("quant_segmentor.tflite", "wb")
           f.write(quant model)
           f.close()
In [24]:
          interpreter_clf = tf.lite.Interpreter(model_path="quant_classifier.tflite")
           interpreter_clf.allocate_tensors()
           interpreter_seg = tf.lite.Interpreter(model_path="quant_segmentor.tflite",num_threads=1)
          interpreter_seg.allocate_tensors()
In [31]: test_image = tf.expand_dims(x_tes, axis=0)
          test image=tf.cast(test image,tf.float32)
           input_index = interpreter_clf.get_input_details()[0]["index"]
           output index = interpreter clf.get output details()[0]["index"]
           # input_index = interpreter_seg.get_input_details()[0]["index"]
# output_index = interpreter_seg.get_output_details()[0]["index"]
           # interpreter_seg.set_tensor(input_index, test_image)
           # predictions = interpreter seg.get tensor(output index)
         Getting predictions using quantized classifier model
In [45]: #Getting the prediction on validation set ofafter quantized model
           scores_val=[]
          X_tr=np.zeros((len(X_valid),128,128,4),dtype=np.float32)
y_tr=np.zeros((len(y_valid),128,128,4),dtype=np.float32)
           pred_val=np.zeros((len(y_valid),128,128,4),dtype=np.float32)
           for i in range(len(X_valid)):
             image_cl=classifier_generator(X_valid.iloc[i])
             image_cl=tf.expand_dims(image_cl,0)
            image_cl=tf.cast(image_cl,tf.float32)
            interpreter_clf.set_tensor(input_index, image_cl)
interpreter_clf.invoke()
            pred clf = interpreter clf.get tensor(output index)
            if pred_clf>=0.4:
               X_tr[i],y_tr[i]=train_generator(X_valid.iloc[i],y_valid.iloc[i])
               pred_seg=model.predict(tf.expand_dims(X_tr[i],0))
               pred_val[i]=pred_seg
               true=tf.expand_dims(y_tr[i],0)
               score=iou coef(true, pred seg)
               scores val.append(score)
               X_tr[i],y_tr[i]=train_generator(X_valid.iloc[i],y_valid.iloc[i])
               pred_seg=np.zeros((1,128,128,4),dtype=np.float32)
               pred val[i]=pred seg
               true=tf.expand_dims(y_tr[i],0)
               score=iou coef(true, pred seg)
               scores_val.append(score)
```

Prediction

```
In [37]: #Getting the prediction on validation set of before quantized model
             scores_val_b=[]
            X_tr=np.zeros((len(X_valid),128,128,4),dtype=np.float32)
             y_tr=np.zeros((len(y_valid),128,128,4),dtype=np.float32)
            pred_val_b=np.zeros((len(y_valid),128,128,4),dtype=np.float32)
            for i in range(len(X_valid)):
   image_cl=classifier_generator(X_valid.iloc[i])
               image cl=tf.expand dims(image cl,0)
               pred_clf=model_clf.predict(image_cl)
               if pred_clf>=0.4:
                 \label{eq:continuous_problem} \textbf{X\_tr[i],y\_tr[i]=} \\ \textbf{train\_generator}(\textbf{X\_valid.iloc[i],y\_valid.iloc[i]})
                 pred_seg=model.predict(tf.expand_dims(X_tr[i],0))
pred_val_b[i]=pred_seg
true=tf.expand_dims(y_tr[i],0)
                  score=iou coef(true,pred seg)
                  scores_val_b.append(score)
               else:
                 \label{eq:continuous_problem} \textbf{X\_tr[i],y\_tr[i]=} \\ \textbf{train\_generator}(\textbf{X\_valid.iloc[i],y\_valid.iloc[i]})
                 pred seg=np.zeros((1,128,128,4),dtype=np.float32)
                 pred_val_b[i]=pred_seg
true=tf.expand_dims(y_tr[i],0)
                  score=iou_coef(true,pred_seg)
                  scores_val_b.append(score)
In [51]: print('Mean validation score before quantization= ',np.mean(scores_val_b))
            print('Mean validation score after quantization= ',np.mean(scores_val))
           Mean validation score before quantization= 0.62266594 Mean validation score after quantization= 0.60848292
In [76]:
            random_gen=[i for i in range(1080)]
            for j in range(10):
              i=random.choice(random gen)
               fig, ax=plt.subplots (1, 4, figsize=(14, 7))
               quer=classifier_generator(X_valid.iloc[i])
               ax[0].imshow(cv2.imread(X_valid.iloc[i],cv2.IMREAD_GRAYSCALE),cmap='gray')
               ax[0].set_title('Original')
               ax[1].imshow(cv2.imread(y_valid.iloc[i],cv2.IMREAD_GRAYSCALE),cmap='gray')
ax[1].set_title('Original mask')
               ax[2].imshow(pred val b[i])
               ax[2].set title('Before quantization predict')
               ax[3].imshow(pred val[i])
               ax[3].set_title('After quantization predict')
                                                                                   Before quantization predict
                                                                                                                     After quantization predict
                                                                                0
                                                                                                                 0
                                                        Original mask
                          Original
              0
                                               0
                                                                                                                 20
                                                                               20
                                                                                40
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            200
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            300
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            400
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                                                                              120
                                                                                                                120
                                                                                   Before quantization predict
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                                                        Original mask
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                                                                                   Before quantization predict
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                                                                                                                                        100
```





```
In [ ]: # https://www.geeksforgeeks.org/how-to-get-size-of-folder-using-python/
          size = 0
          # assign folder path
          Folderpath = '/content/drive/MyDrive/classfier_nerve'
           # get size
          for path, dirs, files in os.walk(Folderpath):
    for f in files:
                  fp = os.path.join(path, f)
size += os.path.getsize(fp)
          # display size
          print("Pre quantization model size= " + str(size/(1024*1024)),'MB')
         Pre quantization model size= 248.4090223312378 MB
In [ ]: # https://www.geeksforgeeks.org/how-to-get-size-of-folder-using-python/
          size = 0
          # assign folder path
          Folderpath = '/content/drive/MyDrive/segmentor_nerve'
          # get size
          for path, dirs, files in os.walk(Folderpath):
    for f in files:
                  fp = os.path.join(path, f)
                  size += os.path.getsize(fp)
          # display size
          print("Pre quantization model size= " + str(size/(1024*1024)),'MB')
         Pre quantization model size= 127.00345516204834 MB
In []: print('Post quantization model size =',os.stat('quant_classifier.tflite').st_size/(1024*1024),'MB')
         Post quantization model size = 59.54229736328125 MB
In []: print('Post quantization model size =',os.stat('quant_segmentor.tflite').st_size/(1024*1024),'MB')
         Post quantization model size = 60.416900634765625 MB
In [50]: print('Reduction in classifier model size = {0:.1f} x'.format(248.40902/59.54229))
          print('Reduction in segmentation model size = {0:.1f} x'.format(127.003455/60.41690))
         Reduction in classifier model size = 4.2 \times 10^{-2}
```

Reduction in validation iou score = 0.014

Reduction in segmentation model size = $2.1 \times$

In [73]: print('Reduction in validation iou score = {0:.3f}'.format(0.62266594-0.60848292))