```
In [1]: from google.colab import drive
    drive.mount('/gdrive')
    %cd /gdrive
```

Go to this URL in a browser: https://accounts.google.com/o/oauth2/auth?client_id=947318989803-6bn6qk8qdgf4n4g 3pfee6491hc0brc4i.apps.googleusercontent.com&redirect_uri=urn%3aietf%3awg%3aoauth%3a2.0%3aoob&response_type=c ode&scope=email%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdocs.test%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly%20https%3a%2f%2fwww.googleapis.com%2fauth%2fpeopleapi.readonly

Enter your authorization code:
.....
Mounted at /gdrive
/gdrive

In [2]: pip install tensorflow-gpu==2.0.0

Collecting tensorflow-gpu==2.0.0

Downloading https://files.pythonhosted.org/packages/25/44/47f0722aea081697143fbcf5d2aa60d1aee4aaacb5869aee2b568974777b/tensorflow_gpu-2.0.0-cp36-cp36m-manylinux2010_x86_64.whl (380.8MB)

| 380.8MB 37kB/s

Collecting tensorflow-estimator<2.1.0,>=2.0.0

Downloading https://files.pythonhosted.org/packages/fc/08/8b927337b7019c374719145d1dceba21a8bb909b93b1ad6f8 fb7d22c1ca1/tensorflow_estimator-2.0.1-py2.py3-none-any.whl (449kB)

450kB 45.2MB/s

Requirement already satisfied: google-pasta>=0.1.6 in /usr/local/lib/python3.6/dist-packages (from tensorflow -gpu==2.0.0) (0.2.0)

Requirement already satisfied: six>=1.10.0 in /usr/local/lib/python3.6/dist-packages (from tensorflow-gpu==2.0.0) (1.12.0)

Requirement already satisfied: wrapt>=1.11.1 in /usr/local/lib/python3.6/dist-packages (from tensorflow-gpu== 2.0.0) (1.12.1)

Requirement already satisfied: numpy<2.0,>=1.16.0 in /usr/local/lib/python3.6/dist-packages (from tensorflow-gpu==2.0.0) (1.18.3)

Requirement already satisfied: termcolor>=1.1.0 in /usr/local/lib/python3.6/dist-packages (from tensorflow-gp u=2.0.0) (1.1.0)

Requirement already satisfied: wheel>=0.26 in /usr/local/lib/python3.6/dist-packages (from tensorflow-gpu==2.0.0) (0.34.2)

Collecting tensorboard<2.1.0,>=2.0.0

Downloading https://files.pythonhosted.org/packages/76/54/99b9d5d52d5cb732f099baaaf7740403e83fe6b0cedde940fabd2b13d75a/tensorboard-2.0.2-py3-none-any.whl (3.8MB)

3.8MB 41.7MB/s

Requirement already satisfied: keras-applications>=1.0.8 in /usr/local/lib/python3.6/dist-packages (from tens orflow-gpu==2.0.0) (1.0.8)

Requirement already satisfied: protobuf>=3.6.1 in /usr/local/lib/python3.6/dist-packages (from tensorflow-gpu ==2.0.0) (3.10.0)

Requirement already satisfied: grpcio>=1.8.6 in /usr/local/lib/python3.6/dist-packages (from tensorflow-gpu== 2.0.0) (1.28.1)

Requirement already satisfied: opt-einsum>=2.3.2 in /usr/local/lib/python3.6/dist-packages (from tensorflow-g pu==2.0.0) (3.2.1)

Collecting gast==0.2.2

Downloading https://files.pythonhosted.org/packages/4e/35/11749bf99b2d4e3cceb4d55ca22590b0d7c2c62b9de38ac4a4a7f4687421/gast-0.2.2.tar.gz

Requirement already satisfied: absl-py>=0.7.0 in /usr/local/lib/python3.6/dist-packages (from tensorflow-gpu= =2.0.0) (0.9.0)

Requirement already satisfied: astor>=0.6.0 in /usr/local/lib/python3.6/dist-packages (from tensorflow-gpu== 2.0.0) (0.8.1)

Requirement already satisfied: keras-preprocessing>=1.0.5 in /usr/local/lib/python3.6/dist-packages (from ten sorflow-gpu==2.0.0) (1.1.0)

Requirement already satisfied: setuptools>=41.0.0 in /usr/local/lib/python3.6/dist-packages (from tensorboard <2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (46.1.3)

Requirement already satisfied: werkzeug>=0.11.15 in /usr/local/lib/python3.6/dist-packages (from tensorboard< 2.1.0, >= 2.0.0 -> tensorflow-gpu== 2.0.0) (1.0.1)Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in /usr/local/lib/python3.6/dist-packages (fr om tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (0.4.1) Requirement already satisfied: requests<3,>=2.21.0 in /usr/local/lib/python3.6/dist-packages (from tensorboar d<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (2.21.0) Requirement already satisfied: google-auth<2,>=1.6.3 in /usr/local/lib/python3.6/dist-packages (from tensorbo ard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (1.7.2) Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.6/dist-packages (from tensorboard<2. 1.0, >= 2.0.0 - tensorflow-gpu == 2.0.0) (3.2.1) Requirement already satisfied: h5py in /usr/local/lib/python3.6/dist-packages (from keras-applications>=1.0.8 ->tensorflow-gpu==2.0.0) (2.10.0) Requirement already satisfied: requests-oauthlib>=0.7.0 in /usr/local/lib/python3.6/dist-packages (from googl e-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (1.3.0) Requirement already satisfied: idna<2.9,>=2.5 in /usr/local/lib/python3.6/dist-packages (from requests<3,>=2. 21.0->tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (2.8) Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.6/dist-packages (from requests<3, >=2.21.0->tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (2020.4.5.1) Requirement already satisfied: urllib3<1.25,>=1.21.1 in /usr/local/lib/python3.6/dist-packages (from requests <3,>=2.21.0->tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (1.24.3) Requirement already satisfied: chardet<3.1.0,>=3.0.2 in /usr/local/lib/python3.6/dist-packages (from requests <3,>=2.21.0->tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (3.0.4) Requirement already satisfied: pyasn1-modules>=0.2.1 in /usr/local/lib/python3.6/dist-packages (from google-a uth<2,>=1.6.3->tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (0.2.8) Requirement already satisfied: rsa<4.1,>=3.1.4 in /usr/local/lib/python3.6/dist-packages (from google-auth<2, >=1.6.3->tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (4.0) Requirement already satisfied: cachetools<3.2,>=2.0.0 in /usr/local/lib/python3.6/dist-packages (from googleauth<2,>=1.6.3->tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (3.1.1) Requirement already satisfied: oauthlib>=3.0.0 in /usr/local/lib/python3.6/dist-packages (from requests-oauth lib>=0.7.0-ygoogle-auth-oauthlib<0.5,>=0.4.1-ytensorboard<2.1.0,>=2.0.0-ytensorflow-gpu==2.0.0) (3.1.0) Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in /usr/local/lib/python3.6/dist-packages (from pyasn1-mo dules>=0.2.1->google-auth<2,>=1.6.3->tensorboard<2.1.0,>=2.0.0->tensorflow-gpu==2.0.0) (0.4.8) Building wheels for collected packages: gast Building wheel for gast (setup.py) ... done Created wheel for gast: filename=gast-0.2.2-cp36-none-any.whl size=7540 sha256=acebd4a5577b451073b798626965

4533b2c226a536ad180e264d6bbd8134b8f4

Stored in directory: /root/.cache/pip/wheels/5c/2e/7e/a1d4d4fcebe6c381f378ce7743a3ced3699feb89bcfbdadadd Successfully built gast

ERROR: tensorflow 2.2.0rc3 has requirement gast==0.3.3, but you'll have gast 0.2.2 which is incompatible. ERROR: tensorflow 2.2.0rc3 has requirement tensorboard<2.3.0,>=2.2.0, but you'll have tensorboard 2.0.2 which is incompatible.

ERROR: tensorflow 2.2.0rc3 has requirement tensorflow-estimator<2.3.0,>=2.2.0rc0, but you'll have tensorflowestimator 2.0.1 which is incompatible.

```
ERROR: tensorflow-probability 0.10.0rc0 has requirement gast>=0.3.2, but you'll have gast 0.2.2 which is inco
        mpatible.
        Installing collected packages: tensorflow-estimator, tensorboard, gast, tensorflow-gpu
          Found existing installation: tensorflow-estimator 2.2.0
            Uninstalling tensorflow-estimator-2.2.0:
              Successfully uninstalled tensorflow-estimator-2.2.0
          Found existing installation: tensorboard 2.2.1
            Uninstalling tensorboard-2.2.1:
              Successfully uninstalled tensorboard-2.2.1
          Found existing installation: gast 0.3.3
            Uninstalling gast-0.3.3:
              Successfully uninstalled gast-0.3.3
        Successfully installed gast-0.2.2 tensorboard-2.0.2 tensorflow-estimator-2.0.1 tensorflow-gpu-2.0.0
In [3]: pip install keras==2.2.4
        Collecting keras==2.2.4
          Downloading https://files.pythonhosted.org/packages/5e/10/aa32dad071ce52b5502266b5c659451cfd6ffcbf14e6c8c4f
        16c0ff5aaab/Keras-2.2.4-py2.py3-none-any.whl (312kB)
                    317kB 3.5MB/s eta 0:00:01
        Requirement already satisfied: numpy>=1.9.1 in /usr/local/lib/python3.6/dist-packages (from keras==2.2.4) (1.
        18.3)
        Requirement already satisfied: pyyaml in /usr/local/lib/python3.6/dist-packages (from keras==2.2.4) (3.13)
        Requirement already satisfied: h5py in /usr/local/lib/python3.6/dist-packages (from keras==2.2.4) (2.10.0)
        Requirement already satisfied: scipy>=0.14 in /usr/local/lib/python3.6/dist-packages (from keras==2.2.4) (1.
        4.1)
        Requirement already satisfied: six>=1.9.0 in /usr/local/lib/python3.6/dist-packages (from keras==2.2.4) (1.1
        2.0)
        Requirement already satisfied: keras-preprocessing>=1.0.5 in /usr/local/lib/python3.6/dist-packages (from ker
        as==2.2.4) (1.1.0)
        Requirement already satisfied: keras-applications>=1.0.6 in /usr/local/lib/python3.6/dist-packages (from kera
        s==2.2.4) (1.0.8)
        Installing collected packages: keras
          Found existing installation: Keras 2.3.1
            Uninstalling Keras-2.3.1:
              Successfully uninstalled Keras-2.3.1
        Successfully installed keras-2.2.4
In [ ]:
```

```
In [4]: gpu info = !nvidia-smi
      gpu info = '\n'.join(gpu info)
      if gpu info.find('failed') >= 0:
        print('Select the Runtime → "Change runtime type" menu to enable a GPU accelerator, ')
        print('and then re-execute this cell.')
       else:
        print(gpu info)
       Fri Apr 24 19:34:45 2020
        NVIDIA-SMI 440.64.00 Driver Version: 418.67 CUDA Version: 10.1
        GPU Name Persistence-M Bus-Id Disp.A | Volatile Uncorr. ECC
        Fan Temp Perf Pwr:Usage/Cap | Memory-Usage | GPU-Util Compute M.
       0 Tesla P100-PCIE... Off | 00000000:04.0 Off |
                                     OMiB / 1628OMiB | 0% Default
                  P0 30W / 250W |
        N/A 42C
                                                           GPU Memory
        Processes:
         GPU
                PID Type Process name
                                                           Usage
       No running processes found
In [5]: from keras.preprocessing.image import ImageDataGenerator
      from keras.preprocessing.image import img to array
      from keras.preprocessing.image import load_img
      import numpy as np
      import argparse
       import cv2
       import os
       from imutils import paths
```

Using TensorFlow backend.

```
In [ ]: from tensorflow.keras.preprocessing.image import ImageDataGenerator
          from tensorflow.keras.applications import VGG16
          from tensorflow.keras.applications.resnet v2 import ResNet152V2
          from tensorflow.keras.layers import AveragePooling2D
          from tensorflow.keras.layers import Dropout
          from tensorflow.keras.layers import Flatten
          from tensorflow.keras.layers import Dense
          from tensorflow.keras.layers import Input
          from tensorflow.keras.models import Model
          from tensorflow.keras.optimizers import Adam
          from tensorflow.keras.utils import to categorical
          from sklearn.preprocessing import LabelBinarizer
          from sklearn.model selection import train test split
          from sklearn.metrics import classification report
          from sklearn.metrics import confusion matrix
          from imutils import paths
          import matplotlib.pvplot as plt
          import numpy as np
          import argparse
          import cv2
          import os
       ]: x train=np.load('My Drive/Colab Notebooks/215-FP Dataset/numpy5/x train.npy')
      1: v train=np.load('My Drive/Colab Notebooks/215-FP Dataset/numpy5/y train.npy')
In [103]:
Out[103]: array([[0.
                         , 0.00392],
                         , 0.00392],
                  [0.
                 [0.
                         , 0.00392],
                  . . . ,
                         , 0.00392],
                 [0.
                 [0.
                         , 0.00392],
                 [0.
                         , 0.00392]], dtype=float16)
  In [ ]:
```

```
In [ ]:
In [24]: y_train
Out[24]: array([[1., 0.],
                [1., 0.],
                [1., 0.],
                 . . . ,
                [1., 0.],
                [1., 0.],
                [1., 0.]])
In [ ]: | x valid=np.load('My Drive/Colab Notebooks/215-FP Dataset/numpy5/x valid.npy')
In [ ]: y_valid=np.load('My Drive/Colab Notebooks/215-FP_Dataset/numpy5/y_valid.npy')
In [ ]:
In [ ]:
In [27]: y_valid
Out[27]: array([[1., 0.],
                [1., 0.],
                [1., 0.],
                 ...,
                [1., 0.],
                [1., 0.],
                [1., 0.]])
In [ ]: y_train2=y_train
         y_valid2=y_valid
In [ ]: y_train2=np.argmax(y_train,axis=1)
         y valid2=np.argmax(y valid,axis=1)
```

```
In [30]: y_train2[:50]
0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0,
              1, 1, 0, 1, 1, 1])
In [31]: y valid2[:50]
Out[31]: array([0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0,
              0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
              0, 0, 1, 0, 1, 1])
In [34]: | summation1=0
        for i in range(len(x valid)):
            if y valid2[i]==0:
               summation1=summation1+1
        summation1
Out[34]: 488
In [35]: summation2=0
        for i in range(len(x train)):
            if y train2[i]==0:
               summation2=summation2+1
        summation2
Out[35]: 1951
In [36]: len(x train)
Out[36]: 6025
In [ ]: #y train=y train2
        y train=((y train)/255.).astype('float16')
In [ ]: #y valid=y valid2
        y valid=((y valid)/255.).astype('float16')
In [ ]:
```

In []:	
In []:	

```
In [44]: from future import print function
         from tensorflow.keras import backend as K
         from tensorflow.keras.layers import Layer
         from tensorflow.keras import activations
         from tensorflow.keras import utils
         from tensorflow.keras.models import Model
         from tensorflow.keras.layers import *
         from tensorflow.keras.preprocessing.image import ImageDataGenerator
         import numpy as np
         import keras
         from tensorflow.keras.callbacks import ModelCheckpoint
         from tensorflow.keras import optimizers
         K.set image data format('channels last')
         def squash(x, axis=-1):
             s squared norm = K.sum(K.square(x), axis, keepdims=True) + K.epsilon()
             scale = K.sqrt(s squared norm) / (1 + s squared norm)
             return scale * x
         def softmax(x, axis=-1):
             ex = K.exp(x - K.max(x, axis=axis, keepdims=True))
             return ex / K.sum(ex, axis=axis, keepdims=True)
         def margin loss(y true, y pred):
             lamb, margin = 0.5, 0.1
             return K.sum((y true * K.square(K.relu(1 - margin - y pred)) + lamb * (
                 1 - y true) * K.square(K.relu(y pred - margin))), axis=-1)
         class Capsule(Laver):
             def init (self,
                          num capsule,
                          dim capsule,
                          routings=3,
                          share weights=True,
```

```
activation='squash',
             **kwargs):
    super(Capsule, self). init (**kwargs)
    self.num capsule = num capsule
    self.dim capsule = dim capsule
    self.routings = routings
    self.share weights = share weights
    if activation == 'squash':
        self.activation = squash
    else:
        self.activation = activations.get(activation)
def get config(self):
    config = super().get config().copy()
    config.update({
    'num capsule': self.num capsule,
    'dim capsule' : self.dim capsule,
    'routings': self.routings,
    'share weight':self.share weights,
    })
   return config
def build(self, input shape):
    input dim capsule = input shape[-1]
   if self.share_weights:
        self.kernel = self.add_weight(
            name='capsule kernel',
            shape=(1, input dim capsule,
                   self.num capsule * self.dim capsule),
            initializer='glorot uniform',
            trainable=True)
    else:
        input num capsule = input shape[-2]
        self.kernel = self.add weight(
            name='capsule kernel',
            shape=(input num capsule, input dim capsule,
                   self.num capsule * self.dim capsule),
            initializer='glorot uniform',
            trainable=True)
```

```
def call(self, inputs):
       if self.share weights:
            hat inputs = K.conv1d(inputs, self.kernel)
        else:
            hat inputs = K.local conv1d(inputs, self.kernel, [1], [1])
       batch size = K.shape(inputs)[0]
       input num capsule = K.shape(inputs)[1]
       hat inputs = K.reshape(hat inputs,
                               (batch size, input num capsule,
                                self.num_capsule, self.dim_capsule))
       hat inputs = K.permute dimensions(hat inputs, (0, 2, 1, 3))
       b = K.zeros like(hat inputs[:, :, :, 0])
       for i in range(self.routings):
            c = softmax(b, 1)
            o = self.activation(keras.backend.batch dot(c, hat inputs, [2, 2]))
            if i < self.routings - 1:</pre>
                b = keras.backend.batch dot(o, hat inputs, [2, 3])
                if K.backend() == 'theano':
                    o = K.sum(o, axis=1)
       return o
   def compute output shape(self, input shape):
       return (None, self.num_capsule, self.dim capsule)
batch size = 16
num classes = 2
epochs = 100
#class weights to handle class imbalance
#class weights = {0: 1-np.count nonzero(y train==0)/len(y train),
                1: 1-np.count nonzero(y train==1)/len(y train)}
```

```
input image = Input(shape=(None, None, 3))
x = Conv2D(64, (3, 3), activation='relu', trainable = False)(input image)
x=BatchNormalization(axis=-1, momentum=0.99, epsilon=0.001, center=True, scale=True, beta initializer='zeros'
, gamma initializer='ones', moving mean initializer='zeros', moving variance initializer='ones', beta regular
izer=None, gamma regularizer=None, beta constraint=None, gamma constraint=None)(x)
x = Conv2D(64, (3, 3), activation='relu', trainable = False)(x)
x = AveragePooling2D((2, 2),trainable = False)(x)
x = Conv2D(128, (3, 3), activation='relu', trainable = False)(x)
x = Conv2D(128, (3, 3), activation='relu', trainable = False)(x)
x = Reshape((-1, 128))(x)
x = Capsule(32, 8, 3, True)(x)
x = Capsule(32, 8, 3, True)(x)
capsule = Capsule(5, 16, 3, True)(x)
output = Lambda(lambda z: K.sqrt(K.sum(K.square(z), 2)))(capsule)
model = Model(inputs=[input image], outputs=[output])
model.load weights('My Drive/Colab Notebooks/215-FP Dataset/pre-train.h5')
capsule2 = Capsule(2, 16, 3, True)(model.layers[-3].output)
output2 = Lambda(lambda z: K.sqrt(K.sum(K.square(z), 2)))(capsule2)
model2 = Model(inputs=[input image], outputs=[output2])
adam = optimizers.Adam(lr=0.001)
model.compile(loss=margin loss, optimizer=adam, metrics=['accuracy'])
model2.compile(loss=margin loss, optimizer=adam, metrics=['accuracy'])
model2.summary()
data augmentation = False
# The best model is selected based on the loss value on the validation set
```

```
filepath='My Drive/Colab Notebooks/215-FP Dataset/after-train.h5'
checkpoint = ModelCheckpoint(filepath, monitor='val loss', verbose=1, save best only=True, mode='min')
callbacks list = [checkpoint]
if not data augmentation:
   print('Not using data augmentation.')
   model2.fit(
       [x train], [y train],
       batch size=batch size,
       epochs=epochs,
       validation data=[[x valid], [y valid]],
       shuffle=True, callbacks = callbacks list)
else:
   print('Using real-time data augmentation.')
   # This will do preprocessing and realtime data augmentation:
   datagen = ImageDataGenerator(
       featurewise center=False, # set input mean to 0 over the dataset
       samplewise center=False, # set each sample mean to 0
       featurewise std normalization=False, # divide inputs by dataset std
       samplewise std normalization=False, # divide each input by its std
       zca whitening=False, # apply ZCA whitening
       zca epsilon=1e-06, # epsilon for ZCA whitening
       rotation range=0.1, # randomly rotate images in 0 to 180 degrees
       width_shift_range=0.1, # randomly shift images horizontally
       height shift range=0.1, # randomly shift images vertically
       brightness range=[0.5,1.5],
       shear range=0.1, # set range for random shear
       zoom range=0.1, # set range for random zoom
       channel shift range=0., # set range for random chann el shifts
       # set mode for filling points outside the input boundaries
       fill mode='nearest',
       cval=0., # value used for fill mode = "constant"
       horizontal_flip=True, # randomly flip images
       vertical flip=True, # randomly flip images
       # set rescaling factor (applied before any other transformation)
       rescale=None,
```

```
# set function that will be applied on each input
preprocessing_function=None,
# image data format, either "channels_first" or "channels_last"
data_format=None,
# fraction of images reserved for validation (strictly between 0 and 1)
validation_split=0.0)

# Compute quantities required for feature-wise normalization
# (std, mean, and principal components if ZCA whitening is applied).
datagen.fit(x_train)

# Fit the model on the batches generated by datagen.flow().
model.fit(
    datagen.flow(x_train, y_train, batch_size=batch_size),
    epochs=epochs,
    validation_data=(x_valid, y_valid),shuffle=True)
```

Model: "model_19"

Layer (type)	Output Shape	Param #
input_10 (InputLayer)	[(None, None, None, 3)]	0
conv2d_36 (Conv2D)	(None, None, None, 64)	1792
batch_normalization_9 (Batch	(None, None, None, 64)	256
conv2d_37 (Conv2D)	(None, None, None, 64)	36928
average_pooling2d_9 (Average	(None, None, None, 64)	0
conv2d_38 (Conv2D)	(None, None, None, 128)	73856
conv2d_39 (Conv2D)	(None, None, None, 128)	147584
reshape_9 (Reshape)	(None, None, 128)	0
capsule_36 (Capsule)	(None, 32, 8)	32768
capsule_37 (Capsule)	(None, 32, 8)	2048
capsule_39 (Capsule)	(None, 2, 16)	256
lambda_19 (Lambda)	(None, 2)	0

Total params: 295,488
Trainable params: 35,200
Non-trainable params: 260,288

Not using data augmentation.

Train on 6025 samples, validate on 1507 samples

Epoch 1/100

Epoch 00001: val_loss improved from inf to 0.00287, saving model to My Drive/Colab Notebooks/215-FP_Dataset/a

fter-train.h5

029 - val_accuracy: 0.6105

Epoch 2/100

Epoch 00002: val loss improved from 0.00287 to 0.00269, saving model to My Drive/Colab Notebooks/215-FP Datas

```
et/after-train.h5
027 - val accuracy: 0.6470
Epoch 3/100
Epoch 00003: val loss improved from 0.00269 to 0.00263, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
026 - val accuracy: 0.6563
Epoch 4/100
Epoch 00004: val loss improved from 0.00263 to 0.00259, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
026 - val accuracy: 0.6735
Epoch 5/100
Epoch 00005: val loss improved from 0.00259 to 0.00257, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
026 - val accuracy: 0.6981
Epoch 6/100
Epoch 00006: val loss improved from 0.00257 to 0.00255, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
026 - val accuracy: 0.7067
Epoch 7/100
Epoch 00007: val loss improved from 0.00255 to 0.00254, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.7259
Epoch 8/100
Epoch 00008: val loss improved from 0.00254 to 0.00253, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.7372
Epoch 9/100
Epoch 00009: val loss improved from 0.00253 to 0.00252, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
```

```
025 - val accuracy: 0.7478
Epoch 10/100
Epoch 00010: val loss improved from 0.00252 to 0.00252, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.7611
Epoch 11/100
Epoch 00011: val loss improved from 0.00252 to 0.00251, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.7585
Epoch 12/100
Epoch 00012: val loss improved from 0.00251 to 0.00251, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.7777
Epoch 13/100
Epoch 00013: val loss improved from 0.00251 to 0.00251, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.7903
Epoch 14/100
Epoch 00014: val loss improved from 0.00251 to 0.00251, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.8069
Epoch 15/100
Epoch 00015: val loss improved from 0.00251 to 0.00250, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.8129
Epoch 16/100
Epoch 00016: val loss improved from 0.00250 to 0.00250, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
```

```
025 - val accuracy: 0.8228
Epoch 17/100
Epoch 00017: val loss improved from 0.00250 to 0.00250, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.8109
Epoch 18/100
Epoch 00018: val loss improved from 0.00250 to 0.00250, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.8646
Epoch 19/100
Epoch 00019: val loss improved from 0.00250 to 0.00250, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.8368
Epoch 20/100
Epoch 00020: val loss improved from 0.00250 to 0.00250, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.8825
Epoch 21/100
Epoch 00021: val loss improved from 0.00250 to 0.00250, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.8832
Epoch 22/100
Epoch 00022: val loss improved from 0.00250 to 0.00250, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9011
Epoch 23/100
Epoch 00023: val loss improved from 0.00250 to 0.00250, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.8945
```

```
Epoch 24/100
Epoch 00024: val loss improved from 0.00250 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9111
Epoch 25/100
Epoch 00025: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9204
Epoch 26/100
Epoch 00026: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9164
Epoch 27/100
Epoch 00027: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9250
Epoch 28/100
Epoch 00028: val loss did not improve from 0.00249
025 - val accuracy: 0.9131
Epoch 29/100
Epoch 00029: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9463
Epoch 30/100
Epoch 00030: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9536
Epoch 31/100
```

```
Epoch 00031: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9575
Epoch 32/100
Epoch 00032: val loss did not improve from 0.00249
025 - val accuracy: 0.9529
Epoch 33/100
Epoch 00033: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9635
Epoch 34/100
Epoch 00034: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9688
Epoch 35/100
Epoch 00035: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9701
Epoch 36/100
Epoch 00036: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9741
Epoch 37/100
Epoch 00037: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9754
Epoch 38/100
Epoch 00038: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
```

```
025 - val accuracy: 0.9768
Epoch 39/100
Epoch 00039: val loss did not improve from 0.00249
025 - val accuracy: 0.9794
Epoch 40/100
Epoch 00040: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9794
Epoch 41/100
Epoch 00041: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9821
Epoch 42/100
Epoch 00042: val loss did not improve from 0.00249
025 - val accuracy: 0.9814
Epoch 43/100
Epoch 00043: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9854
Epoch 44/100
Epoch 00044: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9900
Epoch 45/100
Epoch 00045: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9861
Epoch 46/100
```

```
Epoch 00046: val loss did not improve from 0.00249
025 - val accuracy: 0.9881
Epoch 47/100
Epoch 00047: val loss did not improve from 0.00249
025 - val accuracy: 0.9881
Epoch 48/100
Epoch 00048: val loss did not improve from 0.00249
025 - val accuracy: 0.9861
Epoch 49/100
Epoch 00049: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9907
Epoch 50/100
Epoch 00050: val loss did not improve from 0.00249
025 - val accuracy: 0.9914
Epoch 51/100
Epoch 00051: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP_Datas
et/after-train.h5
025 - val accuracy: 0.9900
Epoch 52/100
Epoch 00052: val loss did not improve from 0.00249
025 - val accuracy: 0.9927
Epoch 53/100
Epoch 00053: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9934
Epoch 54/100
```

```
Epoch 00054: val loss did not improve from 0.00249
025 - val accuracy: 0.9920
Epoch 55/100
Epoch 00055: val loss did not improve from 0.00249
025 - val accuracy: 0.9927
Epoch 56/100
Epoch 00056: val loss did not improve from 0.00249
025 - val accuracy: 0.9914
Epoch 57/100
Epoch 00057: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9940
Epoch 58/100
6000/6025 [=========================>.] - ETA: 0s - loss: 0.0025 - accuracy: 0.9983
Epoch 00058: val loss did not improve from 0.00249
025 - val accuracy: 0.9927
Epoch 59/100
Epoch 00059: val loss did not improve from 0.00249
025 - val accuracy: 0.9940
Epoch 60/100
Epoch 00060: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9940
Epoch 61/100
Epoch 00061: val loss did not improve from 0.00249
025 - val accuracy: 0.9947
Epoch 62/100
```

```
Epoch 00062: val loss did not improve from 0.00249
6025/6025 [=============== ] - 17s 3ms/sample - loss: 0.0025 - accuracy: 0.9968 - val_loss: 0.0
025 - val accuracy: 0.9907
Epoch 63/100
Epoch 00063: val loss did not improve from 0.00249
025 - val accuracy: 0.9940
Epoch 64/100
Epoch 00064: val loss did not improve from 0.00249
025 - val accuracy: 0.9920
Epoch 65/100
Epoch 00065: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9947
Epoch 66/100
Epoch 00066: val_loss did not improve from 0.00249
025 - val accuracy: 0.9927
Epoch 67/100
Epoch 00067: val loss did not improve from 0.00249
6025/6025 [=============== ] - 17s 3ms/sample - loss: 0.0025 - accuracy: 0.9988 - val_loss: 0.0
025 - val accuracy: 0.9940
Epoch 68/100
Epoch 00068: val loss did not improve from 0.00249
025 - val accuracy: 0.9914
Epoch 69/100
Epoch 00069: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9954
Epoch 70/100
Epoch 00070: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
```

```
et/after-train.h5
025 - val accuracy: 0.9967
Epoch 71/100
Epoch 00071: val loss did not improve from 0.00249
025 - val accuracy: 0.9954
Epoch 72/100
Epoch 00072: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9947
Epoch 73/100
Epoch 00073: val loss did not improve from 0.00249
025 - val accuracy: 0.9940
Epoch 74/100
Epoch 00074: val_loss did not improve from 0.00249
025 - val accuracy: 0.9920
Epoch 75/100
Epoch 00075: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9934
Epoch 76/100
Epoch 00076: val loss did not improve from 0.00249
025 - val accuracy: 0.9947
Epoch 77/100
Epoch 00077: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9914
Epoch 78/100
6000/6025 [=========================>.] - ETA: 0s - loss: 0.0025 - accuracy: 0.9993
```

```
Epoch 00078: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9960
Epoch 79/100
Epoch 00079: val loss did not improve from 0.00249
025 - val accuracy: 0.9934
Epoch 80/100
Epoch 00080: val loss did not improve from 0.00249
025 - val accuracy: 0.9954
Epoch 81/100
Epoch 00081: val loss did not improve from 0.00249
025 - val accuracy: 0.9954
Epoch 82/100
Epoch 00082: val loss did not improve from 0.00249
025 - val accuracy: 0.9947
Epoch 83/100
6000/6025 [=========================>.] - ETA: 0s - loss: 0.0025 - accuracy: 0.9987
Epoch 00083: val loss did not improve from 0.00249
025 - val accuracy: 0.9927
Epoch 84/100
Epoch 00084: val loss did not improve from 0.00249
025 - val accuracy: 0.9940
Epoch 85/100
Epoch 00085: val loss did not improve from 0.00249
025 - val accuracy: 0.9934
Epoch 86/100
Epoch 00086: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
```

```
025 - val accuracy: 0.9947
Epoch 87/100
Epoch 00087: val loss did not improve from 0.00249
025 - val accuracy: 0.9954
Epoch 88/100
Epoch 00088: val loss did not improve from 0.00249
025 - val accuracy: 0.9954
Epoch 89/100
Epoch 00089: val loss did not improve from 0.00249
025 - val accuracy: 0.9934
Epoch 90/100
Epoch 00090: val loss improved from 0.00249 to 0.00249, saving model to My Drive/Colab Notebooks/215-FP Datas
et/after-train.h5
025 - val accuracy: 0.9940
Epoch 91/100
Epoch 00091: val loss did not improve from 0.00249
025 - val accuracy: 0.9947
Epoch 92/100
Epoch 00092: val loss did not improve from 0.00249
025 - val accuracy: 0.9960
Epoch 93/100
Epoch 00093: val loss did not improve from 0.00249
025 - val accuracy: 0.9954
Epoch 94/100
Epoch 00094: val loss did not improve from 0.00249
025 - val accuracy: 0.9960
```

```
Epoch 95/100
   Epoch 00095: val loss did not improve from 0.00249
   025 - val accuracy: 0.9954
   Epoch 96/100
   Epoch 00096: val loss did not improve from 0.00249
   025 - val accuracy: 0.9954
   Epoch 97/100
   Epoch 00097: val loss did not improve from 0.00249
   025 - val accuracy: 0.9954
   Epoch 98/100
   Epoch 00098: val loss did not improve from 0.00249
   025 - val accuracy: 0.9960
   Epoch 99/100
   6000/6025 [=========================>.] - ETA: 0s - loss: 0.0025 - accuracy: 0.9990
   Epoch 00099: val_loss did not improve from 0.00249
   025 - val accuracy: 0.9940
   Epoch 100/100
   Epoch 00100: val loss did not improve from 0.00249
   025 - val accuracy: 0.9960
In [ ]:
In [ ]:
```

```
In [ ]: | batch size = 16
        num classes = 2
        epochs = 100
        x test= x valid
        y_test= y_valid
        #model: model without pre-training
        input image = Input(shape=(None, None, 3))
        x = Conv2D(64, (3, 3), activation='relu')(input image)
        x=BatchNormalization(axis=-1, momentum=0.99, epsilon=0.001, center=True, scale=True, beta initializer='zeros'
        , gamma initializer='ones', moving mean initializer='zeros', moving variance initializer='ones', beta regular
        izer=None, gamma regularizer=None, beta constraint=None, gamma constraint=None)(x)
        x = Conv2D(64, (3, 3), activation='relu')(x)
        x = AveragePooling2D((2, 2))(x)
        x = Conv2D(128, (3, 3), activation='relu')(x)
        x = Conv2D(128, (3, 3), activation='relu')(x)
        x = Reshape((-1, 128))(x)
        x = Capsule(32, 8, 3, True)(x)
        x = Capsule(32, 8, 3, True)(x)
        capsule = Capsule(2, 16, 3, True)(x)
        output = Lambda(lambda z: K.sqrt(K.sum(K.square(z), 2)))(capsule)
        #model2: model with pre-training
        input image2 = Input(shape=(None, None, 3))
        x2 = Conv2D(64, (3, 3), activation='relu')(input image2)
        x2=BatchNormalization(axis=-1, momentum=0.99, epsilon=0.001, center=True, scale=True, beta initializer='zero
        s', gamma initializer='ones', moving mean initializer='zeros', moving variance initializer='ones', beta regul
```

```
arizer=None, gamma regularizer=None, beta constraint=None, gamma constraint=None)(x2)
x2 = Conv2D(64, (3, 3), activation='relu')(x2)
x2 = AveragePooling2D((2, 2))(x2)
x2 = Conv2D(128, (3, 3), activation='relu')(x2)
x2 = Conv2D(128, (3, 3), activation='relu')(x2)
x2 = Reshape((-1, 128))(x2)
x2 = Capsule(32, 8, 3, True)(x2)
x2 = Capsule(32, 8, 3, True)(x2)
capsule2 = Capsule(2, 16, 3, True)(x2)
output2 = Lambda(lambda z: K.sqrt(K.sum(K.square(z), 2)))(capsule2)
model = Model(inputs=[input image], outputs=[output])
model2 = Model(inputs=[input_image2], outputs=[output2])
model.load weights('My Drive/Colab Notebooks/215-FP Dataset/after-train.h5')
model2.load weights('My Drive/Colab Notebooks/215-FP Dataset/weights-improvement-binary-after-44.h5')
predict=model.predict([x test])
predict=np.argmax(predict,axis=1)
predict2=model2.predict([x test])
predict2=np.argmax(predict2,axis=1)
summation1=0
```

```
In [47]: | y test=np.argmax(y test,axis=1)
         y test[0:100]
Out[47]: array([0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0,
                0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
                0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0,
                1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1,
                0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1])
In [48]: predict[0:100]
Out[48]: array([0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0,
                0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
                0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0,
                1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1,
                0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1])
         '''from sklearn.preprocessing import LabelEncoder
 In [ ]:
         from sklearn.preprocessing import OneHotEncoder
         predict=np.array(predict)
         label encoder = LabelEncoder()
         predict = label encoder.fit transform(predict)
         predict
         onehot encoder = OneHotEncoder(sparse=False)
         predict = predict.reshape(len(predict), 1)
         predict = onehot encoder.fit transform(predict)
         predict
         predict2=np.array(predict2)
         label encoder = LabelEncoder()
         predict2 = label encoder.fit transform(predict2)
         predict2
         onehot encoder = OneHotEncoder(sparse=False)
         predict2 = predict2.reshape(len(predict2), 1)
         predict2 = onehot encoder.fit transform(predict2)
         predict2'''
 In [ ]:
```

```
In [63]: accuracy before=0
         specificity before=0
         sensitivity before=0
         precision before=0
         f1 before=0
         newaccuracy=0
         from __future__ import print_function
         from tensorflow.keras import backend as K
         from tensorflow.keras.layers import Layer
         from tensorflow.keras import activations
         from tensorflow.keras import utils
         from tensorflow.keras.models import Model
         from tensorflow.keras.layers import *
         import numpy as np
         import keras
         from sklearn.metrics import roc curve, auc
         from matplotlib import pyplot as plt
         summation1=0
         for i in range(len(x test)):
             if predict[i]==y test[i]:
                 summation1=summation1+1
         print('accurate predictions: ',summation1)
         print('Total predictions to be made in testing :', len(x test))
         accuracy before=summation1/len(x test)
         summation1=0
         summation2=0
         summation3=0
         summation4=0
         for i in range(len(x test)):
             if predict[i]==y test[i] and y test[i]==0:
                 summation1=summation1+1
         specificity before=summation1/np.count nonzero(y test==0)
         for i in range(len(x test)):
             if predict[i]==y test[i] and y test[i]==1:
```

```
summation2=summation2+1
        sensitivity before=summation2/np.count nonzero(y test==1)
        for i in range(len(x test)):
            if predict[i]!=y test[i] and y test[i]==1:
                summation3=summation3+1
        for i in range(len(x_test)):
            if predict2[i]!=y test[i] and y test[i]==0:
                summation4=summation4+1
        precision before=summation1/(summation1+summation3)
        newaccuracy=((summation1+summation2)/(summation1+summation2+summation3+summation4))
        f1_before=2*((sensitivity_before*precision_before)/(sensitivity_before + precision_before))
        #print("TN: ",summation1,", TP: ",summation2,", FP:",summation3, ", FN:",summation4)
        print('accuracy : ',accuracy before)
        print('specificity : ',specificity before)
        print('sensitivity or precision : ',sensitivity before)
        print('precision : ', precision before)
        print('f1-score : ', f1 before)
        #print('newaccuracy : ',newaccuracy)
        accurate predictions: 1498
        Total predictions to be made in testing: 1507
        accuracy: 0.9940278699402787
        specificity : 0.9960745829244357
        sensitivity or precision : 0.9897540983606558
        precision : 0.9950980392156863
        f1-score : 0.9924188748760097
In [ ]:
In [ ]:
```