

Covid-19 CT scan identification project code

▼ Installing missing modules

```
#@title Installing missing modules
#installing patool for unzipping & split-folder for splitting the folder into train,test,val
!pip install patool
!pip install split-folders
```

Looking in indexes: <https://pypi.org/simple>, <https://us-python.pkg.dev/colab-wheels/public/simple/>
Collecting patool
 Downloading patool-1.12-py2.py3-none-any.whl (77 kB)
77.5/77.5 kB 4.9 MB/s eta 0:00:00
Installing collected packages: patool
Successfully installed patool-1.12
Looking in indexes: <https://pypi.org/simple>, <https://us-python.pkg.dev/colab-wheels/public/simple/>
Collecting split-folders
 Downloading split_folders-0.5.1-py3-none-any.whl (8.4 kB)
Installing collected packages: split-folders
Successfully installed split-folders-0.5.1

✓ 27s [2] `from google.colab import drive`
`drive.mount('/content/drive')`

Mounted at /content/drive

▼ Importing modules

```
[3] #@title Importing modules
import patoolib
import cv2
import glob
import os
import splitfolders as sf
```

✓ [4] `import tensorflow as tf`
`from tensorflow import keras`
`from tensorflow.keras import layers`
`from tensorflow.python.keras.layers import Dense, Flatten`
`from tensorflow.keras.models import Sequential`
`from tensorflow.keras.optimizers import Adam`
`from keras.callbacks import ModelCheckpoint, EarlyStopping, CSVLogger`
`from keras.preprocessing.image import ImageDataGenerator`

```
✓ [61] import pandas as pd
import numpy as np
from keras.models import load_model
import matplotlib.pyplot as plt
from PIL import Image as imgshow
```

▼ Unzipping files

```
[6] #@title Unzipping files
#set the path where the zip files of images are present
input='/content/drive/MyDrive/CTscan/'
```

```
✓ 4s [7] #unzipping folder containing CT scan images
patoolib.extract_archive(input+"COVID.zip",outdir='/content/CTscan')

patool: Extracting /content/drive/MyDrive/CTscan/COVID.zip ...
patool: running /usr/bin/7z x -o/content/CTscan -- /content/drive/MyDrive/CTscan/COVID.zip
patool: ... /content/drive/MyDrive/CTscan/COVID.zip extracted to `/content/CTscan'.
'/content/CTscan'
```

```
✓ 4s [8] patoolib.extract_archive(input+"nonCOVID.zip",outdir='/content/CTscan')

patool: Extracting /content/drive/MyDrive/CTscan/nonCOVID.zip ...
patool: running /usr/bin/7z x -o/content/CTscan -- /content/drive/MyDrive/CTscan/nonCOVID.zip
patool: ... /content/drive/MyDrive/CTscan/nonCOVID.zip extracted to `/content/CTscan'.
'/content/CTscan'
```

▼ Resizing images

```
[9] #@title Resizing images
curr=os.getcwd()
```

```
✓ [10] #Resizing images to 224x224 for Resnet
inputfolder='/content/CTscan/COVID'
length=len(inputfolder)

path='/content/resize/covid'
os.makedirs(path,mode=0o666,exist_ok=True)

for img in glob.glob(inputfolder+'/*.'):
    image=cv2.imread(img)
    imgRe=cv2.resize(image,(224,224))
    cv2.imwrite(path+img[length:],imgRe)
```

```

✓ [11] inputfolder='/content/CTscan/nonCOVID'
      length=len(inputfolder)

      path='/content/resize/noncovid'
      os.makedirs(path,mode=0o666,exist_ok=True)

      for img in glob.glob(inputfolder+'/*.*'):
          image=cv2.imread(img)
          imgRe=cv2.resize(image,(224,224))
          cv2.imwrite(path+img[length:],imgRe)

```

```

✓ [12] curr
      ↗ '/content'

```

```

✓ [13] input_folder=curr+'/resize'

```

Splitting dataset

```

[14] #@title Splitting dataset
      #splitting folder into train,test,val
      sf.ratio(input_folder,output='/content/final',seed=50,ratio=(.7,.2,.1),group_prefix=None,move=False)

```

Copying files: 2480 files [00:00, 2558.22 files/s]

Data preparation

```

▶ #@title Data preparation
  #Data agumentation
  train_datagen = ImageDataGenerator(rescale=1./255,
                                     rotation_range=45,
                                     width_shift_range=0.2,
                                     height_shift_range=0.2,
                                     shear_range=0.2,
                                     zoom_range=0.2,)

  val_datagen = ImageDataGenerator(rescale=1./255)

```

```

✓ [69] train_dir=curr+'/final/train'
      validation_dir=curr+'/final/val'

```

```
[17] #train & validation dataset generators
train_generator = train_datagen.flow_from_directory(train_dir,
                                                    target_size=(224, 224),
                                                    batch_size=32,
                                                    class_mode='binary',
                                                    color_mode='grayscale',
                                                    shuffle=True,
                                                    seed=50)

validation_generator = val_datagen.flow_from_directory(validation_dir,
                                                        target_size=(224, 224),
                                                        batch_size=32,
                                                        class_mode='binary',
                                                        color_mode='grayscale',
                                                        shuffle=True,
                                                        seed=50)
```

Found 1735 images belonging to 2 classes.
Found 495 images belonging to 2 classes.

Model building

```
[18] #@title Model building
#defining modelcheckpoints & early stopping
filepath='saved_models/weights-improvement--{epoch:02d}--{val_acc:.2f}.hdf5'
checkpoint=ModelCheckpoint(filepath,monitor='val_acc',verbose=1,save_best_only=True,mode='max')
early_stop=EarlyStopping(monitor='val_loss',patience=10,verbose=1)
log_csv=CSVLogger('my_logs.csv',separator=',',append=False)
callbackslist=[checkpoint,early_stop,log_csv]
```

```
[19] #Defining Resnet50 model
resnet_model = Sequential()

pretrained_model= tf.keras.applications.ResNet50(include_top=False,
                                                  input_shape=(224,224,1),
                                                  pooling=None,classes=2,
                                                  weights=None)
for layer in pretrained_model.layers:
    layer.trainable=False
```

```
[20] #Adding layers to resnet model
resnet_model.add(pretrained_model)
resnet_model.add(layers.Conv2D(32, 3, activation='relu',padding='same',
                               input_shape=(224,224,1),data_format='channels_last'))
resnet_model.add(layers.MaxPooling2D(pool_size=2, strides=2,padding='same'))
resnet_model.add(layers.Conv2D(64, 3, activation='relu',padding='same'))
resnet_model.add(layers.MaxPooling2D(pool_size=2, strides=2,padding='same'))
resnet_model.add(layers.Conv2D(128, 3, activation='relu',padding='same'))
resnet_model.add(layers.MaxPooling2D(pool_size=2, strides=2,padding='same'))
resnet_model.add(layers.Dropout(0.3))
resnet_model.add(layers.Conv2D(128, 3, activation='relu',padding='same'))
resnet_model.add(layers.MaxPooling2D(pool_size=2, strides=2,padding='same'))
resnet_model.add(layers.Flatten())
resnet_model.add(layers.Dense(512, activation='relu'))
resnet_model.add(layers.Dropout(0.5))
resnet_model.add(layers.Dense(1, activation='sigmoid'))
```

```
[▶] #model summary
resnet_model.summary()
```

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
=====		
resnet50 (Functional)	(None, 7, 7, 2048)	23581440
conv2d (Conv2D)	(None, 7, 7, 32)	589856
max_pooling2d (MaxPooling2D)	(None, 4, 4, 32)	0
conv2d_1 (Conv2D)	(None, 4, 4, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 2, 2, 64)	0
conv2d_2 (Conv2D)	(None, 2, 2, 128)	73856
max_pooling2d_2 (MaxPooling2D)	(None, 1, 1, 128)	0

dropout (Dropout)	(None, 1, 1, 128)	0
conv2d_3 (Conv2D)	(None, 1, 1, 128)	147584
max_pooling2d_3 (MaxPooling2D)	(None, 1, 1, 128)	0
flatten (Flatten)	(None, 128)	0
dense (Dense)	(None, 512)	66048
dropout_1 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 1)	513

```
=====
Total params: 24,477,793
Trainable params: 896,353
Non-trainable params: 23,581,440
=====
```

```
[22] #compiling model
      resnet_model.compile(loss='binary_crossentropy',
                           optimizer=Adam(learning_rate=0.0001),
                           metrics=['acc','Precision','Recall','AUC'])
```

Model fitting

```
[23] #@title Model fitting
      #model fitting
      batch_size=32
      history=resnet_model.fit(train_generator,
                               steps_per_epoch=1735//batch_size,
                               epochs=100,
                               validation_data=validation_generator,
                               validation_steps=495//batch_size,
                               callbacks=callbackslist)
```

Epoch 1/100

```
Epoch 24/100
54/54 [=====] - ETA: 0s - loss: 0.6010 - acc: 0.6682 - precision: 0.6310 - recall: 0.8017
Epoch 24: val_acc did not improve from 0.70833
54/54 [=====] - 14s 256ms/step - loss: 0.6010 - acc: 0.6682 - precision: 0.6310 - recall:
Epoch 25/100
54/54 [=====] - ETA: 0s - loss: 0.5960 - acc: 0.6782 - precision: 0.6368 - recall: 0.8164
Epoch 25: val_acc did not improve from 0.70833
54/54 [=====] - 13s 235ms/step - loss: 0.5960 - acc: 0.6782 - precision: 0.6368 - recall:
Epoch 26/100
54/54 [=====] - ETA: 0s - loss: 0.5951 - acc: 0.6712 - precision: 0.6279 - recall: 0.8337
Epoch 26: val_acc did not improve from 0.70833
54/54 [=====] - 13s 248ms/step - loss: 0.5951 - acc: 0.6712 - precision: 0.6279 - recall:
Epoch 27/100
54/54 [=====] - ETA: 0s - loss: 0.5885 - acc: 0.6788 - precision: 0.6343 - recall: 0.8384
Epoch 27: val_acc did not improve from 0.70833
54/54 [=====] - 13s 236ms/step - loss: 0.5885 - acc: 0.6788 - precision: 0.6343 - recall:
Epoch 27: early stopping
```

```
[24] hist=pd.DataFrame(history.history)
      hist.tail()
```

	loss	acc	precision	recall	auc	val_loss	val_acc	val_precision	val_recall	val_auc
22	0.587021	0.687610	0.644860	0.819477	0.737783	0.571899	0.695833	0.636103	0.921162	0.779319
23	0.600956	0.668233	0.631041	0.801653	0.712500	0.608055	0.656250	0.593909	0.979079	0.776845
24	0.595966	0.678215	0.636784	0.816351	0.716158	0.570702	0.697917	0.633929	0.906383	0.777195
25	0.595135	0.671169	0.627886	0.833726	0.722451	0.583466	0.675000	0.608696	0.949153	0.777334
26	0.588484	0.678802	0.634255	0.838443	0.733276	0.581687	0.672917	0.610959	0.936975	0.778379

Visualizing accuracy

```
[▶] #@title Visualizing accuracy
acc = hist['acc']
val_acc = hist['val_acc']
loss = hist['loss']
val_loss = hist['val_loss']

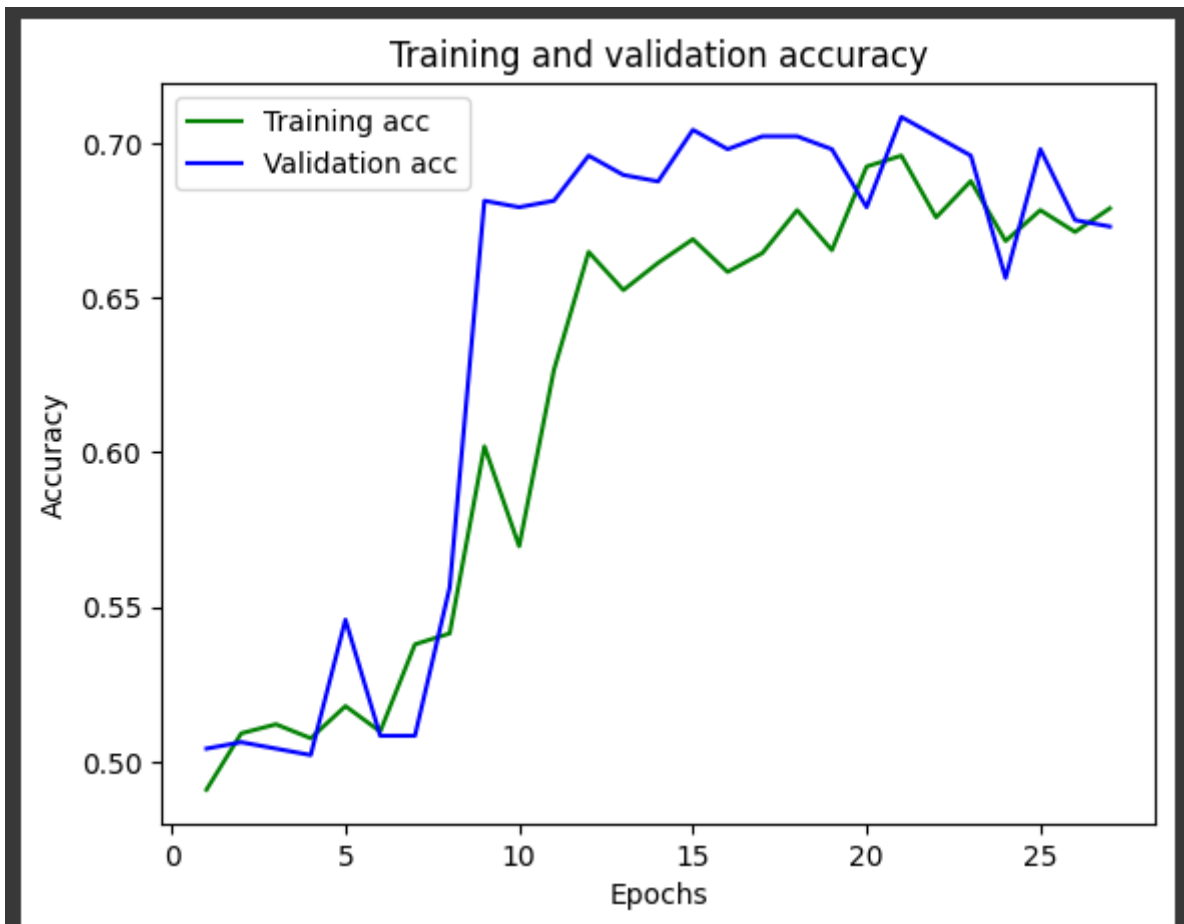
epochs = range(1, len(acc) + 1)

plt.plot(epochs, acc, 'g', label='Training acc')
plt.plot(epochs, val_acc, 'b', label='Validation acc')
plt.title('Training and validation accuracy')
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend()

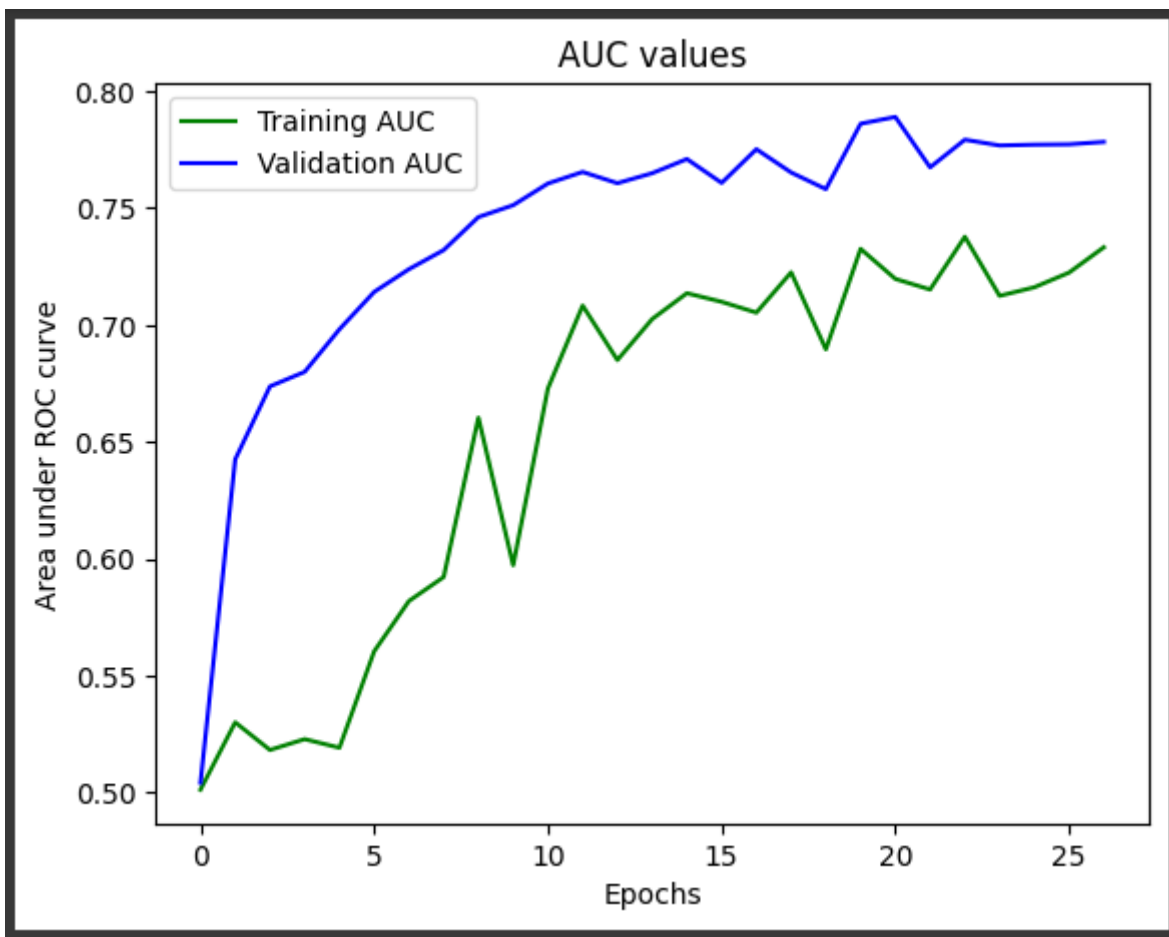
plt.figure()
```

```
plt.plot(epochs, loss, 'g', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
plt.title('Training and validation loss')
plt.xlabel("Epochs")
plt.ylabel("Loss")
plt.legend()

plt.show()
```

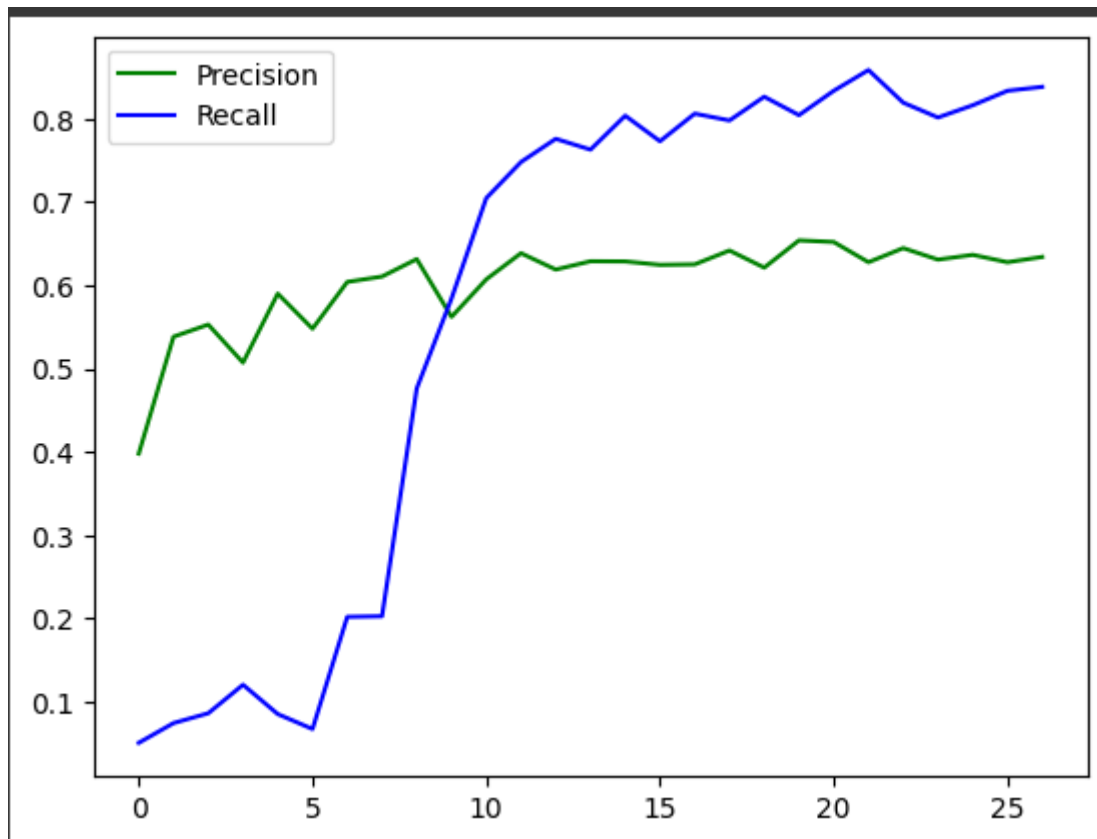



```
+ Code + Text
▶ plt.plot(hist['auc'], 'g', label='Training AUC')
  plt.plot(hist['val_auc'], 'b', label='Validation AUC')
  plt.legend()
  plt.xlabel("Epochs")
  plt.ylabel("Area under ROC curve")
  plt.title("AUC values")
  plt.show()
```



```
[28] # Here, AUC value is closer to 75%
```

```
[29] plt.plot(hist['precision'], 'g', label='Precision')
      plt.plot(hist['recall'], 'b', label='Recall')
      plt.legend()
      plt.show()
```



```
[36] # Here, precision is around 65% & recall is about 85%.
      # It means the models have some chance of giving False positive (slightly low precision),
      # but it is less likely to return false negative (high recall), and maximum positive cases will be caught.
```

▼ Saving model

```
[31] #@title Saving model
      #saving the model for future reference
      resnet_model.save("CovidPredict.h5")
```

▼ Prediction

```
[32] #@title Prediction
      ##          PREDICTION          ##

      # Loading the model
      my_model=load_model('CovidPredict.h5')
```

```
[33] #Generating our test data set
test_dir=curr+ '/final/test'

test_datagen = ImageDataGenerator(rescale=1./255)
test_generator = test_datagen.flow_from_directory(test_dir,
                                                target_size=(224, 224),
                                                batch_size=32,
                                                class_mode='binary',
                                                color_mode='grayscale',
                                                shuffle=False)
```

Found 250 images belonging to 2 classes.

```
[34] #evaluating model for test dataset for the metrics
loss,acc,pre,recall,auc=my_model.evaluate(test_generator, verbose=2)
```

8/8 - 3s - loss: 0.5588 - acc: 0.7240 - precision: 0.6608 - recall: 0.9113 - auc: 0.7776 - 3s/epoch - 369ms/step

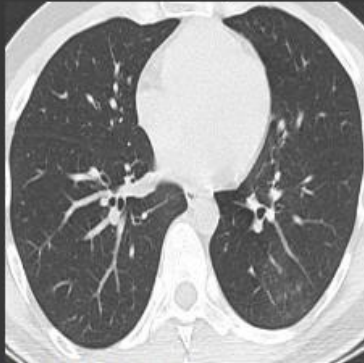
```
[35] # We can also use model.predict to get individual prediction
# this will return array of value between 0 to 1, & after defining a threshold(generally 0.5), we classify it.
predt=my_model.predict(test_generator)
```

8/8 [=====] - 3s 135ms/step

```
[39] files=test_generator.files
class_dict=test_generator.class_indices # a dictionary of the form class name: class index
rev_dict={}
for key, value in class_dict.items():
    rev_dict[value]=key # dictionary of the form class index: class name
```

```
[68] prediction=prede
for i, p in enumerate(prediction):
    index=np.argmax(p)
    klass=rev_dict[index]
    prob=p[index]
    print('for file ', files[i], ' predicted class is ', klass, ' with probability ',(1-prob)*100)
    im = imgshow.open("/content/resize/"+files[i])
    im.show()
    print('*****')
```

for file noncovid/Non-Covid (993).png predicted class is covid with probability 35.43 %



for file noncovid/Non-Covid (998).png predicted class is covid with probability 34.7 %



for file covid/Covid (790).png predicted class is covid with probability 85.03 %



for file covid/Covid (793).png predicted class is covid with probability 73.06 %



✓ 45s completed at 5:06 PM