```
In [ ]:
In [1]: !unzip '/content/drive/My Drive/Covid19Pred/Dataset kaggle.zip'
        Archive: /content/drive/My Drive/Covid19Pred/Dataset kaggle.zip
        replace Dataset_kaggle/COVID/Covid (1).png? [y]es, [n]o, [A]ll, [N]one, [r]enam
        e: N
In [2]:
        import os
        import cv2
        import numpy as np
        from PIL import Image
        import matplotlib.pyplot as plt
        from sklearn.model selection import train test split
        import tensorflow as tf
        from tensorflow import keras
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Conv2D,MaxPooling2D,Dense,Flatten,Dropout
        from tensorflow.keras.layers import BatchNormalization
        from tensorflow.keras.applications.mobilenet import MobileNet
        from tensorflow.keras.applications.resnet50 import ResNet50
        from tensorflow.keras import optimizers
In [3]:
        yes=os.listdir('/content/Dataset kaggle/COVID')
        no=os.listdir('/content/Dataset kaggle/non-COVID')
In [4]: data=np.concatenate([yes,no])
        len(data)==len(yes)+len(no)
Out[4]: True
In [5]: target x=np.full(len(yes),1)
        target y=np.full(len(no),0)
        data_target=np.concatenate([target_x,target_y])
In [6]: yes values=os.listdir('/content/Dataset kaggle/COVID')
        no_values=os.listdir('/content/Dataset_kaggle/non-COVID')
In [7]: | X data =[]
        for file in yes values:
            img = cv2.imread('/content/Dataset_kaggle/COVID/'+file)
            face = cv2.resize(img, (224, 224))
            (b, g, r)=cv2.split(face)
            img=cv2.merge([r,g,b])
            X data.append(img)
```

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In [8]: for file in no values:
              img = cv2.imread('/content/Dataset_kaggle/non-COVID/'+file)
             face = cv2.resize(img, (224, 224) )
              (b, g, r)=cv2.split(face)
              img=cv2.merge([r,g,b])
             X_data.append(img)
 In [9]: | X = np.squeeze(X data)
In [10]: X = X.astype('float32')
         X /= 255
In [11]: train_size = None # number of samples for training
         test_size = None # number of samples for testing
In [12]: x_train,x_test,y_train,y_test=train_test_split(X, data_target, test_size=0.2, rain)
         if train_size:
             x_train = x_train[:train_size]
             y_train = y_train[:train_size]
         if test_size:
             x_test = x_test[:test_size]
             y_test = y_test[:test_size]
         print(x_train.shape, x_test.shape)
         (1984, 224, 224, 3) (497, 224, 224, 3)
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In [13]: mobile_net_model= MobileNet(weights='imagenet', include_top=False, input_shape=()
    mobile_net_model.trainable = False

model =tf.keras.Sequential()
    model.add(mobile_net_model)
    model.add(Flatten())
    model.add(Dense(128, activation='relu'))
    model.add(Dropout(0.2))
    model.add(Dense(64, activation='relu'))
    model.add(Dropout(0.1))
    model.add(Dense(2, activation='softmax'))
    model.summary()
```

## Model: "sequential"

Layer (type)	Output Shape	Param #
mobilenet_1.00_224 (Model)	(None, 7, 7, 1024)	3228864
flatten (Flatten)	(None, 50176)	0
dense (Dense)	(None, 128)	6422656
dropout (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 64)	8256
dropout_1 (Dropout)	(None, 64)	0
dense_2 (Dense)	(None, 2)	130

Total params: 9,659,906 Trainable params: 6,431,042 Non-trainable params: 3,228,864

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In [21]: history = model.fit(
      x train,
      y_train,
      epochs=20,
      batch size=100,
      validation_data=(x_test, y_test),
      verbose=1)
    Epoch 1/20
    8725 - val loss: 0.3426 - val acc: 0.9296
    9254 - val loss: 0.2068 - val acc: 0.9396
    Epoch 3/20
    9556 - val loss: 0.1630 - val acc: 0.9477
    Epoch 4/20
    9723 - val loss: 0.1241 - val acc: 0.9557
    Epoch 5/20
    9803 - val_loss: 0.1119 - val_acc: 0.9557
    Epoch 6/20
    9854 - val loss: 0.1038 - val acc: 0.9537
    Epoch 7/20
    9899 - val loss: 0.1061 - val acc: 0.9557
    Epoch 8/20
    9894 - val loss: 0.1072 - val acc: 0.9577
    Epoch 9/20
    9924 - val_loss: 0.1088 - val_acc: 0.9678
    Epoch 10/20
    9945 - val loss: 0.1135 - val acc: 0.9638
    Epoch 11/20
    9904 - val loss: 0.1354 - val acc: 0.9638
    Epoch 12/20
    9940 - val loss: 0.1084 - val acc: 0.9658
    Epoch 13/20
    9960 - val_loss: 0.1291 - val_acc: 0.9638
    Epoch 14/20
    20/20 [============= ] - 5s 259ms/step - loss: 0.0189 - acc: 0.
    9940 - val_loss: 0.1583 - val_acc: 0.9618
    Epoch 15/20
    9950 - val loss: 0.1118 - val acc: 0.9658
    Epoch 16/20
    9975 - val_loss: 0.1125 - val_acc: 0.9678
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Epoch 17/20
     9980 - val_loss: 0.1154 - val_acc: 0.9658
     Epoch 18/20
     20/20 [============= ] - 5s 258ms/step - loss: 0.0050 - acc: 0.
     9995 - val_loss: 0.1394 - val_acc: 0.9658
     Epoch 19/20
     9980 - val_loss: 0.1287 - val_acc: 0.9678
     Epoch 20/20
     9980 - val_loss: 0.1655 - val_acc: 0.9638
In [22]:
     final_loss, final_acc = model.evaluate(x_test, y_test, verbose=0)
     print('The final accuracy is ',final_acc)
```

The final accuracy is 0.9637826681137085

In [ ]: