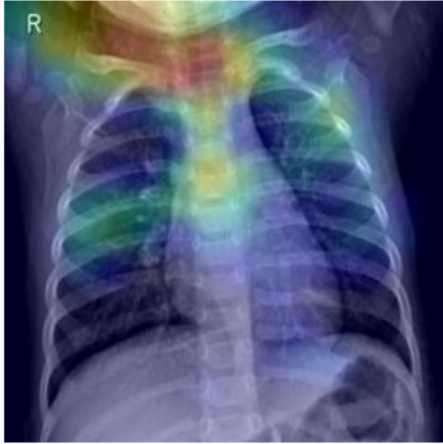

The given X-Ray image is of type = Normal

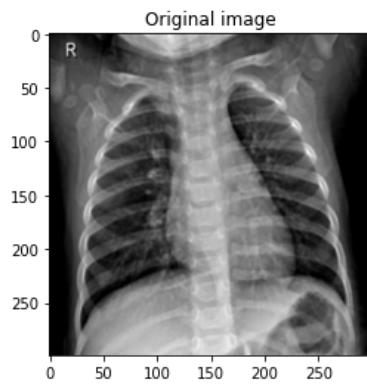
The chances of image being Covid is : 8.306345343589783 %

The chances of image being Normal is : 90.17686247825623 %

image with heatmap representing the covid spot



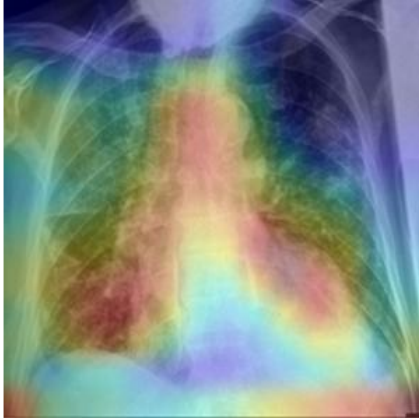
the original input image



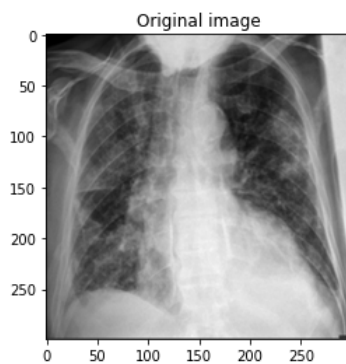
The given X-Ray image is of type = Covid

The chances of image being Covid is : 87.96427249908447 %
The chances of image being Normal is : 9.105725586414337 %

image with heatmap representing the covid spot



the original input image



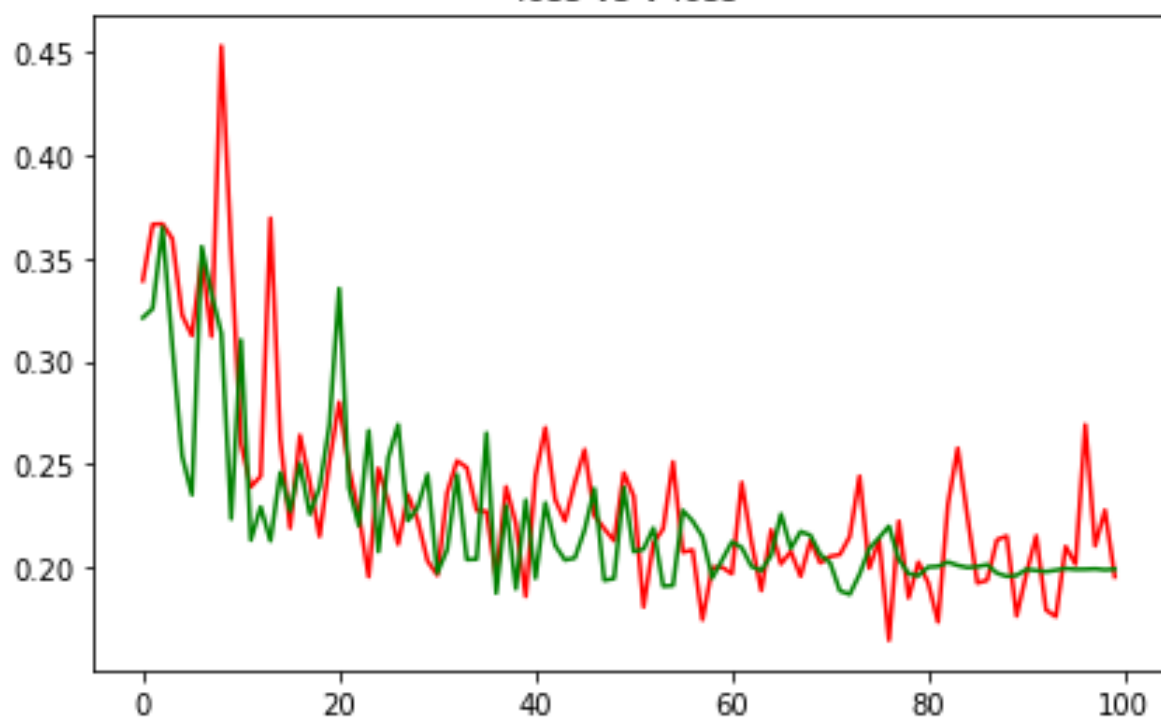
```
#accuracy
```

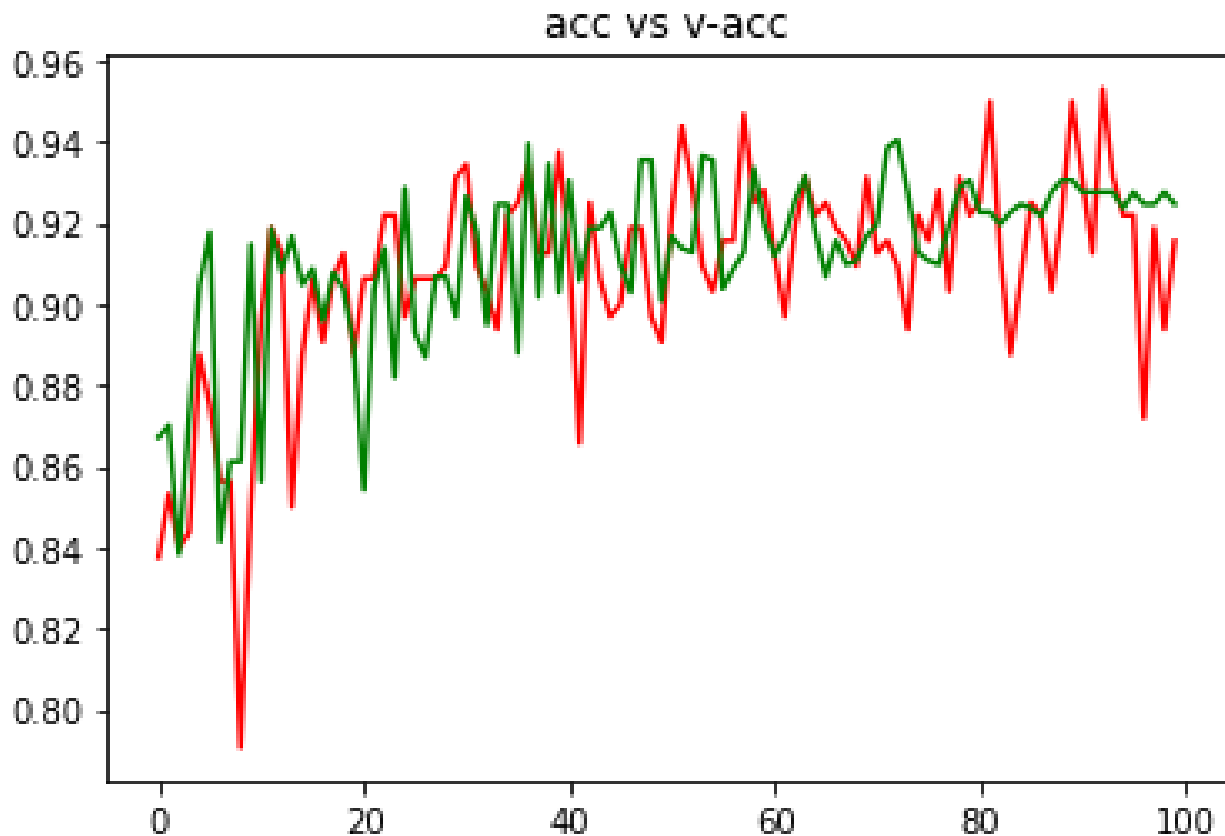
```
acc = model.evaluate_generator(generator= test)[1]
```

```
print(f"The accuracy of your model is = {acc} ")
```

```
The accuracy of your model is = 0.9237499833106995
```

loss vs v-loss





```

Epoch 1/100
10/10 [=====] - ETA: 0s - loss: 0.3392 - accuracy: 0.8375
Epoch 1: accuracy improved from -inf to 0.83750, saving model to bestmodel.h5
10/10 [=====] - 10s 1s/step - loss: 0.3392 - accuracy: 0.8375 - val_loss: 0.3211 - val_accuracy: 0.8671 - lr: 0.0010
Epoch 2/100
10/10 [=====] - ETA: 0s - loss: 0.3664 - accuracy: 0.8531
Epoch 2: accuracy improved from 0.83750 to 0.85312, saving model to bestmodel.h5
10/10 [=====] - 9s 951ms/step - loss: 0.3664 - accuracy: 0.8531 - val_loss: 0.3256 - val_accuracy: 0.8700 - lr: 0.0010
Epoch 3/100
10/10 [=====] - ETA: 0s - loss: 0.3666 - accuracy: 0.8397
Epoch 3: accuracy did not improve from 0.85312
10/10 [=====] - 11s 1s/step - loss: 0.3666 - accuracy: 0.8397 - val_loss: 0.3649 - val_accuracy: 0.8383 - lr: 0.0010
Epoch 4/100
10/10 [=====] - ETA: 0s - loss: 0.3595 - accuracy: 0.8438
Epoch 4: accuracy did not improve from 0.85312
10/10 [=====] - 16s 2s/step - loss: 0.3595 - accuracy: 0.8438 - val_loss: 0.3079 - val_accuracy: 0.8730 - lr: 0.0010
Epoch 5/100
10/10 [=====] - ETA: 0s - loss: 0.3224 - accuracy: 0.8875
Epoch 5: accuracy improved from 0.85312 to 0.88750, saving model to bestmodel.h5

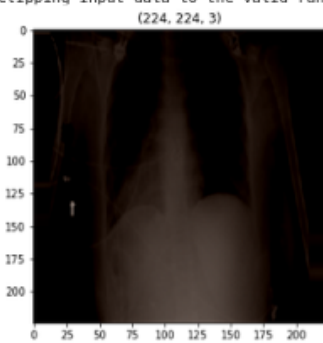
Epoch 98: accuracy did not improve from 0.95312
10/10 [=====] - 9s 980ms/step - loss: 0.2104 - accuracy: 0.9187 - val_loss: 0.1993 - val_accuracy: 0.9246 - lr: 3.3233e-06
Epoch 99/100
10/10 [=====] - ETA: 0s - loss: 0.2277 - accuracy: 0.8938
Epoch 99: accuracy did not improve from 0.95312
10/10 [=====] - 9s 977ms/step - loss: 0.2277 - accuracy: 0.8938 - val_loss: 0.1988 - val_accuracy: 0.9276 - lr: 2.3263e-06
Epoch 100/100
10/10 [=====] - ETA: 0s - loss: 0.1954 - accuracy: 0.9156
Epoch 100: accuracy did not improve from 0.95312
10/10 [=====] - 9s 980ms/step - loss: 0.1954 - accuracy: 0.9156 - val_loss: 0.1991 - val_accuracy: 0.9246 - lr: 2.3263e-06

```

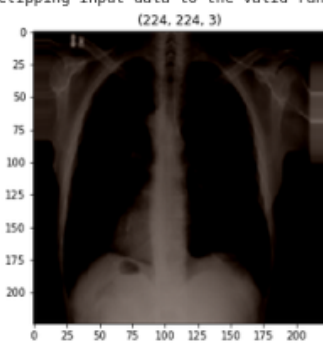
Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 224, 224, 3)]	0
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1180160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2359808
block4_conv3 (Conv2D)	(None, 28, 28, 512)	2359808
block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0
block5_conv1 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv2 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv3 (Conv2D)	(None, 14, 14, 512)	2359808
block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0
flatten (Flatten)	(None, 25088)	0
predictions (Dense)	(None, 2)	50178
=====		
Total params: 14,764,866		
Trainable params: 50,178		
Non-trainable params: 14,714,688		

Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).



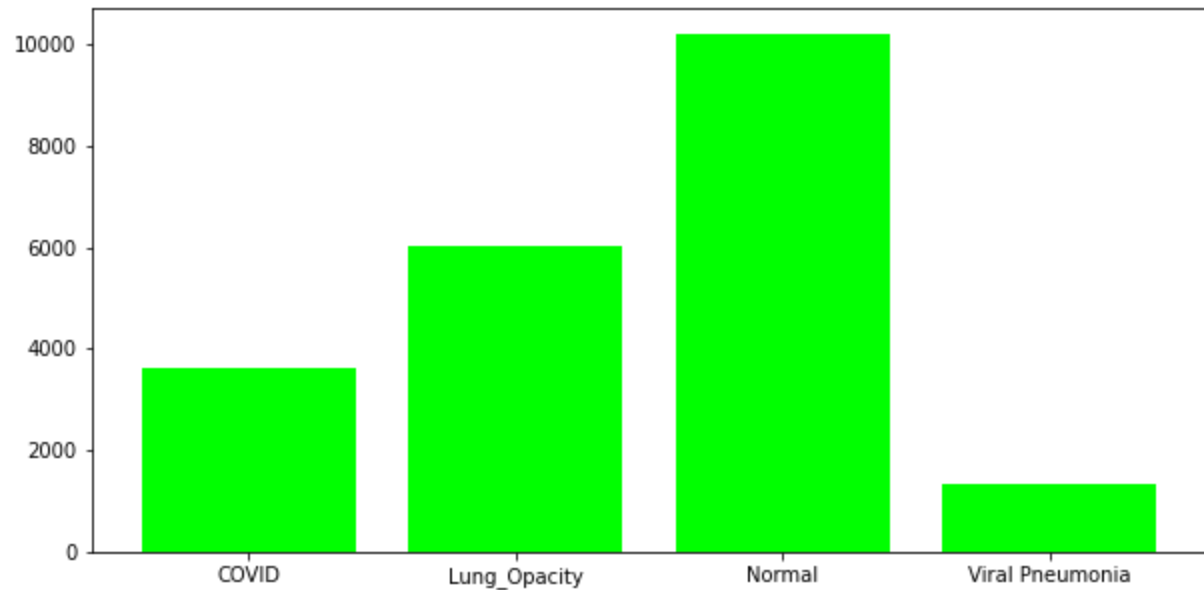
Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).



Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

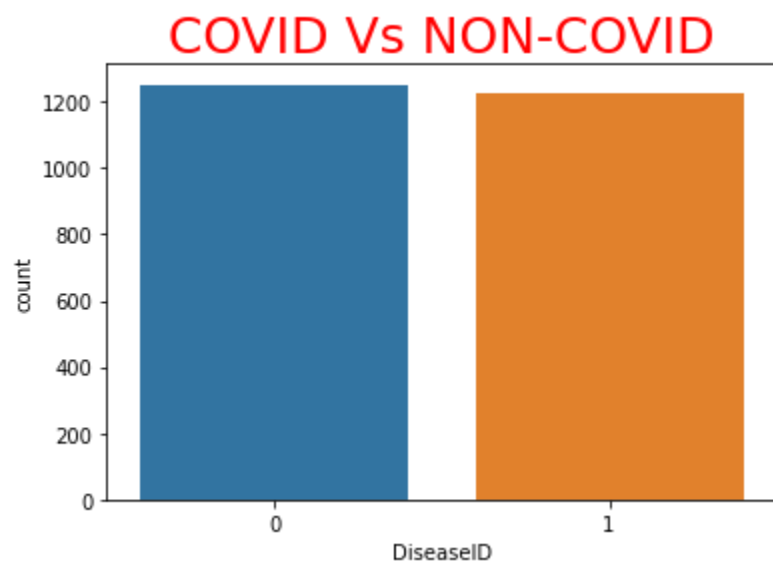


NUMBER OF IMAGES IN EACH CATEGORY



	File	DiseaseID	Disease Type
0	COVID/Covid (1073).png	0	COVID
1	COVID/Covid (593).png	0	COVID
2	COVID/Covid (768).png	0	COVID
3	COVID/Covid (797).png	0	COVID
4	COVID/Covid (207).png	0	COVID
5	COVID/Covid (1225).png	0	COVID
6	COVID/Covid (876).png	0	COVID
7	COVID/Covid (400).png	0	COVID
8	COVID/Covid (408).png	0	COVID
9	COVID/Covid (127).png	0	COVID
10	COVID/Covid (869).png	0	COVID
11	COVID/Covid (523).png	0	COVID
12	COVID/Covid (156).png	0	COVID
13	COVID/Covid (64).png	0	COVID
14	COVID/Covid (569).png	0	COVID
15	COVID/Covid (341).png	0	COVID
16	COVID/Covid (1099).png	0	COVID
17	COVID/Covid (904).png	0	COVID
18	COVID/Covid (1119).png	0	COVID
19	COVID/Covid (956).png	0	COVID

	File	DiseaseID	Disease Type
0	COVID/Covid (986).png	0	COVID
1	COVID/Covid (49).png	0	COVID
2	non-COVID/Non-Covid (1074).png	1	non-COVID
3	non-COVID/Non-Covid (537).png	1	non-COVID
4	COVID/Covid (1251).png	0	COVID
5	non-COVID/Non-Covid (444).png	1	non-COVID
6	COVID/Covid (242).png	0	COVID
7	non-COVID/Non-Covid (543).png	1	non-COVID
8	non-COVID/Non-Covid (436).png	1	non-COVID
9	COVID/Covid (1064).png	0	COVID
10	COVID/Covid (391).png	0	COVID
11	COVID/Covid (1163).png	0	COVID
12	non-COVID/Non-Covid (161).png	1	non-COVID
13	non-COVID/Non-Covid (971).png	1	non-COVID
14	COVID/Covid (365).png	0	COVID
15	COVID/Covid (500).png	0	COVID
16	COVID/Covid (1074).png	0	COVID
17	COVID/Covid (706).png	0	COVID
18	non-COVID/Non-Covid (921).png	1	non-COVID
19	non-COVID/Non-Covid (385).png	1	non-COVID

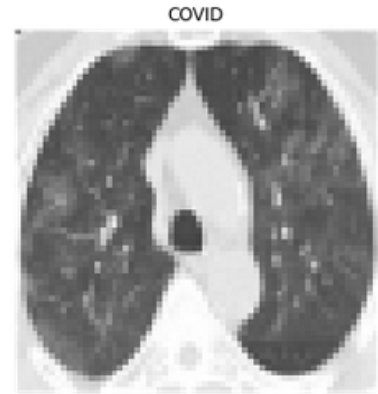
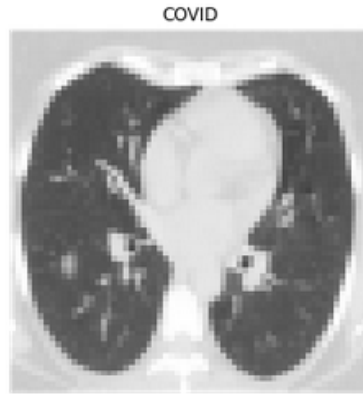
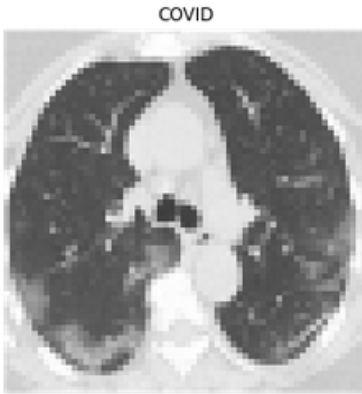


COVID



NON-COVID





Model: "model"

Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, 64, 64, 3)]	0
conv2d (Conv2D)	(None, 64, 64, 3)	84
resnet50 (Functional)	(None, None, None, 2048)	23587712
global_average_pooling2d (Gl	(None, 2048)	0
batch_normalization (BatchNo	(None, 2048)	8192
dropout (Dropout)	(None, 2048)	0
dense (Dense)	(None, 256)	524544
batch_normalization_1 (Batch	(None, 256)	1024
dropout_1 (Dropout)	(None, 256)	0
root (Dense)	(None, 2)	514

Total params: 24,122,070
Trainable params: 24,064,342
Non-trainable params: 57,728

31/31 [=====] - 57s 2s/step - loss: 0.2295 - accuracy: 0.9062 - val_loss: 0.2196 - val_accuracy: 0.9135

Epoch 00147: val_loss did not improve from 0.20534

Epoch 148/150

31/31 [=====] - 57s 2s/step - loss: 0.2514 - accuracy: 0.8992 - val_loss: 0.2190 - val_accuracy: 0.9135

Epoch 00148: val_loss did not improve from 0.20534

Epoch 149/150

31/31 [=====] - 57s 2s/step - loss: 0.2516 - accuracy: 0.8987 - val_loss: 0.2210 - val_accuracy: 0.9135

Epoch 00149: val_loss did not improve from 0.20534

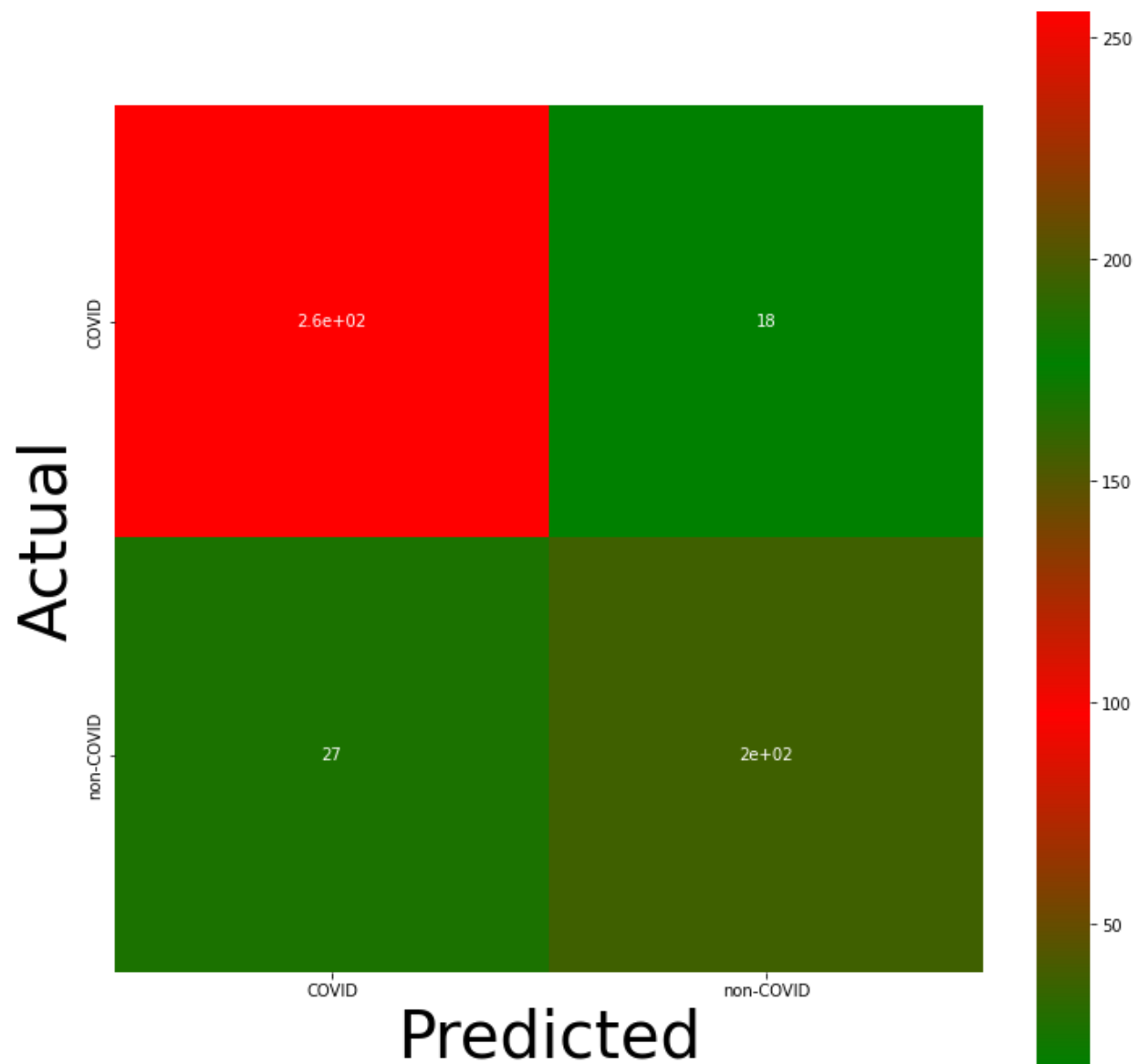
Epoch 150/150

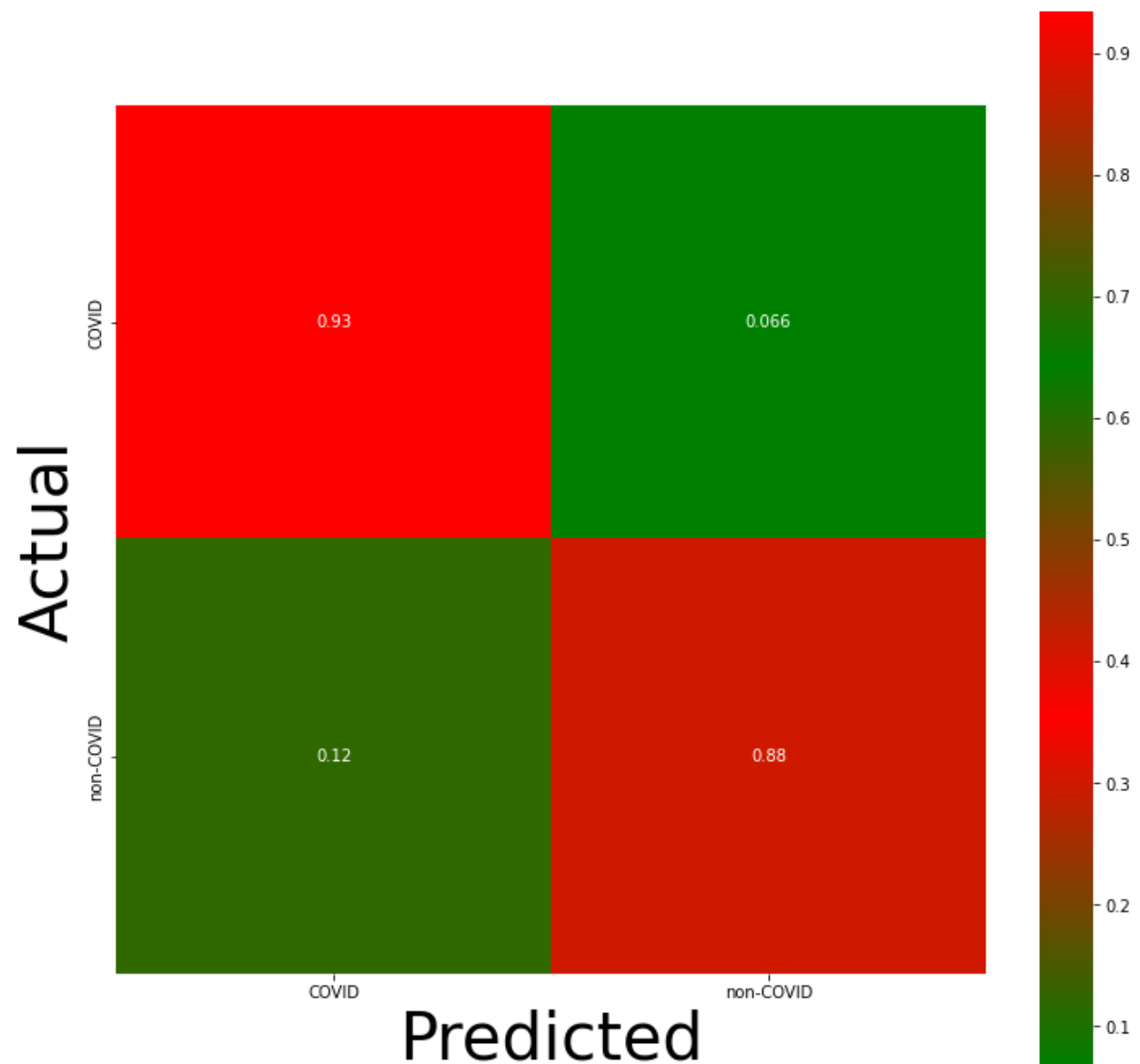
31/31 [=====] - 57s 2s/step - loss: 0.2476 - accuracy: 0.9002 - val_loss: 0.2215 - val_accuracy: 0.9095

Epoch 00150: val_loss did not improve from 0.20534

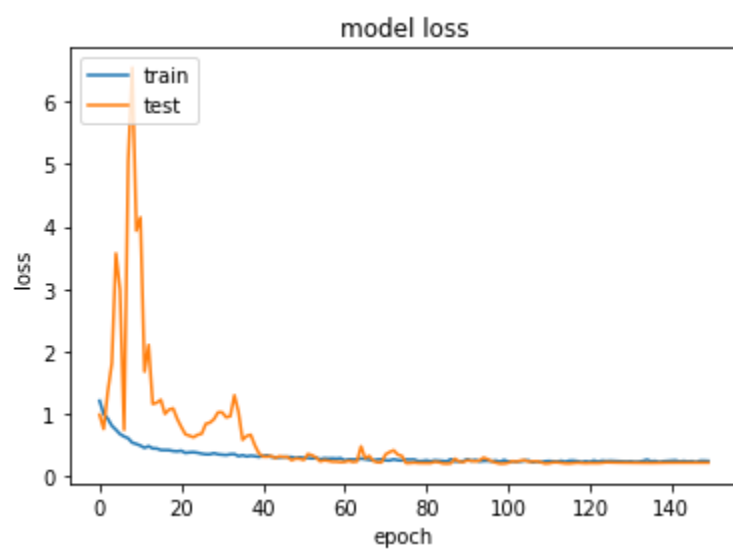
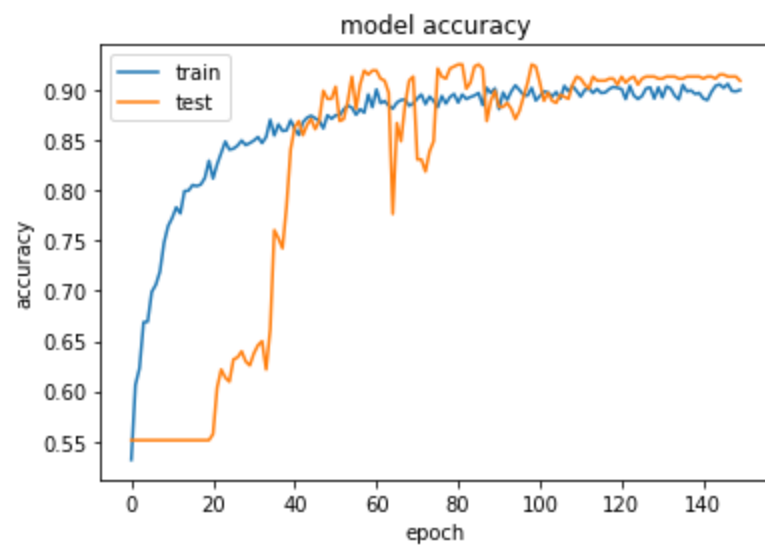
16/16 [=====] - 1s 82ms/step - loss: 0.2215 - accuracy: 0.9095

Final Loss: 0.22152909636497498, Final Accuracy: 0.909456729888916

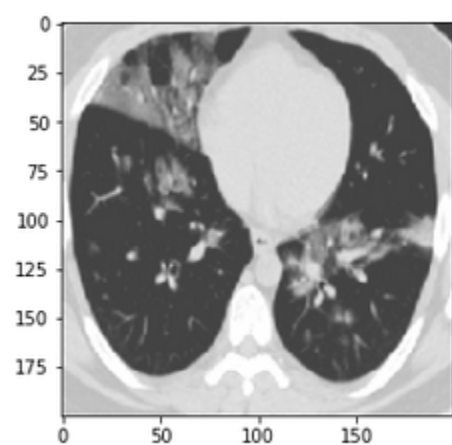




	precision	recall	f1-score	support
0	0.90	0.93	0.92	274
1	0.92	0.88	0.90	223
accuracy			0.91	497
macro avg	0.91	0.91	0.91	497
weighted avg	0.91	0.91	0.91	497

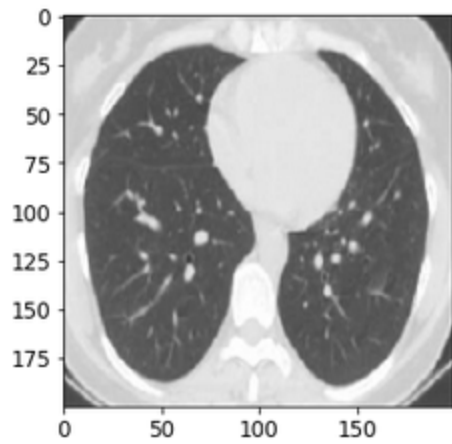


[0.99696785 0.00303217]



Prediction: Covid-19

[0.11544413 0.8845559]



Prediction: Non Covid-19

