# data aug adam val 95 64

October 8, 2021

### 1 Covid Classifier Model

#### 1.0.1 Goals

Classify: - Normal CXR - Viral Pneumonia CXR - COVID CXR

### 1.1 Create Directories for Dataset

Separate the data to use later as generators.

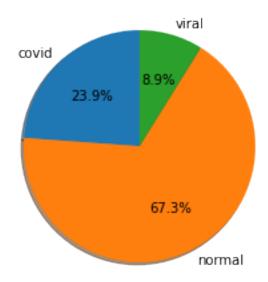
```
[]: # Matriz de confusion, cambiar learnings rates (learning rates dinamicos),
     → dropouts 0.3 & 0.2, Batch Normalization
     # K-Fold o avg de modelos
     import os
     BASE PATH = '/home/hivini/learn/research/new-covid'
     ORIGINAL_DATASET_DIR = os.path.join(BASE_PATH, 'COVID-19_Radiography_Dataset')
     ORIGINAL VIRAL DIR = os.path.join(ORIGINAL DATASET DIR, 'Viral Pneumonia')
     ORIGINAL_COVID_DIR = os.path.join(ORIGINAL_DATASET_DIR, 'COVID')
     ORIGINAL_NORMAL_DIR = os.path.join(ORIGINAL_DATASET_DIR, 'Normal')
     DATASET_DIR = os.path.join(BASE_PATH, 'small_dataset')
     TRAIN_DIR = os.path.join(DATASET_DIR, 'train')
     VALIDATION_DIR = os.path.join(DATASET_DIR, 'validation')
     TEST_DIR = os.path.join(DATASET_DIR, 'test')
     TRAIN_VIRAL_DIR = os.path.join(TRAIN_DIR, 'viral_pneumonia')
     TRAIN_COVID_DIR = os.path.join(TRAIN_DIR, 'covid')
     TRAIN_NORMAL_DIR = os.path.join(TRAIN_DIR, 'normal')
     VALIDATION_VIRAL_DIR = os.path.join(VALIDATION_DIR, 'viral_pneumonia')
     VALIDATION_COVID_DIR = os.path.join(VALIDATION_DIR, 'covid')
     VALIDATION_NORMAL_DIR = os.path.join(VALIDATION_DIR, 'normal')
     TEST_VIRAL_DIR = os.path.join(TEST_DIR, 'viral_pneumonia')
     TEST_COVID_DIR = os.path.join(TEST_DIR, 'covid')
     TEST NORMAL DIR = os.path.join(TEST DIR, 'normal')
     def createDir(path: str) -> None:
         if not os.path.exists(path):
             os.mkdir(path)
```

```
createDir(DATASET_DIR)
createDir(TRAIN_DIR)
createDir(VALIDATION_DIR)
createDir(TEST_DIR)
createDir(TRAIN_VIRAL_DIR)
createDir(TRAIN_COVID_DIR)
createDir(TRAIN_NORMAL_DIR)
createDir(VALIDATION_VIRAL_DIR)
createDir(VALIDATION_COVID_DIR)
createDir(VALIDATION_NORMAL_DIR)
createDir(TEST_VIRAL_DIR)
createDir(TEST_VIRAL_DIR)
createDir(TEST_COVID_DIR)
createDir(TEST_NORMAL_DIR)
```

```
[]: import numpy as np
     import shutil
     def generate_sets(source: str):
         allFiles = os.listdir(source)
         np.random.shuffle(allFiles)
         return np.split(np.array(allFiles), [int(len(allFiles)*0.7),__
      →int(len(allFiles)*0.85)])
     def saveAndSeparateFiles(src_dir: str, train_dir: str, val_dir: str, test_dir):
         train_fnames, val_fnames, test_fnames = generate_sets(src_dir)
         for fname in train_fnames:
             src = os.path.join(src_dir, fname)
             dst = os.path.join(train_dir, fname)
             shutil.copyfile(src, dst)
         for fname in val_fnames:
             src = os.path.join(src_dir, fname)
             dst = os.path.join(val_dir, fname)
             shutil.copyfile(src, dst)
         for fname in test_fnames:
             src = os.path.join(src_dir, fname)
             dst = os.path.join(test_dir, fname)
             shutil.copyfile(src, dst)
     create = False
     if create:
         saveAndSeparateFiles(ORIGINAL_NORMAL_DIR, TRAIN_NORMAL_DIR,
                             VALIDATION_NORMAL_DIR, TEST_NORMAL_DIR)
```

## 1.2 Counting our images

2021-10-07 23:07:23.838838: I tensorflow/stream\_executor/platform/default/dso\_loader.cc:49] Successfully opened dynamic library libcudart.so.10.1



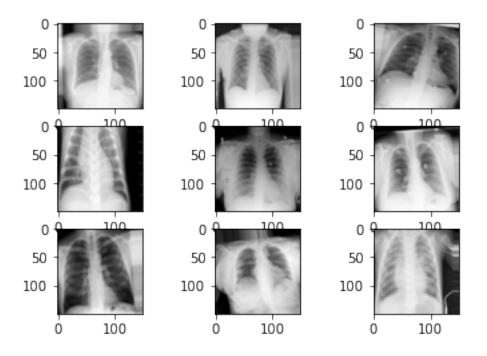
#### 1.3 Create our Covnet Model

In this case we are doing a multi class classification, our total clases are 3: - Viral CXR - Covid CXR - Normal CXR

Our neural network will output neurons as 3 classes that will calculate the probability of being one using the softmax function.

```
[]: from keras.preprocessing.image import ImageDataGenerator
     train_datagen = ImageDataGenerator(
        rescale=1./255,
        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 std of the dataset
        samplewise_std_normalization=False, # divide each input by its std
        zca_whitening=False, # apply ZCA whitening
         # randomly rotate images in the range (degrees, 0 to 180)
        rotation_range=10,
        zoom_range=0.1, # Randomly zoom image
         # randomly shift images horizontally (fraction of total width)
        width shift range=0.1,
         # randomly shift images vertically (fraction of total height)
        height_shift_range=0.1,
        horizontal_flip=False, # randomly flip images
        vertical_flip=False # randomly flip images
     )
     # train_datagen = ImageDataGenerator(rescale=1./255)
     test_datagen = ImageDataGenerator(rescale=1./255)
     evaluate_datagen = ImageDataGenerator(rescale=1./255)
     train_generator = train_datagen.flow_from_directory(
        TRAIN_DIR,
        target_size=(150, 150),
        batch size=32,
         class_mode='categorical',
         color mode='grayscale'
     )
     print(train_generator.class_indices)
     validation_generator = test_datagen.flow_from_directory(
        VALIDATION_DIR,
        target_size=(150, 150),
        batch_size=32,
        class_mode='categorical',
         color_mode='grayscale'
```

```
print(validation_generator.class_indices)
     test_generator = evaluate_datagen.flow_from_directory(
         TEST_DIR,
         target_size=(150, 150),
         batch_size=32,
         class_mode='categorical',
         color_mode='grayscale'
    print(test_generator.class_indices)
    Found 10606 images belonging to 3 classes.
    {'covid': 0, 'normal': 1, 'viral_pneumonia': 2}
    Found 2273 images belonging to 3 classes.
    {'covid': 0, 'normal': 1, 'viral_pneumonia': 2}
    Found 2274 images belonging to 3 classes.
    {'covid': 0, 'normal': 1, 'viral_pneumonia': 2}
[]: for X_batch, y_batch in train_generator:
             # create a grid of 3x3 images
             for i in range(0, 9):
                    plt.subplot(330 + 1 + i)
                     plt.imshow(X_batch[i].reshape(150, 150), cmap=plt.
     →get_cmap('gray'))
             # show the plot
             plt.show()
             break
```



```
[]: from keras.layers import Conv2D, BatchNormalization, MaxPooling2D, Dropout,
     \hookrightarrowFlatten, Dense
     from keras.models import Sequential
     from keras import backend
     backend.clear_session()
     model = Sequential()
     model.add(Conv2D(64, (3, 3), activation='relu', input_shape=(150, 150, 1)))
     model.add(BatchNormalization())
     model.add(MaxPooling2D((2, 2)))
     model.add(Conv2D(64, (3, 3), activation='relu'))
     model.add(BatchNormalization())
     model.add(MaxPooling2D((2, 2)))
     model.add(Conv2D(128, (3, 3), activation='relu'))
     model.add(BatchNormalization())
     model.add(MaxPooling2D((2, 2)))
     model.add(Conv2D(128, (3, 3), activation='relu'))
     model.add(BatchNormalization())
     model.add(MaxPooling2D((2, 2)))
     model.add(Flatten())
     model.add(Dropout(0.5))
     model.add(Dense(512, activation='relu'))
     model.add(Dense(64, activation='relu'))
     model.add(Dense(3, activation='softmax'))
```

# model.summary()

```
2021-10-07 23:07:28.256635: I tensorflow/compiler/jit/xla_cpu_device.cc:41] Not
creating XLA devices, tf_xla_enable_xla_devices not set
2021-10-07 23:07:28.265024: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcuda.so.1
2021-10-07 23:07:28.654163: E
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:927] could not open file to
read NUMA node: /sys/bus/pci/devices/0000:01:00.0/numa_node
Your kernel may have been built without NUMA support.
2021-10-07 23:07:28.654442: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1720] Found device 0 with
properties:
pciBusID: 0000:01:00.0 name: NVIDIA GeForce RTX 2080 with Max-Q Design
computeCapability: 7.5
coreClock: 1.215GHz coreCount: 46 deviceMemorySize: 8.00GiB
deviceMemoryBandwidth: 357.69GiB/s
2021-10-07 23:07:28.654504: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudart.so.10.1
2021-10-07 23:07:28.700452: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcublas.so.10
2021-10-07 23:07:28.700542: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcublasLt.so.10
2021-10-07 23:07:28.727976: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcufft.so.10
2021-10-07 23:07:28.734165: I
tensorflow/stream executor/platform/default/dso loader.cc:49] Successfully
opened dynamic library libcurand.so.10
2021-10-07 23:07:28.781358: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcusolver.so.10
2021-10-07 23:07:28.794976: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcusparse.so.10
2021-10-07 23:07:28.891048: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudnn.so.7
2021-10-07 23:07:28.891989: E
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:927] could not open file to
read NUMA node: /sys/bus/pci/devices/0000:01:00.0/numa_node
Your kernel may have been built without NUMA support.
2021-10-07 23:07:28.893316: E
```

```
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:927] could not open file to
read NUMA node: /sys/bus/pci/devices/0000:01:00.0/numa_node
Your kernel may have been built without NUMA support.
2021-10-07 23:07:28.893738: I
tensorflow/core/common runtime/gpu/gpu device.cc:1862] Adding visible gpu
devices: 0
2021-10-07 23:07:28.894586: I tensorflow/core/platform/cpu feature guard.cc:142]
This TensorFlow binary is optimized with oneAPI Deep Neural Network Library
(oneDNN) to use the following CPU instructions in performance-critical
operations: SSE4.1 SSE4.2 AVX AVX2 FMA
To enable them in other operations, rebuild TensorFlow with the appropriate
compiler flags.
2021-10-07 23:07:28.896693: E
tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:927] could not open file to
read NUMA node: /sys/bus/pci/devices/0000:01:00.0/numa_node
Your kernel may have been built without NUMA support.
2021-10-07 23:07:28.897067: I
tensorflow/core/common runtime/gpu/gpu device.cc:1720] Found device 0 with
properties:
pciBusID: 0000:01:00.0 name: NVIDIA GeForce RTX 2080 with Max-Q Design
computeCapability: 7.5
coreClock: 1.215GHz coreCount: 46 deviceMemorySize: 8.00GiB
deviceMemoryBandwidth: 357.69GiB/s
2021-10-07 23:07:28.897118: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudart.so.10.1
2021-10-07 23:07:28.897162: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcublas.so.10
2021-10-07 23:07:28.897176: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcublasLt.so.10
2021-10-07 23:07:28.897187: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcufft.so.10
2021-10-07 23:07:28.897198: I
tensorflow/stream executor/platform/default/dso loader.cc:49] Successfully
opened dynamic library libcurand.so.10
2021-10-07 23:07:28.897209: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcusolver.so.10
2021-10-07 23:07:28.897221: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcusparse.so.10
2021-10-07 23:07:28.897231: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudnn.so.7
2021-10-07 23:07:28.898158: E
```

tensorflow/stream\_executor/cuda/cuda\_gpu\_executor.cc:927] could not open file to read NUMA node: /sys/bus/pci/devices/0000:01:00.0/numa\_node Your kernel may have been built without NUMA support. 2021-10-07 23:07:28.899312: E tensorflow/stream executor/cuda/cuda gpu executor.cc:927] could not open file to read NUMA node: /sys/bus/pci/devices/0000:01:00.0/numa\_node Your kernel may have been built without NUMA support. 2021-10-07 23:07:28.899742: I tensorflow/core/common\_runtime/gpu/gpu\_device.cc:1862] Adding visible gpu devices: 0 2021-10-07 23:07:28.900428: I tensorflow/stream\_executor/platform/default/dso\_loader.cc:49] Successfully opened dynamic library libcudart.so.10.1 2021-10-07 23:07:30.673375: I tensorflow/core/common\_runtime/gpu/gpu\_device.cc:1261] Device interconnect StreamExecutor with strength 1 edge matrix: 2021-10-07 23:07:30.673401: I tensorflow/core/common\_runtime/gpu/gpu\_device.cc:1267] 2021-10-07 23:07:30.673407: I tensorflow/core/common\_runtime/gpu/gpu\_device.cc:1280] 0: 2021-10-07 23:07:30.675516: E tensorflow/stream\_executor/cuda/cuda\_gpu\_executor.cc:927] could not open file to read NUMA node: /sys/bus/pci/devices/0000:01:00.0/numa\_node Your kernel may have been built without NUMA support. 2021-10-07 23:07:30.675875: I tensorflow/core/common runtime/gpu/gpu\_device.cc:1489] Could not identify NUMA node of platform GPU id 0, defaulting to 0. Your kernel may not have been built with NUMA support. 2021-10-07 23:07:30.676731: E tensorflow/stream\_executor/cuda/cuda\_gpu\_executor.cc:927] could not open file to read NUMA node: /sys/bus/pci/devices/0000:01:00.0/numa\_node Your kernel may have been built without NUMA support. 2021-10-07 23:07:30.677718: E tensorflow/stream\_executor/cuda/cuda\_gpu\_executor.cc:927] could not open file to read NUMA node: /sys/bus/pci/devices/0000:01:00.0/numa node Your kernel may have been built without NUMA support. 2021-10-07 23:07:30.678051: I tensorflow/core/common\_runtime/gpu/gpu\_device.cc:1406] Created TensorFlow device (/job:localhost/replica:0/task:0/device:GPU:0 with 6575 MB memory) -> physical GPU (device: 0, name: NVIDIA GeForce RTX 2080 with Max-Q Design, pci bus id: 0000:01:00.0, compute capability: 7.5) 2021-10-07 23:07:30.681952: I tensorflow/compiler/jit/xla\_gpu\_device.cc:99] Not creating XLA devices, tf\_xla\_enable\_xla\_devices not set Model: "sequential"

9

Param #

Output Shape

\_\_\_\_\_\_

Layer (type)

conv2d (Conv2D)	(None, 148,	148, 64)	640
batch_normalization (BatchNo	(None, 148,	148, 64)	256
max_pooling2d (MaxPooling2D)	(None, 74,	74, 64)	0
conv2d_1 (Conv2D)	(None, 72,	72, 64)	36928
batch_normalization_1 (Batch	(None, 72,	72, 64)	256
max_pooling2d_1 (MaxPooling2	(None, 36,	36, 64)	0
conv2d_2 (Conv2D)	(None, 34,	34, 128)	73856
batch_normalization_2 (Batch	(None, 34,	34, 128)	512
max_pooling2d_2 (MaxPooling2	(None, 17,	17, 128)	0
conv2d_3 (Conv2D)	(None, 15,	15, 128)	147584
batch_normalization_3 (Batch	(None, 15,	15, 128)	512
max_pooling2d_3 (MaxPooling2	(None, 7, 7	7, 128)	0
flatten (Flatten)	(None, 6272	2)	0
dropout (Dropout)	(None, 6272	2)	0
dense (Dense)	(None, 512)	,	3211776
dense_1 (Dense)	(None, 64)		32832
dense_2 (Dense)	(None, 3)		195

Total params: 3,505,347 Trainable params: 3,504,579 Non-trainable params: 768

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```
[]: import numpy as np
     from sklearn.utils import class_weight
     from keras.callbacks import EarlyStopping
     from keras.callbacks import ModelCheckpoint
     classes = train_generator.classes
     class_weights = class_weight.compute_class_weight(None,
                                                      np.unique(classes),
                                                      classes)
     best_model_path = os.path.join(BASE_PATH, 'best_model.h5')
     es = EarlyStopping(monitor='val loss', mode='min', verbose=1, patience=20)
     mc = ModelCheckpoint(best_model_path, monitor='val_accuracy', mode='max',_
     →verbose=1, save_best_only=True)
     history = model.fit(
         train_generator,
         steps_per_epoch=train_generator.n // 32,
         epochs=150,
         validation_data=validation_generator,
         class_weight=dict(zip(np.unique(classes), class_weights)),
         callbacks=[es, mc]
     )
    /home/hivini/anaconda3/envs/tf-gpu/lib/python3.9/site-
    packages/sklearn/utils/validation.py:67: FutureWarning: Pass classes=[0 1 2],
    y=[0 0 0 ... 2 2 2] as keyword args. From version 0.25 passing these as
    positional arguments will result in an error
      warnings.warn("Pass {} as keyword args. From version 0.25 "
    2021-10-07 23:07:31.660305: I
    tensorflow/compiler/mlir/mlir_graph_optimization_pass.cc:116] None of the MLIR
    optimization passes are enabled (registered 2)
    2021-10-07 23:07:31.662515: I
    tensorflow/core/platform/profile utils/cpu utils.cc:112] CPU Frequency:
    2208005000 Hz
    Epoch 1/150
    2021-10-07 23:07:32.451009: I
    tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
    opened dynamic library libcublas.so.10
    2021-10-07 23:07:32.969887: I
    tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
    opened dynamic library libcudnn.so.7
    2021-10-07 23:07:35.163484: W tensorflow/stream_executor/gpu/asm_compiler.cc:63]
    Running ptxas --version returned 256
    2021-10-07 23:07:35.294879: W
    tensorflow/stream executor/gpu/redzone allocator.cc:314] Internal: ptxas exited
    with non-zero error code 256, output:
    Relying on driver to perform ptx compilation.
```

```
Modify $PATH to customize ptxas location.
This message will be only logged once.
accuracy: 0.6052 - val_loss: 1.9275 - val_accuracy: 0.2389
Epoch 00001: val_accuracy improved from -inf to 0.23889, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 2/150
331/331 [=========== ] - 32s 98ms/step - loss: 0.6338 -
accuracy: 0.7207 - val_loss: 0.4280 - val_accuracy: 0.8196
Epoch 00002: val_accuracy improved from 0.23889 to 0.81962, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 3/150
331/331 [============ ] - 32s 97ms/step - loss: 0.5397 -
accuracy: 0.7630 - val_loss: 0.4066 - val_accuracy: 0.8245
Epoch 00003: val_accuracy improved from 0.81962 to 0.82446, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 4/150
accuracy: 0.7832 - val_loss: 0.3740 - val_accuracy: 0.8412
Epoch 00004: val_accuracy improved from 0.82446 to 0.84118, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 5/150
331/331 [============= ] - 32s 97ms/step - loss: 0.4667 -
accuracy: 0.7959 - val_loss: 0.3426 - val_accuracy: 0.8601
Epoch 00005: val_accuracy improved from 0.84118 to 0.86010, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 6/150
331/331 [============= ] - 32s 98ms/step - loss: 0.4236 -
accuracy: 0.8133 - val_loss: 0.3115 - val_accuracy: 0.8715
Epoch 00006: val_accuracy improved from 0.86010 to 0.87154, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 7/150
331/331 [============= ] - 33s 100ms/step - loss: 0.4120 -
accuracy: 0.8213 - val_loss: 0.2977 - val_accuracy: 0.8759
Epoch 00007: val_accuracy improved from 0.87154 to 0.87593, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 8/150
accuracy: 0.8411 - val_loss: 0.2642 - val_accuracy: 0.8927
```

```
Epoch 00008: val_accuracy improved from 0.87593 to 0.89265, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 9/150
accuracy: 0.8509 - val_loss: 0.2773 - val_accuracy: 0.8847
Epoch 00009: val accuracy did not improve from 0.89265
Epoch 10/150
331/331 [============== ] - 33s 100ms/step - loss: 0.3359 -
accuracy: 0.8618 - val_loss: 0.2780 - val_accuracy: 0.8847
Epoch 00010: val_accuracy did not improve from 0.89265
Epoch 11/150
accuracy: 0.8637 - val_loss: 0.2589 - val_accuracy: 0.8883
Epoch 00011: val_accuracy did not improve from 0.89265
Epoch 12/150
accuracy: 0.8742 - val_loss: 0.2393 - val_accuracy: 0.8997
Epoch 00012: val accuracy improved from 0.89265 to 0.89969, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 13/150
accuracy: 0.8746 - val_loss: 0.2263 - val_accuracy: 0.9072
Epoch 00013: val_accuracy improved from 0.89969 to 0.90717, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 14/150
accuracy: 0.8814 - val_loss: 0.2378 - val_accuracy: 0.9041
Epoch 00014: val_accuracy did not improve from 0.90717
Epoch 15/150
accuracy: 0.8827 - val_loss: 0.2155 - val_accuracy: 0.9142
Epoch 00015: val_accuracy improved from 0.90717 to 0.91421, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 16/150
accuracy: 0.8903 - val_loss: 0.2318 - val_accuracy: 0.8997
Epoch 00016: val_accuracy did not improve from 0.91421
Epoch 17/150
accuracy: 0.8964 - val_loss: 0.2304 - val_accuracy: 0.9037
```

```
Epoch 00017: val_accuracy did not improve from 0.91421
Epoch 18/150
331/331 [============ ] - 33s 100ms/step - loss: 0.2742 -
accuracy: 0.8893 - val_loss: 0.2675 - val_accuracy: 0.8883
Epoch 00018: val accuracy did not improve from 0.91421
Epoch 19/150
accuracy: 0.8918 - val_loss: 0.1833 - val_accuracy: 0.9234
Epoch 00019: val_accuracy improved from 0.91421 to 0.92345, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 20/150
331/331 [============== ] - 33s 101ms/step - loss: 0.2544 -
accuracy: 0.8967 - val_loss: 0.2196 - val_accuracy: 0.9085
Epoch 00020: val_accuracy did not improve from 0.92345
Epoch 21/150
accuracy: 0.8976 - val_loss: 0.1889 - val_accuracy: 0.9256
Epoch 00021: val_accuracy improved from 0.92345 to 0.92565, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 22/150
accuracy: 0.9094 - val_loss: 0.2026 - val_accuracy: 0.9173
Epoch 00022: val_accuracy did not improve from 0.92565
Epoch 23/150
accuracy: 0.9131 - val_loss: 0.1975 - val_accuracy: 0.9234
Epoch 00023: val_accuracy did not improve from 0.92565
Epoch 24/150
accuracy: 0.9109 - val_loss: 0.1721 - val_accuracy: 0.9309
Epoch 00024: val_accuracy improved from 0.92565 to 0.93093, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 25/150
accuracy: 0.9112 - val_loss: 0.2292 - val_accuracy: 0.9107
Epoch 00025: val_accuracy did not improve from 0.93093
Epoch 26/150
accuracy: 0.9157 - val_loss: 0.1595 - val_accuracy: 0.9358
```

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Epoch 00026: val_accuracy improved from 0.93093 to 0.93577, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 27/150
accuracy: 0.9155 - val_loss: 0.1712 - val_accuracy: 0.9322
Epoch 00027: val_accuracy did not improve from 0.93577
Epoch 28/150
accuracy: 0.9151 - val_loss: 0.1669 - val_accuracy: 0.9380
Epoch 00028: val_accuracy improved from 0.93577 to 0.93797, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 29/150
accuracy: 0.9181 - val_loss: 0.1617 - val_accuracy: 0.9393
Epoch 00029: val_accuracy improved from 0.93797 to 0.93929, saving model to
/home/hivini/learn/research/new-covid/best model.h5
Epoch 30/150
accuracy: 0.9247 - val_loss: 0.1571 - val_accuracy: 0.9349
Epoch 00030: val_accuracy did not improve from 0.93929
Epoch 31/150
accuracy: 0.9205 - val_loss: 0.1782 - val_accuracy: 0.9287
Epoch 00031: val_accuracy did not improve from 0.93929
Epoch 32/150
accuracy: 0.9233 - val_loss: 0.1520 - val_accuracy: 0.9406
Epoch 00032: val accuracy improved from 0.93929 to 0.94061, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 33/150
accuracy: 0.9200 - val_loss: 0.1519 - val_accuracy: 0.9362
Epoch 00033: val_accuracy did not improve from 0.94061
Epoch 34/150
accuracy: 0.9246 - val_loss: 0.1860 - val_accuracy: 0.9296
Epoch 00034: val_accuracy did not improve from 0.94061
Epoch 35/150
331/331 [============== ] - 33s 101ms/step - loss: 0.1840 -
```

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accuracy: 0.9289 - val_loss: 0.1591 - val_accuracy: 0.9402
Epoch 00035: val_accuracy did not improve from 0.94061
Epoch 36/150
accuracy: 0.9332 - val_loss: 0.1579 - val_accuracy: 0.9419
Epoch 00036: val_accuracy improved from 0.94061 to 0.94193, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 37/150
accuracy: 0.9319 - val_loss: 0.1439 - val_accuracy: 0.9424
Epoch 00037: val_accuracy improved from 0.94193 to 0.94237, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 38/150
accuracy: 0.9369 - val_loss: 0.1482 - val_accuracy: 0.9450
Epoch 00038: val accuracy improved from 0.94237 to 0.94501, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 39/150
accuracy: 0.9354 - val_loss: 0.2238 - val_accuracy: 0.9138
Epoch 00039: val_accuracy did not improve from 0.94501
Epoch 40/150
331/331 [============= ] - 33s 100ms/step - loss: 0.1864 -
accuracy: 0.9284 - val_loss: 0.1533 - val_accuracy: 0.9406
Epoch 00040: val_accuracy did not improve from 0.94501
Epoch 41/150
accuracy: 0.9400 - val_loss: 0.1546 - val_accuracy: 0.9441
Epoch 00041: val_accuracy did not improve from 0.94501
Epoch 42/150
accuracy: 0.9402 - val_loss: 0.2114 - val_accuracy: 0.9212
Epoch 00042: val_accuracy did not improve from 0.94501
Epoch 43/150
331/331 [============== ] - 33s 101ms/step - loss: 0.1644 -
accuracy: 0.9378 - val_loss: 0.1511 - val_accuracy: 0.9384
Epoch 00043: val_accuracy did not improve from 0.94501
Epoch 44/150
```

```
accuracy: 0.9347 - val_loss: 0.1332 - val_accuracy: 0.9490
Epoch 00044: val_accuracy improved from 0.94501 to 0.94897, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 45/150
accuracy: 0.9439 - val_loss: 0.1533 - val_accuracy: 0.9432
Epoch 00045: val_accuracy did not improve from 0.94897
Epoch 46/150
accuracy: 0.9398 - val_loss: 0.1450 - val_accuracy: 0.9419
Epoch 00046: val_accuracy did not improve from 0.94897
Epoch 47/150
accuracy: 0.9500 - val_loss: 0.1610 - val_accuracy: 0.9380
Epoch 00047: val_accuracy did not improve from 0.94897
Epoch 48/150
accuracy: 0.9473 - val_loss: 0.1255 - val_accuracy: 0.9516
Epoch 00048: val_accuracy improved from 0.94897 to 0.95161, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 49/150
accuracy: 0.9434 - val_loss: 0.1383 - val_accuracy: 0.9485
Epoch 00049: val_accuracy did not improve from 0.95161
Epoch 50/150
accuracy: 0.9450 - val_loss: 0.1298 - val_accuracy: 0.9476
Epoch 00050: val_accuracy did not improve from 0.95161
Epoch 51/150
accuracy: 0.9463 - val_loss: 0.1594 - val_accuracy: 0.9393
Epoch 00051: val_accuracy did not improve from 0.95161
Epoch 52/150
331/331 [============ ] - 32s 97ms/step - loss: 0.1460 -
accuracy: 0.9485 - val_loss: 0.1443 - val_accuracy: 0.9481
Epoch 00052: val_accuracy did not improve from 0.95161
Epoch 53/150
331/331 [============= ] - 32s 97ms/step - loss: 0.1426 -
accuracy: 0.9475 - val_loss: 0.1195 - val_accuracy: 0.9569
```

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Epoch 00053: val_accuracy improved from 0.95161 to 0.95689, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 54/150
331/331 [============ ] - 32s 98ms/step - loss: 0.1472 -
accuracy: 0.9427 - val_loss: 0.1168 - val_accuracy: 0.9556
Epoch 00054: val_accuracy did not improve from 0.95689
Epoch 55/150
accuracy: 0.9546 - val_loss: 0.1139 - val_accuracy: 0.9569
Epoch 00055: val_accuracy did not improve from 0.95689
Epoch 56/150
331/331 [============ ] - 34s 103ms/step - loss: 0.1376 -
accuracy: 0.9489 - val_loss: 0.1210 - val_accuracy: 0.9534
Epoch 00056: val_accuracy did not improve from 0.95689
Epoch 57/150
accuracy: 0.9527 - val_loss: 0.1193 - val_accuracy: 0.9551
Epoch 00057: val_accuracy did not improve from 0.95689
Epoch 58/150
accuracy: 0.9536 - val_loss: 0.2349 - val_accuracy: 0.9173
Epoch 00058: val_accuracy did not improve from 0.95689
Epoch 59/150
accuracy: 0.9481 - val_loss: 0.1265 - val_accuracy: 0.9490
Epoch 00059: val_accuracy did not improve from 0.95689
Epoch 60/150
accuracy: 0.9489 - val_loss: 0.1598 - val_accuracy: 0.9406
Epoch 00060: val_accuracy did not improve from 0.95689
Epoch 61/150
accuracy: 0.9513 - val_loss: 0.1068 - val_accuracy: 0.9591
Epoch 00061: val_accuracy improved from 0.95689 to 0.95908, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 62/150
accuracy: 0.9542 - val_loss: 0.1414 - val_accuracy: 0.9454
```

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Epoch 00062: val_accuracy did not improve from 0.95908
Epoch 63/150
accuracy: 0.9546 - val_loss: 0.1711 - val_accuracy: 0.9380
Epoch 00063: val_accuracy did not improve from 0.95908
Epoch 64/150
accuracy: 0.9513 - val_loss: 0.1099 - val_accuracy: 0.9551
Epoch 00064: val_accuracy did not improve from 0.95908
Epoch 65/150
331/331 [============== ] - 33s 101ms/step - loss: 0.1276 -
accuracy: 0.9509 - val_loss: 0.1301 - val_accuracy: 0.9525
Epoch 00065: val_accuracy did not improve from 0.95908
Epoch 66/150
331/331 [============== ] - 33s 101ms/step - loss: 0.1224 -
accuracy: 0.9526 - val_loss: 0.1289 - val_accuracy: 0.9494
Epoch 00066: val_accuracy did not improve from 0.95908
Epoch 67/150
accuracy: 0.9558 - val_loss: 0.1364 - val_accuracy: 0.9520
Epoch 00067: val_accuracy did not improve from 0.95908
Epoch 68/150
accuracy: 0.9583 - val_loss: 0.1098 - val_accuracy: 0.9573
Epoch 00068: val_accuracy did not improve from 0.95908
Epoch 69/150
accuracy: 0.9561 - val_loss: 0.0971 - val_accuracy: 0.9670
Epoch 00069: val_accuracy improved from 0.95908 to 0.96700, saving model to
/home/hivini/learn/research/new-covid/best model.h5
Epoch 70/150
accuracy: 0.9519 - val_loss: 0.1164 - val_accuracy: 0.9547
Epoch 00070: val_accuracy did not improve from 0.96700
Epoch 71/150
accuracy: 0.9568 - val_loss: 0.1377 - val_accuracy: 0.9494
Epoch 00071: val_accuracy did not improve from 0.96700
Epoch 72/150
```

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accuracy: 0.9612 - val_loss: 0.1377 - val_accuracy: 0.9485
Epoch 00072: val_accuracy did not improve from 0.96700
Epoch 73/150
accuracy: 0.9566 - val_loss: 0.1178 - val_accuracy: 0.9573
Epoch 00073: val_accuracy did not improve from 0.96700
Epoch 74/150
accuracy: 0.9554 - val_loss: 0.1238 - val_accuracy: 0.9534
Epoch 00074: val_accuracy did not improve from 0.96700
Epoch 75/150
accuracy: 0.9564 - val_loss: 0.1235 - val_accuracy: 0.9534
Epoch 00075: val_accuracy did not improve from 0.96700
Epoch 76/150
accuracy: 0.9594 - val_loss: 0.0999 - val_accuracy: 0.9626
Epoch 00076: val_accuracy did not improve from 0.96700
Epoch 77/150
accuracy: 0.9576 - val_loss: 0.1245 - val_accuracy: 0.9551
Epoch 00077: val_accuracy did not improve from 0.96700
Epoch 78/150
accuracy: 0.9568 - val_loss: 0.1269 - val_accuracy: 0.9516
Epoch 00078: val_accuracy did not improve from 0.96700
Epoch 79/150
accuracy: 0.9601 - val_loss: 0.1149 - val_accuracy: 0.9542
Epoch 00079: val_accuracy did not improve from 0.96700
Epoch 80/150
accuracy: 0.9603 - val_loss: 0.1155 - val_accuracy: 0.9560
Epoch 00080: val_accuracy did not improve from 0.96700
Epoch 81/150
accuracy: 0.9627 - val_loss: 0.1351 - val_accuracy: 0.9516
```

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Epoch 00081: val_accuracy did not improve from 0.96700
Epoch 82/150
331/331 [============== ] - 33s 101ms/step - loss: 0.1092 -
accuracy: 0.9591 - val_loss: 0.1091 - val_accuracy: 0.9586
Epoch 00082: val_accuracy did not improve from 0.96700
Epoch 83/150
accuracy: 0.9595 - val_loss: 0.1341 - val_accuracy: 0.9551
Epoch 00083: val_accuracy did not improve from 0.96700
Epoch 84/150
accuracy: 0.9590 - val_loss: 0.1094 - val_accuracy: 0.9591
Epoch 00084: val_accuracy did not improve from 0.96700
Epoch 85/150
accuracy: 0.9637 - val_loss: 0.1436 - val_accuracy: 0.9459
Epoch 00085: val_accuracy did not improve from 0.96700
Epoch 86/150
accuracy: 0.9625 - val_loss: 0.1006 - val_accuracy: 0.9622
Epoch 00086: val_accuracy did not improve from 0.96700
Epoch 87/150
accuracy: 0.9630 - val_loss: 0.1466 - val_accuracy: 0.9494
Epoch 00087: val_accuracy did not improve from 0.96700
Epoch 88/150
331/331 [============ ] - 34s 101ms/step - loss: 0.1015 -
accuracy: 0.9611 - val_loss: 0.0953 - val_accuracy: 0.9644
Epoch 00088: val_accuracy did not improve from 0.96700
Epoch 89/150
accuracy: 0.9627 - val_loss: 0.1014 - val_accuracy: 0.9595
Epoch 00089: val_accuracy did not improve from 0.96700
Epoch 90/150
accuracy: 0.9638 - val_loss: 0.1020 - val_accuracy: 0.9639
Epoch 00090: val_accuracy did not improve from 0.96700
Epoch 91/150
```

```
accuracy: 0.9665 - val_loss: 0.1594 - val_accuracy: 0.9393
Epoch 00091: val_accuracy did not improve from 0.96700
Epoch 92/150
accuracy: 0.9616 - val_loss: 0.1715 - val_accuracy: 0.9419
Epoch 00092: val_accuracy did not improve from 0.96700
Epoch 93/150
accuracy: 0.9630 - val_loss: 0.0968 - val_accuracy: 0.9630
Epoch 00093: val_accuracy did not improve from 0.96700
Epoch 94/150
accuracy: 0.9686 - val_loss: 0.1194 - val_accuracy: 0.9534
Epoch 00094: val_accuracy did not improve from 0.96700
Epoch 95/150
331/331 [============ ] - 35s 107ms/step - loss: 0.0908 -
accuracy: 0.9663 - val_loss: 0.1113 - val_accuracy: 0.9591
Epoch 00095: val_accuracy did not improve from 0.96700
Epoch 96/150
accuracy: 0.9640 - val_loss: 0.2489 - val_accuracy: 0.9160
Epoch 00096: val_accuracy did not improve from 0.96700
Epoch 97/150
accuracy: 0.9708 - val_loss: 0.1737 - val_accuracy: 0.9375
Epoch 00097: val_accuracy did not improve from 0.96700
Epoch 98/150
accuracy: 0.9649 - val_loss: 0.1326 - val_accuracy: 0.9516
Epoch 00098: val_accuracy did not improve from 0.96700
Epoch 99/150
accuracy: 0.9664 - val_loss: 0.1795 - val_accuracy: 0.9371
Epoch 00099: val_accuracy did not improve from 0.96700
Epoch 100/150
accuracy: 0.9669 - val_loss: 0.1133 - val_accuracy: 0.9582
Epoch 00100: val_accuracy did not improve from 0.96700
```

```
Epoch 101/150
331/331 [============== ] - 33s 101ms/step - loss: 0.0874 -
accuracy: 0.9695 - val_loss: 0.1043 - val_accuracy: 0.9622
Epoch 00101: val_accuracy did not improve from 0.96700
Epoch 102/150
accuracy: 0.9694 - val_loss: 0.1033 - val_accuracy: 0.9622
Epoch 00102: val_accuracy did not improve from 0.96700
Epoch 103/150
accuracy: 0.9659 - val_loss: 0.1147 - val_accuracy: 0.9591
Epoch 00103: val_accuracy did not improve from 0.96700
Epoch 104/150
331/331 [============ ] - 35s 105ms/step - loss: 0.0798 -
accuracy: 0.9701 - val_loss: 0.1257 - val_accuracy: 0.9556
Epoch 00104: val_accuracy did not improve from 0.96700
Epoch 105/150
accuracy: 0.9704 - val_loss: 0.1098 - val_accuracy: 0.9604
Epoch 00105: val_accuracy did not improve from 0.96700
Epoch 106/150
accuracy: 0.9673 - val_loss: 0.0911 - val_accuracy: 0.9657
Epoch 00106: val_accuracy did not improve from 0.96700
Epoch 107/150
accuracy: 0.9704 - val_loss: 0.2003 - val_accuracy: 0.9314
Epoch 00107: val_accuracy did not improve from 0.96700
Epoch 108/150
accuracy: 0.9718 - val_loss: 0.1037 - val_accuracy: 0.9600
Epoch 00108: val_accuracy did not improve from 0.96700
Epoch 109/150
331/331 [============ ] - 34s 103ms/step - loss: 0.0813 -
accuracy: 0.9702 - val_loss: 0.1549 - val_accuracy: 0.9468
Epoch 00109: val_accuracy did not improve from 0.96700
Epoch 110/150
accuracy: 0.9714 - val_loss: 0.1163 - val_accuracy: 0.9586
```

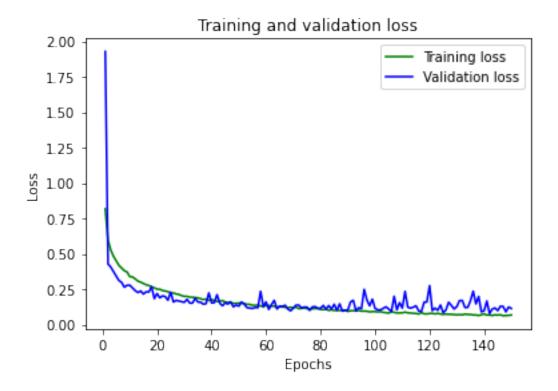
```
Epoch 00110: val_accuracy did not improve from 0.96700
Epoch 111/150
accuracy: 0.9682 - val_loss: 0.2347 - val_accuracy: 0.9261
Epoch 00111: val accuracy did not improve from 0.96700
Epoch 112/150
accuracy: 0.9691 - val_loss: 0.1228 - val_accuracy: 0.9516
Epoch 00112: val_accuracy did not improve from 0.96700
Epoch 113/150
accuracy: 0.9692 - val_loss: 0.1161 - val_accuracy: 0.9569
Epoch 00113: val_accuracy did not improve from 0.96700
Epoch 114/150
accuracy: 0.9724 - val_loss: 0.1227 - val_accuracy: 0.9578
Epoch 00114: val_accuracy did not improve from 0.96700
Epoch 115/150
accuracy: 0.9719 - val_loss: 0.1330 - val_accuracy: 0.9538
Epoch 00115: val_accuracy did not improve from 0.96700
Epoch 116/150
accuracy: 0.9693 - val_loss: 0.0965 - val_accuracy: 0.9648
Epoch 00116: val_accuracy did not improve from 0.96700
Epoch 117/150
accuracy: 0.9700 - val loss: 0.0926 - val accuracy: 0.9661
Epoch 00117: val_accuracy did not improve from 0.96700
Epoch 118/150
accuracy: 0.9688 - val_loss: 0.1577 - val_accuracy: 0.9441
Epoch 00118: val_accuracy did not improve from 0.96700
Epoch 119/150
accuracy: 0.9746 - val_loss: 0.1597 - val_accuracy: 0.9454
Epoch 00119: val_accuracy did not improve from 0.96700
Epoch 120/150
```

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accuracy: 0.9720 - val_loss: 0.2751 - val_accuracy: 0.9164
Epoch 00120: val_accuracy did not improve from 0.96700
Epoch 121/150
accuracy: 0.9705 - val_loss: 0.0990 - val_accuracy: 0.9639
Epoch 00121: val_accuracy did not improve from 0.96700
Epoch 122/150
331/331 [============== ] - 33s 101ms/step - loss: 0.0878 -
accuracy: 0.9678 - val_loss: 0.1162 - val_accuracy: 0.9591
Epoch 00122: val_accuracy did not improve from 0.96700
Epoch 123/150
accuracy: 0.9729 - val_loss: 0.1006 - val_accuracy: 0.9617
Epoch 00123: val_accuracy did not improve from 0.96700
Epoch 124/150
accuracy: 0.9733 - val_loss: 0.1382 - val_accuracy: 0.9512
Epoch 00124: val_accuracy did not improve from 0.96700
Epoch 125/150
accuracy: 0.9738 - val_loss: 0.0838 - val_accuracy: 0.9688
Epoch 00125: val_accuracy improved from 0.96700 to 0.96876, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 126/150
accuracy: 0.9729 - val_loss: 0.1036 - val_accuracy: 0.9591
Epoch 00126: val_accuracy did not improve from 0.96876
Epoch 127/150
accuracy: 0.9751 - val_loss: 0.1578 - val_accuracy: 0.9463
Epoch 00127: val_accuracy did not improve from 0.96876
Epoch 128/150
accuracy: 0.9761 - val_loss: 0.1350 - val_accuracy: 0.9529
Epoch 00128: val_accuracy did not improve from 0.96876
Epoch 129/150
accuracy: 0.9761 - val_loss: 0.1101 - val_accuracy: 0.9604
```

```
Epoch 00129: val_accuracy did not improve from 0.96876
Epoch 130/150
accuracy: 0.9740 - val_loss: 0.1276 - val_accuracy: 0.9560
Epoch 00130: val accuracy did not improve from 0.96876
Epoch 131/150
accuracy: 0.9741 - val_loss: 0.1695 - val_accuracy: 0.9397
Epoch 00131: val_accuracy did not improve from 0.96876
Epoch 132/150
accuracy: 0.9778 - val_loss: 0.1702 - val_accuracy: 0.9406
Epoch 00132: val_accuracy did not improve from 0.96876
Epoch 133/150
accuracy: 0.9718 - val_loss: 0.1204 - val_accuracy: 0.9556
Epoch 00133: val_accuracy did not improve from 0.96876
Epoch 134/150
accuracy: 0.9737 - val_loss: 0.1219 - val_accuracy: 0.9578
Epoch 00134: val_accuracy did not improve from 0.96876
Epoch 135/150
accuracy: 0.9730 - val_loss: 0.1682 - val_accuracy: 0.9437
Epoch 00135: val_accuracy did not improve from 0.96876
Epoch 136/150
accuracy: 0.9743 - val loss: 0.2359 - val accuracy: 0.9309
Epoch 00136: val_accuracy did not improve from 0.96876
Epoch 137/150
accuracy: 0.9751 - val_loss: 0.1461 - val_accuracy: 0.9507
Epoch 00137: val_accuracy did not improve from 0.96876
Epoch 138/150
accuracy: 0.9748 - val_loss: 0.1991 - val_accuracy: 0.9371
Epoch 00138: val_accuracy did not improve from 0.96876
Epoch 139/150
```

```
accuracy: 0.9740 - val_loss: 0.0900 - val_accuracy: 0.9692
Epoch 00139: val_accuracy improved from 0.96876 to 0.96920, saving model to
/home/hivini/learn/research/new-covid/best model.h5
Epoch 140/150
accuracy: 0.9761 - val_loss: 0.0979 - val_accuracy: 0.9595
Epoch 00140: val_accuracy did not improve from 0.96920
Epoch 141/150
accuracy: 0.9747 - val_loss: 0.1665 - val_accuracy: 0.9384
Epoch 00141: val_accuracy did not improve from 0.96920
Epoch 142/150
accuracy: 0.9781 - val_loss: 0.0778 - val_accuracy: 0.9758
Epoch 00142: val accuracy improved from 0.96920 to 0.97580, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 143/150
accuracy: 0.9740 - val_loss: 0.1101 - val_accuracy: 0.9613
Epoch 00143: val_accuracy did not improve from 0.97580
Epoch 144/150
331/331 [============ ] - 33s 98ms/step - loss: 0.0716 -
accuracy: 0.9731 - val_loss: 0.1172 - val_accuracy: 0.9582
Epoch 00144: val_accuracy did not improve from 0.97580
Epoch 145/150
331/331 [============ ] - 32s 98ms/step - loss: 0.0659 -
accuracy: 0.9756 - val_loss: 0.0985 - val_accuracy: 0.9648
Epoch 00145: val_accuracy did not improve from 0.97580
Epoch 146/150
331/331 [============= ] - 32s 98ms/step - loss: 0.0678 -
accuracy: 0.9750 - val_loss: 0.1296 - val_accuracy: 0.9569
Epoch 00146: val_accuracy did not improve from 0.97580
Epoch 147/150
331/331 [============ ] - 32s 98ms/step - loss: 0.0643 -
accuracy: 0.9774 - val_loss: 0.1282 - val_accuracy: 0.9569
Epoch 00147: val_accuracy did not improve from 0.97580
Epoch 148/150
```

```
accuracy: 0.9762 - val_loss: 0.0898 - val_accuracy: 0.9683
    Epoch 00148: val_accuracy did not improve from 0.97580
    Epoch 149/150
    331/331 [============ ] - 32s 98ms/step - loss: 0.0652 -
    accuracy: 0.9760 - val_loss: 0.1243 - val_accuracy: 0.9569
    Epoch 00149: val_accuracy did not improve from 0.97580
    Epoch 150/150
    331/331 [============= ] - 32s 98ms/step - loss: 0.0606 -
    accuracy: 0.9770 - val_loss: 0.1138 - val_accuracy: 0.9617
    Epoch 00150: val_accuracy did not improve from 0.97580
[]: model.save(os.path.join(BASE_PATH, 'covid_classifier_result.h5'))
[]: test loss, test acc = model.evaluate(test generator)
    print("Loss on test set: ", test_loss)
    print("Accuracy on test set: ", test_acc)
    accuracy: 0.9565
    Loss on test set: 0.11710795760154724
    Accuracy on test set: 0.9564643502235413
[]: import matplotlib.pyplot as plt
    acc = history.history['accuracy']
    val_acc = history.history['val_accuracy']
    loss = history.history['loss']
    val loss = history.history['val loss']
    epochs = range(1, len(acc) + 1)
    # bo is for blue dot.
    plt.plot(epochs, loss, 'g', label='Training loss')
    # b is for solid blue line
    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()
```



```
plt.clf()

plt.plot(epochs, acc, 'g', label='Training acc')

plt.plot(epochs, val_acc, 'b', label='Validation acc')

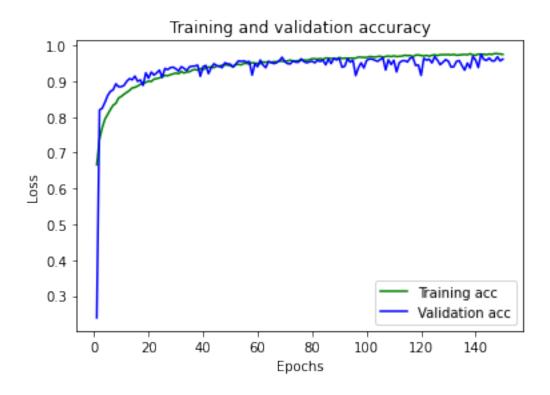
plt.title('Training and validation accuracy')

plt.xlabel('Epochs')

plt.ylabel('Loss')

plt.legend()

plt.show()
```



```
X_{test} = []
     Y_{test} = []
     # Extract the data
     for X, Y in test_generator:
         X_test.append(X)
         Y_test.append(Y)
     X_test = np.array(X_test)
     Y_test = np.array(Y_test)
     predictions = model.predict_classes(X_test)
     predictions = predictions.reshape(1, -1)[0]
     print(classification_report(Y_test, predictions, target_names=[
           'Covid (Class 0)', 'Normal (Class 1)', 'Viral Pneumonia (Class 2)']))
[]: import pandas as pd
     import seaborn as sns
     labels = ['covid', 'normal', 'viral_pneumonia']
     cm = confusion_matrix(Y_test,predictions)
```

[]: from sklearn.metrics import classification\_report, confusion\_matrix