

data\_aug\_adam\_acc\_96\_65

October 10, 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_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)

```

```

saveAndSeparateFiles(ORIGINAL_COVID_DIR, TRAIN_COVID_DIR,
                     VALIDATION_COVID_DIR, TEST_COVID_DIR)
saveAndSeparateFiles(ORIGINAL_VIRAL_DIR, TRAIN_VIRAL_DIR,
                     VALIDATION_VIRAL_DIR, TEST_VIRAL_DIR)

```

## 1.2 Counting our images

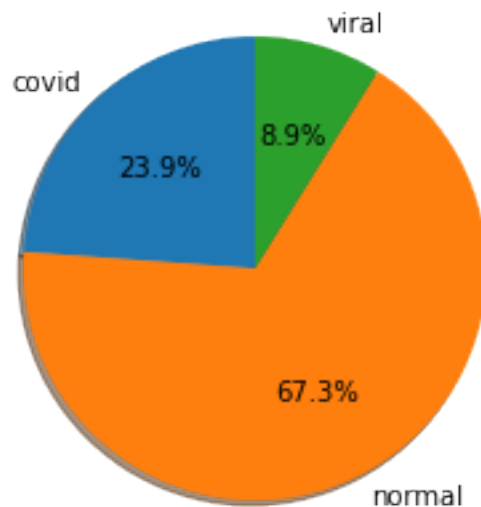
```

[ ]: import tensorflow as tf
import matplotlib.pyplot as plt
normal_train = tf.io.gfile.glob(TRAIN_NORMAL_DIR + '/*')
viral_train = tf.io.gfile.glob(TRAIN_VIRAL_DIR + '/*')
covid_train = tf.io.gfile.glob(TRAIN_COVID_DIR + '/*')

# Plotting Distribution of Each Classes
image_count = {'covid': len(covid_train), 'normal': len(
    normal_train), 'viral': len(viral_train)}
fig1, ax1 = plt.subplots()
ax1.pie(image_count.values(),
        labels=image_count.keys(),
        shadow=True,
        autopct='%1.1f%%',
        startangle=90)
plt.show()

```

2021-10-10 18:47:07.927874: 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 classes 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)

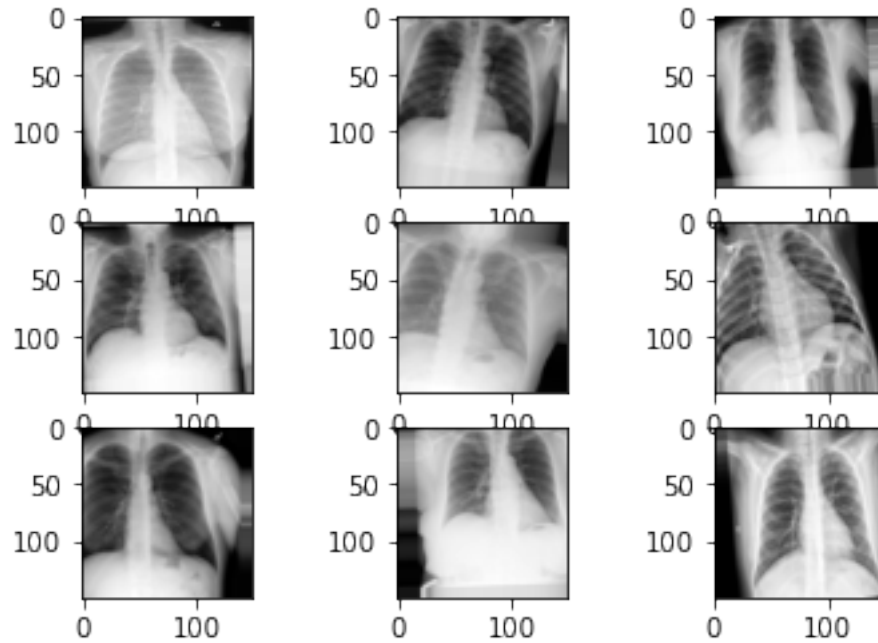
```

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, Flatten, 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()
```

```
2021-10-10 18:47:12.881778: I tensorflow/compiler/jit/xla_cpu_device.cc:41] Not
creating XLA devices, tf_xla_enable_xla_devices not set
2021-10-10 18:47:12.891781: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcuda.so.1
2021-10-10 18:47:13.336200: 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.
2021-10-10 18:47:13.336504: 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
2021-10-10 18:47:13.336537: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudart.so.10.1
2021-10-10 18:47:13.384368: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcublas.so.10
2021-10-10 18:47:13.384462: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcublasLt.so.10
2021-10-10 18:47:13.425423: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcufft.so.10
2021-10-10 18:47:13.434537: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcurand.so.10
2021-10-10 18:47:13.487728: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcusolver.so.10
2021-10-10 18:47:13.507442: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcusparsesparse.so.10
2021-10-10 18:47:13.590740: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudnn.so.7
2021-10-10 18:47:13.591807: 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.
2021-10-10 18:47:13.593207: 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.
2021-10-10 18:47:13.593615: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1862] Adding visible gpu
devices: 0
2021-10-10 18:47:13.596282: 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.
2021-10-10 18:47:13.598129: 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.
2021-10-10 18:47:13.598592: 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
2021-10-10 18:47:13.598636: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudart.so.10.1
2021-10-10 18:47:13.598700: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcublas.so.10
2021-10-10 18:47:13.598720: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcublasLt.so.10
2021-10-10 18:47:13.598732: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcufft.so.10
2021-10-10 18:47:13.598756: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcurand.so.10
2021-10-10 18:47:13.598776: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcusolver.so.10
2021-10-10 18:47:13.598788: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcusparsparse.so.10
2021-10-10 18:47:13.598800: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudnn.so.7
2021-10-10 18:47:13.599627: 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.
2021-10-10 18:47:13.600782: 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.
2021-10-10 18:47:13.601256: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1862] Adding visible gpu
devices: 0
2021-10-10 18:47:13.602433: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudart.so.10.1
2021-10-10 18:47:15.673924: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1261] Device interconnect
StreamExecutor with strength 1 edge matrix:
2021-10-10 18:47:15.673951: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1267]          0
2021-10-10 18:47:15.673957: I
tensorflow/core/common_runtime/gpu/gpu_device.cc:1280] 0:    N
2021-10-10 18:47:15.709348: 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.
2021-10-10 18:47:15.709892: 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.
2021-10-10 18:47:15.710927: 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.
2021-10-10 18:47:15.712131: 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.
2021-10-10 18:47:15.712672: 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)
2021-10-10 18:47:15.715106: I tensorflow/compiler/jit/xla_gpu_device.cc:99] Not
creating XLA devices, tf_xla_enable_xla_devices not set

```

Model: "sequential"

```

-----
Layer (type)                Output Shape                Param #
=====

```

conv2d (Conv2D)	(None, 148, 148, 64)	640
-----		
batch_normalization (Batch Normalization)	(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 Normalization)	(None, 72, 72, 64)	256
-----		
max_pooling2d_1 (MaxPooling2D)	(None, 36, 36, 64)	0
-----		
conv2d_2 (Conv2D)	(None, 34, 34, 128)	73856
-----		
batch_normalization_2 (Batch Normalization)	(None, 34, 34, 128)	512
-----		
max_pooling2d_2 (MaxPooling2D)	(None, 17, 17, 128)	0
-----		
conv2d_3 (Conv2D)	(None, 15, 15, 128)	147584
-----		
batch_normalization_3 (Batch Normalization)	(None, 15, 15, 128)	512
-----		
max_pooling2d_3 (MaxPooling2D)	(None, 7, 7, 128)	0
-----		
flatten (Flatten)	(None, 6272)	0
-----		
dropout (Dropout)	(None, 6272)	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		
-----		

```
[ ]: from keras import optimizers

# opt = RMSprop(lr=0.0001, decay=1e-6)
lr_schedule = optimizers.schedules.ExponentialDecay(
    initial_learning_rate=1e-5,
    decay_steps=2000,
    decay_rate=0.9)
opt = optimizers.Adam(learning_rate=lr_schedule)
```

```
model.compile(loss='categorical_crossentropy', optimizer=opt,
↳metrics=['accuracy'])
```

```
[ ]: 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=10)
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 "
2021-10-10 18:47:16.997222: I
tensorflow/compiler/mlir/mlir_graph_optimization_pass.cc:116] None of the MLIR
optimization passes are enabled (registered 2)
2021-10-10 18:47:17.000353: I
tensorflow/core/platform/profile_utils/cpu_utils.cc:112] CPU Frequency:
2208005000 Hz

Epoch 1/150

2021-10-10 18:47:17.734748: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcublas.so.10
2021-10-10 18:47:18.222118: I
tensorflow/stream_executor/platform/default/dso_loader.cc:49] Successfully
opened dynamic library libcudnn.so.7
2021-10-10 18:47:20.193316: W tensorflow/stream_executor/gpu/asm_compiler.cc:63]
```

```

Running ptexas --version returned 256
2021-10-10 18:47:20.290616: W
tensorflow/stream_executor/gpu/redzone_allocator.cc:314] Internal: ptexas exited
with non-zero error code 256, output:
Relying on driver to perform ptx compilation.
Modify $PATH to customize ptexas location.
This message will be only logged once.

331/331 [=====] - 79s 186ms/step - loss: 1.0379 -
accuracy: 0.6214 - val_loss: 2.2961 - val_accuracy: 0.2415

Epoch 00001: val_accuracy improved from -inf to 0.24153, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 2/150
331/331 [=====] - 33s 100ms/step - loss: 0.6581 -
accuracy: 0.7318 - val_loss: 0.4407 - val_accuracy: 0.8236

Epoch 00002: val_accuracy improved from 0.24153 to 0.82358, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 3/150
331/331 [=====] - 33s 100ms/step - loss: 0.5599 -
accuracy: 0.7663 - val_loss: 0.4212 - val_accuracy: 0.8179

Epoch 00003: val_accuracy did not improve from 0.82358
Epoch 4/150
331/331 [=====] - 33s 101ms/step - loss: 0.5002 -
accuracy: 0.7922 - val_loss: 0.3949 - val_accuracy: 0.8306

Epoch 00004: val_accuracy improved from 0.82358 to 0.83062, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 5/150
331/331 [=====] - 33s 100ms/step - loss: 0.4587 -
accuracy: 0.8035 - val_loss: 0.3935 - val_accuracy: 0.8249

Epoch 00005: val_accuracy did not improve from 0.83062
Epoch 6/150
331/331 [=====] - 33s 100ms/step - loss: 0.4461 -
accuracy: 0.8121 - val_loss: 0.3483 - val_accuracy: 0.8526

Epoch 00006: val_accuracy improved from 0.83062 to 0.85262, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 7/150
331/331 [=====] - 33s 100ms/step - loss: 0.4165 -
accuracy: 0.8248 - val_loss: 0.3109 - val_accuracy: 0.8649

Epoch 00007: val_accuracy improved from 0.85262 to 0.86494, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 8/150

```

331/331 [=====] - 33s 100ms/step - loss: 0.3884 - accuracy: 0.8358 - val\_loss: 0.3000 - val\_accuracy: 0.8698

Epoch 00008: val\_accuracy improved from 0.86494 to 0.86978, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 9/150

331/331 [=====] - 34s 101ms/step - loss: 0.3888 - accuracy: 0.8371 - val\_loss: 0.2910 - val\_accuracy: 0.8715

Epoch 00009: val\_accuracy improved from 0.86978 to 0.87154, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 10/150

331/331 [=====] - 33s 101ms/step - loss: 0.3389 - accuracy: 0.8605 - val\_loss: 0.2661 - val\_accuracy: 0.8843

Epoch 00010: val\_accuracy improved from 0.87154 to 0.88429, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 11/150

331/331 [=====] - 33s 101ms/step - loss: 0.3507 - accuracy: 0.8537 - val\_loss: 0.2513 - val\_accuracy: 0.8918

Epoch 00011: val\_accuracy improved from 0.88429 to 0.89177, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 12/150

331/331 [=====] - 33s 101ms/step - loss: 0.3416 - accuracy: 0.8605 - val\_loss: 0.2367 - val\_accuracy: 0.9063

Epoch 00012: val\_accuracy improved from 0.89177 to 0.90629, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 13/150

331/331 [=====] - 33s 101ms/step - loss: 0.3347 - accuracy: 0.8572 - val\_loss: 0.2352 - val\_accuracy: 0.9001

Epoch 00013: val\_accuracy did not improve from 0.90629

Epoch 14/150

331/331 [=====] - 33s 101ms/step - loss: 0.3093 - accuracy: 0.8768 - val\_loss: 0.2175 - val\_accuracy: 0.9147

Epoch 00014: val\_accuracy improved from 0.90629 to 0.91465, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 15/150

331/331 [=====] - 33s 100ms/step - loss: 0.3085 - accuracy: 0.8725 - val\_loss: 0.2058 - val\_accuracy: 0.9190

Epoch 00015: val\_accuracy improved from 0.91465 to 0.91905, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 16/150

331/331 [=====] - 33s 100ms/step - loss: 0.2984 -

accuracy: 0.8789 - val\_loss: 0.2003 - val\_accuracy: 0.9265

Epoch 00016: val\_accuracy improved from 0.91905 to 0.92653, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 17/150

331/331 [=====] - 34s 102ms/step - loss: 0.2922 - accuracy: 0.8787 - val\_loss: 0.1950 - val\_accuracy: 0.9239

Epoch 00017: val\_accuracy did not improve from 0.92653

Epoch 18/150

331/331 [=====] - 34s 101ms/step - loss: 0.2848 - accuracy: 0.8857 - val\_loss: 0.1960 - val\_accuracy: 0.9252

Epoch 00018: val\_accuracy did not improve from 0.92653

Epoch 19/150

331/331 [=====] - 34s 103ms/step - loss: 0.2720 - accuracy: 0.8909 - val\_loss: 0.2024 - val\_accuracy: 0.9186

Epoch 00019: val\_accuracy did not improve from 0.92653

Epoch 20/150

331/331 [=====] - 32s 97ms/step - loss: 0.2753 - accuracy: 0.8935 - val\_loss: 0.1841 - val\_accuracy: 0.9292

Epoch 00020: val\_accuracy improved from 0.92653 to 0.92917, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 21/150

331/331 [=====] - 32s 98ms/step - loss: 0.2523 - accuracy: 0.9006 - val\_loss: 0.1826 - val\_accuracy: 0.9248

Epoch 00021: val\_accuracy did not improve from 0.92917

Epoch 22/150

331/331 [=====] - 34s 103ms/step - loss: 0.2642 - accuracy: 0.8969 - val\_loss: 0.1786 - val\_accuracy: 0.9305

Epoch 00022: val\_accuracy improved from 0.92917 to 0.93049, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 23/150

331/331 [=====] - 33s 99ms/step - loss: 0.2428 - accuracy: 0.9064 - val\_loss: 0.1756 - val\_accuracy: 0.9314

Epoch 00023: val\_accuracy improved from 0.93049 to 0.93137, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 24/150

331/331 [=====] - 33s 99ms/step - loss: 0.2398 - accuracy: 0.9072 - val\_loss: 0.1682 - val\_accuracy: 0.9358

Epoch 00024: val\_accuracy improved from 0.93137 to 0.93577, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 25/150  
331/331 [=====] - 36s 110ms/step - loss: 0.2311 - accuracy: 0.9082 - val\_loss: 0.1665 - val\_accuracy: 0.9358

Epoch 00025: val\_accuracy did not improve from 0.93577

Epoch 26/150  
331/331 [=====] - 36s 107ms/step - loss: 0.2412 - accuracy: 0.9067 - val\_loss: 0.1699 - val\_accuracy: 0.9309

Epoch 00026: val\_accuracy did not improve from 0.93577

Epoch 27/150  
331/331 [=====] - 34s 103ms/step - loss: 0.2193 - accuracy: 0.9122 - val\_loss: 0.1679 - val\_accuracy: 0.9344

Epoch 00027: val\_accuracy did not improve from 0.93577

Epoch 28/150  
331/331 [=====] - 34s 103ms/step - loss: 0.2217 - accuracy: 0.9108 - val\_loss: 0.1592 - val\_accuracy: 0.9384

Epoch 00028: val\_accuracy improved from 0.93577 to 0.93841, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 29/150  
331/331 [=====] - 35s 107ms/step - loss: 0.2206 - accuracy: 0.9167 - val\_loss: 0.1628 - val\_accuracy: 0.9375

Epoch 00029: val\_accuracy did not improve from 0.93841

Epoch 30/150  
331/331 [=====] - 35s 106ms/step - loss: 0.2190 - accuracy: 0.9137 - val\_loss: 0.1602 - val\_accuracy: 0.9371

Epoch 00030: val\_accuracy did not improve from 0.93841

Epoch 31/150  
331/331 [=====] - 35s 107ms/step - loss: 0.2123 - accuracy: 0.9156 - val\_loss: 0.1563 - val\_accuracy: 0.9419

Epoch 00031: val\_accuracy improved from 0.93841 to 0.94193, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 32/150  
331/331 [=====] - 34s 104ms/step - loss: 0.2154 - accuracy: 0.9150 - val\_loss: 0.1546 - val\_accuracy: 0.9384

Epoch 00032: val\_accuracy did not improve from 0.94193

Epoch 33/150  
331/331 [=====] - 34s 102ms/step - loss: 0.2079 - accuracy: 0.9177 - val\_loss: 0.1536 - val\_accuracy: 0.9432

Epoch 00033: val\_accuracy improved from 0.94193 to 0.94325, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 34/150  
331/331 [=====] - 35s 104ms/step - loss: 0.2005 - accuracy: 0.9222 - val\_loss: 0.1469 - val\_accuracy: 0.9424

Epoch 00034: val\_accuracy did not improve from 0.94325  
Epoch 35/150  
331/331 [=====] - 35s 104ms/step - loss: 0.2063 - accuracy: 0.9198 - val\_loss: 0.1630 - val\_accuracy: 0.9388

Epoch 00035: val\_accuracy did not improve from 0.94325  
Epoch 36/150  
331/331 [=====] - 34s 104ms/step - loss: 0.2009 - accuracy: 0.9217 - val\_loss: 0.1411 - val\_accuracy: 0.9450

Epoch 00036: val\_accuracy improved from 0.94325 to 0.94501, saving model to /home/hivini/learn/research/new-covid/best\_model.h5  
Epoch 37/150  
331/331 [=====] - 33s 99ms/step - loss: 0.1833 - accuracy: 0.9304 - val\_loss: 0.1496 - val\_accuracy: 0.9441

Epoch 00037: val\_accuracy did not improve from 0.94501  
Epoch 38/150  
331/331 [=====] - 35s 104ms/step - loss: 0.1866 - accuracy: 0.9314 - val\_loss: 0.1480 - val\_accuracy: 0.9406

Epoch 00038: val\_accuracy did not improve from 0.94501  
Epoch 39/150  
331/331 [=====] - 35s 107ms/step - loss: 0.1996 - accuracy: 0.9232 - val\_loss: 0.1488 - val\_accuracy: 0.9406

Epoch 00039: val\_accuracy did not improve from 0.94501  
Epoch 40/150  
331/331 [=====] - 38s 113ms/step - loss: 0.1913 - accuracy: 0.9254 - val\_loss: 0.1385 - val\_accuracy: 0.9463

Epoch 00040: val\_accuracy improved from 0.94501 to 0.94633, saving model to /home/hivini/learn/research/new-covid/best\_model.h5  
Epoch 41/150  
331/331 [=====] - 36s 109ms/step - loss: 0.1961 - accuracy: 0.9281 - val\_loss: 0.1367 - val\_accuracy: 0.9468

Epoch 00041: val\_accuracy improved from 0.94633 to 0.94677, saving model to /home/hivini/learn/research/new-covid/best\_model.h5  
Epoch 42/150  
331/331 [=====] - 36s 107ms/step - loss: 0.1815 - accuracy: 0.9281 - val\_loss: 0.1570 - val\_accuracy: 0.9380

Epoch 00042: val\_accuracy did not improve from 0.94677



Epoch 43/150  
 331/331 [=====] - 36s 108ms/step - loss: 0.1729 - accuracy: 0.9368 - val\_loss: 0.1413 - val\_accuracy: 0.9424

Epoch 00043: val\_accuracy did not improve from 0.94677

Epoch 44/150  
 331/331 [=====] - 34s 103ms/step - loss: 0.1856 - accuracy: 0.9309 - val\_loss: 0.1263 - val\_accuracy: 0.9534

Epoch 00044: val\_accuracy improved from 0.94677 to 0.95337, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 45/150  
 331/331 [=====] - 34s 101ms/step - loss: 0.1770 - accuracy: 0.9326 - val\_loss: 0.1377 - val\_accuracy: 0.9446

Epoch 00045: val\_accuracy did not improve from 0.95337

Epoch 46/150  
 331/331 [=====] - 34s 102ms/step - loss: 0.1764 - accuracy: 0.9316 - val\_loss: 0.1402 - val\_accuracy: 0.9437

Epoch 00046: val\_accuracy did not improve from 0.95337

Epoch 47/150  
 331/331 [=====] - 40s 122ms/step - loss: 0.1673 - accuracy: 0.9363 - val\_loss: 0.1317 - val\_accuracy: 0.9503

Epoch 00047: val\_accuracy did not improve from 0.95337

Epoch 48/150  
 331/331 [=====] - 35s 105ms/step - loss: 0.1790 - accuracy: 0.9332 - val\_loss: 0.1731 - val\_accuracy: 0.9344

Epoch 00048: val\_accuracy did not improve from 0.95337

Epoch 49/150  
 331/331 [=====] - 33s 100ms/step - loss: 0.1714 - accuracy: 0.9341 - val\_loss: 0.1279 - val\_accuracy: 0.9503

Epoch 00049: val\_accuracy did not improve from 0.95337

Epoch 50/150  
 331/331 [=====] - 33s 100ms/step - loss: 0.1713 - accuracy: 0.9387 - val\_loss: 0.1343 - val\_accuracy: 0.9476

Epoch 00050: val\_accuracy did not improve from 0.95337

Epoch 51/150  
 331/331 [=====] - 33s 101ms/step - loss: 0.1696 - accuracy: 0.9350 - val\_loss: 0.1480 - val\_accuracy: 0.9437

Epoch 00051: val\_accuracy did not improve from 0.95337

Epoch 52/150  
 331/331 [=====] - 35s 104ms/step - loss: 0.1719 -

accuracy: 0.9337 - val\_loss: 0.1330 - val\_accuracy: 0.9472

Epoch 00052: val\_accuracy did not improve from 0.95337

Epoch 53/150

331/331 [=====] - 35s 106ms/step - loss: 0.1686 -  
accuracy: 0.9350 - val\_loss: 0.1314 - val\_accuracy: 0.9481

Epoch 00053: val\_accuracy did not improve from 0.95337

Epoch 54/150

331/331 [=====] - 32s 97ms/step - loss: 0.1645 -  
accuracy: 0.9395 - val\_loss: 0.1253 - val\_accuracy: 0.9512

Epoch 00054: val\_accuracy did not improve from 0.95337

Epoch 55/150

331/331 [=====] - 32s 97ms/step - loss: 0.1766 -  
accuracy: 0.9315 - val\_loss: 0.1413 - val\_accuracy: 0.9410

Epoch 00055: val\_accuracy did not improve from 0.95337

Epoch 56/150

331/331 [=====] - 33s 100ms/step - loss: 0.1639 -  
accuracy: 0.9405 - val\_loss: 0.1280 - val\_accuracy: 0.9520

Epoch 00056: val\_accuracy did not improve from 0.95337

Epoch 57/150

331/331 [=====] - 32s 97ms/step - loss: 0.1708 -  
accuracy: 0.9367 - val\_loss: 0.1230 - val\_accuracy: 0.9542

Epoch 00057: val\_accuracy improved from 0.95337 to 0.95425, saving model to  
/home/hivini/learn/research/new-covid/best\_model.h5

Epoch 58/150

331/331 [=====] - 32s 97ms/step - loss: 0.1575 -  
accuracy: 0.9419 - val\_loss: 0.1349 - val\_accuracy: 0.9476

Epoch 00058: val\_accuracy did not improve from 0.95425

Epoch 59/150

331/331 [=====] - 32s 97ms/step - loss: 0.1686 -  
accuracy: 0.9365 - val\_loss: 0.1223 - val\_accuracy: 0.9547

Epoch 00059: val\_accuracy improved from 0.95425 to 0.95469, saving model to  
/home/hivini/learn/research/new-covid/best\_model.h5

Epoch 60/150

331/331 [=====] - 32s 97ms/step - loss: 0.1551 -  
accuracy: 0.9404 - val\_loss: 0.1195 - val\_accuracy: 0.9551

Epoch 00060: val\_accuracy improved from 0.95469 to 0.95513, saving model to  
/home/hivini/learn/research/new-covid/best\_model.h5

Epoch 61/150

331/331 [=====] - 36s 108ms/step - loss: 0.1661 -

accuracy: 0.9369 - val\_loss: 0.1217 - val\_accuracy: 0.9534

Epoch 00061: val\_accuracy did not improve from 0.95513

Epoch 62/150

331/331 [=====] - 37s 113ms/step - loss: 0.1656 -  
accuracy: 0.9390 - val\_loss: 0.1193 - val\_accuracy: 0.9556

Epoch 00062: val\_accuracy improved from 0.95513 to 0.95557, saving model to  
/home/hivini/learn/research/new-covid/best\_model.h5

Epoch 63/150

331/331 [=====] - 32s 97ms/step - loss: 0.1553 -  
accuracy: 0.9432 - val\_loss: 0.1307 - val\_accuracy: 0.9468

Epoch 00063: val\_accuracy did not improve from 0.95557

Epoch 64/150

331/331 [=====] - 32s 98ms/step - loss: 0.1575 -  
accuracy: 0.9403 - val\_loss: 0.1202 - val\_accuracy: 0.9534

Epoch 00064: val\_accuracy did not improve from 0.95557

Epoch 65/150

331/331 [=====] - 33s 98ms/step - loss: 0.1578 -  
accuracy: 0.9409 - val\_loss: 0.1177 - val\_accuracy: 0.9556

Epoch 00065: val\_accuracy did not improve from 0.95557

Epoch 66/150

331/331 [=====] - 32s 97ms/step - loss: 0.1601 -  
accuracy: 0.9387 - val\_loss: 0.1276 - val\_accuracy: 0.9485

Epoch 00066: val\_accuracy did not improve from 0.95557

Epoch 67/150

331/331 [=====] - 34s 102ms/step - loss: 0.1482 -  
accuracy: 0.9458 - val\_loss: 0.1223 - val\_accuracy: 0.9503

Epoch 00067: val\_accuracy did not improve from 0.95557

Epoch 68/150

331/331 [=====] - 32s 97ms/step - loss: 0.1555 -  
accuracy: 0.9427 - val\_loss: 0.1191 - val\_accuracy: 0.9534

Epoch 00068: val\_accuracy did not improve from 0.95557

Epoch 69/150

331/331 [=====] - 32s 97ms/step - loss: 0.1644 -  
accuracy: 0.9368 - val\_loss: 0.1170 - val\_accuracy: 0.9569

Epoch 00069: val\_accuracy improved from 0.95557 to 0.95689, saving model to  
/home/hivini/learn/research/new-covid/best\_model.h5

Epoch 70/150

331/331 [=====] - 34s 102ms/step - loss: 0.1564 -  
accuracy: 0.9407 - val\_loss: 0.1134 - val\_accuracy: 0.9573

Epoch 00070: val\_accuracy improved from 0.95689 to 0.95733, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 71/150

331/331 [=====] - 33s 101ms/step - loss: 0.1389 - accuracy: 0.9456 - val\_loss: 0.1205 - val\_accuracy: 0.9556

Epoch 00071: val\_accuracy did not improve from 0.95733

Epoch 72/150

331/331 [=====] - 33s 101ms/step - loss: 0.1539 - accuracy: 0.9447 - val\_loss: 0.1164 - val\_accuracy: 0.9529

Epoch 00072: val\_accuracy did not improve from 0.95733

Epoch 73/150

331/331 [=====] - 33s 101ms/step - loss: 0.1309 - accuracy: 0.9499 - val\_loss: 0.1155 - val\_accuracy: 0.9529

Epoch 00073: val\_accuracy did not improve from 0.95733

Epoch 74/150

331/331 [=====] - 34s 102ms/step - loss: 0.1501 - accuracy: 0.9448 - val\_loss: 0.1141 - val\_accuracy: 0.9564

Epoch 00074: val\_accuracy did not improve from 0.95733

Epoch 75/150

331/331 [=====] - 35s 106ms/step - loss: 0.1426 - accuracy: 0.9490 - val\_loss: 0.1130 - val\_accuracy: 0.9556

Epoch 00075: val\_accuracy did not improve from 0.95733

Epoch 76/150

331/331 [=====] - 37s 113ms/step - loss: 0.1450 - accuracy: 0.9429 - val\_loss: 0.1219 - val\_accuracy: 0.9529

Epoch 00076: val\_accuracy did not improve from 0.95733

Epoch 77/150

331/331 [=====] - 32s 98ms/step - loss: 0.1500 - accuracy: 0.9442 - val\_loss: 0.1183 - val\_accuracy: 0.9560

Epoch 00077: val\_accuracy did not improve from 0.95733

Epoch 78/150

331/331 [=====] - 32s 98ms/step - loss: 0.1507 - accuracy: 0.9499 - val\_loss: 0.1196 - val\_accuracy: 0.9498

Epoch 00078: val\_accuracy did not improve from 0.95733

Epoch 79/150

331/331 [=====] - 33s 99ms/step - loss: 0.1474 - accuracy: 0.9455 - val\_loss: 0.1119 - val\_accuracy: 0.9586

Epoch 00079: val\_accuracy improved from 0.95733 to 0.95864, saving model to

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/home/hivini/learn/research/new-covid/best_model.h5
Epoch 80/150
331/331 [=====] - 33s 98ms/step - loss: 0.1515 -
accuracy: 0.9447 - val_loss: 0.1222 - val_accuracy: 0.9556

Epoch 00080: val_accuracy did not improve from 0.95864
Epoch 81/150
331/331 [=====] - 33s 99ms/step - loss: 0.1508 -
accuracy: 0.9429 - val_loss: 0.1131 - val_accuracy: 0.9569

Epoch 00081: val_accuracy did not improve from 0.95864
Epoch 82/150
331/331 [=====] - 33s 99ms/step - loss: 0.1451 -
accuracy: 0.9433 - val_loss: 0.1158 - val_accuracy: 0.9534

Epoch 00082: val_accuracy did not improve from 0.95864
Epoch 83/150
331/331 [=====] - 33s 99ms/step - loss: 0.1402 -
accuracy: 0.9469 - val_loss: 0.1143 - val_accuracy: 0.9569

Epoch 00083: val_accuracy did not improve from 0.95864
Epoch 84/150
331/331 [=====] - 32s 98ms/step - loss: 0.1425 -
accuracy: 0.9498 - val_loss: 0.1212 - val_accuracy: 0.9529

Epoch 00084: val_accuracy did not improve from 0.95864
Epoch 85/150
331/331 [=====] - 33s 99ms/step - loss: 0.1567 -
accuracy: 0.9397 - val_loss: 0.1152 - val_accuracy: 0.9569

Epoch 00085: val_accuracy did not improve from 0.95864
Epoch 86/150
331/331 [=====] - 36s 109ms/step - loss: 0.1392 -
accuracy: 0.9473 - val_loss: 0.1152 - val_accuracy: 0.9564

Epoch 00086: val_accuracy did not improve from 0.95864
Epoch 87/150
331/331 [=====] - 34s 103ms/step - loss: 0.1412 -
accuracy: 0.9468 - val_loss: 0.1115 - val_accuracy: 0.9591

Epoch 00087: val_accuracy improved from 0.95864 to 0.95908, saving model to
/home/hivini/learn/research/new-covid/best_model.h5
Epoch 88/150
331/331 [=====] - 34s 102ms/step - loss: 0.1403 -
accuracy: 0.9457 - val_loss: 0.1182 - val_accuracy: 0.9547

Epoch 00088: val_accuracy did not improve from 0.95908
Epoch 89/150

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331/331 [=====] - 40s 121ms/step - loss: 0.1408 - accuracy: 0.9465 - val\_loss: 0.1092 - val\_accuracy: 0.9595

Epoch 00089: val\_accuracy improved from 0.95908 to 0.95952, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 90/150

331/331 [=====] - 40s 120ms/step - loss: 0.1496 - accuracy: 0.9442 - val\_loss: 0.1119 - val\_accuracy: 0.9560

Epoch 00090: val\_accuracy did not improve from 0.95952

Epoch 91/150

331/331 [=====] - 36s 109ms/step - loss: 0.1339 - accuracy: 0.9518 - val\_loss: 0.1075 - val\_accuracy: 0.9595

Epoch 00091: val\_accuracy did not improve from 0.95952

Epoch 92/150

331/331 [=====] - 37s 111ms/step - loss: 0.1412 - accuracy: 0.9500 - val\_loss: 0.1149 - val\_accuracy: 0.9569

Epoch 00092: val\_accuracy did not improve from 0.95952

Epoch 93/150

331/331 [=====] - 38s 116ms/step - loss: 0.1395 - accuracy: 0.9472 - val\_loss: 0.1102 - val\_accuracy: 0.9578

Epoch 00093: val\_accuracy did not improve from 0.95952

Epoch 94/150

331/331 [=====] - 39s 117ms/step - loss: 0.1402 - accuracy: 0.9483 - val\_loss: 0.1083 - val\_accuracy: 0.9600

Epoch 00094: val\_accuracy improved from 0.95952 to 0.95996, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 95/150

331/331 [=====] - 40s 120ms/step - loss: 0.1352 - accuracy: 0.9490 - val\_loss: 0.1085 - val\_accuracy: 0.9582

Epoch 00095: val\_accuracy did not improve from 0.95996

Epoch 96/150

331/331 [=====] - 37s 113ms/step - loss: 0.1379 - accuracy: 0.9484 - val\_loss: 0.1101 - val\_accuracy: 0.9595

Epoch 00096: val\_accuracy did not improve from 0.95996

Epoch 97/150

331/331 [=====] - 39s 116ms/step - loss: 0.1451 - accuracy: 0.9462 - val\_loss: 0.1174 - val\_accuracy: 0.9551

Epoch 00097: val\_accuracy did not improve from 0.95996

Epoch 98/150

331/331 [=====] - 40s 120ms/step - loss: 0.1315 -

accuracy: 0.9518 - val\_loss: 0.1147 - val\_accuracy: 0.9569

Epoch 00098: val\_accuracy did not improve from 0.95996  
Epoch 99/150  
331/331 [=====] - 41s 122ms/step - loss: 0.1459 -  
accuracy: 0.9461 - val\_loss: 0.1133 - val\_accuracy: 0.9578

Epoch 00099: val\_accuracy did not improve from 0.95996  
Epoch 100/150  
331/331 [=====] - 43s 130ms/step - loss: 0.1314 -  
accuracy: 0.9550 - val\_loss: 0.1161 - val\_accuracy: 0.9551

Epoch 00100: val\_accuracy did not improve from 0.95996  
Epoch 101/150  
331/331 [=====] - 41s 124ms/step - loss: 0.1417 -  
accuracy: 0.9488 - val\_loss: 0.1068 - val\_accuracy: 0.9600

Epoch 00101: val\_accuracy did not improve from 0.95996  
Epoch 102/150  
331/331 [=====] - 44s 133ms/step - loss: 0.1371 -  
accuracy: 0.9473 - val\_loss: 0.1077 - val\_accuracy: 0.9604

Epoch 00102: val\_accuracy improved from 0.95996 to 0.96040, saving model to  
/home/hivini/learn/research/new-covid/best\_model.h5  
Epoch 103/150  
331/331 [=====] - 41s 123ms/step - loss: 0.1460 -  
accuracy: 0.9452 - val\_loss: 0.1085 - val\_accuracy: 0.9617

Epoch 00103: val\_accuracy improved from 0.96040 to 0.96172, saving model to  
/home/hivini/learn/research/new-covid/best\_model.h5  
Epoch 104/150  
331/331 [=====] - 44s 133ms/step - loss: 0.1357 -  
accuracy: 0.9517 - val\_loss: 0.1049 - val\_accuracy: 0.9617

Epoch 00104: val\_accuracy did not improve from 0.96172  
Epoch 105/150  
331/331 [=====] - 41s 123ms/step - loss: 0.1283 -  
accuracy: 0.9552 - val\_loss: 0.1106 - val\_accuracy: 0.9591

Epoch 00105: val\_accuracy did not improve from 0.96172  
Epoch 106/150  
331/331 [=====] - 42s 127ms/step - loss: 0.1355 -  
accuracy: 0.9521 - val\_loss: 0.1078 - val\_accuracy: 0.9617

Epoch 00106: val\_accuracy did not improve from 0.96172  
Epoch 107/150  
331/331 [=====] - 42s 127ms/step - loss: 0.1392 -  
accuracy: 0.9481 - val\_loss: 0.1058 - val\_accuracy: 0.9608

Epoch 00107: val\_accuracy did not improve from 0.96172  
Epoch 108/150  
331/331 [=====] - 40s 121ms/step - loss: 0.1395 -  
accuracy: 0.9495 - val\_loss: 0.1079 - val\_accuracy: 0.9586

Epoch 00108: val\_accuracy did not improve from 0.96172  
Epoch 109/150  
331/331 [=====] - 36s 110ms/step - loss: 0.1325 -  
accuracy: 0.9515 - val\_loss: 0.1076 - val\_accuracy: 0.9560

Epoch 00109: val\_accuracy did not improve from 0.96172  
Epoch 110/150  
331/331 [=====] - 37s 111ms/step - loss: 0.1414 -  
accuracy: 0.9479 - val\_loss: 0.1046 - val\_accuracy: 0.9613

Epoch 00110: val\_accuracy did not improve from 0.96172  
Epoch 111/150  
331/331 [=====] - 40s 119ms/step - loss: 0.1261 -  
accuracy: 0.9538 - val\_loss: 0.1058 - val\_accuracy: 0.9604

Epoch 00111: val\_accuracy did not improve from 0.96172  
Epoch 112/150  
331/331 [=====] - 37s 111ms/step - loss: 0.1325 -  
accuracy: 0.9502 - val\_loss: 0.1087 - val\_accuracy: 0.9600

Epoch 00112: val\_accuracy did not improve from 0.96172  
Epoch 113/150  
331/331 [=====] - 36s 110ms/step - loss: 0.1369 -  
accuracy: 0.9494 - val\_loss: 0.1082 - val\_accuracy: 0.9617

Epoch 00113: val\_accuracy did not improve from 0.96172  
Epoch 114/150  
331/331 [=====] - 37s 112ms/step - loss: 0.1370 -  
accuracy: 0.9499 - val\_loss: 0.1044 - val\_accuracy: 0.9622

Epoch 00114: val\_accuracy improved from 0.96172 to 0.96216, saving model to  
/home/hivini/learn/research/new-covid/best\_model.h5  
Epoch 115/150  
331/331 [=====] - 39s 117ms/step - loss: 0.1287 -  
accuracy: 0.9527 - val\_loss: 0.1057 - val\_accuracy: 0.9617

Epoch 00115: val\_accuracy did not improve from 0.96216  
Epoch 116/150  
331/331 [=====] - 36s 109ms/step - loss: 0.1240 -  
accuracy: 0.9562 - val\_loss: 0.1070 - val\_accuracy: 0.9600

Epoch 00116: val\_accuracy did not improve from 0.96216



Epoch 117/150  
331/331 [=====] - 36s 109ms/step - loss: 0.1352 -  
accuracy: 0.9480 - val\_loss: 0.1106 - val\_accuracy: 0.9547

Epoch 00117: val\_accuracy did not improve from 0.96216

Epoch 118/150  
331/331 [=====] - 39s 119ms/step - loss: 0.1260 -  
accuracy: 0.9544 - val\_loss: 0.1086 - val\_accuracy: 0.9591

Epoch 00118: val\_accuracy did not improve from 0.96216

Epoch 119/150  
331/331 [=====] - 38s 115ms/step - loss: 0.1278 -  
accuracy: 0.9541 - val\_loss: 0.1049 - val\_accuracy: 0.9622

Epoch 00119: val\_accuracy did not improve from 0.96216

Epoch 120/150  
331/331 [=====] - 40s 120ms/step - loss: 0.1296 -  
accuracy: 0.9505 - val\_loss: 0.1034 - val\_accuracy: 0.9613

Epoch 00120: val\_accuracy did not improve from 0.96216

Epoch 121/150  
331/331 [=====] - 41s 123ms/step - loss: 0.1284 -  
accuracy: 0.9525 - val\_loss: 0.1028 - val\_accuracy: 0.9617

Epoch 00121: val\_accuracy did not improve from 0.96216

Epoch 122/150  
331/331 [=====] - 40s 121ms/step - loss: 0.1353 -  
accuracy: 0.9495 - val\_loss: 0.1053 - val\_accuracy: 0.9604

Epoch 00122: val\_accuracy did not improve from 0.96216

Epoch 123/150  
331/331 [=====] - 36s 110ms/step - loss: 0.1262 -  
accuracy: 0.9539 - val\_loss: 0.1014 - val\_accuracy: 0.9622

Epoch 00123: val\_accuracy did not improve from 0.96216

Epoch 124/150  
331/331 [=====] - 35s 105ms/step - loss: 0.1243 -  
accuracy: 0.9556 - val\_loss: 0.1042 - val\_accuracy: 0.9608

Epoch 00124: val\_accuracy did not improve from 0.96216

Epoch 125/150  
331/331 [=====] - 35s 106ms/step - loss: 0.1416 -  
accuracy: 0.9507 - val\_loss: 0.1035 - val\_accuracy: 0.9617

Epoch 00125: val\_accuracy did not improve from 0.96216

Epoch 126/150  
331/331 [=====] - 35s 105ms/step - loss: 0.1387 -  
accuracy: 0.9480 - val\_loss: 0.1040 - val\_accuracy: 0.9608

Epoch 00126: val\_accuracy did not improve from 0.96216  
Epoch 127/150  
331/331 [=====] - 34s 104ms/step - loss: 0.1351 - accuracy: 0.9505 - val\_loss: 0.1038 - val\_accuracy: 0.9608

Epoch 00127: val\_accuracy did not improve from 0.96216  
Epoch 128/150  
331/331 [=====] - 36s 110ms/step - loss: 0.1318 - accuracy: 0.9492 - val\_loss: 0.1033 - val\_accuracy: 0.9613

Epoch 00128: val\_accuracy did not improve from 0.96216  
Epoch 129/150  
331/331 [=====] - 36s 109ms/step - loss: 0.1226 - accuracy: 0.9545 - val\_loss: 0.1013 - val\_accuracy: 0.9622

Epoch 00129: val\_accuracy did not improve from 0.96216  
Epoch 130/150  
331/331 [=====] - 36s 109ms/step - loss: 0.1320 - accuracy: 0.9506 - val\_loss: 0.1029 - val\_accuracy: 0.9608

Epoch 00130: val\_accuracy did not improve from 0.96216  
Epoch 131/150  
331/331 [=====] - 36s 108ms/step - loss: 0.1341 - accuracy: 0.9499 - val\_loss: 0.1031 - val\_accuracy: 0.9613

Epoch 00131: val\_accuracy did not improve from 0.96216  
Epoch 132/150  
331/331 [=====] - 35s 106ms/step - loss: 0.1368 - accuracy: 0.9482 - val\_loss: 0.1037 - val\_accuracy: 0.9595

Epoch 00132: val\_accuracy did not improve from 0.96216  
Epoch 133/150  
331/331 [=====] - 36s 109ms/step - loss: 0.1157 - accuracy: 0.9594 - val\_loss: 0.1022 - val\_accuracy: 0.9622

Epoch 00133: val\_accuracy did not improve from 0.96216  
Epoch 134/150  
331/331 [=====] - 36s 110ms/step - loss: 0.1273 - accuracy: 0.9552 - val\_loss: 0.1036 - val\_accuracy: 0.9613

Epoch 00134: val\_accuracy did not improve from 0.96216  
Epoch 135/150  
331/331 [=====] - 35s 105ms/step - loss: 0.1234 - accuracy: 0.9571 - val\_loss: 0.1049 - val\_accuracy: 0.9586

Epoch 00135: val\_accuracy did not improve from 0.96216  
Epoch 136/150

331/331 [=====] - 36s 110ms/step - loss: 0.1201 - accuracy: 0.9564 - val\_loss: 0.1052 - val\_accuracy: 0.9604

Epoch 00136: val\_accuracy did not improve from 0.96216

Epoch 137/150

331/331 [=====] - 36s 110ms/step - loss: 0.1246 - accuracy: 0.9552 - val\_loss: 0.1028 - val\_accuracy: 0.9639

Epoch 00137: val\_accuracy improved from 0.96216 to 0.96392, saving model to /home/hivini/learn/research/new-covid/best\_model.h5

Epoch 138/150

331/331 [=====] - 35s 106ms/step - loss: 0.1216 - accuracy: 0.9523 - val\_loss: 0.1029 - val\_accuracy: 0.9622

Epoch 00138: val\_accuracy did not improve from 0.96392

Epoch 139/150

331/331 [=====] - 36s 110ms/step - loss: 0.1237 - accuracy: 0.9551 - val\_loss: 0.1023 - val\_accuracy: 0.9608

Epoch 00139: val\_accuracy did not improve from 0.96392

Epoch 00139: early stopping

```
[ ]: 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)
```

72/72 [=====] - 12s 171ms/step - loss: 0.0858 - accuracy: 0.9666  
Loss on test set: 0.08580382913351059  
Accuracy on test set: 0.9665787220001221

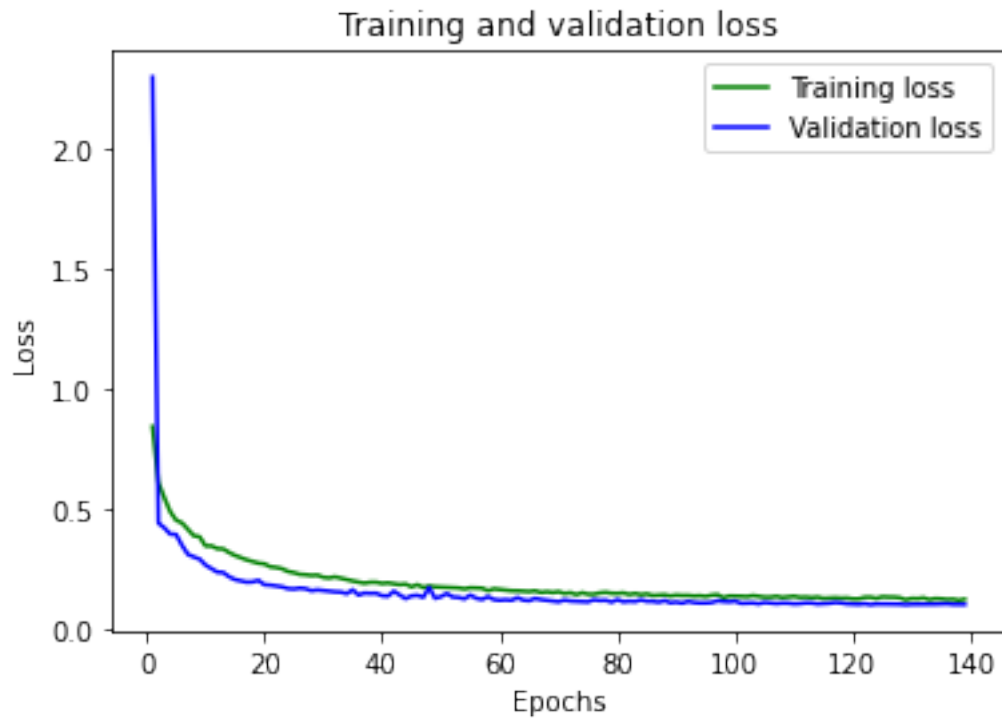
```
[ ]: import matplotlib.pyplot as plt

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()
```

