

DenseNet_CIFAR_resubmission

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1 Implement DenseNet on CIFAR-10

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1.2 1. Assignment instructions

1. Please visit this link to access the state-of-art DenseNet code for reference - DenseNet - cifar10 notebook link
2. You need to create a copy of this and “retrain” this model to achieve 90+ test accuracy.
3. You cannot use DropOut layers.
4. You MUST use Image Augmentation Techniques.
5. You cannot use an already trained model as a beginning points, you have to initilize as your own
6. You cannot run the program for more than 300 Epochs, and it should be clear from your log, that you have only used 300 Epochs
7. You cannot use test images for training the model.
8. You cannot change the general architecture of DenseNet (which means you must use Dense Block, Transition and Output blocks as mentioned in the code)
9. You are free to change Convolution types (e.g. from 3x3 normal convolution to Depthwise Separable, etc)
10. You cannot have more than 1 Million parameters in total
11. You are free to move the code from Keras to Tensorflow, Pytorch, MXNET etc.
12. You can use any optimization algorithm you need.
13. You can checkpoint your model and retrain the model from that checkpoint so that no need of training the model from first if you lost at any epoch while training. You can directly load that model and Train from that epoch.

1.3 2. Assignment

```
[ ]: # import keras
# from keras.datasets import cifar10
# from keras.models import Model, Sequential
# from keras.layers import Dense, Dropout, Flatten, Input, AveragePooling2D,
    ↳merge, Activation
# from keras.layers import Conv2D, MaxPooling2D, BatchNormalization
# from keras.layers import Concatenate
# from keras.optimizers import Adam
from tensorflow.keras import models, layers
from tensorflow.keras.models import Model
from tensorflow.keras.layers import BatchNormalization, Activation, Flatten
from tensorflow.keras.optimizers import Adam
```

```
[ ]: # this part will prevent tensorflow to allocate all the available GPU Memory
# backend
import tensorflow as tf
```

```
[ ]: # Load CIFAR10 Data
(X_train, y_train), (X_test, y_test) = tf.keras.datasets.cifar10.load_data()
img_height, img_width, channel = X_train.shape[1], X_train.shape[2], X_train.
    ↳shape[3]

#scale the data (images) to [0,1] range
X_train = X_train.astype("float32")/255
X_test = X_test.astype("float32")/255

# convert to one hot encoding
num_classes = 10
y_train = tf.keras.utils.to_categorical(y_train, num_classes)
y_test = tf.keras.utils.to_categorical(y_test, num_classes)
```

Downloading data from <https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz>
170500096/170498071 [=====] - 2s 0us/step

```
[ ]: X_train.shape
```

```
[ ]: (50000, 32, 32, 3)
```

```
[ ]: X_test.shape
```

```
[ ]: (10000, 32, 32, 3)
```

1.3.1 2.1 Defining Dense Block, Transition Block and Output Layer

```
[ ]: # Dense Block
def denseblock(input, num_filter = 12, dropout_rate = 0.2):
    global compression
    temp = input
    for _ in range(1):
        BatchNorm = layers.BatchNormalization()(temp)
        relu = layers.Activation('relu')(BatchNorm)
        Conv2D_1_1 = layers.Conv2D(int(num_filter*compression), (1,1),
        ↪(1,1), use_bias=False, padding='same')(relu)
        BatchNorm1 = layers.BatchNormalization()(Conv2D_1_1)
        relu1 = layers.Activation('relu')(BatchNorm1)
        Conv2D_3_3 = layers.Conv2D(int(num_filter*compression), (3,3),
        ↪use_bias=False, padding='same')(relu1)
        if dropout_rate>0:
            Conv2D_3_3 = layers.Dropout(dropout_rate)(Conv2D_3_3)
        concat = layers.Concatenate(axis=-1)([temp, Conv2D_3_3])

        temp = concat

    return temp

## transition Block
def transition(input, num_filter = 12, dropout_rate = 0.2):
    global compression
    BatchNorm = layers.BatchNormalization()(input)
    relu = layers.Activation('relu')(BatchNorm)
    Conv2D_BottleNeck = layers.Conv2D(int(num_filter*compression), (1,1),
    ↪use_bias=False, padding='same')(relu)
    if dropout_rate>0:
        Conv2D_BottleNeck = layers.Dropout(dropout_rate)(Conv2D_BottleNeck)
    avg = layers.AveragePooling2D(pool_size=(2,2))(Conv2D_BottleNeck)
    return avg

#output layer
def output_layer(input):
    global compression
    BatchNorm = layers.BatchNormalization()(input)
    relu = layers.Activation('relu')(BatchNorm)
    AvgPooling = layers.AveragePooling2D(pool_size=(2,2))(relu)
    flat = layers.Flatten()(AvgPooling)
    output = layers.Dense(num_classes, activation='softmax')(flat)
    return output
```

1.3.2 2.2 Using Data Augmentation for training the DenseNet

```
[ ]: #https://www.pyimagesearch.com/2018/12/24/  
      ↪how-to-use-keras-fit-and-fit_generator-a-hands-on-tutorial/  
from tensorflow.keras.preprocessing.image import ImageDataGenerator  
#creating a training image generator for data augmentation  
aug = ImageDataGenerator(rotation_range=0.20,width_shift_range=0.  
      ↪20,height_shift_range=0.15,shear_range=0.15,  
                           zoom_range=0.30,horizontal_flip=True)
```

1.3.3 2.3 Using LearningRateScheduler, ReduceLRonPlateau,CSVLogger in callbacks

```
[ ]: #https://keras.io/api/callbacks/reduce_lr_on_plateau/  
from tensorflow.keras.callbacks import ReduceLRonPlateau  
  
reduce_lr = ReduceLRonPlateau(monitor='val_loss', factor=0.2, patience=5,  
      ↪min_lr=0.001)
```

```
[ ]: #https://keras.io/api/callbacks/learning_rate_scheduler/  
from tensorflow.keras.callbacks import LearningRateScheduler  
lr_list = [0.01,0.001,0.0001]  
def scheduler(epoch,lr):  
    if epoch<25:  
        return lr_list[0]  
    if epoch>=25 and epoch<50:  
        return lr_list[1]  
    else:  
        return lr_list[2]  
  
lr_scheduler = LearningRateScheduler(scheduler)
```

```
[ ]: #https://keras.io/api/callbacks/csv_logger/  
from tensorflow.keras.callbacks import CSVLogger  
  
csv_logger = CSVLogger('training.log')
```

```
[ ]: #https://keras.io/api/callbacks/early_stopping/  
from tensorflow.keras.callbacks import EarlyStopping  
  
early_stop = EarlyStopping(monitor='loss',patience=10)
```

1.3.4 2.4 Growth Rate(num_filter) = 36, compression = 0.7, Number of blocks = 12

```
[ ]: # Hyperparameters
```

```
batch_size = 64
num_classes = 10
nb_epoch = 100
l = 12
num_filter = 36
compression = 0.7
dropout_rate = 0
```

```
[ ]: from tensorflow.keras.callbacks import ModelCheckpoint
```

```
#https://machinelearningmastery.com/check-point-deep-learning-models-keras/
filepath="model5_weights.best.hdf5"
model_checkpoint =
    ↳ModelCheckpoint(filepath,monitor='val_accuracy',save_best_only=True,verbose=1)
```

```
[ ]: input = layers.Input(shape=(img_height, img_width, channel,))
```

```
First_Conv2D = layers.Conv2D(num_filter, (3,3), use_bias=False,
    ↳padding='same')(input)
```

```
First_Block = denseblock(First_Conv2D, num_filter, dropout_rate)
```

```
First_Transition = transition(First_Block, num_filter, dropout_rate)
```

```
Second_Block = denseblock(First_Transition, num_filter, dropout_rate)
```

```
Second_Transition = transition(Second_Block, num_filter, dropout_rate)
```

```
Third_Block = denseblock(Second_Transition, num_filter, dropout_rate)
```

```
Third_Transition = transition(Third_Block, num_filter, dropout_rate)
```

```
Last_Block = denseblock(Third_Transition, num_filter, dropout_rate)
```

```
output = output_layer>Last_Block)
```

```
[16]: model5 = Model(inputs = [input], outputs = [output])
```

```
model5.summary()
```

```
Model: "model"
```

```
-----
Layer (type)                Output Shape          Param #   Connected to
=====
input_1 (InputLayer)        [(None, 32, 32, 3)]  0
-----
conv2d (Conv2D)             (None, 32, 32, 36)   972      input_1[0][0]
```

```

-----
batch_normalization (BatchNorm (None, 32, 32, 36) 144 conv2d[0][0]
-----
activation (Activation) (None, 32, 32, 36) 0
batch_normalization[0][0]
-----
conv2d_1 (Conv2D) (None, 32, 32, 25) 900
activation[0][0]
-----
batch_normalization_1 (BatchNor (None, 32, 32, 25) 100 conv2d_1[0][0]
-----
activation_1 (Activation) (None, 32, 32, 25) 0
batch_normalization_1[0][0]
-----
conv2d_2 (Conv2D) (None, 32, 32, 25) 5625
activation_1[0][0]
-----
concatenate (Concatenate) (None, 32, 32, 61) 0 conv2d[0][0]
conv2d_2[0][0]
-----
batch_normalization_2 (BatchNor (None, 32, 32, 61) 244
concatenate[0][0]
-----
activation_2 (Activation) (None, 32, 32, 61) 0
batch_normalization_2[0][0]
-----
conv2d_3 (Conv2D) (None, 32, 32, 25) 1525
activation_2[0][0]
-----
batch_normalization_3 (BatchNor (None, 32, 32, 25) 100 conv2d_3[0][0]
-----
activation_3 (Activation) (None, 32, 32, 25) 0
batch_normalization_3[0][0]
-----
conv2d_4 (Conv2D) (None, 32, 32, 25) 5625
activation_3[0][0]

```

```

-----
concatenate_1 (Concatenate)      (None, 32, 32, 86)    0
concatenate[0][0]

conv2d_4[0][0]
-----
batch_normalization_4 (BatchNor (None, 32, 32, 86)    344
concatenate_1[0][0]
-----
activation_4 (Activation)        (None, 32, 32, 86)    0
batch_normalization_4[0][0]
-----
conv2d_5 (Conv2D)                (None, 32, 32, 25)    2150
activation_4[0][0]
-----
batch_normalization_5 (BatchNor (None, 32, 32, 25)    100      conv2d_5[0][0]
-----
activation_5 (Activation)        (None, 32, 32, 25)    0
batch_normalization_5[0][0]
-----
conv2d_6 (Conv2D)                (None, 32, 32, 25)    5625
activation_5[0][0]
-----
concatenate_2 (Concatenate)      (None, 32, 32, 111)  0
concatenate_1[0][0]

conv2d_6[0][0]
-----
batch_normalization_6 (BatchNor (None, 32, 32, 111)  444
concatenate_2[0][0]
-----
activation_6 (Activation)        (None, 32, 32, 111)  0
batch_normalization_6[0][0]
-----
conv2d_7 (Conv2D)                (None, 32, 32, 25)    2775
activation_6[0][0]
-----
batch_normalization_7 (BatchNor (None, 32, 32, 25)    100      conv2d_7[0][0]

```

```

-----
activation_7 (Activation)      (None, 32, 32, 25)    0
batch_normalization_7[0][0]

-----

conv2d_8 (Conv2D)             (None, 32, 32, 25)    5625
activation_7[0][0]

-----

concatenate_3 (Concatenate)    (None, 32, 32, 136)   0
concatenate_2[0][0]

conv2d_8[0][0]

-----

batch_normalization_8 (BatchNor (None, 32, 32, 136)  544
concatenate_3[0][0]

-----

activation_8 (Activation)      (None, 32, 32, 136)   0
batch_normalization_8[0][0]

-----

conv2d_9 (Conv2D)             (None, 32, 32, 25)    3400
activation_8[0][0]

-----

batch_normalization_9 (BatchNor (None, 32, 32, 25)    100
conv2d_9[0][0]

-----

activation_9 (Activation)      (None, 32, 32, 25)    0
batch_normalization_9[0][0]

-----

conv2d_10 (Conv2D)            (None, 32, 32, 25)    5625
activation_9[0][0]

-----

concatenate_4 (Concatenate)    (None, 32, 32, 161)   0
concatenate_3[0][0]

conv2d_10[0][0]

-----

batch_normalization_10 (BatchNo (None, 32, 32, 161)  644
concatenate_4[0][0]

-----

activation_10 (Activation)      (None, 32, 32, 161)   0

```



```

batch_normalization_10[0][0]
-----
-----
conv2d_11 (Conv2D)          (None, 32, 32, 25)  4025
activation_10[0][0]
-----
-----
batch_normalization_11 (BatchNo (None, 32, 32, 25)  100          conv2d_11[0][0]
-----
-----
activation_11 (Activation)    (None, 32, 32, 25)  0
batch_normalization_11[0][0]
-----
-----
conv2d_12 (Conv2D)          (None, 32, 32, 25)  5625
activation_11[0][0]
-----
-----
concatenate_5 (Concatenate)  (None, 32, 32, 186)  0
concatenate_4[0][0]
-----
-----
batch_normalization_12 (BatchNo (None, 32, 32, 186)  744          conv2d_12[0][0]
concatenate_5[0][0]
-----
-----
activation_12 (Activation)    (None, 32, 32, 186)  0
batch_normalization_12[0][0]
-----
-----
conv2d_13 (Conv2D)          (None, 32, 32, 25)  4650
activation_12[0][0]
-----
-----
batch_normalization_13 (BatchNo (None, 32, 32, 25)  100          conv2d_13[0][0]
-----
-----
activation_13 (Activation)    (None, 32, 32, 25)  0
batch_normalization_13[0][0]
-----
-----
conv2d_14 (Conv2D)          (None, 32, 32, 25)  5625
activation_13[0][0]
-----
-----
concatenate_6 (Concatenate)  (None, 32, 32, 211)  0
concatenate_5[0][0]

```

```

conv2d_14[0][0]
-----
batch_normalization_14 (BatchNo (None, 32, 32, 211) 844
concatenate_6[0][0]
-----
activation_14 (Activation) (None, 32, 32, 211) 0
batch_normalization_14[0][0]
-----
conv2d_15 (Conv2D) (None, 32, 32, 25) 5275
activation_14[0][0]
-----
batch_normalization_15 (BatchNo (None, 32, 32, 25) 100 conv2d_15[0][0]
-----
activation_15 (Activation) (None, 32, 32, 25) 0
batch_normalization_15[0][0]
-----
conv2d_16 (Conv2D) (None, 32, 32, 25) 5625
activation_15[0][0]
-----
concatenate_7 (Concatenate) (None, 32, 32, 236) 0
concatenate_6[0][0]
conv2d_16[0][0]
-----
batch_normalization_16 (BatchNo (None, 32, 32, 236) 944
concatenate_7[0][0]
-----
activation_16 (Activation) (None, 32, 32, 236) 0
batch_normalization_16[0][0]
-----
conv2d_17 (Conv2D) (None, 32, 32, 25) 5900
activation_16[0][0]
-----
batch_normalization_17 (BatchNo (None, 32, 32, 25) 100 conv2d_17[0][0]
-----
activation_17 (Activation) (None, 32, 32, 25) 0
batch_normalization_17[0][0]

```

```

-----
conv2d_18 (Conv2D)          (None, 32, 32, 25)    5625
activation_17[0][0]
-----
conv2d_18[0][0]
-----
concatenate_8 (Concatenate) (None, 32, 32, 261)    0
concatenate_7[0][0]
-----
conv2d_18[0][0]
-----
batch_normalization_18 (BatchNo (None, 32, 32, 261)    1044
concatenate_8[0][0]
-----
activation_18 (Activation)   (None, 32, 32, 261)    0
batch_normalization_18[0][0]
-----
conv2d_19 (Conv2D)          (None, 32, 32, 25)    6525
activation_18[0][0]
-----
batch_normalization_19 (BatchNo (None, 32, 32, 25)    100
conv2d_19[0][0]
-----
activation_19 (Activation)   (None, 32, 32, 25)    0
batch_normalization_19[0][0]
-----
conv2d_20 (Conv2D)          (None, 32, 32, 25)    5625
activation_19[0][0]
-----
concatenate_9 (Concatenate) (None, 32, 32, 286)    0
concatenate_8[0][0]
-----
conv2d_20[0][0]
-----
batch_normalization_20 (BatchNo (None, 32, 32, 286)    1144
concatenate_9[0][0]
-----
activation_20 (Activation)   (None, 32, 32, 286)    0
batch_normalization_20[0][0]
-----
conv2d_21 (Conv2D)          (None, 32, 32, 25)    7150

```

```

activation_20[0][0]
-----
-----
batch_normalization_21 (BatchNo (None, 32, 32, 25) 100 conv2d_21[0][0]
-----
-----
activation_21 (Activation) (None, 32, 32, 25) 0
batch_normalization_21[0][0]
-----
-----
conv2d_22 (Conv2D) (None, 32, 32, 25) 5625
activation_21[0][0]
-----
-----
concatenate_10 (Concatenate) (None, 32, 32, 311) 0
concatenate_9[0][0]
conv2d_22[0][0]
-----
-----
batch_normalization_22 (BatchNo (None, 32, 32, 311) 1244
concatenate_10[0][0]
-----
-----
activation_22 (Activation) (None, 32, 32, 311) 0
batch_normalization_22[0][0]
-----
-----
conv2d_23 (Conv2D) (None, 32, 32, 25) 7775
activation_22[0][0]
-----
-----
batch_normalization_23 (BatchNo (None, 32, 32, 25) 100 conv2d_23[0][0]
-----
-----
activation_23 (Activation) (None, 32, 32, 25) 0
batch_normalization_23[0][0]
-----
-----
conv2d_24 (Conv2D) (None, 32, 32, 25) 5625
activation_23[0][0]
-----
-----
concatenate_11 (Concatenate) (None, 32, 32, 336) 0
concatenate_10[0][0]
conv2d_24[0][0]
-----
-----
batch_normalization_24 (BatchNo (None, 32, 32, 336) 1344

```

```

concatenate_11[0][0]

-----

activation_24 (Activation)      (None, 32, 32, 336)  0
batch_normalization_24[0][0]

-----

conv2d_25 (Conv2D)              (None, 32, 32, 25)   8400
activation_24[0][0]

-----

average_pooling2d (AveragePooli (None, 16, 16, 25)  0          conv2d_25[0][0]

-----

batch_normalization_25 (BatchNo (None, 16, 16, 25)  100
average_pooling2d[0][0]

-----

activation_25 (Activation)      (None, 16, 16, 25)  0
batch_normalization_25[0][0]

-----

conv2d_26 (Conv2D)              (None, 16, 16, 25)   625
activation_25[0][0]

-----

batch_normalization_26 (BatchNo (None, 16, 16, 25)  100          conv2d_26[0][0]

-----

activation_26 (Activation)      (None, 16, 16, 25)  0
batch_normalization_26[0][0]

-----

conv2d_27 (Conv2D)              (None, 16, 16, 25)   5625
activation_26[0][0]

-----

concatenate_12 (Concatenate)    (None, 16, 16, 50)   0
average_pooling2d[0][0]

                                          conv2d_27[0][0]

-----

batch_normalization_27 (BatchNo (None, 16, 16, 50)  200
concatenate_12[0][0]

-----

activation_27 (Activation)      (None, 16, 16, 50)   0
batch_normalization_27[0][0]

```

```

-----
conv2d_28 (Conv2D)          (None, 16, 16, 25)  1250
activation_27[0][0]

-----
batch_normalization_28 (BatchNormaliz (None, 16, 16, 25)  100      conv2d_28[0][0]
-----
activation_28 (Activation)    (None, 16, 16, 25)   0
batch_normalization_28[0][0]

-----
conv2d_29 (Conv2D)          (None, 16, 16, 25)  5625
activation_28[0][0]

-----
concatenate_13 (Concatenate)  (None, 16, 16, 75)   0
concatenate_12[0][0]
                                         conv2d_29[0][0]

-----
batch_normalization_29 (BatchNormaliz (None, 16, 16, 75)  300
concatenate_13[0][0]

-----
activation_29 (Activation)    (None, 16, 16, 75)   0
batch_normalization_29[0][0]

-----
conv2d_30 (Conv2D)          (None, 16, 16, 25)  1875
activation_29[0][0]

-----
batch_normalization_30 (BatchNormaliz (None, 16, 16, 25)  100      conv2d_30[0][0]
-----
activation_30 (Activation)    (None, 16, 16, 25)   0
batch_normalization_30[0][0]

-----
conv2d_31 (Conv2D)          (None, 16, 16, 25)  5625
activation_30[0][0]

-----
concatenate_14 (Concatenate)  (None, 16, 16, 100)  0
concatenate_13[0][0]
                                         conv2d_31[0][0]

```

```

-----
batch_normalization_31 (BatchNo (None, 16, 16, 100) 400
concatenate_14[0][0]
-----

activation_31 (Activation) (None, 16, 16, 100) 0
batch_normalization_31[0][0]
-----

conv2d_32 (Conv2D) (None, 16, 16, 25) 2500
activation_31[0][0]
-----

batch_normalization_32 (BatchNo (None, 16, 16, 25) 100 conv2d_32[0][0]
-----

activation_32 (Activation) (None, 16, 16, 25) 0
batch_normalization_32[0][0]
-----

conv2d_33 (Conv2D) (None, 16, 16, 25) 5625
activation_32[0][0]
-----

concatenate_15 (Concatenate) (None, 16, 16, 125) 0
concatenate_14[0][0]
conv2d_33[0][0]
-----

batch_normalization_33 (BatchNo (None, 16, 16, 125) 500
concatenate_15[0][0]
-----

activation_33 (Activation) (None, 16, 16, 125) 0
batch_normalization_33[0][0]
-----

conv2d_34 (Conv2D) (None, 16, 16, 25) 3125
activation_33[0][0]
-----

batch_normalization_34 (BatchNo (None, 16, 16, 25) 100 conv2d_34[0][0]
-----

activation_34 (Activation) (None, 16, 16, 25) 0
batch_normalization_34[0][0]
-----

```

```

-----
conv2d_35 (Conv2D)                (None, 16, 16, 25)    5625
activation_34[0][0]
-----

-----
concatenate_16 (Concatenate)      (None, 16, 16, 150)   0
concatenate_15[0][0]
conv2d_35[0][0]
-----

-----
batch_normalization_35 (BatchNo (None, 16, 16, 150)  600
concatenate_16[0][0]
-----

-----
activation_35 (Activation)        (None, 16, 16, 150)   0
batch_normalization_35[0][0]
-----

-----
conv2d_36 (Conv2D)                (None, 16, 16, 25)    3750
activation_35[0][0]
-----

-----
batch_normalization_36 (BatchNo (None, 16, 16, 25)    100
conv2d_36[0][0]
-----

-----
activation_36 (Activation)        (None, 16, 16, 25)    0
batch_normalization_36[0][0]
-----

-----
conv2d_37 (Conv2D)                (None, 16, 16, 25)    5625
activation_36[0][0]
-----

-----
concatenate_17 (Concatenate)      (None, 16, 16, 175)   0
concatenate_16[0][0]
conv2d_37[0][0]
-----

-----
batch_normalization_37 (BatchNo (None, 16, 16, 175)  700
concatenate_17[0][0]
-----

-----
activation_37 (Activation)        (None, 16, 16, 175)   0
batch_normalization_37[0][0]
-----

-----
conv2d_38 (Conv2D)                (None, 16, 16, 25)    4375
activation_37[0][0]

```



```

-----
batch_normalization_38 (BatchNo (None, 16, 16, 25) 100 conv2d_38[0] [0]
-----

activation_38 (Activation) (None, 16, 16, 25) 0
batch_normalization_38[0] [0]
-----

conv2d_39 (Conv2D) (None, 16, 16, 25) 5625
activation_38[0] [0]
-----

concatenate_18 (Concatenate) (None, 16, 16, 200) 0
concatenate_17[0] [0]
conv2d_39[0] [0]
-----

batch_normalization_39 (BatchNo (None, 16, 16, 200) 800
concatenate_18[0] [0]
-----

activation_39 (Activation) (None, 16, 16, 200) 0
batch_normalization_39[0] [0]
-----

conv2d_40 (Conv2D) (None, 16, 16, 25) 5000
activation_39[0] [0]
-----

batch_normalization_40 (BatchNo (None, 16, 16, 25) 100 conv2d_40[0] [0]
-----

activation_40 (Activation) (None, 16, 16, 25) 0
batch_normalization_40[0] [0]
-----

conv2d_41 (Conv2D) (None, 16, 16, 25) 5625
activation_40[0] [0]
-----

concatenate_19 (Concatenate) (None, 16, 16, 225) 0
concatenate_18[0] [0]
conv2d_41[0] [0]
-----

batch_normalization_41 (BatchNo (None, 16, 16, 225) 900
concatenate_19[0] [0]

```

```

-----
activation_41 (Activation)      (None, 16, 16, 225)  0
batch_normalization_41[0][0]

-----

conv2d_42 (Conv2D)             (None, 16, 16, 25)   5625
activation_41[0][0]

-----

batch_normalization_42 (BatchNo (None, 16, 16, 25)   100          conv2d_42[0][0]
-----

activation_42 (Activation)      (None, 16, 16, 25)   0
batch_normalization_42[0][0]

-----

conv2d_43 (Conv2D)             (None, 16, 16, 25)   5625
activation_42[0][0]

-----

concatenate_20 (Concatenate)    (None, 16, 16, 250)  0
concatenate_19[0][0]

                                          conv2d_43[0][0]

-----

batch_normalization_43 (BatchNo (None, 16, 16, 250)  1000
concatenate_20[0][0]

-----

activation_43 (Activation)      (None, 16, 16, 250)  0
batch_normalization_43[0][0]

-----

conv2d_44 (Conv2D)             (None, 16, 16, 25)   6250
activation_43[0][0]

-----

batch_normalization_44 (BatchNo (None, 16, 16, 25)   100          conv2d_44[0][0]
-----

activation_44 (Activation)      (None, 16, 16, 25)   0
batch_normalization_44[0][0]

-----

conv2d_45 (Conv2D)             (None, 16, 16, 25)   5625
activation_44[0][0]
-----

```

```

-----
concatenate_21 (Concatenate)      (None, 16, 16, 275)  0
concatenate_20[0][0]

conv2d_45[0][0]

-----
batch_normalization_45 (BatchNo (None, 16, 16, 275)  1100
concatenate_21[0][0]

-----
activation_45 (Activation)         (None, 16, 16, 275)  0
batch_normalization_45[0][0]

-----
conv2d_46 (Conv2D)                (None, 16, 16, 25)   6875
activation_45[0][0]

-----
batch_normalization_46 (BatchNo (None, 16, 16, 25)   100      conv2d_46[0][0]

-----
activation_46 (Activation)         (None, 16, 16, 25)   0
batch_normalization_46[0][0]

-----
conv2d_47 (Conv2D)                (None, 16, 16, 25)   5625
activation_46[0][0]

-----
concatenate_22 (Concatenate)      (None, 16, 16, 300)  0
concatenate_21[0][0]

conv2d_47[0][0]

-----
batch_normalization_47 (BatchNo (None, 16, 16, 300)  1200
concatenate_22[0][0]

-----
activation_47 (Activation)         (None, 16, 16, 300)  0
batch_normalization_47[0][0]

-----
conv2d_48 (Conv2D)                (None, 16, 16, 25)   7500
activation_47[0][0]

-----
batch_normalization_48 (BatchNo (None, 16, 16, 25)   100      conv2d_48[0][0]

```

```

-----
activation_48 (Activation)      (None, 16, 16, 25)    0
batch_normalization_48[0][0]

-----

conv2d_49 (Conv2D)             (None, 16, 16, 25)    5625
activation_48[0][0]

-----

concatenate_23 (Concatenate)   (None, 16, 16, 325)   0
concatenate_22[0][0]

conv2d_49[0][0]

-----

batch_normalization_49 (BatchNo (None, 16, 16, 325) 1300
concatenate_23[0][0]

-----

activation_49 (Activation)      (None, 16, 16, 325)   0
batch_normalization_49[0][0]

-----

conv2d_50 (Conv2D)             (None, 16, 16, 25)    8125
activation_49[0][0]

-----

average_pooling2d_1 (AveragePoo (None, 8, 8, 25)      0
conv2d_50[0][0]

-----

batch_normalization_50 (BatchNo (None, 8, 8, 25)      100
average_pooling2d_1[0][0]

-----

activation_50 (Activation)      (None, 8, 8, 25)      0
batch_normalization_50[0][0]

-----

conv2d_51 (Conv2D)             (None, 8, 8, 25)      625
activation_50[0][0]

-----

batch_normalization_51 (BatchNo (None, 8, 8, 25)      100
conv2d_51[0][0]

-----

activation_51 (Activation)      (None, 8, 8, 25)      0
batch_normalization_51[0][0]

-----

```

conv2d_52 (Conv2D)	(None, 8, 8, 25)	5625	
activation_51[0][0]			

concatenate_24 (Concatenate)	(None, 8, 8, 50)	0	
average_pooling2d_1[0][0]			
			conv2d_52[0][0]

batch_normalization_52 (Batch Normalization)	(None, 8, 8, 50)	200	
concatenate_24[0][0]			

activation_52 (Activation)	(None, 8, 8, 50)	0	
batch_normalization_52[0][0]			

conv2d_53 (Conv2D)	(None, 8, 8, 25)	1250	
activation_52[0][0]			

batch_normalization_53 (Batch Normalization)	(None, 8, 8, 25)	100	
			conv2d_53[0][0]

activation_53 (Activation)	(None, 8, 8, 25)	0	
batch_normalization_53[0][0]			

conv2d_54 (Conv2D)	(None, 8, 8, 25)	5625	
activation_53[0][0]			

concatenate_25 (Concatenate)	(None, 8, 8, 75)	0	
concatenate_24[0][0]			
			conv2d_54[0][0]

batch_normalization_54 (Batch Normalization)	(None, 8, 8, 75)	300	
concatenate_25[0][0]			

activation_54 (Activation)	(None, 8, 8, 75)	0	
batch_normalization_54[0][0]			

conv2d_55 (Conv2D)	(None, 8, 8, 25)	1875	
activation_54[0][0]			

```

-----
batch_normalization_55 (BatchNo (None, 8, 8, 25)      100      conv2d_55[0] [0]
-----

-----
activation_55 (Activation)      (None, 8, 8, 25)      0
batch_normalization_55[0] [0]
-----

-----
conv2d_56 (Conv2D)              (None, 8, 8, 25)      5625
activation_55[0] [0]
-----

-----
concatenate_26 (Concatenate)    (None, 8, 8, 100)     0
concatenate_25[0] [0]
                                         conv2d_56[0] [0]
-----

-----
batch_normalization_56 (BatchNo (None, 8, 8, 100)     400
concatenate_26[0] [0]
-----

-----
activation_56 (Activation)      (None, 8, 8, 100)     0
batch_normalization_56[0] [0]
-----

-----
conv2d_57 (Conv2D)              (None, 8, 8, 25)      2500
activation_56[0] [0]
-----

-----
batch_normalization_57 (BatchNo (None, 8, 8, 25)      100      conv2d_57[0] [0]
-----

-----
activation_57 (Activation)      (None, 8, 8, 25)      0
batch_normalization_57[0] [0]
-----

-----
conv2d_58 (Conv2D)              (None, 8, 8, 25)      5625
activation_57[0] [0]
-----

-----
concatenate_27 (Concatenate)    (None, 8, 8, 125)     0
concatenate_26[0] [0]
                                         conv2d_58[0] [0]
-----

-----
batch_normalization_58 (BatchNo (None, 8, 8, 125)     500
concatenate_27[0] [0]
-----

```

```

-----
activation_58 (Activation)      (None, 8, 8, 125)    0
batch_normalization_58[0][0]

-----

conv2d_59 (Conv2D)             (None, 8, 8, 25)     3125
activation_58[0][0]

-----

batch_normalization_59 (BatchNo (None, 8, 8, 25)    100          conv2d_59[0][0]
-----

activation_59 (Activation)      (None, 8, 8, 25)     0
batch_normalization_59[0][0]

-----

conv2d_60 (Conv2D)             (None, 8, 8, 25)     5625
activation_59[0][0]

-----

concatenate_28 (Concatenate)    (None, 8, 8, 150)    0
concatenate_27[0][0]

                                          conv2d_60[0][0]
-----

batch_normalization_60 (BatchNo (None, 8, 8, 150)    600
concatenate_28[0][0]

-----

activation_60 (Activation)      (None, 8, 8, 150)    0
batch_normalization_60[0][0]

-----

conv2d_61 (Conv2D)             (None, 8, 8, 25)     3750
activation_60[0][0]

-----

batch_normalization_61 (BatchNo (None, 8, 8, 25)    100          conv2d_61[0][0]
-----

activation_61 (Activation)      (None, 8, 8, 25)     0
batch_normalization_61[0][0]

-----

conv2d_62 (Conv2D)             (None, 8, 8, 25)     5625
activation_61[0][0]

-----

```

```

concatenate_29 (Concatenate)      (None, 8, 8, 175)    0
concatenate_28[0][0]
conv2d_62[0][0]

-----

batch_normalization_62 (BatchNo (None, 8, 8, 175)    700
concatenate_29[0][0]

-----

activation_62 (Activation)        (None, 8, 8, 175)    0
batch_normalization_62[0][0]

-----

conv2d_63 (Conv2D)                (None, 8, 8, 25)     4375
activation_62[0][0]

-----

batch_normalization_63 (BatchNo (None, 8, 8, 25)     100      conv2d_63[0][0]

-----

activation_63 (Activation)        (None, 8, 8, 25)     0
batch_normalization_63[0][0]

-----

conv2d_64 (Conv2D)                (None, 8, 8, 25)     5625
activation_63[0][0]

-----

concatenate_30 (Concatenate)      (None, 8, 8, 200)    0
concatenate_29[0][0]
conv2d_64[0][0]

-----

batch_normalization_64 (BatchNo (None, 8, 8, 200)    800
concatenate_30[0][0]

-----

activation_64 (Activation)        (None, 8, 8, 200)    0
batch_normalization_64[0][0]

-----

conv2d_65 (Conv2D)                (None, 8, 8, 25)     5000
activation_64[0][0]

-----

batch_normalization_65 (BatchNo (None, 8, 8, 25)     100      conv2d_65[0][0]

-----

```


activation_65 (Activation)	(None, 8, 8, 25)	0	
batch_normalization_65[0][0]			

conv2d_66 (Conv2D)	(None, 8, 8, 25)	5625	
activation_65[0][0]			

concatenate_31 (Concatenate)	(None, 8, 8, 225)	0	
concatenate_30[0][0]			
			conv2d_66[0][0]

batch_normalization_66 (BatchNo	(None, 8, 8, 225)	900	
concatenate_31[0][0]			

activation_66 (Activation)	(None, 8, 8, 225)	0	
batch_normalization_66[0][0]			

conv2d_67 (Conv2D)	(None, 8, 8, 25)	5625	
activation_66[0][0]			

batch_normalization_67 (BatchNo	(None, 8, 8, 25)	100	conv2d_67[0][0]

activation_67 (Activation)	(None, 8, 8, 25)	0	
batch_normalization_67[0][0]			

conv2d_68 (Conv2D)	(None, 8, 8, 25)	5625	
activation_67[0][0]			

concatenate_32 (Concatenate)	(None, 8, 8, 250)	0	
concatenate_31[0][0]			
			conv2d_68[0][0]

batch_normalization_68 (BatchNo	(None, 8, 8, 250)	1000	
concatenate_32[0][0]			

activation_68 (Activation)	(None, 8, 8, 250)	0	
batch_normalization_68[0][0]			

```

-----
conv2d_69 (Conv2D)                (None, 8, 8, 25)    6250
activation_68[0][0]
-----
-----
batch_normalization_69 (BatchNo (None, 8, 8, 25)    100          conv2d_69[0][0]
-----
-----
activation_69 (Activation)         (None, 8, 8, 25)    0
batch_normalization_69[0][0]
-----
-----
conv2d_70 (Conv2D)                (None, 8, 8, 25)    5625
activation_69[0][0]
-----
-----
concatenate_33 (Concatenate)      (None, 8, 8, 275)   0
concatenate_32[0][0]
                                         conv2d_70[0][0]
-----
-----
batch_normalization_70 (BatchNo (None, 8, 8, 275)   1100
concatenate_33[0][0]
-----
-----
activation_70 (Activation)         (None, 8, 8, 275)   0
batch_normalization_70[0][0]
-----
-----
conv2d_71 (Conv2D)                (None, 8, 8, 25)    6875
activation_70[0][0]
-----
-----
batch_normalization_71 (BatchNo (None, 8, 8, 25)    100          conv2d_71[0][0]
-----
-----
activation_71 (Activation)         (None, 8, 8, 25)    0
batch_normalization_71[0][0]
-----
-----
conv2d_72 (Conv2D)                (None, 8, 8, 25)    5625
activation_71[0][0]
-----
-----
concatenate_34 (Concatenate)      (None, 8, 8, 300)   0
concatenate_33[0][0]
                                         conv2d_72[0][0]
-----

```

```

-----
batch_normalization_72 (BatchNo (None, 8, 8, 300)    1200
concatenate_34[0][0]
-----
-----
activation_72 (Activation)      (None, 8, 8, 300)    0
batch_normalization_72[0][0]
-----
-----
conv2d_73 (Conv2D)              (None, 8, 8, 25)     7500
activation_72[0][0]
-----
-----
batch_normalization_73 (BatchNo (None, 8, 8, 25)     100          conv2d_73[0][0]
-----
-----
activation_73 (Activation)      (None, 8, 8, 25)     0
batch_normalization_73[0][0]
-----
-----
conv2d_74 (Conv2D)              (None, 8, 8, 25)     5625
activation_73[0][0]
-----
-----
concatenate_35 (Concatenate)    (None, 8, 8, 325)    0
concatenate_34[0][0]
                                         conv2d_74[0][0]
-----
-----
batch_normalization_74 (BatchNo (None, 8, 8, 325)    1300
concatenate_35[0][0]
-----
-----
activation_74 (Activation)      (None, 8, 8, 325)    0
batch_normalization_74[0][0]
-----
-----
conv2d_75 (Conv2D)              (None, 8, 8, 25)     8125
activation_74[0][0]
-----
-----
average_pooling2d_2 (AveragePoo (None, 4, 4, 25)     0          conv2d_75[0][0]
-----
-----
batch_normalization_75 (BatchNo (None, 4, 4, 25)     100
average_pooling2d_2[0][0]
-----
-----

```

activation_75 (Activation)	(None, 4, 4, 25)	0	
batch_normalization_75[0][0]			

conv2d_76 (Conv2D)	(None, 4, 4, 25)	625	
activation_75[0][0]			

batch_normalization_76 (BatchNo	(None, 4, 4, 25)	100	conv2d_76[0][0]

activation_76 (Activation)	(None, 4, 4, 25)	0	
batch_normalization_76[0][0]			

conv2d_77 (Conv2D)	(None, 4, 4, 25)	5625	
activation_76[0][0]			

concatenate_36 (Concatenate)	(None, 4, 4, 50)	0	
average_pooling2d_2[0][0]			conv2d_77[0][0]

batch_normalization_77 (BatchNo	(None, 4, 4, 50)	200	
concatenate_36[0][0]			

activation_77 (Activation)	(None, 4, 4, 50)	0	
batch_normalization_77[0][0]			

conv2d_78 (Conv2D)	(None, 4, 4, 25)	1250	
activation_77[0][0]			

batch_normalization_78 (BatchNo	(None, 4, 4, 25)	100	conv2d_78[0][0]

activation_78 (Activation)	(None, 4, 4, 25)	0	
batch_normalization_78[0][0]			

conv2d_79 (Conv2D)	(None, 4, 4, 25)	5625	
activation_78[0][0]			

concatenate_37 (Concatenate)	(None, 4, 4, 75)	0	

```

concatenate_36[0][0]
conv2d_79[0][0]

-----
batch_normalization_79 (BatchNo (None, 4, 4, 75) 300
concatenate_37[0][0]

-----
activation_79 (Activation) (None, 4, 4, 75) 0
batch_normalization_79[0][0]

-----
conv2d_80 (Conv2D) (None, 4, 4, 25) 1875
activation_79[0][0]

-----
batch_normalization_80 (BatchNo (None, 4, 4, 25) 100 conv2d_80[0][0]

-----
activation_80 (Activation) (None, 4, 4, 25) 0
batch_normalization_80[0][0]

-----
conv2d_81 (Conv2D) (None, 4, 4, 25) 5625
activation_80[0][0]

-----
concatenate_38 (Concatenate) (None, 4, 4, 100) 0
concatenate_37[0][0]
conv2d_81[0][0]

-----
batch_normalization_81 (BatchNo (None, 4, 4, 100) 400
concatenate_38[0][0]

-----
activation_81 (Activation) (None, 4, 4, 100) 0
batch_normalization_81[0][0]

-----
conv2d_82 (Conv2D) (None, 4, 4, 25) 2500
activation_81[0][0]

-----
batch_normalization_82 (BatchNo (None, 4, 4, 25) 100 conv2d_82[0][0]

-----
activation_82 (Activation) (None, 4, 4, 25) 0

```

```

batch_normalization_82[0][0]
-----
conv2d_83 (Conv2D)          (None, 4, 4, 25)      5625
activation_82[0][0]
-----
concatenate_39 (Concatenate) (None, 4, 4, 125)      0
concatenate_38[0][0]
conv2d_83[0][0]
-----
batch_normalization_83 (BatchNo (None, 4, 4, 125)      500
concatenate_39[0][0]
-----
activation_83 (Activation)    (None, 4, 4, 125)      0
batch_normalization_83[0][0]
-----
conv2d_84 (Conv2D)          (None, 4, 4, 25)      3125
activation_83[0][0]
-----
batch_normalization_84 (BatchNo (None, 4, 4, 25)      100
conv2d_84[0][0]
-----
activation_84 (Activation)    (None, 4, 4, 25)      0
batch_normalization_84[0][0]
-----
conv2d_85 (Conv2D)          (None, 4, 4, 25)      5625
activation_84[0][0]
-----
concatenate_40 (Concatenate) (None, 4, 4, 150)      0
concatenate_39[0][0]
conv2d_85[0][0]
-----
batch_normalization_85 (BatchNo (None, 4, 4, 150)      600
concatenate_40[0][0]
-----
activation_85 (Activation)    (None, 4, 4, 150)      0
batch_normalization_85[0][0]
-----

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```

conv2d_86 (Conv2D)                (None, 4, 4, 25)    3750
activation_85[0][0]
-----
batch_normalization_86 (BatchNo (None, 4, 4, 25)    100          conv2d_86[0][0]
-----
activation_86 (Activation)        (None, 4, 4, 25)    0
batch_normalization_86[0][0]
-----
conv2d_87 (Conv2D)                (None, 4, 4, 25)    5625
activation_86[0][0]
-----
concatenate_41 (Concatenate)      (None, 4, 4, 175)   0
concatenate_40[0][0]
                                         conv2d_87[0][0]
-----
batch_normalization_87 (BatchNo (None, 4, 4, 175)   700
concatenate_41[0][0]
-----
activation_87 (Activation)        (None, 4, 4, 175)   0
batch_normalization_87[0][0]
-----
conv2d_88 (Conv2D)                (None, 4, 4, 25)    4375
activation_87[0][0]
-----
batch_normalization_88 (BatchNo (None, 4, 4, 25)    100          conv2d_88[0][0]
-----
activation_88 (Activation)        (None, 4, 4, 25)    0
batch_normalization_88[0][0]
-----
conv2d_89 (Conv2D)                (None, 4, 4, 25)    5625
activation_88[0][0]
-----
concatenate_42 (Concatenate)      (None, 4, 4, 200)   0
concatenate_41[0][0]
                                         conv2d_89[0][0]
-----

```

batch_normalization_89 (BatchNo	(None, 4, 4, 200)	800	
concatenate_42[0][0]			

activation_89 (Activation)	(None, 4, 4, 200)	0	
batch_normalization_89[0][0]			

conv2d_90 (Conv2D)	(None, 4, 4, 25)	5000	
activation_89[0][0]			

batch_normalization_90 (BatchNo	(None, 4, 4, 25)	100	conv2d_90[0][0]

activation_90 (Activation)	(None, 4, 4, 25)	0	
batch_normalization_90[0][0]			

conv2d_91 (Conv2D)	(None, 4, 4, 25)	5625	
activation_90[0][0]			

concatenate_43 (Concatenate)	(None, 4, 4, 225)	0	
concatenate_42[0][0]			
			conv2d_91[0][0]

batch_normalization_91 (BatchNo	(None, 4, 4, 225)	900	
concatenate_43[0][0]			

activation_91 (Activation)	(None, 4, 4, 225)	0	
batch_normalization_91[0][0]			

conv2d_92 (Conv2D)	(None, 4, 4, 25)	5625	
activation_91[0][0]			

batch_normalization_92 (BatchNo	(None, 4, 4, 25)	100	conv2d_92[0][0]

activation_92 (Activation)	(None, 4, 4, 25)	0	
batch_normalization_92[0][0]			

conv2d_93 (Conv2D)	(None, 4, 4, 25)	5625	


```

activation_92[0][0]
-----
-----
concatenate_44 (Concatenate)      (None, 4, 4, 250)      0
concatenate_43[0][0]
                                                    conv2d_93[0][0]
-----
-----
batch_normalization_93 (BatchNo (None, 4, 4, 250)      1000
concatenate_44[0][0]
-----
-----
activation_93 (Activation)          (None, 4, 4, 250)      0
batch_normalization_93[0][0]
-----
-----
conv2d_94 (Conv2D)                  (None, 4, 4, 25)       6250
activation_93[0][0]
-----
-----
batch_normalization_94 (BatchNo (None, 4, 4, 25)       100      conv2d_94[0][0]
-----
-----
activation_94 (Activation)          (None, 4, 4, 25)       0
batch_normalization_94[0][0]
-----
-----
conv2d_95 (Conv2D)                  (None, 4, 4, 25)       5625
activation_94[0][0]
-----
-----
concatenate_45 (Concatenate)      (None, 4, 4, 275)      0
concatenate_44[0][0]
                                                    conv2d_95[0][0]
-----
-----
batch_normalization_95 (BatchNo (None, 4, 4, 275)      1100
concatenate_45[0][0]
-----
-----
activation_95 (Activation)          (None, 4, 4, 275)      0
batch_normalization_95[0][0]
-----
-----
conv2d_96 (Conv2D)                  (None, 4, 4, 25)       6875
activation_95[0][0]
-----
-----

```

```

batch_normalization_96 (BatchNo (None, 4, 4, 25)      100      conv2d_96[0] [0]
-----
activation_96 (Activation)      (None, 4, 4, 25)      0
batch_normalization_96[0] [0]
-----
conv2d_97 (Conv2D)      (None, 4, 4, 25)      5625
activation_96[0] [0]
-----
concatenate_46 (Concatenate)      (None, 4, 4, 300)      0
concatenate_45[0] [0]
conv2d_97[0] [0]
-----
batch_normalization_97 (BatchNo (None, 4, 4, 300)      1200
concatenate_46[0] [0]
-----
activation_97 (Activation)      (None, 4, 4, 300)      0
batch_normalization_97[0] [0]
-----
conv2d_98 (Conv2D)      (None, 4, 4, 25)      7500
activation_97[0] [0]
-----
batch_normalization_98 (BatchNo (None, 4, 4, 25)      100      conv2d_98[0] [0]
-----
activation_98 (Activation)      (None, 4, 4, 25)      0
batch_normalization_98[0] [0]
-----
conv2d_99 (Conv2D)      (None, 4, 4, 25)      5625
activation_98[0] [0]
-----
concatenate_47 (Concatenate)      (None, 4, 4, 325)      0
concatenate_46[0] [0]
conv2d_99[0] [0]
-----
batch_normalization_99 (BatchNo (None, 4, 4, 325)      1300
concatenate_47[0] [0]
-----

```

```

activation_99 (Activation)          (None, 4, 4, 325)    0
batch_normalization_99[0][0]
-----
average_pooling2d_3 (AveragePool) (None, 2, 2, 325)    0
activation_99[0][0]
-----
flatten (Flatten)                  (None, 1300)          0
average_pooling2d_3[0][0]
-----
dense (Dense)                      (None, 10)            13010   flatten[0][0]
=====
Total params: 548,704
Trainable params: 527,818
Non-trainable params: 20,886
-----

```

```
[18]: print(len(model5.layers))
```

355

```
[19]: model5.
      ↪ compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])

model5.fit(aug.
      ↪ flow(X_train, y_train, batch_size=batch_size), epochs=nb_epoch, batch_size=batch_size, verbose=1,
          steps_per_epoch=(len(X_train)//batch_size),
          ↵
      ↪ callbacks=[reduce_lr, lr_scheduler, csv_logger, early_stop, model_checkpoint],
          validation_data=(X_test, y_test))
```

Epoch 1/100

781/781 [=====] - 160s 153ms/step - loss: 1.8645 - accuracy: 0.3217 - val_loss: 2.1654 - val_accuracy: 0.2920

Epoch 00001: val_accuracy improved from -inf to 0.29200, saving model to model5_weights.best.hdf5

Epoch 2/100

781/781 [=====] - 118s 151ms/step - loss: 1.5126 - accuracy: 0.4451 - val_loss: 2.6054 - val_accuracy: 0.3066

Epoch 00002: val_accuracy improved from 0.29200 to 0.30660, saving model to model5_weights.best.hdf5

Epoch 3/100

781/781 [=====] - 119s 152ms/step - loss: 1.2816 -
accuracy: 0.5356 - val_loss: 1.7561 - val_accuracy: 0.5004

Epoch 00003: val_accuracy improved from 0.30660 to 0.50040, saving model to
model5_weights.best.hdf5

Epoch 4/100

781/781 [=====] - 120s 153ms/step - loss: 1.0908 -
accuracy: 0.6122 - val_loss: 1.1677 - val_accuracy: 0.6171

Epoch 00004: val_accuracy improved from 0.50040 to 0.61710, saving model to
model5_weights.best.hdf5

Epoch 5/100

781/781 [=====] - 119s 153ms/step - loss: 0.9554 -
accuracy: 0.6589 - val_loss: 1.0298 - val_accuracy: 0.6521

Epoch 00005: val_accuracy improved from 0.61710 to 0.65210, saving model to
model5_weights.best.hdf5

Epoch 6/100

781/781 [=====] - 120s 154ms/step - loss: 0.8619 -
accuracy: 0.6957 - val_loss: 1.2446 - val_accuracy: 0.6327

Epoch 00006: val_accuracy did not improve from 0.65210

Epoch 7/100

781/781 [=====] - 119s 153ms/step - loss: 0.7958 -
accuracy: 0.7197 - val_loss: 0.9859 - val_accuracy: 0.6887

Epoch 00007: val_accuracy improved from 0.65210 to 0.68870, saving model to
model5_weights.best.hdf5

Epoch 8/100

781/781 [=====] - 120s 153ms/step - loss: 0.7362 -
accuracy: 0.7413 - val_loss: 0.9728 - val_accuracy: 0.6962

Epoch 00008: val_accuracy improved from 0.68870 to 0.69620, saving model to
model5_weights.best.hdf5

Epoch 9/100

781/781 [=====] - 119s 153ms/step - loss: 0.6962 -
accuracy: 0.7567 - val_loss: 1.1996 - val_accuracy: 0.6552

Epoch 00009: val_accuracy did not improve from 0.69620

Epoch 10/100

781/781 [=====] - 120s 153ms/step - loss: 0.6580 -
accuracy: 0.7704 - val_loss: 0.8988 - val_accuracy: 0.7022

Epoch 00010: val_accuracy improved from 0.69620 to 0.70220, saving model to
model5_weights.best.hdf5

Epoch 11/100

781/781 [=====] - 120s 153ms/step - loss: 0.6234 -
accuracy: 0.7828 - val_loss: 0.9954 - val_accuracy: 0.7033

Epoch 00011: val_accuracy improved from 0.70220 to 0.70330, saving model to model5_weights.best.hdf5

Epoch 12/100

781/781 [=====] - 120s 153ms/step - loss: 0.5976 - accuracy: 0.7914 - val_loss: 0.8875 - val_accuracy: 0.7239

Epoch 00012: val_accuracy improved from 0.70330 to 0.72390, saving model to model5_weights.best.hdf5

Epoch 13/100

781/781 [=====] - 119s 153ms/step - loss: 0.5776 - accuracy: 0.8001 - val_loss: 0.8790 - val_accuracy: 0.7306

Epoch 00013: val_accuracy improved from 0.72390 to 0.73060, saving model to model5_weights.best.hdf5

Epoch 14/100

781/781 [=====] - 120s 153ms/step - loss: 0.5595 - accuracy: 0.8053 - val_loss: 1.1846 - val_accuracy: 0.6676

Epoch 00014: val_accuracy did not improve from 0.73060

Epoch 15/100

781/781 [=====] - 120s 153ms/step - loss: 0.5326 - accuracy: 0.8162 - val_loss: 0.6720 - val_accuracy: 0.7863

Epoch 00015: val_accuracy improved from 0.73060 to 0.78630, saving model to model5_weights.best.hdf5

Epoch 16/100

781/781 [=====] - 120s 153ms/step - loss: 0.5182 - accuracy: 0.8208 - val_loss: 0.6716 - val_accuracy: 0.7794

Epoch 00016: val_accuracy did not improve from 0.78630

Epoch 17/100

781/781 [=====] - 119s 153ms/step - loss: 0.5030 - accuracy: 0.8265 - val_loss: 0.9736 - val_accuracy: 0.7291

Epoch 00017: val_accuracy did not improve from 0.78630

Epoch 18/100

781/781 [=====] - 120s 153ms/step - loss: 0.4860 - accuracy: 0.8317 - val_loss: 0.7766 - val_accuracy: 0.7814

Epoch 00018: val_accuracy did not improve from 0.78630

Epoch 19/100

781/781 [=====] - 120s 153ms/step - loss: 0.4732 - accuracy: 0.8361 - val_loss: 1.1807 - val_accuracy: 0.6949

Epoch 00019: val_accuracy did not improve from 0.78630

Epoch 20/100

781/781 [=====] - 120s 153ms/step - loss: 0.4582 -

accuracy: 0.8396 - val_loss: 0.6134 - val_accuracy: 0.8037

Epoch 00020: val_accuracy improved from 0.78630 to 0.80370, saving model to model5_weights.best.hdf5

Epoch 21/100

781/781 [=====] - 119s 153ms/step - loss: 0.4470 - accuracy: 0.8440 - val_loss: 0.6685 - val_accuracy: 0.7864

Epoch 00021: val_accuracy did not improve from 0.80370

Epoch 22/100

781/781 [=====] - 120s 153ms/step - loss: 0.4347 - accuracy: 0.8498 - val_loss: 0.5627 - val_accuracy: 0.8141

Epoch 00022: val_accuracy improved from 0.80370 to 0.81410, saving model to model5_weights.best.hdf5

Epoch 23/100

781/781 [=====] - 119s 153ms/step - loss: 0.4277 - accuracy: 0.8514 - val_loss: 0.6146 - val_accuracy: 0.8029

Epoch 00023: val_accuracy did not improve from 0.81410

Epoch 24/100

781/781 [=====] - 119s 152ms/step - loss: 0.4198 - accuracy: 0.8548 - val_loss: 0.5680 - val_accuracy: 0.8143

Epoch 00024: val_accuracy improved from 0.81410 to 0.81430, saving model to model5_weights.best.hdf5

Epoch 25/100

781/781 [=====] - 120s 154ms/step - loss: 0.4065 - accuracy: 0.8597 - val_loss: 0.7392 - val_accuracy: 0.7805

Epoch 00025: val_accuracy did not improve from 0.81430

Epoch 26/100

781/781 [=====] - 120s 153ms/step - loss: 0.3311 - accuracy: 0.8833 - val_loss: 0.3738 - val_accuracy: 0.8762

Epoch 00026: val_accuracy improved from 0.81430 to 0.87620, saving model to model5_weights.best.hdf5

Epoch 27/100

781/781 [=====] - 120s 153ms/step - loss: 0.3031 - accuracy: 0.8946 - val_loss: 0.3818 - val_accuracy: 0.8785

Epoch 00027: val_accuracy improved from 0.87620 to 0.87850, saving model to model5_weights.best.hdf5

Epoch 28/100

781/781 [=====] - 120s 153ms/step - loss: 0.2953 - accuracy: 0.8979 - val_loss: 0.3645 - val_accuracy: 0.8801

Epoch 00028: val_accuracy improved from 0.87850 to 0.88010, saving model to

```

model5_weights.best.hdf5
Epoch 29/100
781/781 [=====] - 120s 154ms/step - loss: 0.2857 -
accuracy: 0.9000 - val_loss: 0.3744 - val_accuracy: 0.8788

Epoch 00029: val_accuracy did not improve from 0.88010
Epoch 30/100
781/781 [=====] - 120s 154ms/step - loss: 0.2830 -
accuracy: 0.9021 - val_loss: 0.3479 - val_accuracy: 0.8904

Epoch 00030: val_accuracy improved from 0.88010 to 0.89040, saving model to
model5_weights.best.hdf5
Epoch 31/100
781/781 [=====] - 120s 153ms/step - loss: 0.2742 -
accuracy: 0.9052 - val_loss: 0.3228 - val_accuracy: 0.8932

Epoch 00031: val_accuracy improved from 0.89040 to 0.89320, saving model to
model5_weights.best.hdf5
Epoch 32/100
781/781 [=====] - 121s 154ms/step - loss: 0.2747 -
accuracy: 0.9045 - val_loss: 0.3600 - val_accuracy: 0.8826

Epoch 00032: val_accuracy did not improve from 0.89320
Epoch 33/100
781/781 [=====] - 120s 154ms/step - loss: 0.2678 -
accuracy: 0.9067 - val_loss: 0.3407 - val_accuracy: 0.8885

Epoch 00033: val_accuracy did not improve from 0.89320
Epoch 34/100
781/781 [=====] - 120s 154ms/step - loss: 0.2657 -
accuracy: 0.9079 - val_loss: 0.3401 - val_accuracy: 0.8903

Epoch 00034: val_accuracy did not improve from 0.89320
Epoch 35/100
781/781 [=====] - 122s 156ms/step - loss: 0.2603 -
accuracy: 0.9085 - val_loss: 0.3519 - val_accuracy: 0.8881

Epoch 00035: val_accuracy did not improve from 0.89320
Epoch 36/100
781/781 [=====] - 122s 156ms/step - loss: 0.2561 -
accuracy: 0.9113 - val_loss: 0.3397 - val_accuracy: 0.8891

Epoch 00036: val_accuracy did not improve from 0.89320
Epoch 37/100
781/781 [=====] - 122s 156ms/step - loss: 0.2539 -
accuracy: 0.9115 - val_loss: 0.3564 - val_accuracy: 0.8879

Epoch 00037: val_accuracy did not improve from 0.89320

```

Epoch 38/100
781/781 [=====] - 121s 155ms/step - loss: 0.2542 - accuracy: 0.9126 - val_loss: 0.3587 - val_accuracy: 0.8870

Epoch 00038: val_accuracy did not improve from 0.89320

Epoch 39/100
781/781 [=====] - 121s 155ms/step - loss: 0.2510 - accuracy: 0.9121 - val_loss: 0.3576 - val_accuracy: 0.8875

Epoch 00039: val_accuracy did not improve from 0.89320

Epoch 40/100
781/781 [=====] - 121s 155ms/step - loss: 0.2467 - accuracy: 0.9143 - val_loss: 0.3505 - val_accuracy: 0.8896

Epoch 00040: val_accuracy did not improve from 0.89320

Epoch 41/100
781/781 [=====] - 121s 155ms/step - loss: 0.2480 - accuracy: 0.9129 - val_loss: 0.3604 - val_accuracy: 0.8866

Epoch 00041: val_accuracy did not improve from 0.89320

Epoch 42/100
781/781 [=====] - 121s 155ms/step - loss: 0.2457 - accuracy: 0.9144 - val_loss: 0.3695 - val_accuracy: 0.8819

Epoch 00042: val_accuracy did not improve from 0.89320

Epoch 43/100
781/781 [=====] - 121s 155ms/step - loss: 0.2406 - accuracy: 0.9162 - val_loss: 0.3325 - val_accuracy: 0.8966

Epoch 00043: val_accuracy improved from 0.89320 to 0.89660, saving model to model5_weights.best.hdf5

Epoch 44/100
781/781 [=====] - 121s 155ms/step - loss: 0.2363 - accuracy: 0.9167 - val_loss: 0.3364 - val_accuracy: 0.8912

Epoch 00044: val_accuracy did not improve from 0.89660

Epoch 45/100
781/781 [=====] - 121s 155ms/step - loss: 0.2369 - accuracy: 0.9161 - val_loss: 0.3447 - val_accuracy: 0.8910

Epoch 00045: val_accuracy did not improve from 0.89660

Epoch 46/100
781/781 [=====] - 121s 155ms/step - loss: 0.2347 - accuracy: 0.9184 - val_loss: 0.3308 - val_accuracy: 0.8934

Epoch 00046: val_accuracy did not improve from 0.89660

Epoch 47/100
781/781 [=====] - 121s 155ms/step - loss: 0.2294 -

accuracy: 0.9199 - val_loss: 0.3310 - val_accuracy: 0.8966

Epoch 00047: val_accuracy did not improve from 0.89660

Epoch 48/100

781/781 [=====] - 122s 156ms/step - loss: 0.2279 -
accuracy: 0.9204 - val_loss: 0.3455 - val_accuracy: 0.8918

Epoch 00048: val_accuracy did not improve from 0.89660

Epoch 49/100

781/781 [=====] - 122s 156ms/step - loss: 0.2311 -
accuracy: 0.9192 - val_loss: 0.3227 - val_accuracy: 0.8988

Epoch 00049: val_accuracy improved from 0.89660 to 0.89880, saving model to
model5_weights.best.hdf5

Epoch 50/100

781/781 [=====] - 121s 155ms/step - loss: 0.2281 -
accuracy: 0.9198 - val_loss: 0.3356 - val_accuracy: 0.8942

Epoch 00050: val_accuracy did not improve from 0.89880

Epoch 51/100

781/781 [=====] - 121s 155ms/step - loss: 0.2194 -
accuracy: 0.9224 - val_loss: 0.3308 - val_accuracy: 0.8966

Epoch 00051: val_accuracy did not improve from 0.89880

Epoch 52/100

781/781 [=====] - 121s 155ms/step - loss: 0.2182 -
accuracy: 0.9234 - val_loss: 0.3265 - val_accuracy: 0.8978

Epoch 00052: val_accuracy did not improve from 0.89880

Epoch 53/100

781/781 [=====] - 121s 155ms/step - loss: 0.2164 -
accuracy: 0.9233 - val_loss: 0.3286 - val_accuracy: 0.8979

Epoch 00053: val_accuracy did not improve from 0.89880

Epoch 54/100

781/781 [=====] - 121s 154ms/step - loss: 0.2166 -
accuracy: 0.9241 - val_loss: 0.3309 - val_accuracy: 0.8973

Epoch 00054: val_accuracy did not improve from 0.89880

Epoch 55/100

781/781 [=====] - 121s 155ms/step - loss: 0.2165 -
accuracy: 0.9248 - val_loss: 0.3279 - val_accuracy: 0.8978

Epoch 00055: val_accuracy did not improve from 0.89880

Epoch 56/100

781/781 [=====] - 121s 155ms/step - loss: 0.2141 -
accuracy: 0.9245 - val_loss: 0.3245 - val_accuracy: 0.8983

Epoch 00056: val_accuracy did not improve from 0.89880
Epoch 57/100
781/781 [=====] - 121s 155ms/step - loss: 0.2153 -
accuracy: 0.9244 - val_loss: 0.3288 - val_accuracy: 0.8982

Epoch 00057: val_accuracy did not improve from 0.89880
Epoch 58/100
781/781 [=====] - 121s 155ms/step - loss: 0.2150 -
accuracy: 0.9241 - val_loss: 0.3253 - val_accuracy: 0.8985

Epoch 00058: val_accuracy did not improve from 0.89880
Epoch 59/100
781/781 [=====] - 121s 155ms/step - loss: 0.2099 -
accuracy: 0.9261 - val_loss: 0.3230 - val_accuracy: 0.8998

Epoch 00059: val_accuracy improved from 0.89880 to 0.89980, saving model to
model5_weights.best.hdf5
Epoch 60/100
781/781 [=====] - 121s 155ms/step - loss: 0.2105 -
accuracy: 0.9263 - val_loss: 0.3285 - val_accuracy: 0.8978

Epoch 00060: val_accuracy did not improve from 0.89980
Epoch 61/100
781/781 [=====] - 121s 155ms/step - loss: 0.2122 -
accuracy: 0.9259 - val_loss: 0.3254 - val_accuracy: 0.8990

Epoch 00061: val_accuracy did not improve from 0.89980
Epoch 62/100
781/781 [=====] - 121s 155ms/step - loss: 0.2066 -
accuracy: 0.9275 - val_loss: 0.3238 - val_accuracy: 0.8991

Epoch 00062: val_accuracy did not improve from 0.89980
Epoch 63/100
781/781 [=====] - 121s 155ms/step - loss: 0.2088 -
accuracy: 0.9268 - val_loss: 0.3285 - val_accuracy: 0.8968

Epoch 00063: val_accuracy did not improve from 0.89980
Epoch 64/100
781/781 [=====] - 122s 156ms/step - loss: 0.2140 -
accuracy: 0.9231 - val_loss: 0.3257 - val_accuracy: 0.8979

Epoch 00064: val_accuracy did not improve from 0.89980
Epoch 65/100
781/781 [=====] - 121s 155ms/step - loss: 0.2067 -
accuracy: 0.9280 - val_loss: 0.3275 - val_accuracy: 0.8980

Epoch 00065: val_accuracy did not improve from 0.89980
Epoch 66/100

```
781/781 [=====] - 121s 155ms/step - loss: 0.2089 -  
accuracy: 0.9256 - val_loss: 0.3267 - val_accuracy: 0.8980
```

Epoch 00066: val_accuracy did not improve from 0.89980

Epoch 67/100

```
781/781 [=====] - 121s 155ms/step - loss: 0.2127 -  
accuracy: 0.9258 - val_loss: 0.3208 - val_accuracy: 0.8991
```

Epoch 00067: val_accuracy did not improve from 0.89980

Epoch 68/100

```
781/781 [=====] - 121s 155ms/step - loss: 0.2119 -  
accuracy: 0.9255 - val_loss: 0.3262 - val_accuracy: 0.8983
```

Epoch 00068: val_accuracy did not improve from 0.89980

Epoch 69/100

```
781/781 [=====] - 121s 155ms/step - loss: 0.2091 -  
accuracy: 0.9264 - val_loss: 0.3237 - val_accuracy: 0.8991
```

Epoch 00069: val_accuracy did not improve from 0.89980

Epoch 70/100

```
781/781 [=====] - 121s 155ms/step - loss: 0.2090 -  
accuracy: 0.9274 - val_loss: 0.3212 - val_accuracy: 0.9015
```

Epoch 00070: val_accuracy improved from 0.89980 to 0.90150, saving model to
model5_weights.best.hdf5

Epoch 71/100

```
781/781 [=====] - 121s 155ms/step - loss: 0.2102 -  
accuracy: 0.9264 - val_loss: 0.3249 - val_accuracy: 0.8994
```

Epoch 00071: val_accuracy did not improve from 0.90150

Epoch 72/100

```
781/781 [=====] - 121s 154ms/step - loss: 0.2115 -  
accuracy: 0.9262 - val_loss: 0.3298 - val_accuracy: 0.8973
```

Epoch 00072: val_accuracy did not improve from 0.90150

[19]: <tensorflow.python.keras.callbacks.History at 0x7f80f632c350>

```
[20]: import pandas as pd  
import numpy as np  
training_log = pd.read_csv('/content/training.log')  
training_log.head(100)
```

```
[20]:      epoch  accuracy      loss      lr  val_accuracy  val_loss  
0         0  0.321692  1.864464  0.0100         0.2920  2.165388  
1         1  0.445070  1.512568  0.0100         0.3066  2.605386  
2         2  0.535646  1.281641  0.0100         0.5004  1.756118
```

3	3	0.612224	1.090826	0.0100	0.6171	1.167669
4	4	0.658863	0.955366	0.0100	0.6521	1.029768
..
67	67	0.925485	0.211873	0.0001	0.8983	0.326231
68	68	0.926426	0.209126	0.0001	0.8991	0.323680
69	69	0.927447	0.209021	0.0001	0.9015	0.321206
70	70	0.926446	0.210156	0.0001	0.8994	0.324877
71	71	0.926206	0.211468	0.0001	0.8973	0.329835

[72 rows x 6 columns]

```
[21]: model5.load_weights('/content/model5_weights.best.hdf5')
model5.
      ↪ compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
print("Model created and weights loaded from file")
```

Model created and weights loaded from file

```
[22]: score = model5.evaluate(X_test, y_test, verbose=0)
print("Test loss = ", score[0])
print("Test accuracy = ", score[1])
```

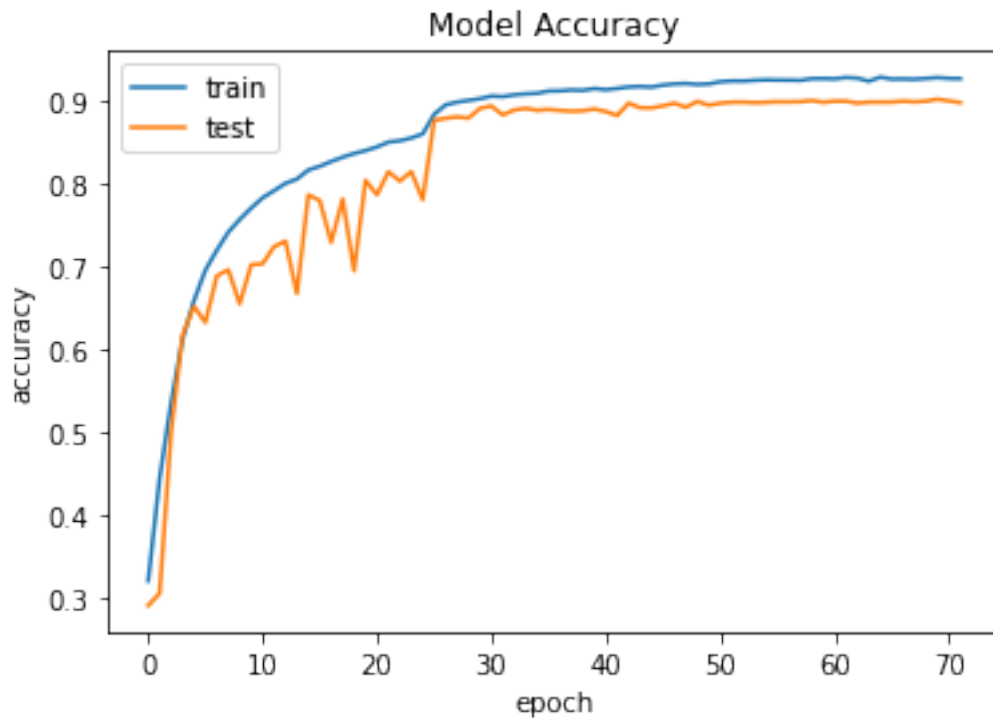
Test loss = 0.32120639085769653

Test accuracy = 0.9014999866485596

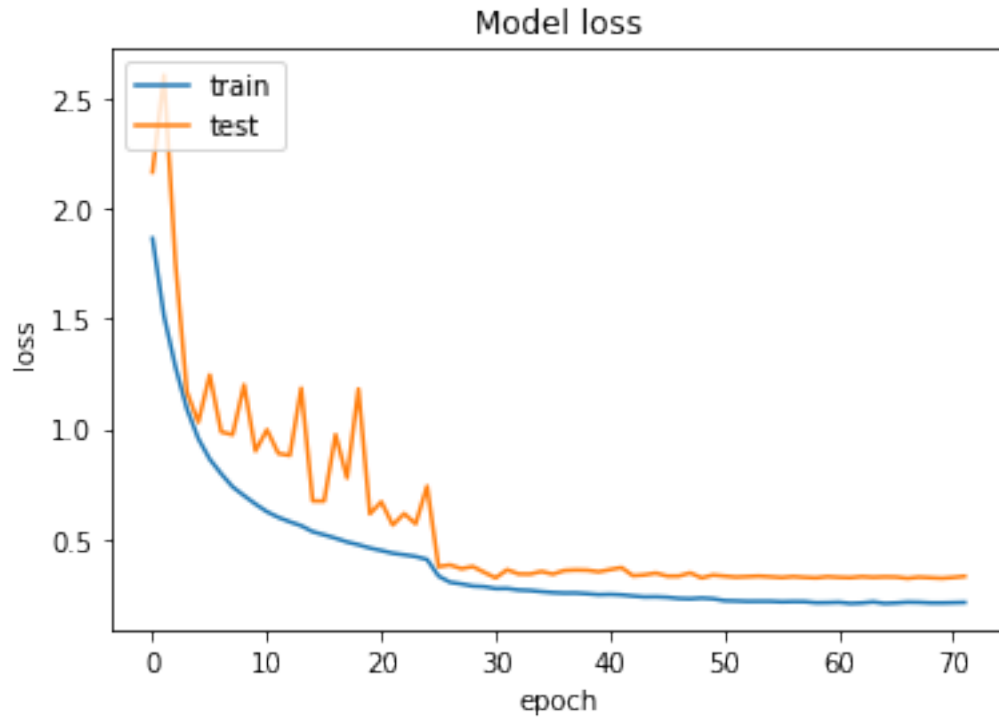
1.3.5 2.5 Plotting loss and accuracy of Model 5 above

```
[ ]: %matplotlib notebook
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import time
```

```
[24]: #https://machinelearningmastery.com/
      ↪ display-deep-learning-model-training-history-in-keras/
plt.plot(training_log['accuracy'])
plt.plot(training_log['val_accuracy'])
plt.title('Model Accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```



```
[25]: #https://machinelearningmastery.com/  
      ↪display-deep-learning-model-training-history-in-keras/  
plt.plot(training_log['loss'])  
plt.plot(training_log['val_loss'])  
plt.title('Model loss')  
plt.xlabel('epoch')  
plt.ylabel('loss')  
plt.legend(['train', 'test'], loc='upper left')  
plt.show()
```



1.4 3. Conclusion

```
[1]: from prettytable import PrettyTable

x = PrettyTable()
x.field_names = ['Growth Rate', 'Compression', '# of Blocks', 'Test Accuracy']
x.add_row([36, 0.7, 12, 90.149])

print(x)
```

```
+-----+-----+-----+-----+
| Growth Rate | Compression | # of Blocks | Test Accuracy |
+-----+-----+-----+-----+
|      36      |      0.7      |      12      |      90.149      |
+-----+-----+-----+-----+
```

Summary:

I have used Keras callbacks to adjust the learning rate as per the performance of the model (ReduceLROnPlateau, LearningRate Scheduler).

Additional links and resuorces:

1. 2016 DenseNet paper summary: https://www.youtube.com/watch?v=hSC_0S8Zf9s

2. Separable Depth wise convolutions: <https://towardsdatascience.com/a-basic-introduction-to-separable-convolutions-b99ec3102728>
3. Review DenseNet image classification: <https://towardsdatascience.com/review-densenet-image-classification-b6631a8ef803>

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