

DenseNet_CIFAR_Final

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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 initialize 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 Block

```

[ ]: # 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)
```

```
[ ]: from tensorflow.keras.callbacks import ModelCheckpoint  
#https://machinelearningmastery.com/check-point-deep-learning-models-keras/  
filepath="weights.best.hdf5"
```

```
model_checkpoint = ↳ModelCheckpoint(filepath,monitor='val_accuracy',save_best_only=True,verbose=1)
```

1.3.4 2.4 Growth rate(num_filter)=24, compression = 0.5, number of blocks = 12

```
[ ]: # Hyperparameters
batch_size = 128
num_classes = 10
nb_epoch = 100
l = 12
num_filter = 24
compression = 0.5
dropout_rate = 0.2
```

```
[ ]: 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)
```

```
[ ]: model3 = Model(inputs = [input], outputs = [output])
model3.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 32, 32, 3)]	0	
conv2d (Conv2D)	(None, 32, 32, 24)	648	input_1[0][0]
batch_normalization (BatchNorma	(None, 32, 32, 24)	96	conv2d[0][0]

```

-----
activation (Activation)          (None, 32, 32, 24)    0
batch_normalization[0][0]
-----

conv2d_1 (Conv2D)                (None, 32, 32, 12)   288
activation[0][0]
-----

batch_normalization_1 (BatchNor (None, 32, 32, 12)   48          conv2d_1[0][0]
-----

activation_1 (Activation)        (None, 32, 32, 12)    0
batch_normalization_1[0][0]
-----

conv2d_2 (Conv2D)                (None, 32, 32, 12)  1296
activation_1[0][0]
-----

dropout (Dropout)                (None, 32, 32, 12)    0          conv2d_2[0][0]
-----

concatenate (Concatenate)        (None, 32, 32, 36)    0          conv2d[0][0]
                                         dropout[0][0]
-----

batch_normalization_2 (BatchNor (None, 32, 32, 36)   144
concatenate[0][0]
-----

activation_2 (Activation)        (None, 32, 32, 36)    0
batch_normalization_2[0][0]
-----

conv2d_3 (Conv2D)                (None, 32, 32, 12)   432
activation_2[0][0]
-----

batch_normalization_3 (BatchNor (None, 32, 32, 12)   48          conv2d_3[0][0]
-----

activation_3 (Activation)        (None, 32, 32, 12)    0
batch_normalization_3[0][0]
-----

conv2d_4 (Conv2D)                (None, 32, 32, 12)  1296

```

```

activation_3[0][0]
-----
dropout_1 (Dropout)          (None, 32, 32, 12)    0          conv2d_4[0][0]
-----
concatenate_1 (Concatenate)  (None, 32, 32, 48)    0
concatenate[0][0]
                                         dropout_1[0][0]
-----
batch_normalization_4 (BatchNor (None, 32, 32, 48)    192
concatenate_1[0][0]
-----
activation_4 (Activation)    (None, 32, 32, 48)    0
batch_normalization_4[0][0]
-----
conv2d_5 (Conv2D)           (None, 32, 32, 12)    576
activation_4[0][0]
-----
batch_normalization_5 (BatchNor (None, 32, 32, 12)    48          conv2d_5[0][0]
-----
activation_5 (Activation)    (None, 32, 32, 12)    0
batch_normalization_5[0][0]
-----
conv2d_6 (Conv2D)           (None, 32, 32, 12)    1296
activation_5[0][0]
-----
dropout_2 (Dropout)          (None, 32, 32, 12)    0          conv2d_6[0][0]
-----
concatenate_2 (Concatenate)  (None, 32, 32, 60)    0
concatenate_1[0][0]
                                         dropout_2[0][0]
-----
batch_normalization_6 (BatchNor (None, 32, 32, 60)    240
concatenate_2[0][0]
-----
activation_6 (Activation)    (None, 32, 32, 60)    0
batch_normalization_6[0][0]

```

conv2d_7 (Conv2D)	(None, 32, 32, 12)	720	
activation_6[0][0]			

batch_normalization_7 (BatchNor	(None, 32, 32, 12)	48	conv2d_7[0][0]

activation_7 (Activation)	(None, 32, 32, 12)	0	
batch_normalization_7[0][0]			

conv2d_8 (Conv2D)	(None, 32, 32, 12)	1296	
activation_7[0][0]			

dropout_3 (Dropout)	(None, 32, 32, 12)	0	conv2d_8[0][0]

concatenate_3 (Concatenate)	(None, 32, 32, 72)	0	
concatenate_2[0][0]			

			dropout_3[0][0]

batch_normalization_8 (BatchNor	(None, 32, 32, 72)	288	
concatenate_3[0][0]			

activation_8 (Activation)	(None, 32, 32, 72)	0	
batch_normalization_8[0][0]			

conv2d_9 (Conv2D)	(None, 32, 32, 12)	864	
activation_8[0][0]			

batch_normalization_9 (BatchNor	(None, 32, 32, 12)	48	conv2d_9[0][0]

activation_9 (Activation)	(None, 32, 32, 12)	0	
batch_normalization_9[0][0]			

conv2d_10 (Conv2D)	(None, 32, 32, 12)	1296	
activation_9[0][0]			

dropout_4 (Dropout)	(None, 32, 32, 12)	0	conv2d_10[0][0]

concatenate_4 (Concatenate)	(None, 32, 32, 84)	0	
concatenate_3[0][0]			dropout_4[0][0]

batch_normalization_10 (Batch Normalization)	(None, 32, 32, 84)	336	
concatenate_4[0][0]			

activation_10 (Activation)	(None, 32, 32, 84)	0	
batch_normalization_10[0][0]			

conv2d_11 (Conv2D)	(None, 32, 32, 12)	1008	
activation_10[0][0]			

batch_normalization_11 (Batch Normalization)	(None, 32, 32, 12)	48	conv2d_11[0][0]

activation_11 (Activation)	(None, 32, 32, 12)	0	
batch_normalization_11[0][0]			

conv2d_12 (Conv2D)	(None, 32, 32, 12)	1296	
activation_11[0][0]			

dropout_5 (Dropout)	(None, 32, 32, 12)	0	conv2d_12[0][0]

concatenate_5 (Concatenate)	(None, 32, 32, 96)	0	
concatenate_4[0][0]			dropout_5[0][0]

batch_normalization_12 (Batch Normalization)	(None, 32, 32, 96)	384	
concatenate_5[0][0]			

activation_12 (Activation)	(None, 32, 32, 96)	0	
batch_normalization_12[0][0]			

conv2d_13 (Conv2D)	(None, 32, 32, 12)	1152	

```

activation_12[0][0]
-----
batch_normalization_13 (BatchNo (None, 32, 32, 12) 48 conv2d_13[0][0]
-----
activation_13 (Activation) (None, 32, 32, 12) 0
batch_normalization_13[0][0]
-----
conv2d_14 (Conv2D) (None, 32, 32, 12) 1296
activation_13[0][0]
-----
dropout_6 (Dropout) (None, 32, 32, 12) 0 conv2d_14[0][0]
-----
concatenate_6 (Concatenate) (None, 32, 32, 108) 0
concatenate_5[0][0]
dropout_6[0][0]
-----
batch_normalization_14 (BatchNo (None, 32, 32, 108) 432
concatenate_6[0][0]
-----
activation_14 (Activation) (None, 32, 32, 108) 0
batch_normalization_14[0][0]
-----
conv2d_15 (Conv2D) (None, 32, 32, 12) 1296
activation_14[0][0]
-----
batch_normalization_15 (BatchNo (None, 32, 32, 12) 48 conv2d_15[0][0]
-----
activation_15 (Activation) (None, 32, 32, 12) 0
batch_normalization_15[0][0]
-----
conv2d_16 (Conv2D) (None, 32, 32, 12) 1296
activation_15[0][0]
-----
dropout_7 (Dropout) (None, 32, 32, 12) 0 conv2d_16[0][0]
-----

```

```

concatenate_7 (Concatenate)      (None, 32, 32, 120)  0
concatenate_6[0][0]

dropout_7[0][0]

-----

batch_normalization_16 (BatchNo (None, 32, 32, 120)  480
concatenate_7[0][0]

-----

activation_16 (Activation)        (None, 32, 32, 120)  0
batch_normalization_16[0][0]

-----

conv2d_17 (Conv2D)                (None, 32, 32, 12)   1440
activation_16[0][0]

-----

batch_normalization_17 (BatchNo (None, 32, 32, 12)   48      conv2d_17[0][0]

-----

activation_17 (Activation)        (None, 32, 32, 12)   0
batch_normalization_17[0][0]

-----

conv2d_18 (Conv2D)                (None, 32, 32, 12)   1296
activation_17[0][0]

-----

dropout_8 (Dropout)               (None, 32, 32, 12)   0      conv2d_18[0][0]

-----

concatenate_8 (Concatenate)      (None, 32, 32, 132)  0
concatenate_7[0][0]

dropout_8[0][0]

-----

batch_normalization_18 (BatchNo (None, 32, 32, 132)  528
concatenate_8[0][0]

-----

activation_18 (Activation)        (None, 32, 32, 132)  0
batch_normalization_18[0][0]

-----

conv2d_19 (Conv2D)                (None, 32, 32, 12)   1584
activation_18[0][0]

-----

```

```

batch_normalization_19 (BatchNo (None, 32, 32, 12) 48 conv2d_19[0][0]
-----
activation_19 (Activation) (None, 32, 32, 12) 0
batch_normalization_19[0][0]
-----
conv2d_20 (Conv2D) (None, 32, 32, 12) 1296
activation_19[0][0]
-----
dropout_9 (Dropout) (None, 32, 32, 12) 0 conv2d_20[0][0]
-----
concatenate_9 (Concatenate) (None, 32, 32, 144) 0
concatenate_8[0][0]
dropout_9[0][0]
-----
batch_normalization_20 (BatchNo (None, 32, 32, 144) 576
concatenate_9[0][0]
-----
activation_20 (Activation) (None, 32, 32, 144) 0
batch_normalization_20[0][0]
-----
conv2d_21 (Conv2D) (None, 32, 32, 12) 1728
activation_20[0][0]
-----
batch_normalization_21 (BatchNo (None, 32, 32, 12) 48 conv2d_21[0][0]
-----
activation_21 (Activation) (None, 32, 32, 12) 0
batch_normalization_21[0][0]
-----
conv2d_22 (Conv2D) (None, 32, 32, 12) 1296
activation_21[0][0]
-----
dropout_10 (Dropout) (None, 32, 32, 12) 0 conv2d_22[0][0]
-----
concatenate_10 (Concatenate) (None, 32, 32, 156) 0
concatenate_9[0][0]
dropout_10[0][0]

```

```

-----
batch_normalization_22 (BatchNo (None, 32, 32, 156) 624
concatenate_10[0][0]
-----
activation_22 (Activation) (None, 32, 32, 156) 0
batch_normalization_22[0][0]
-----
conv2d_23 (Conv2D) (None, 32, 32, 12) 1872
activation_22[0][0]
-----
batch_normalization_23 (BatchNo (None, 32, 32, 12) 48 conv2d_23[0][0]
-----
activation_23 (Activation) (None, 32, 32, 12) 0
batch_normalization_23[0][0]
-----
conv2d_24 (Conv2D) (None, 32, 32, 12) 1296
activation_23[0][0]
-----
dropout_11 (Dropout) (None, 32, 32, 12) 0 conv2d_24[0][0]
-----
concatenate_11 (Concatenate) (None, 32, 32, 168) 0
concatenate_10[0][0]
dropout_11[0][0]
-----
batch_normalization_24 (BatchNo (None, 32, 32, 168) 672
concatenate_11[0][0]
-----
activation_24 (Activation) (None, 32, 32, 168) 0
batch_normalization_24[0][0]
-----
conv2d_25 (Conv2D) (None, 32, 32, 12) 2016
activation_24[0][0]
-----
dropout_12 (Dropout) (None, 32, 32, 12) 0 conv2d_25[0][0]
-----

```

average_pooling2d (AveragePooli	(None, 16, 16, 12)	0	
dropout_12[0][0]			

batch_normalization_25 (BatchNo	(None, 16, 16, 12)	48	
average_pooling2d[0][0]			

activation_25 (Activation)	(None, 16, 16, 12)	0	
batch_normalization_25[0][0]			

conv2d_26 (Conv2D)	(None, 16, 16, 12)	144	
activation_25[0][0]			

batch_normalization_26 (BatchNo	(None, 16, 16, 12)	48	conv2d_26[0][0]

activation_26 (Activation)	(None, 16, 16, 12)	0	
batch_normalization_26[0][0]			

conv2d_27 (Conv2D)	(None, 16, 16, 12)	1296	
activation_26[0][0]			

dropout_13 (Dropout)	(None, 16, 16, 12)	0	conv2d_27[0][0]

concatenate_12 (Concatenate)	(None, 16, 16, 24)	0	
average_pooling2d[0][0]			
dropout_13[0][0]			

batch_normalization_27 (BatchNo	(None, 16, 16, 24)	96	
concatenate_12[0][0]			

activation_27 (Activation)	(None, 16, 16, 24)	0	
batch_normalization_27[0][0]			

conv2d_28 (Conv2D)	(None, 16, 16, 12)	288	
activation_27[0][0]			

batch_normalization_28 (BatchNo	(None, 16, 16, 12)	48	conv2d_28[0][0]

```

-----
activation_28 (Activation)      (None, 16, 16, 12)    0
batch_normalization_28[0][0]

-----

conv2d_29 (Conv2D)             (None, 16, 16, 12)    1296
activation_28[0][0]

-----

dropout_14 (Dropout)           (None, 16, 16, 12)    0          conv2d_29[0][0]

-----

concatenate_13 (Concatenate)   (None, 16, 16, 36)    0
concatenate_12[0][0]
dropout_14[0][0]

-----

batch_normalization_29 (BatchNo (None, 16, 16, 36)    144
concatenate_13[0][0]

-----

activation_29 (Activation)      (None, 16, 16, 36)    0
batch_normalization_29[0][0]

-----

conv2d_30 (Conv2D)             (None, 16, 16, 12)    432
activation_29[0][0]

-----

batch_normalization_30 (BatchNo (None, 16, 16, 12)    48          conv2d_30[0][0]

-----

activation_30 (Activation)      (None, 16, 16, 12)    0
batch_normalization_30[0][0]

-----

conv2d_31 (Conv2D)             (None, 16, 16, 12)    1296
activation_30[0][0]

-----

dropout_15 (Dropout)           (None, 16, 16, 12)    0          conv2d_31[0][0]

-----

concatenate_14 (Concatenate)   (None, 16, 16, 48)    0
concatenate_13[0][0]
dropout_15[0][0]
-----

```



```

-----
batch_normalization_31 (BatchNo (None, 16, 16, 48)    192
concatenate_14[0][0]
-----
-----
activation_31 (Activation)      (None, 16, 16, 48)    0
batch_normalization_31[0][0]
-----
-----
conv2d_32 (Conv2D)              (None, 16, 16, 12)    576
activation_31[0][0]
-----
-----
batch_normalization_32 (BatchNo (None, 16, 16, 12)    48          conv2d_32[0][0]
-----
-----
activation_32 (Activation)      (None, 16, 16, 12)    0
batch_normalization_32[0][0]
-----
-----
conv2d_33 (Conv2D)              (None, 16, 16, 12)    1296
activation_32[0][0]
-----
-----
dropout_16 (Dropout)            (None, 16, 16, 12)    0          conv2d_33[0][0]
-----
-----
concatenate_15 (Concatenate)    (None, 16, 16, 60)    0
concatenate_14[0][0]
dropout_16[0][0]
-----
-----
batch_normalization_33 (BatchNo (None, 16, 16, 60)    240
concatenate_15[0][0]
-----
-----
activation_33 (Activation)      (None, 16, 16, 60)    0
batch_normalization_33[0][0]
-----
-----
conv2d_34 (Conv2D)              (None, 16, 16, 12)    720
activation_33[0][0]
-----
-----
batch_normalization_34 (BatchNo (None, 16, 16, 12)    48          conv2d_34[0][0]
-----
-----
activation_34 (Activation)      (None, 16, 16, 12)    0

```

batch_normalization_34[0][0]

conv2d_35 (Conv2D) (None, 16, 16, 12) 1296
activation_34[0][0]

dropout_17 (Dropout) (None, 16, 16, 12) 0 conv2d_35[0][0]

concatenate_16 (Concatenate) (None, 16, 16, 72) 0
concatenate_15[0][0]
dropout_17[0][0]

batch_normalization_35 (BatchNormaliz (None, 16, 16, 72) 288
concatenate_16[0][0]

activation_35 (Activation) (None, 16, 16, 72) 0
batch_normalization_35[0][0]

conv2d_36 (Conv2D) (None, 16, 16, 12) 864
activation_35[0][0]

batch_normalization_36 (BatchNormaliz (None, 16, 16, 12) 48 conv2d_36[0][0]

activation_36 (Activation) (None, 16, 16, 12) 0
batch_normalization_36[0][0]

conv2d_37 (Conv2D) (None, 16, 16, 12) 1296
activation_36[0][0]

dropout_18 (Dropout) (None, 16, 16, 12) 0 conv2d_37[0][0]

concatenate_17 (Concatenate) (None, 16, 16, 84) 0
concatenate_16[0][0]
dropout_18[0][0]

batch_normalization_37 (BatchNormaliz (None, 16, 16, 84) 336
concatenate_17[0][0]

```

-----
activation_37 (Activation)      (None, 16, 16, 84)    0
batch_normalization_37[0][0]

-----

conv2d_38 (Conv2D)             (None, 16, 16, 12)   1008
activation_37[0][0]

-----

batch_normalization_38 (BatchNo (None, 16, 16, 12)    48          conv2d_38[0][0]
-----

activation_38 (Activation)      (None, 16, 16, 12)    0
batch_normalization_38[0][0]

-----

conv2d_39 (Conv2D)             (None, 16, 16, 12)   1296
activation_38[0][0]

-----

dropout_19 (Dropout)           (None, 16, 16, 12)    0          conv2d_39[0][0]
-----

concatenate_18 (Concatenate)    (None, 16, 16, 96)    0
concatenate_17[0][0]
dropout_19[0][0]

-----

batch_normalization_39 (BatchNo (None, 16, 16, 96)    384
concatenate_18[0][0]

-----

activation_39 (Activation)      (None, 16, 16, 96)    0
batch_normalization_39[0][0]

-----

conv2d_40 (Conv2D)             (None, 16, 16, 12)   1152
activation_39[0][0]

-----

batch_normalization_40 (BatchNo (None, 16, 16, 12)    48          conv2d_40[0][0]
-----

activation_40 (Activation)      (None, 16, 16, 12)    0
batch_normalization_40[0][0]
-----

```

conv2d_41 (Conv2D)	(None, 16, 16, 12)	1296	
activation_40[0][0]			

dropout_20 (Dropout)	(None, 16, 16, 12)	0	conv2d_41[0][0]

concatenate_19 (Concatenate)	(None, 16, 16, 108)	0	
concatenate_18[0][0]			
dropout_20[0][0]			

batch_normalization_41 (BatchNo	(None, 16, 16, 108)	432	
concatenate_19[0][0]			

activation_41 (Activation)	(None, 16, 16, 108)	0	
batch_normalization_41[0][0]			

conv2d_42 (Conv2D)	(None, 16, 16, 12)	1296	
activation_41[0][0]			

batch_normalization_42 (BatchNo	(None, 16, 16, 12)	48	conv2d_42[0][0]

activation_42 (Activation)	(None, 16, 16, 12)	0	
batch_normalization_42[0][0]			

conv2d_43 (Conv2D)	(None, 16, 16, 12)	1296	
activation_42[0][0]			

dropout_21 (Dropout)	(None, 16, 16, 12)	0	conv2d_43[0][0]

concatenate_20 (Concatenate)	(None, 16, 16, 120)	0	
concatenate_19[0][0]			
dropout_21[0][0]			

batch_normalization_43 (BatchNo	(None, 16, 16, 120)	480	
concatenate_20[0][0]			

activation_43 (Activation)	(None, 16, 16, 120)	0	

batch_normalization_43[0][0]

conv2d_44 (Conv2D) (None, 16, 16, 12) 1440
activation_43[0][0]

batch_normalization_44 (BatchNo (None, 16, 16, 12) 48 conv2d_44[0][0]

activation_44 (Activation) (None, 16, 16, 12) 0
batch_normalization_44[0][0]

conv2d_45 (Conv2D) (None, 16, 16, 12) 1296
activation_44[0][0]

dropout_22 (Dropout) (None, 16, 16, 12) 0 conv2d_45[0][0]

concatenate_21 (Concatenate) (None, 16, 16, 132) 0
concatenate_20[0][0]
dropout_22[0][0]

batch_normalization_45 (BatchNo (None, 16, 16, 132) 528
concatenate_21[0][0]

activation_45 (Activation) (None, 16, 16, 132) 0
batch_normalization_45[0][0]

conv2d_46 (Conv2D) (None, 16, 16, 12) 1584
activation_45[0][0]

batch_normalization_46 (BatchNo (None, 16, 16, 12) 48 conv2d_46[0][0]

activation_46 (Activation) (None, 16, 16, 12) 0
batch_normalization_46[0][0]

conv2d_47 (Conv2D) (None, 16, 16, 12) 1296
activation_46[0][0]

dropout_23 (Dropout) (None, 16, 16, 12) 0 conv2d_47[0][0]

concatenate_22 (Concatenate) (None, 16, 16, 144) 0
concatenate_21[0][0]
dropout_23[0][0]

batch_normalization_47 (BatchNo (None, 16, 16, 144) 576
concatenate_22[0][0]

activation_47 (Activation) (None, 16, 16, 144) 0
batch_normalization_47[0][0]

conv2d_48 (Conv2D) (None, 16, 16, 12) 1728
activation_47[0][0]

batch_normalization_48 (BatchNo (None, 16, 16, 12) 48 conv2d_48[0][0]

activation_48 (Activation) (None, 16, 16, 12) 0
batch_normalization_48[0][0]

conv2d_49 (Conv2D) (None, 16, 16, 12) 1296
activation_48[0][0]

dropout_24 (Dropout) (None, 16, 16, 12) 0 conv2d_49[0][0]

concatenate_23 (Concatenate) (None, 16, 16, 156) 0
concatenate_22[0][0]
dropout_24[0][0]

batch_normalization_49 (BatchNo (None, 16, 16, 156) 624
concatenate_23[0][0]

activation_49 (Activation) (None, 16, 16, 156) 0
batch_normalization_49[0][0]

conv2d_50 (Conv2D)	(None, 16, 16, 12)	1872	
activation_49[0][0]			

dropout_25 (Dropout)	(None, 16, 16, 12)	0	conv2d_50[0][0]

average_pooling2d_1 (AveragePool)	(None, 8, 8, 12)	0	
dropout_25[0][0]			

batch_normalization_50 (Batch Normalization)	(None, 8, 8, 12)	48	
average_pooling2d_1[0][0]			

activation_50 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_50[0][0]			

conv2d_51 (Conv2D)	(None, 8, 8, 12)	144	
activation_50[0][0]			

batch_normalization_51 (Batch Normalization)	(None, 8, 8, 12)	48	conv2d_51[0][0]

activation_51 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_51[0][0]			

conv2d_52 (Conv2D)	(None, 8, 8, 12)	1296	
activation_51[0][0]			

dropout_26 (Dropout)	(None, 8, 8, 12)	0	conv2d_52[0][0]

concatenate_24 (Concatenate)	(None, 8, 8, 24)	0	
average_pooling2d_1[0][0]			
dropout_26[0][0]			

batch_normalization_52 (Batch Normalization)	(None, 8, 8, 24)	96	
concatenate_24[0][0]			

activation_52 (Activation)	(None, 8, 8, 24)	0	
batch_normalization_52[0][0]			

conv2d_53 (Conv2D)	(None, 8, 8, 12)	288	
activation_52[0][0]			
batch_normalization_53 (BatchNo	(None, 8, 8, 12)	48	conv2d_53[0][0]
activation_53 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_53[0][0]			
conv2d_54 (Conv2D)	(None, 8, 8, 12)	1296	
activation_53[0][0]			
dropout_27 (Dropout)	(None, 8, 8, 12)	0	conv2d_54[0][0]
concatenate_25 (Concatenate)	(None, 8, 8, 36)	0	
concatenate_24[0][0]			
dropout_27[0][0]			
batch_normalization_54 (BatchNo	(None, 8, 8, 36)	144	
concatenate_25[0][0]			
activation_54 (Activation)	(None, 8, 8, 36)	0	
batch_normalization_54[0][0]			
conv2d_55 (Conv2D)	(None, 8, 8, 12)	432	
activation_54[0][0]			
batch_normalization_55 (BatchNo	(None, 8, 8, 12)	48	conv2d_55[0][0]
activation_55 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_55[0][0]			
conv2d_56 (Conv2D)	(None, 8, 8, 12)	1296	
activation_55[0][0]			

dropout_28 (Dropout)	(None, 8, 8, 12)	0	conv2d_56[0][0]

concatenate_26 (Concatenate)	(None, 8, 8, 48)	0	
concatenate_25[0][0]			
dropout_28[0][0]			

batch_normalization_56 (BatchNo	(None, 8, 8, 48)	192	
concatenate_26[0][0]			

activation_56 (Activation)	(None, 8, 8, 48)	0	
batch_normalization_56[0][0]			

conv2d_57 (Conv2D)	(None, 8, 8, 12)	576	
activation_56[0][0]			

batch_normalization_57 (BatchNo	(None, 8, 8, 12)	48	conv2d_57[0][0]

activation_57 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_57[0][0]			

conv2d_58 (Conv2D)	(None, 8, 8, 12)	1296	
activation_57[0][0]			

dropout_29 (Dropout)	(None, 8, 8, 12)	0	conv2d_58[0][0]

concatenate_27 (Concatenate)	(None, 8, 8, 60)	0	
concatenate_26[0][0]			
dropout_29[0][0]			

batch_normalization_58 (BatchNo	(None, 8, 8, 60)	240	
concatenate_27[0][0]			

activation_58 (Activation)	(None, 8, 8, 60)	0	
batch_normalization_58[0][0]			

conv2d_59 (Conv2D)	(None, 8, 8, 12)	720	

```

activation_58[0][0]
-----
-----
batch_normalization_59 (BatchNo (None, 8, 8, 12)      48      conv2d_59[0][0]
-----
-----
activation_59 (Activation)      (None, 8, 8, 12)      0
batch_normalization_59[0][0]
-----
-----
conv2d_60 (Conv2D)      (None, 8, 8, 12)      1296
activation_59[0][0]
-----
-----
dropout_30 (Dropout)      (None, 8, 8, 12)      0      conv2d_60[0][0]
-----
-----
concatenate_28 (Concatenate)      (None, 8, 8, 72)      0
concatenate_27[0][0]
dropout_30[0][0]
-----
-----
batch_normalization_60 (BatchNo (None, 8, 8, 72)      288
concatenate_28[0][0]
-----
-----
activation_60 (Activation)      (None, 8, 8, 72)      0
batch_normalization_60[0][0]
-----
-----
conv2d_61 (Conv2D)      (None, 8, 8, 12)      864
activation_60[0][0]
-----
-----
batch_normalization_61 (BatchNo (None, 8, 8, 12)      48      conv2d_61[0][0]
-----
-----
activation_61 (Activation)      (None, 8, 8, 12)      0
batch_normalization_61[0][0]
-----
-----
conv2d_62 (Conv2D)      (None, 8, 8, 12)      1296
activation_61[0][0]
-----
-----
dropout_31 (Dropout)      (None, 8, 8, 12)      0      conv2d_62[0][0]
-----
-----

```

concatenate_29 (Concatenate)	(None, 8, 8, 84)	0	
concatenate_28[0][0]			
dropout_31[0][0]			

batch_normalization_62 (Batch Normalization)	(None, 8, 8, 84)	336	
concatenate_29[0][0]			

activation_62 (Activation)	(None, 8, 8, 84)	0	
batch_normalization_62[0][0]			

conv2d_63 (Conv2D)	(None, 8, 8, 12)	1008	
activation_62[0][0]			

batch_normalization_63 (Batch Normalization)	(None, 8, 8, 12)	48	conv2d_63[0][0]

activation_63 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_63[0][0]			

conv2d_64 (Conv2D)	(None, 8, 8, 12)	1296	
activation_63[0][0]			

dropout_32 (Dropout)	(None, 8, 8, 12)	0	conv2d_64[0][0]

concatenate_30 (Concatenate)	(None, 8, 8, 96)	0	
concatenate_29[0][0]			
dropout_32[0][0]			

batch_normalization_64 (Batch Normalization)	(None, 8, 8, 96)	384	
concatenate_30[0][0]			

activation_64 (Activation)	(None, 8, 8, 96)	0	
batch_normalization_64[0][0]			

conv2d_65 (Conv2D)	(None, 8, 8, 12)	1152	
activation_64[0][0]			

batch_normalization_65 (BatchNo	(None, 8, 8, 12)	48	conv2d_65[0][0]

activation_65 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_65[0][0]			

conv2d_66 (Conv2D)	(None, 8, 8, 12)	1296	
activation_65[0][0]			

dropout_33 (Dropout)	(None, 8, 8, 12)	0	conv2d_66[0][0]

concatenate_31 (Concatenate)	(None, 8, 8, 108)	0	
concatenate_30[0][0]			
dropout_33[0][0]			

batch_normalization_66 (BatchNo	(None, 8, 8, 108)	432	
concatenate_31[0][0]			

activation_66 (Activation)	(None, 8, 8, 108)	0	
batch_normalization_66[0][0]			

conv2d_67 (Conv2D)	(None, 8, 8, 12)	1296	
activation_66[0][0]			

batch_normalization_67 (BatchNo	(None, 8, 8, 12)	48	conv2d_67[0][0]

activation_67 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_67[0][0]			

conv2d_68 (Conv2D)	(None, 8, 8, 12)	1296	
activation_67[0][0]			

dropout_34 (Dropout)	(None, 8, 8, 12)	0	conv2d_68[0][0]

concatenate_32 (Concatenate)	(None, 8, 8, 120)	0	
concatenate_31[0][0]			
dropout_34[0][0]			

```

-----
batch_normalization_68 (BatchNo (None, 8, 8, 120)    480
concatenate_32[0][0]
-----

activation_68 (Activation)      (None, 8, 8, 120)    0
batch_normalization_68[0][0]
-----

conv2d_69 (Conv2D)              (None, 8, 8, 12)     1440
activation_68[0][0]
-----

batch_normalization_69 (BatchNo (None, 8, 8, 12)     48          conv2d_69[0][0]
-----

activation_69 (Activation)      (None, 8, 8, 12)     0
batch_normalization_69[0][0]
-----

conv2d_70 (Conv2D)              (None, 8, 8, 12)     1296
activation_69[0][0]
-----

dropout_35 (Dropout)            (None, 8, 8, 12)     0          conv2d_70[0][0]
-----

concatenate_33 (Concatenate)    (None, 8, 8, 132)    0
concatenate_32[0][0]
dropout_35[0][0]
-----

batch_normalization_70 (BatchNo (None, 8, 8, 132)    528
concatenate_33[0][0]
-----

activation_70 (Activation)      (None, 8, 8, 132)    0
batch_normalization_70[0][0]
-----

conv2d_71 (Conv2D)              (None, 8, 8, 12)     1584
activation_70[0][0]
-----

batch_normalization_71 (BatchNo (None, 8, 8, 12)     48          conv2d_71[0][0]
-----

```

activation_71 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_71[0][0]			

conv2d_72 (Conv2D)	(None, 8, 8, 12)	1296	
activation_71[0][0]			

dropout_36 (Dropout)	(None, 8, 8, 12)	0	conv2d_72[0][0]

concatenate_34 (Concatenate)	(None, 8, 8, 144)	0	
concatenate_33[0][0]			
dropout_36[0][0]			

batch_normalization_72 (BatchNo	(None, 8, 8, 144)	576	
concatenate_34[0][0]			

activation_72 (Activation)	(None, 8, 8, 144)	0	
batch_normalization_72[0][0]			

conv2d_73 (Conv2D)	(None, 8, 8, 12)	1728	
activation_72[0][0]			

batch_normalization_73 (BatchNo	(None, 8, 8, 12)	48	conv2d_73[0][0]

activation_73 (Activation)	(None, 8, 8, 12)	0	
batch_normalization_73[0][0]			

conv2d_74 (Conv2D)	(None, 8, 8, 12)	1296	
activation_73[0][0]			

dropout_37 (Dropout)	(None, 8, 8, 12)	0	conv2d_74[0][0]

concatenate_35 (Concatenate)	(None, 8, 8, 156)	0	
concatenate_34[0][0]			
dropout_37[0][0]			

batch_normalization_74 (BatchNo	(None, 8, 8, 156)	624	

concatenate_35[0][0]

activation_74 (Activation) (None, 8, 8, 156) 0
batch_normalization_74[0][0]

conv2d_75 (Conv2D) (None, 8, 8, 12) 1872
activation_74[0][0]

dropout_38 (Dropout) (None, 8, 8, 12) 0 conv2d_75[0][0]

average_pooling2d_2 (AveragePool) (None, 4, 4, 12) 0
dropout_38[0][0]

batch_normalization_75 (BatchNormalizatio (None, 4, 4, 12) 48
average_pooling2d_2[0][0]

activation_75 (Activation) (None, 4, 4, 12) 0
batch_normalization_75[0][0]

conv2d_76 (Conv2D) (None, 4, 4, 12) 144
activation_75[0][0]

batch_normalization_76 (BatchNormalizatio (None, 4, 4, 12) 48 conv2d_76[0][0]

activation_76 (Activation) (None, 4, 4, 12) 0
batch_normalization_76[0][0]

conv2d_77 (Conv2D) (None, 4, 4, 12) 1296
activation_76[0][0]

dropout_39 (Dropout) (None, 4, 4, 12) 0 conv2d_77[0][0]

concatenate_36 (Concatenate) (None, 4, 4, 24) 0
average_pooling2d_2[0][0]
dropout_39[0][0]

```

-----
batch_normalization_77 (BatchNo (None, 4, 4, 24)      96
concatenate_36[0][0]
-----
-----
activation_77 (Activation)      (None, 4, 4, 24)      0
batch_normalization_77[0][0]
-----
-----
conv2d_78 (Conv2D)              (None, 4, 4, 12)      288
activation_77[0][0]
-----
-----
batch_normalization_78 (BatchNo (None, 4, 4, 12)      48      conv2d_78[0][0]
-----
-----
activation_78 (Activation)      (None, 4, 4, 12)      0
batch_normalization_78[0][0]
-----
-----
conv2d_79 (Conv2D)              (None, 4, 4, 12)      1296
activation_78[0][0]
-----
-----
dropout_40 (Dropout)            (None, 4, 4, 12)      0      conv2d_79[0][0]
-----
-----
concatenate_37 (Concatenate)    (None, 4, 4, 36)      0
concatenate_36[0][0]
dropout_40[0][0]
-----
-----
batch_normalization_79 (BatchNo (None, 4, 4, 36)      144
concatenate_37[0][0]
-----
-----
activation_79 (Activation)      (None, 4, 4, 36)      0
batch_normalization_79[0][0]
-----
-----
conv2d_80 (Conv2D)              (None, 4, 4, 12)      432
activation_79[0][0]
-----
-----
batch_normalization_80 (BatchNo (None, 4, 4, 12)      48      conv2d_80[0][0]
-----
-----
activation_80 (Activation)      (None, 4, 4, 12)      0

```


batch_normalization_80[0][0]

conv2d_81 (Conv2D) (None, 4, 4, 12) 1296
activation_80[0][0]

dropout_41 (Dropout) (None, 4, 4, 12) 0 conv2d_81[0][0]

concatenate_38 (Concatenate) (None, 4, 4, 48) 0
concatenate_37[0][0]
dropout_41[0][0]

batch_normalization_81 (BatchNormalizatio (None, 4, 4, 48) 192
concatenate_38[0][0]

activation_81 (Activation) (None, 4, 4, 48) 0
batch_normalization_81[0][0]

conv2d_82 (Conv2D) (None, 4, 4, 12) 576
activation_81[0][0]

batch_normalization_82 (BatchNormalizatio (None, 4, 4, 12) 48 conv2d_82[0][0]

activation_82 (Activation) (None, 4, 4, 12) 0
batch_normalization_82[0][0]

conv2d_83 (Conv2D) (None, 4, 4, 12) 1296
activation_82[0][0]

dropout_42 (Dropout) (None, 4, 4, 12) 0 conv2d_83[0][0]

concatenate_39 (Concatenate) (None, 4, 4, 60) 0
concatenate_38[0][0]
dropout_42[0][0]

batch_normalization_83 (BatchNormalizatio (None, 4, 4, 60) 240
concatenate_39[0][0]

activation_83 (Activation)	(None, 4, 4, 60)	0	
batch_normalization_83[0][0]			

conv2d_84 (Conv2D)	(None, 4, 4, 12)	720	
activation_83[0][0]			

batch_normalization_84 (BatchNo	(None, 4, 4, 12)	48	conv2d_84[0][0]

activation_84 (Activation)	(None, 4, 4, 12)	0	
batch_normalization_84[0][0]			

conv2d_85 (Conv2D)	(None, 4, 4, 12)	1296	
activation_84[0][0]			

dropout_43 (Dropout)	(None, 4, 4, 12)	0	conv2d_85[0][0]

concatenate_40 (Concatenate)	(None, 4, 4, 72)	0	
concatenate_39[0][0]			
dropout_43[0][0]			

batch_normalization_85 (BatchNo	(None, 4, 4, 72)	288	
concatenate_40[0][0]			

activation_85 (Activation)	(None, 4, 4, 72)	0	
batch_normalization_85[0][0]			

conv2d_86 (Conv2D)	(None, 4, 4, 12)	864	
activation_85[0][0]			

batch_normalization_86 (BatchNo	(None, 4, 4, 12)	48	conv2d_86[0][0]

activation_86 (Activation)	(None, 4, 4, 12)	0	
batch_normalization_86[0][0]			

conv2d_87 (Conv2D)	(None, 4, 4, 12)	1296	
activation_86[0][0]			

dropout_44 (Dropout)	(None, 4, 4, 12)	0	conv2d_87[0][0]

concatenate_41 (Concatenate)	(None, 4, 4, 84)	0	
concatenate_40[0][0]			
dropout_44[0][0]			

batch_normalization_87 (BatchNo	(None, 4, 4, 84)	336	
concatenate_41[0][0]			

activation_87 (Activation)	(None, 4, 4, 84)	0	
batch_normalization_87[0][0]			

conv2d_88 (Conv2D)	(None, 4, 4, 12)	1008	
activation_87[0][0]			

batch_normalization_88 (BatchNo	(None, 4, 4, 12)	48	conv2d_88[0][0]

activation_88 (Activation)	(None, 4, 4, 12)	0	
batch_normalization_88[0][0]			

conv2d_89 (Conv2D)	(None, 4, 4, 12)	1296	
activation_88[0][0]			

dropout_45 (Dropout)	(None, 4, 4, 12)	0	conv2d_89[0][0]

concatenate_42 (Concatenate)	(None, 4, 4, 96)	0	
concatenate_41[0][0]			
dropout_45[0][0]			

batch_normalization_89 (BatchNo	(None, 4, 4, 96)	384	
concatenate_42[0][0]			

activation_89 (Activation)	(None, 4, 4, 96)	0	

batch_normalization_89[0][0]

conv2d_90 (Conv2D) (None, 4, 4, 12) 1152
activation_89[0][0]

batch_normalization_90 (BatchNo (None, 4, 4, 12) 48 conv2d_90[0][0]

activation_90 (Activation) (None, 4, 4, 12) 0
batch_normalization_90[0][0]

conv2d_91 (Conv2D) (None, 4, 4, 12) 1296
activation_90[0][0]

dropout_46 (Dropout) (None, 4, 4, 12) 0 conv2d_91[0][0]

concatenate_43 (Concatenate) (None, 4, 4, 108) 0
concatenate_42[0][0]
dropout_46[0][0]

batch_normalization_91 (BatchNo (None, 4, 4, 108) 432
concatenate_43[0][0]

activation_91 (Activation) (None, 4, 4, 108) 0
batch_normalization_91[0][0]

conv2d_92 (Conv2D) (None, 4, 4, 12) 1296
activation_91[0][0]

batch_normalization_92 (BatchNo (None, 4, 4, 12) 48 conv2d_92[0][0]

activation_92 (Activation) (None, 4, 4, 12) 0
batch_normalization_92[0][0]

conv2d_93 (Conv2D) (None, 4, 4, 12) 1296
activation_92[0][0]

```

-----
dropout_47 (Dropout)          (None, 4, 4, 12)    0          conv2d_93[0] [0]
-----

-----
concatenate_44 (Concatenate)  (None, 4, 4, 120)   0
concatenate_43[0] [0]
dropout_47[0] [0]
-----

-----
batch_normalization_93 (BatchNo (None, 4, 4, 120)   480
concatenate_44[0] [0]
-----

-----
activation_93 (Activation)    (None, 4, 4, 120)   0
batch_normalization_93[0] [0]
-----

-----
conv2d_94 (Conv2D)           (None, 4, 4, 12)    1440
activation_93[0] [0]
-----

-----
batch_normalization_94 (BatchNo (None, 4, 4, 12)    48          conv2d_94[0] [0]
-----

-----
activation_94 (Activation)    (None, 4, 4, 12)    0
batch_normalization_94[0] [0]
-----

-----
conv2d_95 (Conv2D)           (None, 4, 4, 12)    1296
activation_94[0] [0]
-----

-----
dropout_48 (Dropout)          (None, 4, 4, 12)    0          conv2d_95[0] [0]
-----

-----
concatenate_45 (Concatenate)  (None, 4, 4, 132)   0
concatenate_44[0] [0]
dropout_48[0] [0]
-----

-----
batch_normalization_95 (BatchNo (None, 4, 4, 132)   528
concatenate_45[0] [0]
-----

-----
activation_95 (Activation)    (None, 4, 4, 132)   0
batch_normalization_95[0] [0]
-----

-----

```

conv2d_96 (Conv2D)	(None, 4, 4, 12)	1584	
activation_95[0][0]			

batch_normalization_96 (BatchNo	(None, 4, 4, 12)	48	conv2d_96[0][0]

activation_96 (Activation)	(None, 4, 4, 12)	0	
batch_normalization_96[0][0]			

conv2d_97 (Conv2D)	(None, 4, 4, 12)	1296	
activation_96[0][0]			

dropout_49 (Dropout)	(None, 4, 4, 12)	0	conv2d_97[0][0]

concatenate_46 (Concatenate)	(None, 4, 4, 144)	0	
concatenate_45[0][0]			
dropout_49[0][0]			

batch_normalization_97 (BatchNo	(None, 4, 4, 144)	576	
concatenate_46[0][0]			

activation_97 (Activation)	(None, 4, 4, 144)	0	
batch_normalization_97[0][0]			

conv2d_98 (Conv2D)	(None, 4, 4, 12)	1728	
activation_97[0][0]			

batch_normalization_98 (BatchNo	(None, 4, 4, 12)	48	conv2d_98[0][0]

activation_98 (Activation)	(None, 4, 4, 12)	0	
batch_normalization_98[0][0]			

conv2d_99 (Conv2D)	(None, 4, 4, 12)	1296	
activation_98[0][0]			

dropout_50 (Dropout)	(None, 4, 4, 12)	0	conv2d_99[0][0]

```

-----
concatenate_47 (Concatenate)      (None, 4, 4, 156)      0
concatenate_46[0][0]
dropout_50[0][0]
-----

batch_normalization_99 (BatchNo (None, 4, 4, 156)      624
concatenate_47[0][0]
-----

activation_99 (Activation)        (None, 4, 4, 156)      0
batch_normalization_99[0][0]
-----

average_pooling2d_3 (AveragePoo (None, 2, 2, 156)      0
activation_99[0][0]
-----

flatten (Flatten)                (None, 624)            0
average_pooling2d_3[0][0]
-----

dense (Dense)                    (None, 10)             6250      flatten[0][0]
=====
=====
Total params: 141,922
Trainable params: 131,722
Non-trainable params: 10,200
-----
-----

```

```
[ ]: print(len(model3.layers))
```

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```
[ ]: model3.
    ↳ compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])

model3.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

390/390 [=====] - 76s 174ms/step - loss: 1.8787 - accuracy: 0.3063 - val_loss: 6.2094 - val_accuracy: 0.1329

Epoch 00001: val_accuracy improved from -inf to 0.13290, saving model to weights.best.hdf5
Epoch 2/100
390/390 [=====] - 65s 166ms/step - loss: 1.5245 - accuracy: 0.4403 - val_loss: 1.9439 - val_accuracy: 0.4002

Epoch 00002: val_accuracy improved from 0.13290 to 0.40020, saving model to weights.best.hdf5
Epoch 3/100
390/390 [=====] - 65s 167ms/step - loss: 1.3421 - accuracy: 0.5108 - val_loss: 3.8294 - val_accuracy: 0.3237

Epoch 00003: val_accuracy did not improve from 0.40020
Epoch 4/100
390/390 [=====] - 65s 167ms/step - loss: 1.2104 - accuracy: 0.5635 - val_loss: 2.2493 - val_accuracy: 0.4519

Epoch 00004: val_accuracy improved from 0.40020 to 0.45190, saving model to weights.best.hdf5
Epoch 5/100
390/390 [=====] - 65s 166ms/step - loss: 1.1075 - accuracy: 0.6057 - val_loss: 1.7633 - val_accuracy: 0.5227

Epoch 00005: val_accuracy improved from 0.45190 to 0.52270, saving model to weights.best.hdf5
Epoch 6/100
390/390 [=====] - 65s 166ms/step - loss: 1.0290 - accuracy: 0.6315 - val_loss: 1.4483 - val_accuracy: 0.5767

Epoch 00006: val_accuracy improved from 0.52270 to 0.57670, saving model to weights.best.hdf5
Epoch 7/100
390/390 [=====] - 65s 165ms/step - loss: 0.9777 - accuracy: 0.6519 - val_loss: 2.5660 - val_accuracy: 0.4636

Epoch 00007: val_accuracy did not improve from 0.57670
Epoch 8/100
390/390 [=====] - 64s 164ms/step - loss: 0.9328 - accuracy: 0.6688 - val_loss: 1.8276 - val_accuracy: 0.5384

Epoch 00008: val_accuracy did not improve from 0.57670
Epoch 9/100
390/390 [=====] - 65s 165ms/step - loss: 0.8982 - accuracy: 0.6822 - val_loss: 1.6682 - val_accuracy: 0.5601

Epoch 00009: val_accuracy did not improve from 0.57670
Epoch 10/100

390/390 [=====] - 64s 165ms/step - loss: 0.8749 - accuracy: 0.6896 - val_loss: 1.3930 - val_accuracy: 0.6313

Epoch 00010: val_accuracy improved from 0.57670 to 0.63130, saving model to weights.best.hdf5

Epoch 11/100

390/390 [=====] - 64s 165ms/step - loss: 0.8522 - accuracy: 0.6970 - val_loss: 1.1654 - val_accuracy: 0.6492

Epoch 00011: val_accuracy improved from 0.63130 to 0.64920, saving model to weights.best.hdf5

Epoch 12/100

390/390 [=====] - 64s 165ms/step - loss: 0.8241 - accuracy: 0.7086 - val_loss: 1.5537 - val_accuracy: 0.6059

Epoch 00012: val_accuracy did not improve from 0.64920

Epoch 13/100

390/390 [=====] - 65s 166ms/step - loss: 0.8022 - accuracy: 0.7167 - val_loss: 1.3158 - val_accuracy: 0.5879

Epoch 00013: val_accuracy did not improve from 0.64920

Epoch 14/100

390/390 [=====] - 65s 165ms/step - loss: 0.7882 - accuracy: 0.7229 - val_loss: 1.4757 - val_accuracy: 0.6051

Epoch 00014: val_accuracy did not improve from 0.64920

Epoch 15/100

390/390 [=====] - 65s 166ms/step - loss: 0.7644 - accuracy: 0.7320 - val_loss: 0.9961 - val_accuracy: 0.6942

Epoch 00015: val_accuracy improved from 0.64920 to 0.69420, saving model to weights.best.hdf5

Epoch 16/100

390/390 [=====] - 65s 166ms/step - loss: 0.7446 - accuracy: 0.7390 - val_loss: 1.0217 - val_accuracy: 0.7059

Epoch 00016: val_accuracy improved from 0.69420 to 0.70590, saving model to weights.best.hdf5

Epoch 17/100

390/390 [=====] - 65s 166ms/step - loss: 0.7307 - accuracy: 0.7432 - val_loss: 0.9654 - val_accuracy: 0.6975

Epoch 00017: val_accuracy did not improve from 0.70590

Epoch 18/100

390/390 [=====] - 65s 166ms/step - loss: 0.7159 - accuracy: 0.7480 - val_loss: 1.2995 - val_accuracy: 0.6513

Epoch 00018: val_accuracy did not improve from 0.70590

Epoch 19/100
390/390 [=====] - 64s 165ms/step - loss: 0.7013 - accuracy: 0.7530 - val_loss: 1.0214 - val_accuracy: 0.6978

Epoch 00019: val_accuracy did not improve from 0.70590
Epoch 20/100
390/390 [=====] - 64s 164ms/step - loss: 0.6920 - accuracy: 0.7567 - val_loss: 1.3233 - val_accuracy: 0.6280

Epoch 00020: val_accuracy did not improve from 0.70590
Epoch 21/100
390/390 [=====] - 64s 163ms/step - loss: 0.6844 - accuracy: 0.7609 - val_loss: 0.9723 - val_accuracy: 0.7109

Epoch 00021: val_accuracy improved from 0.70590 to 0.71090, saving model to weights.best.hdf5
Epoch 22/100
390/390 [=====] - 64s 163ms/step - loss: 0.6724 - accuracy: 0.7660 - val_loss: 0.7707 - val_accuracy: 0.7574

Epoch 00022: val_accuracy improved from 0.71090 to 0.75740, saving model to weights.best.hdf5
Epoch 23/100
390/390 [=====] - 64s 163ms/step - loss: 0.6639 - accuracy: 0.7681 - val_loss: 0.9687 - val_accuracy: 0.7260

Epoch 00023: val_accuracy did not improve from 0.75740
Epoch 24/100
390/390 [=====] - 64s 164ms/step - loss: 0.6538 - accuracy: 0.7721 - val_loss: 0.9126 - val_accuracy: 0.7317

Epoch 00024: val_accuracy did not improve from 0.75740
Epoch 25/100
390/390 [=====] - 64s 163ms/step - loss: 0.6481 - accuracy: 0.7726 - val_loss: 1.8253 - val_accuracy: 0.6180

Epoch 00025: val_accuracy did not improve from 0.75740
Epoch 26/100
390/390 [=====] - 64s 163ms/step - loss: 0.5803 - accuracy: 0.7983 - val_loss: 0.6085 - val_accuracy: 0.8065

Epoch 00026: val_accuracy improved from 0.75740 to 0.80650, saving model to weights.best.hdf5
Epoch 27/100
390/390 [=====] - 64s 164ms/step - loss: 0.5542 - accuracy: 0.8054 - val_loss: 0.6685 - val_accuracy: 0.7945

Epoch 00027: val_accuracy did not improve from 0.80650

Epoch 28/100
390/390 [=====] - 64s 163ms/step - loss: 0.5452 - accuracy: 0.8079 - val_loss: 0.7349 - val_accuracy: 0.7816

Epoch 00028: val_accuracy did not improve from 0.80650

Epoch 29/100
390/390 [=====] - 63s 163ms/step - loss: 0.5400 - accuracy: 0.8101 - val_loss: 0.7039 - val_accuracy: 0.7900

Epoch 00029: val_accuracy did not improve from 0.80650

Epoch 30/100
390/390 [=====] - 63s 163ms/step - loss: 0.5333 - accuracy: 0.8130 - val_loss: 0.6732 - val_accuracy: 0.7933

Epoch 00030: val_accuracy did not improve from 0.80650

Epoch 31/100
390/390 [=====] - 63s 162ms/step - loss: 0.5295 - accuracy: 0.8153 - val_loss: 0.5967 - val_accuracy: 0.8126

Epoch 00031: val_accuracy improved from 0.80650 to 0.81260, saving model to weights.best.hdf5

Epoch 32/100
390/390 [=====] - 64s 163ms/step - loss: 0.5266 - accuracy: 0.8149 - val_loss: 0.6517 - val_accuracy: 0.8036

Epoch 00032: val_accuracy did not improve from 0.81260

Epoch 33/100
390/390 [=====] - 63s 162ms/step - loss: 0.5259 - accuracy: 0.8179 - val_loss: 0.7121 - val_accuracy: 0.7878

Epoch 00033: val_accuracy did not improve from 0.81260

Epoch 34/100
390/390 [=====] - 63s 162ms/step - loss: 0.5200 - accuracy: 0.8164 - val_loss: 0.6949 - val_accuracy: 0.7905

Epoch 00034: val_accuracy did not improve from 0.81260

Epoch 35/100
390/390 [=====] - 64s 164ms/step - loss: 0.5119 - accuracy: 0.8211 - val_loss: 0.6087 - val_accuracy: 0.8097

Epoch 00035: val_accuracy did not improve from 0.81260

Epoch 36/100
390/390 [=====] - 64s 164ms/step - loss: 0.5189 - accuracy: 0.8174 - val_loss: 0.5921 - val_accuracy: 0.8145

Epoch 00036: val_accuracy improved from 0.81260 to 0.81450, saving model to weights.best.hdf5

Epoch 37/100

390/390 [=====] - 64s 164ms/step - loss: 0.5131 -
accuracy: 0.8210 - val_loss: 0.6484 - val_accuracy: 0.8044

Epoch 00037: val_accuracy did not improve from 0.81450

Epoch 38/100

390/390 [=====] - 64s 165ms/step - loss: 0.5070 -
accuracy: 0.8235 - val_loss: 0.6497 - val_accuracy: 0.8017

Epoch 00038: val_accuracy did not improve from 0.81450

Epoch 39/100

390/390 [=====] - 64s 165ms/step - loss: 0.5071 -
accuracy: 0.8247 - val_loss: 0.6591 - val_accuracy: 0.8039

Epoch 00039: val_accuracy did not improve from 0.81450

Epoch 40/100

390/390 [=====] - 64s 165ms/step - loss: 0.5062 -
accuracy: 0.8222 - val_loss: 0.6059 - val_accuracy: 0.8115

Epoch 00040: val_accuracy did not improve from 0.81450

Epoch 41/100

390/390 [=====] - 64s 163ms/step - loss: 0.5033 -
accuracy: 0.8236 - val_loss: 0.6382 - val_accuracy: 0.8071

Epoch 00041: val_accuracy did not improve from 0.81450

Epoch 42/100

390/390 [=====] - 63s 162ms/step - loss: 0.5019 -
accuracy: 0.8264 - val_loss: 0.6656 - val_accuracy: 0.8002

Epoch 00042: val_accuracy did not improve from 0.81450

Epoch 43/100

390/390 [=====] - 64s 163ms/step - loss: 0.5012 -
accuracy: 0.8257 - val_loss: 0.6749 - val_accuracy: 0.7962

Epoch 00043: val_accuracy did not improve from 0.81450

Epoch 44/100

390/390 [=====] - 64s 163ms/step - loss: 0.5013 -
accuracy: 0.8232 - val_loss: 0.6785 - val_accuracy: 0.7997

Epoch 00044: val_accuracy did not improve from 0.81450

Epoch 45/100

390/390 [=====] - 64s 163ms/step - loss: 0.4997 -
accuracy: 0.8261 - val_loss: 0.5985 - val_accuracy: 0.8147

Epoch 00045: val_accuracy improved from 0.81450 to 0.81470, saving model to
weights.best.hdf5

Epoch 46/100

390/390 [=====] - 63s 162ms/step - loss: 0.4949 -
accuracy: 0.8271 - val_loss: 0.6376 - val_accuracy: 0.8057

Epoch 00046: val_accuracy did not improve from 0.81470
Epoch 47/100
390/390 [=====] - 64s 163ms/step - loss: 0.4920 -
accuracy: 0.8272 - val_loss: 0.6769 - val_accuracy: 0.8001

Epoch 00047: val_accuracy did not improve from 0.81470
Epoch 48/100
390/390 [=====] - 64s 163ms/step - loss: 0.4907 -
accuracy: 0.8269 - val_loss: 0.5599 - val_accuracy: 0.8259

Epoch 00048: val_accuracy improved from 0.81470 to 0.82590, saving model to
weights.best.hdf5
Epoch 49/100
390/390 [=====] - 64s 163ms/step - loss: 0.4904 -
accuracy: 0.8281 - val_loss: 0.5944 - val_accuracy: 0.8191

Epoch 00049: val_accuracy did not improve from 0.82590
Epoch 50/100
390/390 [=====] - 64s 163ms/step - loss: 0.4923 -
accuracy: 0.8295 - val_loss: 0.6196 - val_accuracy: 0.8115

Epoch 00050: val_accuracy did not improve from 0.82590
Epoch 51/100
390/390 [=====] - 64s 163ms/step - loss: 0.4771 -
accuracy: 0.8336 - val_loss: 0.5877 - val_accuracy: 0.8199

Epoch 00051: val_accuracy did not improve from 0.82590
Epoch 52/100
390/390 [=====] - 64s 164ms/step - loss: 0.4797 -
accuracy: 0.8317 - val_loss: 0.5976 - val_accuracy: 0.8197

Epoch 00052: val_accuracy did not improve from 0.82590
Epoch 53/100
390/390 [=====] - 63s 163ms/step - loss: 0.4778 -
accuracy: 0.8332 - val_loss: 0.6047 - val_accuracy: 0.8172

Epoch 00053: val_accuracy did not improve from 0.82590
Epoch 54/100
390/390 [=====] - 64s 163ms/step - loss: 0.4803 -
accuracy: 0.8328 - val_loss: 0.5944 - val_accuracy: 0.8193

Epoch 00054: val_accuracy did not improve from 0.82590
Epoch 55/100
390/390 [=====] - 64s 164ms/step - loss: 0.4740 -
accuracy: 0.8350 - val_loss: 0.6027 - val_accuracy: 0.8182

Epoch 00055: val_accuracy did not improve from 0.82590

Epoch 56/100
 390/390 [=====] - 63s 162ms/step - loss: 0.4762 - accuracy: 0.8324 - val_loss: 0.5926 - val_accuracy: 0.8200

Epoch 00056: val_accuracy did not improve from 0.82590

Epoch 57/100
 390/390 [=====] - 64s 163ms/step - loss: 0.4754 - accuracy: 0.8337 - val_loss: 0.5978 - val_accuracy: 0.8182

Epoch 00057: val_accuracy did not improve from 0.82590

Epoch 58/100
 390/390 [=====] - 64s 164ms/step - loss: 0.4789 - accuracy: 0.8337 - val_loss: 0.5979 - val_accuracy: 0.8188

Epoch 00058: val_accuracy did not improve from 0.82590

Epoch 59/100
 390/390 [=====] - 64s 164ms/step - loss: 0.4787 - accuracy: 0.8340 - val_loss: 0.6087 - val_accuracy: 0.8176

Epoch 00059: val_accuracy did not improve from 0.82590

Epoch 60/100
 390/390 [=====] - 64s 165ms/step - loss: 0.4762 - accuracy: 0.8338 - val_loss: 0.6094 - val_accuracy: 0.8173

Epoch 00060: val_accuracy did not improve from 0.82590

Epoch 61/100
 390/390 [=====] - 64s 165ms/step - loss: 0.4764 - accuracy: 0.8326 - val_loss: 0.6118 - val_accuracy: 0.8167

Epoch 00061: val_accuracy did not improve from 0.82590

Epoch 62/100
 390/390 [=====] - 64s 164ms/step - loss: 0.4737 - accuracy: 0.8348 - val_loss: 0.5979 - val_accuracy: 0.8186

Epoch 00062: val_accuracy did not improve from 0.82590

Epoch 63/100
 390/390 [=====] - 64s 164ms/step - loss: 0.4763 - accuracy: 0.8335 - val_loss: 0.5987 - val_accuracy: 0.8196

Epoch 00063: val_accuracy did not improve from 0.82590

Epoch 64/100
 390/390 [=====] - 64s 163ms/step - loss: 0.4751 - accuracy: 0.8336 - val_loss: 0.5927 - val_accuracy: 0.8203

Epoch 00064: val_accuracy did not improve from 0.82590

Epoch 65/100
 390/390 [=====] - 63s 162ms/step - loss: 0.4737 - accuracy: 0.8345 - val_loss: 0.6007 - val_accuracy: 0.8183

Epoch 00065: val_accuracy did not improve from 0.82590
Epoch 66/100
390/390 [=====] - 63s 162ms/step - loss: 0.4755 -
accuracy: 0.8351 - val_loss: 0.6137 - val_accuracy: 0.8152

Epoch 00066: val_accuracy did not improve from 0.82590
Epoch 67/100
390/390 [=====] - 63s 162ms/step - loss: 0.4724 -
accuracy: 0.8345 - val_loss: 0.6001 - val_accuracy: 0.8194

Epoch 00067: val_accuracy did not improve from 0.82590
Epoch 68/100
390/390 [=====] - 64s 163ms/step - loss: 0.4731 -
accuracy: 0.8344 - val_loss: 0.5974 - val_accuracy: 0.8187

Epoch 00068: val_accuracy did not improve from 0.82590
Epoch 69/100
390/390 [=====] - 63s 162ms/step - loss: 0.4721 -
accuracy: 0.8328 - val_loss: 0.6013 - val_accuracy: 0.8185

Epoch 00069: val_accuracy did not improve from 0.82590
Epoch 70/100
390/390 [=====] - 64s 163ms/step - loss: 0.4708 -
accuracy: 0.8332 - val_loss: 0.5925 - val_accuracy: 0.8204

Epoch 00070: val_accuracy did not improve from 0.82590
Epoch 71/100
390/390 [=====] - 64s 164ms/step - loss: 0.4703 -
accuracy: 0.8352 - val_loss: 0.5956 - val_accuracy: 0.8200

Epoch 00071: val_accuracy did not improve from 0.82590
Epoch 72/100
390/390 [=====] - 63s 163ms/step - loss: 0.4741 -
accuracy: 0.8340 - val_loss: 0.6019 - val_accuracy: 0.8192

Epoch 00072: val_accuracy did not improve from 0.82590
Epoch 73/100
390/390 [=====] - 64s 163ms/step - loss: 0.4769 -
accuracy: 0.8342 - val_loss: 0.5974 - val_accuracy: 0.8196

Epoch 00073: val_accuracy did not improve from 0.82590
Epoch 74/100
390/390 [=====] - 64s 163ms/step - loss: 0.4732 -
accuracy: 0.8349 - val_loss: 0.6004 - val_accuracy: 0.8193

Epoch 00074: val_accuracy did not improve from 0.82590
Epoch 75/100

390/390 [=====] - 64s 164ms/step - loss: 0.4754 -
accuracy: 0.8328 - val_loss: 0.6164 - val_accuracy: 0.8165

Epoch 00075: val_accuracy did not improve from 0.82590

Epoch 76/100

390/390 [=====] - 64s 163ms/step - loss: 0.4711 -
accuracy: 0.8346 - val_loss: 0.5940 - val_accuracy: 0.8194

Epoch 00076: val_accuracy did not improve from 0.82590

Epoch 77/100

390/390 [=====] - 64s 163ms/step - loss: 0.4659 -
accuracy: 0.8373 - val_loss: 0.6167 - val_accuracy: 0.8164

Epoch 00077: val_accuracy did not improve from 0.82590

Epoch 78/100

390/390 [=====] - 64s 163ms/step - loss: 0.4644 -
accuracy: 0.8376 - val_loss: 0.5916 - val_accuracy: 0.8205

Epoch 00078: val_accuracy did not improve from 0.82590

Epoch 79/100

390/390 [=====] - 64s 164ms/step - loss: 0.4671 -
accuracy: 0.8366 - val_loss: 0.5899 - val_accuracy: 0.8204

Epoch 00079: val_accuracy did not improve from 0.82590

Epoch 80/100

390/390 [=====] - 64s 165ms/step - loss: 0.4735 -
accuracy: 0.8352 - val_loss: 0.5945 - val_accuracy: 0.8196

Epoch 00080: val_accuracy did not improve from 0.82590

Epoch 81/100

390/390 [=====] - 64s 165ms/step - loss: 0.4684 -
accuracy: 0.8353 - val_loss: 0.5949 - val_accuracy: 0.8207

Epoch 00081: val_accuracy did not improve from 0.82590

Epoch 82/100

390/390 [=====] - 64s 163ms/step - loss: 0.4709 -
accuracy: 0.8374 - val_loss: 0.5980 - val_accuracy: 0.8203

Epoch 00082: val_accuracy did not improve from 0.82590

Epoch 83/100

390/390 [=====] - 64s 163ms/step - loss: 0.4705 -
accuracy: 0.8367 - val_loss: 0.5977 - val_accuracy: 0.8198

Epoch 00083: val_accuracy did not improve from 0.82590

Epoch 84/100

390/390 [=====] - 64s 164ms/step - loss: 0.4679 -
accuracy: 0.8338 - val_loss: 0.5871 - val_accuracy: 0.8220

Epoch 00084: val_accuracy did not improve from 0.82590
 Epoch 85/100
 390/390 [=====] - 63s 162ms/step - loss: 0.4693 - accuracy: 0.8369 - val_loss: 0.5832 - val_accuracy: 0.8221

Epoch 00085: val_accuracy did not improve from 0.82590
 Epoch 86/100
 390/390 [=====] - 64s 163ms/step - loss: 0.4712 - accuracy: 0.8348 - val_loss: 0.5976 - val_accuracy: 0.8199

Epoch 00086: val_accuracy did not improve from 0.82590
 Epoch 87/100
 390/390 [=====] - 64s 164ms/step - loss: 0.4734 - accuracy: 0.8344 - val_loss: 0.5842 - val_accuracy: 0.8223

Epoch 00087: val_accuracy did not improve from 0.82590
 Epoch 88/100
 390/390 [=====] - 64s 163ms/step - loss: 0.4667 - accuracy: 0.8361 - val_loss: 0.6003 - val_accuracy: 0.8187

Epoch 00088: val_accuracy did not improve from 0.82590

[]: <tensorflow.python.keras.callbacks.History at 0x7f58c765c350>

1.3.5 2.5 Growth rate(num_filter)=32, compression = 0.7, Number of blocks = 12

```
[ ]: # Hyperparameters
batch_size = 64
num_classes = 10
nb_epoch = 100
l = 12
num_filter = 32
compression = 0.7
dropout_rate = 0.2
```

```
[ ]: from tensorflow.keras.callbacks import ModelCheckpoint
#https://machinelearningmastery.com/check-point-deep-learning-models-keras/
filepath="model4_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)

```

```

[ ]: model4 = Model(inputs = [input], outputs = [output])
model4.summary()

```

Model: "model"

```

-----
Layer (type)                 Output Shape          Param #   Connected to
=====
input_2 (InputLayer)         [(None, 32, 32, 3)]  0
-----
conv2d (Conv2D)              (None, 32, 32, 32)   864       input_2[0][0]
-----
batch_normalization (BatchNorm (None, 32, 32, 32)   128       conv2d[0][0]
-----
activation (Activation)      (None, 32, 32, 32)   0
batch_normalization[0][0]
-----
conv2d_1 (Conv2D)            (None, 32, 32, 22)   704
activation[0][0]
-----
batch_normalization_1 (BatchNor (None, 32, 32, 22)   88        conv2d_1[0][0]
-----
activation_1 (Activation)    (None, 32, 32, 22)   0
batch_normalization_1[0][0]
-----
conv2d_2 (Conv2D)            (None, 32, 32, 22)   4356
activation_1[0][0]
-----

```

dropout (Dropout)	(None, 32, 32, 22)	0	conv2d_2[0][0]
concatenate (Concatenate)	(None, 32, 32, 54)	0	conv2d[0][0] dropout[0][0]
batch_normalization_2 (BatchNor concatenate[0][0])	(None, 32, 32, 54)	216	
activation_2 (Activation) batch_normalization_2[0][0]	(None, 32, 32, 54)	0	
conv2d_3 (Conv2D) activation_2[0][0]	(None, 32, 32, 22)	1188	
batch_normalization_3 (BatchNor	(None, 32, 32, 22)	88	conv2d_3[0][0]
activation_3 (Activation) batch_normalization_3[0][0]	(None, 32, 32, 22)	0	
conv2d_4 (Conv2D) activation_3[0][0]	(None, 32, 32, 22)	4356	
dropout_1 (Dropout)	(None, 32, 32, 22)	0	conv2d_4[0][0]
concatenate_1 (Concatenate) concatenate[0][0]	(None, 32, 32, 76)	0	dropout_1[0][0]
batch_normalization_4 (BatchNor concatenate_1[0][0])	(None, 32, 32, 76)	304	
activation_4 (Activation) batch_normalization_4[0][0]	(None, 32, 32, 76)	0	
conv2d_5 (Conv2D)	(None, 32, 32, 22)	1672	

```

activation_4[0][0]
-----
batch_normalization_5 (BatchNor (None, 32, 32, 22) 88 conv2d_5[0][0]
-----
activation_5 (Activation) (None, 32, 32, 22) 0
batch_normalization_5[0][0]
-----
conv2d_6 (Conv2D) (None, 32, 32, 22) 4356
activation_5[0][0]
-----
dropout_2 (Dropout) (None, 32, 32, 22) 0 conv2d_6[0][0]
-----
concatenate_2 (Concatenate) (None, 32, 32, 98) 0
concatenate_1[0][0]
dropout_2[0][0]
-----
batch_normalization_6 (BatchNor (None, 32, 32, 98) 392
concatenate_2[0][0]
-----
activation_6 (Activation) (None, 32, 32, 98) 0
batch_normalization_6[0][0]
-----
conv2d_7 (Conv2D) (None, 32, 32, 22) 2156
activation_6[0][0]
-----
batch_normalization_7 (BatchNor (None, 32, 32, 22) 88 conv2d_7[0][0]
-----
activation_7 (Activation) (None, 32, 32, 22) 0
batch_normalization_7[0][0]
-----
conv2d_8 (Conv2D) (None, 32, 32, 22) 4356
activation_7[0][0]
-----
dropout_3 (Dropout) (None, 32, 32, 22) 0 conv2d_8[0][0]
-----

```

```

concatenate_3 (Concatenate)      (None, 32, 32, 120)  0
concatenate_2[0][0]

dropout_3[0][0]

-----

batch_normalization_8 (BatchNor (None, 32, 32, 120)  480
concatenate_3[0][0]

-----

activation_8 (Activation)        (None, 32, 32, 120)  0
batch_normalization_8[0][0]

-----

conv2d_9 (Conv2D)                (None, 32, 32, 22)   2640
activation_8[0][0]

-----

batch_normalization_9 (BatchNor (None, 32, 32, 22)   88      conv2d_9[0][0]

-----

activation_9 (Activation)        (None, 32, 32, 22)   0
batch_normalization_9[0][0]

-----

conv2d_10 (Conv2D)              (None, 32, 32, 22)   4356
activation_9[0][0]

-----

dropout_4 (Dropout)             (None, 32, 32, 22)   0      conv2d_10[0][0]

-----

concatenate_4 (Concatenate)      (None, 32, 32, 142)  0
concatenate_3[0][0]

dropout_4[0][0]

-----

batch_normalization_10 (BatchNo (None, 32, 32, 142)  568
concatenate_4[0][0]

-----

activation_10 (Activation)       (None, 32, 32, 142)  0
batch_normalization_10[0][0]

-----

conv2d_11 (Conv2D)              (None, 32, 32, 22)   3124
activation_10[0][0]

```

```

batch_normalization_11 (BatchNo (None, 32, 32, 22) 88 conv2d_11[0][0]
-----
activation_11 (Activation) (None, 32, 32, 22) 0
batch_normalization_11[0][0]
-----
conv2d_12 (Conv2D) (None, 32, 32, 22) 4356
activation_11[0][0]
-----
dropout_5 (Dropout) (None, 32, 32, 22) 0 conv2d_12[0][0]
-----
concatenate_5 (Concatenate) (None, 32, 32, 164) 0
concatenate_4[0][0]
dropout_5[0][0]
-----
batch_normalization_12 (BatchNo (None, 32, 32, 164) 656
concatenate_5[0][0]
-----
activation_12 (Activation) (None, 32, 32, 164) 0
batch_normalization_12[0][0]
-----
conv2d_13 (Conv2D) (None, 32, 32, 22) 3608
activation_12[0][0]
-----
batch_normalization_13 (BatchNo (None, 32, 32, 22) 88 conv2d_13[0][0]
-----
activation_13 (Activation) (None, 32, 32, 22) 0
batch_normalization_13[0][0]
-----
conv2d_14 (Conv2D) (None, 32, 32, 22) 4356
activation_13[0][0]
-----
dropout_6 (Dropout) (None, 32, 32, 22) 0 conv2d_14[0][0]
-----
concatenate_6 (Concatenate) (None, 32, 32, 186) 0
concatenate_5[0][0]
dropout_6[0][0]

```

```

-----
batch_normalization_14 (BatchNo (None, 32, 32, 186) 744
concatenate_6[0][0]
-----

activation_14 (Activation) (None, 32, 32, 186) 0
batch_normalization_14[0][0]
-----

conv2d_15 (Conv2D) (None, 32, 32, 22) 4092
activation_14[0][0]
-----

batch_normalization_15 (BatchNo (None, 32, 32, 22) 88 conv2d_15[0][0]
-----

activation_15 (Activation) (None, 32, 32, 22) 0
batch_normalization_15[0][0]
-----

conv2d_16 (Conv2D) (None, 32, 32, 22) 4356
activation_15[0][0]
-----

dropout_7 (Dropout) (None, 32, 32, 22) 0 conv2d_16[0][0]
-----

concatenate_7 (Concatenate) (None, 32, 32, 208) 0
concatenate_6[0][0]
dropout_7[0][0]
-----

batch_normalization_16 (BatchNo (None, 32, 32, 208) 832
concatenate_7[0][0]
-----

activation_16 (Activation) (None, 32, 32, 208) 0
batch_normalization_16[0][0]
-----

conv2d_17 (Conv2D) (None, 32, 32, 22) 4576
activation_16[0][0]
-----

batch_normalization_17 (BatchNo (None, 32, 32, 22) 88 conv2d_17[0][0]
-----

```

activation_17 (Activation)	(None, 32, 32, 22)	0	
batch_normalization_17[0][0]			

conv2d_18 (Conv2D)	(None, 32, 32, 22)	4356	
activation_17[0][0]			

dropout_8 (Dropout)	(None, 32, 32, 22)	0	conv2d_18[0][0]

concatenate_8 (Concatenate)	(None, 32, 32, 230)	0	
concatenate_7[0][0]			
			dropout_8[0][0]

batch_normalization_18 (BatchNo	(None, 32, 32, 230)	920	
concatenate_8[0][0]			

activation_18 (Activation)	(None, 32, 32, 230)	0	
batch_normalization_18[0][0]			

conv2d_19 (Conv2D)	(None, 32, 32, 22)	5060	
activation_18[0][0]			

batch_normalization_19 (BatchNo	(None, 32, 32, 22)	88	conv2d_19[0][0]

activation_19 (Activation)	(None, 32, 32, 22)	0	
batch_normalization_19[0][0]			

conv2d_20 (Conv2D)	(None, 32, 32, 22)	4356	
activation_19[0][0]			

dropout_9 (Dropout)	(None, 32, 32, 22)	0	conv2d_20[0][0]

concatenate_9 (Concatenate)	(None, 32, 32, 252)	0	
concatenate_8[0][0]			
			dropout_9[0][0]

batch_normalization_20 (BatchNo	(None, 32, 32, 252)	1008	


```

concatenate_9[0][0]
-----
-----
activation_20 (Activation)      (None, 32, 32, 252)  0
batch_normalization_20[0][0]
-----
-----
conv2d_21 (Conv2D)              (None, 32, 32, 22)   5544
activation_20[0][0]
-----
-----
batch_normalization_21 (BatchNo (None, 32, 32, 22)   88           conv2d_21[0][0]
-----
-----
activation_21 (Activation)      (None, 32, 32, 22)   0
batch_normalization_21[0][0]
-----
-----
conv2d_22 (Conv2D)              (None, 32, 32, 22)   4356
activation_21[0][0]
-----
-----
dropout_10 (Dropout)            (None, 32, 32, 22)   0           conv2d_22[0][0]
-----
-----
concatenate_10 (Concatenate)    (None, 32, 32, 274)  0
concatenate_9[0][0]
dropout_10[0][0]
-----
-----
batch_normalization_22 (BatchNo (None, 32, 32, 274)  1096
concatenate_10[0][0]
-----
-----
activation_22 (Activation)      (None, 32, 32, 274)  0
batch_normalization_22[0][0]
-----
-----
conv2d_23 (Conv2D)              (None, 32, 32, 22)   6028
activation_22[0][0]
-----
-----
batch_normalization_23 (BatchNo (None, 32, 32, 22)   88           conv2d_23[0][0]
-----
-----
activation_23 (Activation)      (None, 32, 32, 22)   0
batch_normalization_23[0][0]
-----

```

conv2d_24 (Conv2D)	(None, 32, 32, 22)	4356	
activation_23[0][0]			

dropout_11 (Dropout)	(None, 32, 32, 22)	0	conv2d_24[0][0]

concatenate_11 (Concatenate)	(None, 32, 32, 296)	0	
concatenate_10[0][0]			
dropout_11[0][0]			

batch_normalization_24 (Batch Normalization)	(None, 32, 32, 296)	1184	
concatenate_11[0][0]			

activation_24 (Activation)	(None, 32, 32, 296)	0	
batch_normalization_24[0][0]			

conv2d_25 (Conv2D)	(None, 32, 32, 22)	6512	
activation_24[0][0]			

dropout_12 (Dropout)	(None, 32, 32, 22)	0	conv2d_25[0][0]

average_pooling2d (AveragePooling2D)	(None, 16, 16, 22)	0	
dropout_12[0][0]			

batch_normalization_25 (Batch Normalization)	(None, 16, 16, 22)	88	
average_pooling2d[0][0]			

activation_25 (Activation)	(None, 16, 16, 22)	0	
batch_normalization_25[0][0]			

conv2d_26 (Conv2D)	(None, 16, 16, 22)	484	
activation_25[0][0]			

batch_normalization_26 (Batch Normalization)	(None, 16, 16, 22)	88	conv2d_26[0][0]

activation_26 (Activation)	(None, 16, 16, 22)	0	

batch_normalization_26[0][0]

conv2d_27 (Conv2D) (None, 16, 16, 22) 4356
activation_26[0][0]

dropout_13 (Dropout) (None, 16, 16, 22) 0 conv2d_27[0][0]

concatenate_12 (Concatenate) (None, 16, 16, 44) 0
average_pooling2d[0][0]
dropout_13[0][0]

batch_normalization_27 (BatchNormaliz (None, 16, 16, 44) 176
concatenate_12[0][0]

activation_27 (Activation) (None, 16, 16, 44) 0
batch_normalization_27[0][0]

conv2d_28 (Conv2D) (None, 16, 16, 22) 968
activation_27[0][0]

batch_normalization_28 (BatchNormaliz (None, 16, 16, 22) 88 conv2d_28[0][0]

activation_28 (Activation) (None, 16, 16, 22) 0
batch_normalization_28[0][0]

conv2d_29 (Conv2D) (None, 16, 16, 22) 4356
activation_28[0][0]

dropout_14 (Dropout) (None, 16, 16, 22) 0 conv2d_29[0][0]

concatenate_13 (Concatenate) (None, 16, 16, 66) 0
concatenate_12[0][0]
dropout_14[0][0]

batch_normalization_29 (BatchNormaliz (None, 16, 16, 66) 264
concatenate_13[0][0]

```

-----
activation_29 (Activation)      (None, 16, 16, 66)    0
batch_normalization_29[0][0]

-----

conv2d_30 (Conv2D)             (None, 16, 16, 22)    1452
activation_29[0][0]

-----

batch_normalization_30 (BatchNo (None, 16, 16, 22)    88          conv2d_30[0][0]
-----

activation_30 (Activation)      (None, 16, 16, 22)    0
batch_normalization_30[0][0]

-----

conv2d_31 (Conv2D)             (None, 16, 16, 22)    4356
activation_30[0][0]

-----

dropout_15 (Dropout)           (None, 16, 16, 22)    0          conv2d_31[0][0]
-----

concatenate_14 (Concatenate)    (None, 16, 16, 88)    0
concatenate_13[0][0]
dropout_15[0][0]

-----

batch_normalization_31 (BatchNo (None, 16, 16, 88)    352
concatenate_14[0][0]

-----

activation_31 (Activation)      (None, 16, 16, 88)    0
batch_normalization_31[0][0]

-----

conv2d_32 (Conv2D)             (None, 16, 16, 22)    1936
activation_31[0][0]

-----

batch_normalization_32 (BatchNo (None, 16, 16, 22)    88          conv2d_32[0][0]
-----

activation_32 (Activation)      (None, 16, 16, 22)    0
batch_normalization_32[0][0]
-----

```

conv2d_33 (Conv2D)	(None, 16, 16, 22)	4356	
activation_32[0][0]			

dropout_16 (Dropout)	(None, 16, 16, 22)	0	conv2d_33[0][0]

concatenate_15 (Concatenate)	(None, 16, 16, 110)	0	
concatenate_14[0][0]			
dropout_16[0][0]			

batch_normalization_33 (BatchNo	(None, 16, 16, 110)	440	
concatenate_15[0][0]			

activation_33 (Activation)	(None, 16, 16, 110)	0	
batch_normalization_33[0][0]			

conv2d_34 (Conv2D)	(None, 16, 16, 22)	2420	
activation_33[0][0]			

batch_normalization_34 (BatchNo	(None, 16, 16, 22)	88	conv2d_34[0][0]

activation_34 (Activation)	(None, 16, 16, 22)	0	
batch_normalization_34[0][0]			

conv2d_35 (Conv2D)	(None, 16, 16, 22)	4356	
activation_34[0][0]			

dropout_17 (Dropout)	(None, 16, 16, 22)	0	conv2d_35[0][0]

concatenate_16 (Concatenate)	(None, 16, 16, 132)	0	
concatenate_15[0][0]			
dropout_17[0][0]			

batch_normalization_35 (BatchNo	(None, 16, 16, 132)	528	
concatenate_16[0][0]			

activation_35 (Activation)	(None, 16, 16, 132)	0	

batch_normalization_35[0][0]

conv2d_36 (Conv2D) (None, 16, 16, 22) 2904
activation_35[0][0]

batch_normalization_36 (BatchNo (None, 16, 16, 22) 88 conv2d_36[0][0]

activation_36 (Activation) (None, 16, 16, 22) 0
batch_normalization_36[0][0]

conv2d_37 (Conv2D) (None, 16, 16, 22) 4356
activation_36[0][0]

dropout_18 (Dropout) (None, 16, 16, 22) 0 conv2d_37[0][0]

concatenate_17 (Concatenate) (None, 16, 16, 154) 0
concatenate_16[0][0]
dropout_18[0][0]

batch_normalization_37 (BatchNo (None, 16, 16, 154) 616
concatenate_17[0][0]

activation_37 (Activation) (None, 16, 16, 154) 0
batch_normalization_37[0][0]

conv2d_38 (Conv2D) (None, 16, 16, 22) 3388
activation_37[0][0]

batch_normalization_38 (BatchNo (None, 16, 16, 22) 88 conv2d_38[0][0]

activation_38 (Activation) (None, 16, 16, 22) 0
batch_normalization_38[0][0]

conv2d_39 (Conv2D) (None, 16, 16, 22) 4356
activation_38[0][0]

```

-----
dropout_19 (Dropout)          (None, 16, 16, 22)    0          conv2d_39[0][0]
-----

-----
concatenate_18 (Concatenate)  (None, 16, 16, 176)  0
concatenate_17[0][0]
dropout_19[0][0]
-----

-----
batch_normalization_39 (BatchNo (None, 16, 16, 176)  704
concatenate_18[0][0]
-----

-----
activation_39 (Activation)    (None, 16, 16, 176)  0
batch_normalization_39[0][0]
-----

-----
conv2d_40 (Conv2D)           (None, 16, 16, 22)   3872
activation_39[0][0]
-----

-----
batch_normalization_40 (BatchNo (None, 16, 16, 22)   88          conv2d_40[0][0]
-----

-----
activation_40 (Activation)    (None, 16, 16, 22)   0
batch_normalization_40[0][0]
-----

-----
conv2d_41 (Conv2D)           (None, 16, 16, 22)   4356
activation_40[0][0]
-----

-----
dropout_20 (Dropout)          (None, 16, 16, 22)    0          conv2d_41[0][0]
-----

-----
concatenate_19 (Concatenate)  (None, 16, 16, 198)  0
concatenate_18[0][0]
dropout_20[0][0]
-----

-----
batch_normalization_41 (BatchNo (None, 16, 16, 198)  792
concatenate_19[0][0]
-----

-----
activation_41 (Activation)    (None, 16, 16, 198)  0
batch_normalization_41[0][0]
-----

-----

```

conv2d_42 (Conv2D)	(None, 16, 16, 22)	4356	
activation_41[0][0]			

batch_normalization_42 (BatchNo	(None, 16, 16, 22)	88	conv2d_42[0][0]

activation_42 (Activation)	(None, 16, 16, 22)	0	
batch_normalization_42[0][0]			

conv2d_43 (Conv2D)	(None, 16, 16, 22)	4356	
activation_42[0][0]			

dropout_21 (Dropout)	(None, 16, 16, 22)	0	conv2d_43[0][0]

concatenate_20 (Concatenate)	(None, 16, 16, 220)	0	
concatenate_19[0][0]			
dropout_21[0][0]			

batch_normalization_43 (BatchNo	(None, 16, 16, 220)	880	
concatenate_20[0][0]			

activation_43 (Activation)	(None, 16, 16, 220)	0	
batch_normalization_43[0][0]			

conv2d_44 (Conv2D)	(None, 16, 16, 22)	4840	
activation_43[0][0]			

batch_normalization_44 (BatchNo	(None, 16, 16, 22)	88	conv2d_44[0][0]

activation_44 (Activation)	(None, 16, 16, 22)	0	
batch_normalization_44[0][0]			

conv2d_45 (Conv2D)	(None, 16, 16, 22)	4356	
activation_44[0][0]			

dropout_22 (Dropout)	(None, 16, 16, 22)	0	conv2d_45[0][0]

```

-----
concatenate_21 (Concatenate)      (None, 16, 16, 242)  0
concatenate_20[0][0]
dropout_22[0][0]
-----

-----
batch_normalization_45 (BatchNo (None, 16, 16, 242)  968
concatenate_21[0][0]
-----

-----
activation_45 (Activation)         (None, 16, 16, 242)  0
batch_normalization_45[0][0]
-----

-----
conv2d_46 (Conv2D)                (None, 16, 16, 22)   5324
activation_45[0][0]
-----

-----
batch_normalization_46 (BatchNo (None, 16, 16, 22)   88           conv2d_46[0][0]
-----

-----
activation_46 (Activation)         (None, 16, 16, 22)   0
batch_normalization_46[0][0]
-----

-----
conv2d_47 (Conv2D)                (None, 16, 16, 22)   4356
activation_46[0][0]
-----

-----
dropout_23 (Dropout)              (None, 16, 16, 22)   0           conv2d_47[0][0]
-----

-----
concatenate_22 (Concatenate)      (None, 16, 16, 264)  0
concatenate_21[0][0]
dropout_23[0][0]
-----

-----
batch_normalization_47 (BatchNo (None, 16, 16, 264)  1056
concatenate_22[0][0]
-----

-----
activation_47 (Activation)         (None, 16, 16, 264)  0
batch_normalization_47[0][0]
-----

-----
conv2d_48 (Conv2D)                (None, 16, 16, 22)   5808
activation_47[0][0]
-----

```

```

-----
batch_normalization_48 (BatchNo (None, 16, 16, 22) 88 conv2d_48[0] [0]
-----
-----
activation_48 (Activation) (None, 16, 16, 22) 0
batch_normalization_48[0] [0]
-----
-----
conv2d_49 (Conv2D) (None, 16, 16, 22) 4356
activation_48[0] [0]
-----
-----
dropout_24 (Dropout) (None, 16, 16, 22) 0 conv2d_49[0] [0]
-----
-----
concatenate_23 (Concatenate) (None, 16, 16, 286) 0
concatenate_22[0] [0]
dropout_24[0] [0]
-----
-----
batch_normalization_49 (BatchNo (None, 16, 16, 286) 1144
concatenate_23[0] [0]
-----
-----
activation_49 (Activation) (None, 16, 16, 286) 0
batch_normalization_49[0] [0]
-----
-----
conv2d_50 (Conv2D) (None, 16, 16, 22) 6292
activation_49[0] [0]
-----
-----
dropout_25 (Dropout) (None, 16, 16, 22) 0 conv2d_50[0] [0]
-----
-----
average_pooling2d_1 (AveragePoo (None, 8, 8, 22) 0
dropout_25[0] [0]
-----
-----
batch_normalization_50 (BatchNo (None, 8, 8, 22) 88
average_pooling2d_1[0] [0]
-----
-----
activation_50 (Activation) (None, 8, 8, 22) 0
batch_normalization_50[0] [0]
-----
-----
conv2d_51 (Conv2D) (None, 8, 8, 22) 484

```

activation_50[0][0]

batch_normalization_51 (BatchNo (None, 8, 8, 22) 88 conv2d_51[0][0]

activation_51 (Activation) (None, 8, 8, 22) 0
batch_normalization_51[0][0]

conv2d_52 (Conv2D) (None, 8, 8, 22) 4356
activation_51[0][0]

dropout_26 (Dropout) (None, 8, 8, 22) 0 conv2d_52[0][0]

concatenate_24 (Concatenate) (None, 8, 8, 44) 0
average_pooling2d_1[0][0]
dropout_26[0][0]

batch_normalization_52 (BatchNo (None, 8, 8, 44) 176
concatenate_24[0][0]

activation_52 (Activation) (None, 8, 8, 44) 0
batch_normalization_52[0][0]

conv2d_53 (Conv2D) (None, 8, 8, 22) 968
activation_52[0][0]

batch_normalization_53 (BatchNo (None, 8, 8, 22) 88 conv2d_53[0][0]

activation_53 (Activation) (None, 8, 8, 22) 0
batch_normalization_53[0][0]

conv2d_54 (Conv2D) (None, 8, 8, 22) 4356
activation_53[0][0]

dropout_27 (Dropout) (None, 8, 8, 22) 0 conv2d_54[0][0]

concatenate_25 (Concatenate)	(None, 8, 8, 66)	0	
concatenate_24[0][0]			
dropout_27[0][0]			

batch_normalization_54 (BatchNo	(None, 8, 8, 66)	264	
concatenate_25[0][0]			

activation_54 (Activation)	(None, 8, 8, 66)	0	
batch_normalization_54[0][0]			

conv2d_55 (Conv2D)	(None, 8, 8, 22)	1452	
activation_54[0][0]			

batch_normalization_55 (BatchNo	(None, 8, 8, 22)	88	conv2d_55[0][0]

activation_55 (Activation)	(None, 8, 8, 22)	0	
batch_normalization_55[0][0]			

conv2d_56 (Conv2D)	(None, 8, 8, 22)	4356	
activation_55[0][0]			

dropout_28 (Dropout)	(None, 8, 8, 22)	0	conv2d_56[0][0]

concatenate_26 (Concatenate)	(None, 8, 8, 88)	0	
concatenate_25[0][0]			
dropout_28[0][0]			

batch_normalization_56 (BatchNo	(None, 8, 8, 88)	352	
concatenate_26[0][0]			

activation_56 (Activation)	(None, 8, 8, 88)	0	
batch_normalization_56[0][0]			

conv2d_57 (Conv2D)	(None, 8, 8, 22)	1936	
activation_56[0][0]			

batch_normalization_57 (BatchNo	(None, 8, 8, 22)	88	conv2d_57[0][0]

activation_57 (Activation)	(None, 8, 8, 22)	0	
batch_normalization_57[0][0]			

conv2d_58 (Conv2D)	(None, 8, 8, 22)	4356	
activation_57[0][0]			

dropout_29 (Dropout)	(None, 8, 8, 22)	0	conv2d_58[0][0]

concatenate_27 (Concatenate)	(None, 8, 8, 110)	0	
concatenate_26[0][0]			
dropout_29[0][0]			

batch_normalization_58 (BatchNo	(None, 8, 8, 110)	440	
concatenate_27[0][0]			

activation_58 (Activation)	(None, 8, 8, 110)	0	
batch_normalization_58[0][0]			

conv2d_59 (Conv2D)	(None, 8, 8, 22)	2420	
activation_58[0][0]			

batch_normalization_59 (BatchNo	(None, 8, 8, 22)	88	conv2d_59[0][0]

activation_59 (Activation)	(None, 8, 8, 22)	0	
batch_normalization_59[0][0]			

conv2d_60 (Conv2D)	(None, 8, 8, 22)	4356	
activation_59[0][0]			

dropout_30 (Dropout)	(None, 8, 8, 22)	0	conv2d_60[0][0]

concatenate_28 (Concatenate)	(None, 8, 8, 132)	0	
concatenate_27[0][0]			
dropout_30[0][0]			

```

-----
batch_normalization_60 (BatchNo (None, 8, 8, 132)    528
concatenate_28[0][0]
-----

activation_60 (Activation)      (None, 8, 8, 132)    0
batch_normalization_60[0][0]
-----

conv2d_61 (Conv2D)              (None, 8, 8, 22)     2904
activation_60[0][0]
-----

batch_normalization_61 (BatchNo (None, 8, 8, 22)     88          conv2d_61[0][0]
-----

activation_61 (Activation)      (None, 8, 8, 22)     0
batch_normalization_61[0][0]
-----

conv2d_62 (Conv2D)              (None, 8, 8, 22)     4356
activation_61[0][0]
-----

dropout_31 (Dropout)            (None, 8, 8, 22)     0          conv2d_62[0][0]
-----

concatenate_29 (Concatenate)    (None, 8, 8, 154)    0
concatenate_28[0][0]
dropout_31[0][0]
-----

batch_normalization_62 (BatchNo (None, 8, 8, 154)    616
concatenate_29[0][0]
-----

activation_62 (Activation)      (None, 8, 8, 154)    0
batch_normalization_62[0][0]
-----

conv2d_63 (Conv2D)              (None, 8, 8, 22)     3388
activation_62[0][0]
-----

batch_normalization_63 (BatchNo (None, 8, 8, 22)     88          conv2d_63[0][0]
-----

```

activation_63 (Activation)	(None, 8, 8, 22)	0	
batch_normalization_63[0][0]			

conv2d_64 (Conv2D)	(None, 8, 8, 22)	4356	
activation_63[0][0]			

dropout_32 (Dropout)	(None, 8, 8, 22)	0	conv2d_64[0][0]

concatenate_30 (Concatenate)	(None, 8, 8, 176)	0	
concatenate_29[0][0]			
dropout_32[0][0]			

batch_normalization_64 (BatchNo	(None, 8, 8, 176)	704	
concatenate_30[0][0]			

activation_64 (Activation)	(None, 8, 8, 176)	0	
batch_normalization_64[0][0]			

conv2d_65 (Conv2D)	(None, 8, 8, 22)	3872	
activation_64[0][0]			

batch_normalization_65 (BatchNo	(None, 8, 8, 22)	88	conv2d_65[0][0]

activation_65 (Activation)	(None, 8, 8, 22)	0	
batch_normalization_65[0][0]			

conv2d_66 (Conv2D)	(None, 8, 8, 22)	4356	
activation_65[0][0]			

dropout_33 (Dropout)	(None, 8, 8, 22)	0	conv2d_66[0][0]

concatenate_31 (Concatenate)	(None, 8, 8, 198)	0	
concatenate_30[0][0]			
dropout_33[0][0]			

batch_normalization_66 (BatchNo	(None, 8, 8, 198)	792	

```

concatenate_31[0][0]
-----
-----
activation_66 (Activation)      (None, 8, 8, 198)      0
batch_normalization_66[0][0]
-----
-----
conv2d_67 (Conv2D)              (None, 8, 8, 22)      4356
activation_66[0][0]
-----
-----
batch_normalization_67 (BatchNo (None, 8, 8, 22)      88      conv2d_67[0][0]
-----
-----
activation_67 (Activation)      (None, 8, 8, 22)      0
batch_normalization_67[0][0]
-----
-----
conv2d_68 (Conv2D)              (None, 8, 8, 22)      4356
activation_67[0][0]
-----
-----
dropout_34 (Dropout)            (None, 8, 8, 22)      0      conv2d_68[0][0]
-----
-----
concatenate_32 (Concatenate)    (None, 8, 8, 220)      0
concatenate_31[0][0]
dropout_34[0][0]
-----
-----
batch_normalization_68 (BatchNo (None, 8, 8, 220)      880
concatenate_32[0][0]
-----
-----
activation_68 (Activation)      (None, 8, 8, 220)      0
batch_normalization_68[0][0]
-----
-----
conv2d_69 (Conv2D)              (None, 8, 8, 22)      4840
activation_68[0][0]
-----
-----
batch_normalization_69 (BatchNo (None, 8, 8, 22)      88      conv2d_69[0][0]
-----
-----
activation_69 (Activation)      (None, 8, 8, 22)      0
batch_normalization_69[0][0]
-----

```


conv2d_70 (Conv2D) activation_69[0][0]	(None, 8, 8, 22)	4356	
dropout_35 (Dropout)	(None, 8, 8, 22)	0	conv2d_70[0][0]
concatenate_33 (Concatenate) concatenate_32[0][0] dropout_35[0][0]	(None, 8, 8, 242)	0	
batch_normalization_70 (Batch Normalization) concatenate_33[0][0]	(None, 8, 8, 242)	968	
activation_70 (Activation) batch_normalization_70[0][0]	(None, 8, 8, 242)	0	
conv2d_71 (Conv2D) activation_70[0][0]	(None, 8, 8, 22)	5324	
batch_normalization_71 (Batch Normalization) conv2d_71[0][0]	(None, 8, 8, 22)	88	conv2d_71[0][0]
activation_71 (Activation) batch_normalization_71[0][0]	(None, 8, 8, 22)	0	
conv2d_72 (Conv2D) activation_71[0][0]	(None, 8, 8, 22)	4356	
dropout_36 (Dropout)	(None, 8, 8, 22)	0	conv2d_72[0][0]
concatenate_34 (Concatenate) concatenate_33[0][0] dropout_36[0][0]	(None, 8, 8, 264)	0	
batch_normalization_72 (Batch Normalization) concatenate_34[0][0]	(None, 8, 8, 264)	1056	

activation_72 (Activation)	(None, 8, 8, 264)	0	
batch_normalization_72[0][0]			

conv2d_73 (Conv2D)	(None, 8, 8, 22)	5808	
activation_72[0][0]			

batch_normalization_73 (BatchNo	(None, 8, 8, 22)	88	conv2d_73[0][0]

activation_73 (Activation)	(None, 8, 8, 22)	0	
batch_normalization_73[0][0]			

conv2d_74 (Conv2D)	(None, 8, 8, 22)	4356	
activation_73[0][0]			

dropout_37 (Dropout)	(None, 8, 8, 22)	0	conv2d_74[0][0]

concatenate_35 (Concatenate)	(None, 8, 8, 286)	0	
concatenate_34[0][0]			
dropout_37[0][0]			

batch_normalization_74 (BatchNo	(None, 8, 8, 286)	1144	
concatenate_35[0][0]			

activation_74 (Activation)	(None, 8, 8, 286)	0	
batch_normalization_74[0][0]			

conv2d_75 (Conv2D)	(None, 8, 8, 22)	6292	
activation_74[0][0]			

dropout_38 (Dropout)	(None, 8, 8, 22)	0	conv2d_75[0][0]

average_pooling2d_2 (AveragePoo	(None, 4, 4, 22)	0	
dropout_38[0][0]			

batch_normalization_75 (BatchNo	(None, 4, 4, 22)	88	
average_pooling2d_2[0][0]			

activation_75 (Activation)	(None, 4, 4, 22)	0	
batch_normalization_75[0][0]			

conv2d_76 (Conv2D)	(None, 4, 4, 22)	484	
activation_75[0][0]			

batch_normalization_76 (BatchNo	(None, 4, 4, 22)	88	conv2d_76[0][0]

activation_76 (Activation)	(None, 4, 4, 22)	0	
batch_normalization_76[0][0]			

conv2d_77 (Conv2D)	(None, 4, 4, 22)	4356	
activation_76[0][0]			

dropout_39 (Dropout)	(None, 4, 4, 22)	0	conv2d_77[0][0]

concatenate_36 (Concatenate)	(None, 4, 4, 44)	0	
average_pooling2d_2[0][0]			
dropout_39[0][0]			

batch_normalization_77 (BatchNo	(None, 4, 4, 44)	176	
concatenate_36[0][0]			

activation_77 (Activation)	(None, 4, 4, 44)	0	
batch_normalization_77[0][0]			

conv2d_78 (Conv2D)	(None, 4, 4, 22)	968	
activation_77[0][0]			

batch_normalization_78 (BatchNo	(None, 4, 4, 22)	88	conv2d_78[0][0]

activation_78 (Activation)	(None, 4, 4, 22)	0	
batch_normalization_78[0][0]			

conv2d_79 (Conv2D)	(None, 4, 4, 22)	4356	
activation_78[0][0]			

dropout_40 (Dropout)	(None, 4, 4, 22)	0	conv2d_79[0][0]

concatenate_37 (Concatenate)	(None, 4, 4, 66)	0	
concatenate_36[0][0]			
dropout_40[0][0]			

batch_normalization_79 (BatchNo	(None, 4, 4, 66)	264	
concatenate_37[0][0]			

activation_79 (Activation)	(None, 4, 4, 66)	0	
batch_normalization_79[0][0]			

conv2d_80 (Conv2D)	(None, 4, 4, 22)	1452	
activation_79[0][0]			

batch_normalization_80 (BatchNo	(None, 4, 4, 22)	88	conv2d_80[0][0]

activation_80 (Activation)	(None, 4, 4, 22)	0	
batch_normalization_80[0][0]			

conv2d_81 (Conv2D)	(None, 4, 4, 22)	4356	
activation_80[0][0]			

dropout_41 (Dropout)	(None, 4, 4, 22)	0	conv2d_81[0][0]

concatenate_38 (Concatenate)	(None, 4, 4, 88)	0	
concatenate_37[0][0]			
dropout_41[0][0]			

batch_normalization_81 (BatchNo	(None, 4, 4, 88)	352	
concatenate_38[0][0]			

activation_81 (Activation)	(None, 4, 4, 88)	0	

batch_normalization_81[0][0]

conv2d_82 (Conv2D) (None, 4, 4, 22) 1936
activation_81[0][0]

batch_normalization_82 (BatchNo (None, 4, 4, 22) 88 conv2d_82[0][0]

activation_82 (Activation) (None, 4, 4, 22) 0
batch_normalization_82[0][0]

conv2d_83 (Conv2D) (None, 4, 4, 22) 4356
activation_82[0][0]

dropout_42 (Dropout) (None, 4, 4, 22) 0 conv2d_83[0][0]

concatenate_39 (Concatenate) (None, 4, 4, 110) 0
concatenate_38[0][0]
dropout_42[0][0]

batch_normalization_83 (BatchNo (None, 4, 4, 110) 440
concatenate_39[0][0]

activation_83 (Activation) (None, 4, 4, 110) 0
batch_normalization_83[0][0]

conv2d_84 (Conv2D) (None, 4, 4, 22) 2420
activation_83[0][0]

batch_normalization_84 (BatchNo (None, 4, 4, 22) 88 conv2d_84[0][0]

activation_84 (Activation) (None, 4, 4, 22) 0
batch_normalization_84[0][0]

conv2d_85 (Conv2D) (None, 4, 4, 22) 4356
activation_84[0][0]

```

-----
dropout_43 (Dropout)          (None, 4, 4, 22)    0          conv2d_85[0][0]
-----

-----
concatenate_40 (Concatenate)  (None, 4, 4, 132)   0
concatenate_39[0][0]
dropout_43[0][0]
-----

-----
batch_normalization_85 (BatchNo (None, 4, 4, 132)    528
concatenate_40[0][0]
-----

-----
activation_85 (Activation)    (None, 4, 4, 132)   0
batch_normalization_85[0][0]
-----

-----
conv2d_86 (Conv2D)           (None, 4, 4, 22)    2904
activation_85[0][0]
-----

-----
batch_normalization_86 (BatchNo (None, 4, 4, 22)    88          conv2d_86[0][0]
-----

-----
activation_86 (Activation)    (None, 4, 4, 22)    0
batch_normalization_86[0][0]
-----

-----
conv2d_87 (Conv2D)           (None, 4, 4, 22)    4356
activation_86[0][0]
-----

-----
dropout_44 (Dropout)          (None, 4, 4, 22)    0          conv2d_87[0][0]
-----

-----
concatenate_41 (Concatenate)  (None, 4, 4, 154)   0
concatenate_40[0][0]
dropout_44[0][0]
-----

-----
batch_normalization_87 (BatchNo (None, 4, 4, 154)    616
concatenate_41[0][0]
-----

-----
activation_87 (Activation)    (None, 4, 4, 154)   0
batch_normalization_87[0][0]
-----

```

conv2d_88 (Conv2D)	(None, 4, 4, 22)	3388	
activation_87[0][0]			

batch_normalization_88 (BatchNo	(None, 4, 4, 22)	88	conv2d_88[0][0]

activation_88 (Activation)	(None, 4, 4, 22)	0	
batch_normalization_88[0][0]			

conv2d_89 (Conv2D)	(None, 4, 4, 22)	4356	
activation_88[0][0]			

dropout_45 (Dropout)	(None, 4, 4, 22)	0	conv2d_89[0][0]

concatenate_42 (Concatenate)	(None, 4, 4, 176)	0	
concatenate_41[0][0]			
dropout_45[0][0]			

batch_normalization_89 (BatchNo	(None, 4, 4, 176)	704	
concatenate_42[0][0]			

activation_89 (Activation)	(None, 4, 4, 176)	0	
batch_normalization_89[0][0]			

conv2d_90 (Conv2D)	(None, 4, 4, 22)	3872	
activation_89[0][0]			

batch_normalization_90 (BatchNo	(None, 4, 4, 22)	88	conv2d_90[0][0]

activation_90 (Activation)	(None, 4, 4, 22)	0	
batch_normalization_90[0][0]			

conv2d_91 (Conv2D)	(None, 4, 4, 22)	4356	
activation_90[0][0]			

dropout_46 (Dropout)	(None, 4, 4, 22)	0	conv2d_91[0][0]

```

-----
concatenate_43 (Concatenate)      (None, 4, 4, 198)      0
concatenate_42[0][0]
dropout_46[0][0]
-----

-----
batch_normalization_91 (BatchNo (None, 4, 4, 198)      792
concatenate_43[0][0]
-----

-----
activation_91 (Activation)          (None, 4, 4, 198)      0
batch_normalization_91[0][0]
-----

-----
conv2d_92 (Conv2D)                  (None, 4, 4, 22)       4356
activation_91[0][0]
-----

-----
batch_normalization_92 (BatchNo (None, 4, 4, 22)       88          conv2d_92[0][0]
-----

-----
activation_92 (Activation)          (None, 4, 4, 22)       0
batch_normalization_92[0][0]
-----

-----
conv2d_93 (Conv2D)                  (None, 4, 4, 22)       4356
activation_92[0][0]
-----

-----
dropout_47 (Dropout)                (None, 4, 4, 22)       0          conv2d_93[0][0]
-----

-----
concatenate_44 (Concatenate)      (None, 4, 4, 220)      0
concatenate_43[0][0]
dropout_47[0][0]
-----

-----
batch_normalization_93 (BatchNo (None, 4, 4, 220)      880
concatenate_44[0][0]
-----

-----
activation_93 (Activation)          (None, 4, 4, 220)      0
batch_normalization_93[0][0]
-----

-----
conv2d_94 (Conv2D)                  (None, 4, 4, 22)       4840
activation_93[0][0]
-----

```


batch_normalization_94 (BatchNo	(None, 4, 4, 22)	88	conv2d_94[0][0]

activation_94 (Activation)	(None, 4, 4, 22)	0	
batch_normalization_94[0][0]			

conv2d_95 (Conv2D)	(None, 4, 4, 22)	4356	
activation_94[0][0]			

dropout_48 (Dropout)	(None, 4, 4, 22)	0	conv2d_95[0][0]

concatenate_45 (Concatenate)	(None, 4, 4, 242)	0	
concatenate_44[0][0]			
dropout_48[0][0]			

batch_normalization_95 (BatchNo	(None, 4, 4, 242)	968	
concatenate_45[0][0]			

activation_95 (Activation)	(None, 4, 4, 242)	0	
batch_normalization_95[0][0]			

conv2d_96 (Conv2D)	(None, 4, 4, 22)	5324	
activation_95[0][0]			

batch_normalization_96 (BatchNo	(None, 4, 4, 22)	88	conv2d_96[0][0]

activation_96 (Activation)	(None, 4, 4, 22)	0	
batch_normalization_96[0][0]			

conv2d_97 (Conv2D)	(None, 4, 4, 22)	4356	
activation_96[0][0]			

dropout_49 (Dropout)	(None, 4, 4, 22)	0	conv2d_97[0][0]

concatenate_46 (Concatenate)	(None, 4, 4, 264)	0	
concatenate_45[0][0]			

dropout_49[0][0]

batch_normalization_97 (BatchNo (None, 4, 4, 264) 1056
concatenate_46[0][0]

activation_97 (Activation) (None, 4, 4, 264) 0
batch_normalization_97[0][0]

conv2d_98 (Conv2D) (None, 4, 4, 22) 5808
activation_97[0][0]

batch_normalization_98 (BatchNo (None, 4, 4, 22) 88 conv2d_98[0][0]

activation_98 (Activation) (None, 4, 4, 22) 0
batch_normalization_98[0][0]

conv2d_99 (Conv2D) (None, 4, 4, 22) 4356
activation_98[0][0]

dropout_50 (Dropout) (None, 4, 4, 22) 0 conv2d_99[0][0]

concatenate_47 (Concatenate) (None, 4, 4, 286) 0
concatenate_46[0][0]
dropout_50[0][0]

batch_normalization_99 (BatchNo (None, 4, 4, 286) 1144
concatenate_47[0][0]

activation_99 (Activation) (None, 4, 4, 286) 0
batch_normalization_99[0][0]

average_pooling2d_3 (AveragePoo (None, 2, 2, 286) 0
activation_99[0][0]

flatten (Flatten) (None, 1144) 0
average_pooling2d_3[0][0]

```

-----
-----
dense (Dense)                (None, 10)                11450                flatten[0][0]
=====
=====
Total params: 430,922
Trainable params: 412,534
Non-trainable params: 18,388
-----
-----

```

```
[ ]: print(len(model4.layers))
```

406

```
[ ]: model4.
      ↳ compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])

model4.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 [=====] - 156s 147ms/step - loss: 1.8745 - accuracy: 0.3166 - val_loss: 1.5755 - val_accuracy: 0.4120

Epoch 00001: val_accuracy improved from -inf to 0.41200, saving model to model4_weights.best.hdf5

Epoch 2/100

781/781 [=====] - 113s 144ms/step - loss: 1.5134 - accuracy: 0.4476 - val_loss: 2.1495 - val_accuracy: 0.3897

Epoch 00002: val_accuracy did not improve from 0.41200

Epoch 3/100

781/781 [=====] - 113s 145ms/step - loss: 1.2878 - accuracy: 0.5322 - val_loss: 1.3345 - val_accuracy: 0.5544

Epoch 00003: val_accuracy improved from 0.41200 to 0.55440, saving model to model4_weights.best.hdf5

Epoch 4/100

781/781 [=====] - 113s 144ms/step - loss: 1.1293 - accuracy: 0.5937 - val_loss: 1.5858 - val_accuracy: 0.5304

Epoch 00004: val_accuracy did not improve from 0.55440

Epoch 5/100

781/781 [=====] - 112s 144ms/step - loss: 1.0299 -
accuracy: 0.6319 - val_loss: 1.9975 - val_accuracy: 0.5153

Epoch 00005: val_accuracy did not improve from 0.55440

Epoch 6/100

781/781 [=====] - 111s 142ms/step - loss: 0.9488 -
accuracy: 0.6613 - val_loss: 1.9263 - val_accuracy: 0.5524

Epoch 00006: val_accuracy did not improve from 0.55440

Epoch 7/100

781/781 [=====] - 113s 145ms/step - loss: 0.8844 -
accuracy: 0.6872 - val_loss: 1.7518 - val_accuracy: 0.5706

Epoch 00007: val_accuracy improved from 0.55440 to 0.57060, saving model to
model4_weights.best.hdf5

Epoch 8/100

781/781 [=====] - 112s 144ms/step - loss: 0.8365 -
accuracy: 0.7077 - val_loss: 1.0898 - val_accuracy: 0.6690

Epoch 00008: val_accuracy improved from 0.57060 to 0.66900, saving model to
model4_weights.best.hdf5

Epoch 9/100

781/781 [=====] - 112s 143ms/step - loss: 0.7971 -
accuracy: 0.7210 - val_loss: 1.0606 - val_accuracy: 0.6947

Epoch 00009: val_accuracy improved from 0.66900 to 0.69470, saving model to
model4_weights.best.hdf5

Epoch 10/100

781/781 [=====] - 112s 143ms/step - loss: 0.7677 -
accuracy: 0.7309 - val_loss: 1.2691 - val_accuracy: 0.6506

Epoch 00010: val_accuracy did not improve from 0.69470

Epoch 11/100

781/781 [=====] - 111s 143ms/step - loss: 0.7344 -
accuracy: 0.7426 - val_loss: 1.2606 - val_accuracy: 0.6762

Epoch 00011: val_accuracy did not improve from 0.69470

Epoch 12/100

781/781 [=====] - 111s 142ms/step - loss: 0.6964 -
accuracy: 0.7563 - val_loss: 1.0189 - val_accuracy: 0.7016

Epoch 00012: val_accuracy improved from 0.69470 to 0.70160, saving model to
model4_weights.best.hdf5

Epoch 13/100

781/781 [=====] - 111s 141ms/step - loss: 0.6829 -
accuracy: 0.7622 - val_loss: 1.1461 - val_accuracy: 0.6758

Epoch 00013: val_accuracy did not improve from 0.70160

Epoch 14/100
781/781 [=====] - 110s 141ms/step - loss: 0.6550 -
accuracy: 0.7730 - val_loss: 0.6706 - val_accuracy: 0.7818

Epoch 00014: val_accuracy improved from 0.70160 to 0.78180, saving model to
model4_weights.best.hdf5

Epoch 15/100
781/781 [=====] - 111s 143ms/step - loss: 0.6381 -
accuracy: 0.7790 - val_loss: 0.8606 - val_accuracy: 0.7516

Epoch 00015: val_accuracy did not improve from 0.78180

Epoch 16/100
781/781 [=====] - 110s 141ms/step - loss: 0.6196 -
accuracy: 0.7846 - val_loss: 0.7987 - val_accuracy: 0.7640

Epoch 00016: val_accuracy did not improve from 0.78180

Epoch 17/100
781/781 [=====] - 111s 142ms/step - loss: 0.5988 -
accuracy: 0.7919 - val_loss: 1.0603 - val_accuracy: 0.7108

Epoch 00017: val_accuracy did not improve from 0.78180

Epoch 18/100
781/781 [=====] - 110s 141ms/step - loss: 0.5874 -
accuracy: 0.7963 - val_loss: 0.7823 - val_accuracy: 0.7798

Epoch 00018: val_accuracy did not improve from 0.78180

Epoch 19/100
781/781 [=====] - 110s 141ms/step - loss: 0.5753 -
accuracy: 0.8027 - val_loss: 1.0639 - val_accuracy: 0.7091

Epoch 00019: val_accuracy did not improve from 0.78180

Epoch 20/100
781/781 [=====] - 111s 142ms/step - loss: 0.5601 -
accuracy: 0.8059 - val_loss: 0.8634 - val_accuracy: 0.7545

Epoch 00020: val_accuracy did not improve from 0.78180

Epoch 21/100
781/781 [=====] - 110s 141ms/step - loss: 0.5575 -
accuracy: 0.8075 - val_loss: 0.5594 - val_accuracy: 0.8232

Epoch 00021: val_accuracy improved from 0.78180 to 0.82320, saving model to
model4_weights.best.hdf5

Epoch 22/100
781/781 [=====] - 110s 141ms/step - loss: 0.5381 -
accuracy: 0.8136 - val_loss: 1.3817 - val_accuracy: 0.6963

Epoch 00022: val_accuracy did not improve from 0.82320

Epoch 23/100

781/781 [=====] - 111s 143ms/step - loss: 0.5261 -
accuracy: 0.8193 - val_loss: 0.7018 - val_accuracy: 0.7932

Epoch 00023: val_accuracy did not improve from 0.82320

Epoch 24/100

781/781 [=====] - 111s 142ms/step - loss: 0.5167 -
accuracy: 0.8212 - val_loss: 0.5503 - val_accuracy: 0.8296

Epoch 00024: val_accuracy improved from 0.82320 to 0.82960, saving model to
model4_weights.best.hdf5

Epoch 25/100

781/781 [=====] - 111s 143ms/step - loss: 0.5086 -
accuracy: 0.8253 - val_loss: 0.6031 - val_accuracy: 0.8230

Epoch 00025: val_accuracy did not improve from 0.82960

Epoch 26/100

781/781 [=====] - 111s 142ms/step - loss: 0.4321 -
accuracy: 0.8503 - val_loss: 0.5011 - val_accuracy: 0.8493

Epoch 00026: val_accuracy improved from 0.82960 to 0.84930, saving model to
model4_weights.best.hdf5

Epoch 27/100

781/781 [=====] - 111s 142ms/step - loss: 0.4076 -
accuracy: 0.8589 - val_loss: 0.5119 - val_accuracy: 0.8483

Epoch 00027: val_accuracy did not improve from 0.84930

Epoch 28/100

781/781 [=====] - 111s 142ms/step - loss: 0.4057 -
accuracy: 0.8594 - val_loss: 0.5179 - val_accuracy: 0.8472

Epoch 00028: val_accuracy did not improve from 0.84930

Epoch 29/100

781/781 [=====] - 112s 143ms/step - loss: 0.3879 -
accuracy: 0.8648 - val_loss: 0.4711 - val_accuracy: 0.8602

Epoch 00029: val_accuracy improved from 0.84930 to 0.86020, saving model to
model4_weights.best.hdf5

Epoch 30/100

781/781 [=====] - 111s 142ms/step - loss: 0.3886 -
accuracy: 0.8649 - val_loss: 0.5071 - val_accuracy: 0.8506

Epoch 00030: val_accuracy did not improve from 0.86020

Epoch 31/100

781/781 [=====] - 112s 144ms/step - loss: 0.3812 -
accuracy: 0.8668 - val_loss: 0.4773 - val_accuracy: 0.8555

Epoch 00031: val_accuracy did not improve from 0.86020

Epoch 32/100

781/781 [=====] - 112s 143ms/step - loss: 0.3778 -
accuracy: 0.8697 - val_loss: 0.4644 - val_accuracy: 0.8615

Epoch 00032: val_accuracy improved from 0.86020 to 0.86150, saving model to
model4_weights.best.hdf5

Epoch 33/100

781/781 [=====] - 113s 144ms/step - loss: 0.3708 -
accuracy: 0.8713 - val_loss: 0.4852 - val_accuracy: 0.8567

Epoch 00033: val_accuracy did not improve from 0.86150

Epoch 34/100

781/781 [=====] - 112s 143ms/step - loss: 0.3701 -
accuracy: 0.8716 - val_loss: 0.4682 - val_accuracy: 0.8628

Epoch 00034: val_accuracy improved from 0.86150 to 0.86280, saving model to
model4_weights.best.hdf5

Epoch 35/100

781/781 [=====] - 112s 143ms/step - loss: 0.3693 -
accuracy: 0.8707 - val_loss: 0.4320 - val_accuracy: 0.8683

Epoch 00035: val_accuracy improved from 0.86280 to 0.86830, saving model to
model4_weights.best.hdf5

Epoch 36/100

781/781 [=====] - 113s 144ms/step - loss: 0.3677 -
accuracy: 0.8723 - val_loss: 0.4543 - val_accuracy: 0.8612

Epoch 00036: val_accuracy did not improve from 0.86830

Epoch 37/100

781/781 [=====] - 111s 142ms/step - loss: 0.3629 -
accuracy: 0.8740 - val_loss: 0.4707 - val_accuracy: 0.8607

Epoch 00037: val_accuracy did not improve from 0.86830

Epoch 38/100

781/781 [=====] - 111s 142ms/step - loss: 0.3639 -
accuracy: 0.8720 - val_loss: 0.4179 - val_accuracy: 0.8723

Epoch 00038: val_accuracy improved from 0.86830 to 0.87230, saving model to
model4_weights.best.hdf5

Epoch 39/100

781/781 [=====] - 111s 142ms/step - loss: 0.3587 -
accuracy: 0.8753 - val_loss: 0.4980 - val_accuracy: 0.8545

Epoch 00039: val_accuracy did not improve from 0.87230

Epoch 40/100

781/781 [=====] - 111s 142ms/step - loss: 0.3590 -
accuracy: 0.8754 - val_loss: 0.4487 - val_accuracy: 0.8654

Epoch 00040: val_accuracy did not improve from 0.87230

Epoch 41/100
781/781 [=====] - 112s 143ms/step - loss: 0.3584 - accuracy: 0.8753 - val_loss: 0.4502 - val_accuracy: 0.8675

Epoch 00041: val_accuracy did not improve from 0.87230

Epoch 42/100
781/781 [=====] - 112s 143ms/step - loss: 0.3542 - accuracy: 0.8765 - val_loss: 0.4410 - val_accuracy: 0.8695

Epoch 00042: val_accuracy did not improve from 0.87230

Epoch 43/100
781/781 [=====] - 112s 143ms/step - loss: 0.3493 - accuracy: 0.8794 - val_loss: 0.4599 - val_accuracy: 0.8645

Epoch 00043: val_accuracy did not improve from 0.87230

Epoch 44/100
781/781 [=====] - 112s 143ms/step - loss: 0.3514 - accuracy: 0.8775 - val_loss: 0.4182 - val_accuracy: 0.8739

Epoch 00044: val_accuracy improved from 0.87230 to 0.87390, saving model to model4_weights.best.hdf5

Epoch 45/100
781/781 [=====] - 111s 142ms/step - loss: 0.3435 - accuracy: 0.8812 - val_loss: 0.4750 - val_accuracy: 0.8621

Epoch 00045: val_accuracy did not improve from 0.87390

Epoch 46/100
781/781 [=====] - 112s 143ms/step - loss: 0.3432 - accuracy: 0.8806 - val_loss: 0.4735 - val_accuracy: 0.8647

Epoch 00046: val_accuracy did not improve from 0.87390

Epoch 47/100
781/781 [=====] - 112s 144ms/step - loss: 0.3459 - accuracy: 0.8795 - val_loss: 0.4101 - val_accuracy: 0.8783

Epoch 00047: val_accuracy improved from 0.87390 to 0.87830, saving model to model4_weights.best.hdf5

Epoch 48/100
781/781 [=====] - 112s 143ms/step - loss: 0.3429 - accuracy: 0.8806 - val_loss: 0.4195 - val_accuracy: 0.8758

Epoch 00048: val_accuracy did not improve from 0.87830

Epoch 49/100
781/781 [=====] - 111s 143ms/step - loss: 0.3411 - accuracy: 0.8811 - val_loss: 0.4149 - val_accuracy: 0.8765

Epoch 00049: val_accuracy did not improve from 0.87830

Epoch 50/100

781/781 [=====] - 111s 143ms/step - loss: 0.3400 -
accuracy: 0.8812 - val_loss: 0.4378 - val_accuracy: 0.8727

Epoch 00050: val_accuracy did not improve from 0.87830

Epoch 51/100

781/781 [=====] - 112s 144ms/step - loss: 0.3305 -
accuracy: 0.8839 - val_loss: 0.4342 - val_accuracy: 0.8747

Epoch 00051: val_accuracy did not improve from 0.87830

Epoch 52/100

781/781 [=====] - 112s 143ms/step - loss: 0.3309 -
accuracy: 0.8839 - val_loss: 0.4376 - val_accuracy: 0.8740

Epoch 00052: val_accuracy did not improve from 0.87830

Epoch 53/100

781/781 [=====] - 112s 143ms/step - loss: 0.3282 -
accuracy: 0.8860 - val_loss: 0.4272 - val_accuracy: 0.8764

Epoch 00053: val_accuracy did not improve from 0.87830

Epoch 54/100

781/781 [=====] - 112s 143ms/step - loss: 0.3270 -
accuracy: 0.8854 - val_loss: 0.4354 - val_accuracy: 0.8747

Epoch 00054: val_accuracy did not improve from 0.87830

Epoch 55/100

781/781 [=====] - 112s 144ms/step - loss: 0.3290 -
accuracy: 0.8861 - val_loss: 0.4283 - val_accuracy: 0.8758

Epoch 00055: val_accuracy did not improve from 0.87830

Epoch 56/100

781/781 [=====] - 113s 144ms/step - loss: 0.3254 -
accuracy: 0.8854 - val_loss: 0.4209 - val_accuracy: 0.8769

Epoch 00056: val_accuracy did not improve from 0.87830

Epoch 57/100

781/781 [=====] - 113s 144ms/step - loss: 0.3272 -
accuracy: 0.8857 - val_loss: 0.4203 - val_accuracy: 0.8762

Epoch 00057: val_accuracy did not improve from 0.87830

Epoch 58/100

781/781 [=====] - 113s 144ms/step - loss: 0.3252 -
accuracy: 0.8862 - val_loss: 0.4251 - val_accuracy: 0.8750

Epoch 00058: val_accuracy did not improve from 0.87830

Epoch 59/100

781/781 [=====] - 112s 144ms/step - loss: 0.3272 -
accuracy: 0.8863 - val_loss: 0.4227 - val_accuracy: 0.8766

Epoch 00059: val_accuracy did not improve from 0.87830
Epoch 60/100
781/781 [=====] - 112s 143ms/step - loss: 0.3245 -
accuracy: 0.8865 - val_loss: 0.4110 - val_accuracy: 0.8793

Epoch 00060: val_accuracy improved from 0.87830 to 0.87930, saving model to
model4_weights.best.hdf5
Epoch 61/100
781/781 [=====] - 112s 143ms/step - loss: 0.3223 -
accuracy: 0.8853 - val_loss: 0.4218 - val_accuracy: 0.8763

Epoch 00061: val_accuracy did not improve from 0.87930
Epoch 62/100
781/781 [=====] - 112s 143ms/step - loss: 0.3241 -
accuracy: 0.8875 - val_loss: 0.4214 - val_accuracy: 0.8776

Epoch 00062: val_accuracy did not improve from 0.87930
Epoch 63/100
781/781 [=====] - 113s 144ms/step - loss: 0.3159 -
accuracy: 0.8897 - val_loss: 0.4203 - val_accuracy: 0.8772

Epoch 00063: val_accuracy did not improve from 0.87930
Epoch 64/100
781/781 [=====] - 113s 144ms/step - loss: 0.3170 -
accuracy: 0.8894 - val_loss: 0.4326 - val_accuracy: 0.8756

Epoch 00064: val_accuracy did not improve from 0.87930
Epoch 65/100
781/781 [=====] - 112s 143ms/step - loss: 0.3238 -
accuracy: 0.8863 - val_loss: 0.4230 - val_accuracy: 0.8780

Epoch 00065: val_accuracy did not improve from 0.87930
Epoch 66/100
781/781 [=====] - 112s 144ms/step - loss: 0.3202 -
accuracy: 0.8885 - val_loss: 0.4199 - val_accuracy: 0.8773

Epoch 00066: val_accuracy did not improve from 0.87930
Epoch 67/100
781/781 [=====] - 113s 145ms/step - loss: 0.3236 -
accuracy: 0.8879 - val_loss: 0.4251 - val_accuracy: 0.8763

Epoch 00067: val_accuracy did not improve from 0.87930
Epoch 68/100
781/781 [=====] - 112s 143ms/step - loss: 0.3223 -
accuracy: 0.8871 - val_loss: 0.4196 - val_accuracy: 0.8784

Epoch 00068: val_accuracy did not improve from 0.87930
Epoch 69/100

781/781 [=====] - 112s 143ms/step - loss: 0.3236 -
accuracy: 0.8870 - val_loss: 0.4162 - val_accuracy: 0.8787

Epoch 00069: val_accuracy did not improve from 0.87930

Epoch 70/100

781/781 [=====] - 112s 143ms/step - loss: 0.3185 -
accuracy: 0.8889 - val_loss: 0.4195 - val_accuracy: 0.8776

Epoch 00070: val_accuracy did not improve from 0.87930

Epoch 71/100

781/781 [=====] - 111s 143ms/step - loss: 0.3236 -
accuracy: 0.8878 - val_loss: 0.4258 - val_accuracy: 0.8761

Epoch 00071: val_accuracy did not improve from 0.87930

Epoch 72/100

781/781 [=====] - 113s 144ms/step - loss: 0.3148 -
accuracy: 0.8906 - val_loss: 0.4257 - val_accuracy: 0.8770

Epoch 00072: val_accuracy did not improve from 0.87930

Epoch 73/100

781/781 [=====] - 113s 145ms/step - loss: 0.3217 -
accuracy: 0.8883 - val_loss: 0.4205 - val_accuracy: 0.8779

Epoch 00073: val_accuracy did not improve from 0.87930

Epoch 74/100

781/781 [=====] - 114s 146ms/step - loss: 0.3227 -
accuracy: 0.8872 - val_loss: 0.4192 - val_accuracy: 0.8773

Epoch 00074: val_accuracy did not improve from 0.87930

Epoch 75/100

781/781 [=====] - 113s 144ms/step - loss: 0.3201 -
accuracy: 0.8880 - val_loss: 0.4136 - val_accuracy: 0.8799

Epoch 00075: val_accuracy improved from 0.87930 to 0.87990, saving model to
model4_weights.best.hdf5

Epoch 76/100

781/781 [=====] - 113s 144ms/step - loss: 0.3209 -
accuracy: 0.8875 - val_loss: 0.4124 - val_accuracy: 0.8793

Epoch 00076: val_accuracy did not improve from 0.87990

Epoch 77/100

781/781 [=====] - 113s 145ms/step - loss: 0.3179 -
accuracy: 0.8881 - val_loss: 0.4118 - val_accuracy: 0.8793

Epoch 00077: val_accuracy did not improve from 0.87990

Epoch 78/100

781/781 [=====] - 112s 143ms/step - loss: 0.3236 -
accuracy: 0.8860 - val_loss: 0.4153 - val_accuracy: 0.8780

```
Epoch 00078: val_accuracy did not improve from 0.87990
Epoch 79/100
781/781 [=====] - 113s 144ms/step - loss: 0.3209 -
accuracy: 0.8888 - val_loss: 0.4224 - val_accuracy: 0.8780
```

```
Epoch 00079: val_accuracy did not improve from 0.87990
Epoch 80/100
781/781 [=====] - 112s 143ms/step - loss: 0.3206 -
accuracy: 0.8882 - val_loss: 0.4182 - val_accuracy: 0.8777
```

```
Epoch 00080: val_accuracy did not improve from 0.87990
Epoch 81/100
781/781 [=====] - 112s 144ms/step - loss: 0.3177 -
accuracy: 0.8892 - val_loss: 0.4137 - val_accuracy: 0.8793
```

```
Epoch 00081: val_accuracy did not improve from 0.87990
Epoch 82/100
781/781 [=====] - 112s 143ms/step - loss: 0.3198 -
accuracy: 0.8883 - val_loss: 0.4240 - val_accuracy: 0.8772
```

```
Epoch 00082: val_accuracy did not improve from 0.87990
```

```
[ ]: <tensorflow.python.keras.callbacks.History at 0x7f01140b5b90>
```

1.3.6 2.6 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 = 0.2
```

```
[ ]: 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)

```

```

[ ]: 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 (BatchNorma (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]

```

dropout (Dropout)	(None, 32, 32, 25)	0	conv2d_2[0][0]
concatenate (Concatenate)	(None, 32, 32, 61)	0	conv2d[0][0] dropout[0][0]
batch_normalization_2 (BatchNor concatenate[0][0])	(None, 32, 32, 61)	244	
activation_2 (Activation) batch_normalization_2[0][0]	(None, 32, 32, 61)	0	
conv2d_3 (Conv2D) activation_2[0][0]	(None, 32, 32, 25)	1525	
batch_normalization_3 (BatchNor	(None, 32, 32, 25)	100	conv2d_3[0][0]
activation_3 (Activation) batch_normalization_3[0][0]	(None, 32, 32, 25)	0	
conv2d_4 (Conv2D) activation_3[0][0]	(None, 32, 32, 25)	5625	
dropout_1 (Dropout)	(None, 32, 32, 25)	0	conv2d_4[0][0]
concatenate_1 (Concatenate) concatenate[0][0]	(None, 32, 32, 86)	0	dropout_1[0][0]
batch_normalization_4 (BatchNor concatenate_1[0][0])	(None, 32, 32, 86)	344	
activation_4 (Activation) batch_normalization_4[0][0]	(None, 32, 32, 86)	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]			

dropout_2 (Dropout)	(None, 32, 32, 25)	0	conv2d_6[0][0]

concatenate_2 (Concatenate)	(None, 32, 32, 111)	0	
concatenate_1[0][0]			
			dropout_2[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]			

dropout_3 (Dropout)	(None, 32, 32, 25)	0	conv2d_8[0][0]

```

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

dropout_3[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]

-----

dropout_4 (Dropout)              (None, 32, 32, 25)   0      conv2d_10[0][0]

-----

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

dropout_4[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]
-----
-----
dropout_5 (Dropout) (None, 32, 32, 25) 0 conv2d_12[0] [0]
-----
-----
concatenate_5 (Concatenate) (None, 32, 32, 186) 0
concatenate_4[0] [0]
dropout_5[0] [0]
-----
-----
batch_normalization_12 (BatchNo (None, 32, 32, 186) 744
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]
-----
-----
dropout_6 (Dropout) (None, 32, 32, 25) 0 conv2d_14[0] [0]
-----
-----
concatenate_6 (Concatenate) (None, 32, 32, 211) 0
concatenate_5[0] [0]

```

```

dropout_6[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]

-----
dropout_7 (Dropout) (None, 32, 32, 25) 0 conv2d_16[0][0]

-----
concatenate_7 (Concatenate) (None, 32, 32, 236) 0
concatenate_6[0][0]

dropout_7[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]

-----

dropout_8 (Dropout)            (None, 32, 32, 25)    0          conv2d_18[0][0]

-----

concatenate_8 (Concatenate)    (None, 32, 32, 261)   0
concatenate_7[0][0]

                                          dropout_8[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]

-----

dropout_9 (Dropout)            (None, 32, 32, 25)    0          conv2d_20[0][0]

-----

concatenate_9 (Concatenate)    (None, 32, 32, 286)   0
concatenate_8[0][0]

                                          dropout_9[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]
-----
dropout_10 (Dropout) (None, 32, 32, 25) 0 conv2d_22[0][0]
-----
concatenate_10 (Concatenate) (None, 32, 32, 311) 0
concatenate_9[0][0]
dropout_10[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]			
dropout_11 (Dropout)	(None, 32, 32, 25)	0	conv2d_24[0][0]
concatenate_11 (Concatenate)	(None, 32, 32, 336)	0	
concatenate_10[0][0]			
dropout_11[0][0]			
batch_normalization_24 (Batch Normalization)	(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]			
dropout_12 (Dropout)	(None, 32, 32, 25)	0	conv2d_25[0][0]
average_pooling2d (AveragePooling2D)	(None, 16, 16, 25)	0	
dropout_12[0][0]			
batch_normalization_25 (Batch Normalization)	(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 (Batch Normalization)	(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]			

dropout_13 (Dropout)	(None, 16, 16, 25)	0	conv2d_27[0][0]

concatenate_12 (Concatenate)	(None, 16, 16, 50)	0	
average_pooling2d[0][0]			
dropout_13[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 (BatchNo	(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]			

dropout_14 (Dropout)	(None, 16, 16, 25)	0	conv2d_29[0][0]

concatenate_13 (Concatenate)	(None, 16, 16, 75)	0	
concatenate_12[0][0]			
dropout_14[0][0]			

batch_normalization_29 (BatchNo	(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 (BatchNo (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]
-----
-----
dropout_15 (Dropout)           (None, 16, 16, 25)    0          conv2d_31[0][0]
-----
-----
concatenate_14 (Concatenate)    (None, 16, 16, 100)   0
concatenate_13[0][0]
dropout_15[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]
-----
-----
dropout_16 (Dropout)              (None, 16, 16, 25)    0          conv2d_33[0][0]
-----
-----
concatenate_15 (Concatenate)      (None, 16, 16, 125)   0
concatenate_14[0][0]
dropout_16[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]
-----
-----
dropout_17 (Dropout)              (None, 16, 16, 25)    0          conv2d_35[0][0]
-----
-----
concatenate_16 (Concatenate)      (None, 16, 16, 150)   0
concatenate_15[0][0]
dropout_17[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]			

dropout_18 (Dropout)	(None, 16, 16, 25)	0	conv2d_37[0][0]

concatenate_17 (Concatenate)	(None, 16, 16, 175)	0	
concatenate_16[0][0]			
dropout_18[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]			

```

-----
-----
dropout_19 (Dropout)          (None, 16, 16, 25)    0          conv2d_39[0][0]
-----
-----
concatenate_18 (Concatenate)  (None, 16, 16, 200)   0
concatenate_17[0][0]
dropout_19[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]
-----
-----
dropout_20 (Dropout)          (None, 16, 16, 25)    0          conv2d_41[0][0]
-----
-----
concatenate_19 (Concatenate)  (None, 16, 16, 225)   0
concatenate_18[0][0]
dropout_20[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]
-----

-----
dropout_21 (Dropout)              (None, 16, 16, 25)    0          conv2d_43[0][0]
-----

-----
concatenate_20 (Concatenate)      (None, 16, 16, 250)   0
concatenate_19[0][0]
dropout_21[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]
-----

-----
dropout_22 (Dropout)              (None, 16, 16, 25)    0          conv2d_45[0][0]

```

```

-----
concatenate_21 (Concatenate)      (None, 16, 16, 275)  0
concatenate_20[0][0]
dropout_22[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]
-----

dropout_23 (Dropout)                (None, 16, 16, 25)   0          conv2d_47[0][0]
-----

concatenate_22 (Concatenate)      (None, 16, 16, 300)  0
concatenate_21[0][0]
dropout_23[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]
-----

dropout_24 (Dropout) (None, 16, 16, 25) 0 conv2d_49[0] [0]
-----

concatenate_23 (Concatenate) (None, 16, 16, 325) 0
concatenate_22[0] [0]
dropout_24[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]
-----

dropout_25 (Dropout) (None, 16, 16, 25) 0 conv2d_50[0] [0]
-----

average_pooling2d_1 (AveragePoo (None, 8, 8, 25) 0
dropout_25[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]			

dropout_26 (Dropout)	(None, 8, 8, 25)	0	conv2d_52[0][0]

concatenate_24 (Concatenate)	(None, 8, 8, 50)	0	
average_pooling2d_1[0][0]			
dropout_26[0][0]			

batch_normalization_52 (BatchNo	(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 (BatchNo	(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]			

dropout_27 (Dropout)	(None, 8, 8, 25)	0	conv2d_54[0][0]

```

-----
concatenate_25 (Concatenate)      (None, 8, 8, 75)      0
concatenate_24[0][0]
dropout_27[0][0]
-----

batch_normalization_54 (BatchNo (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]
-----

dropout_28 (Dropout)                (None, 8, 8, 25)      0      conv2d_56[0][0]
-----

concatenate_26 (Concatenate)      (None, 8, 8, 100)     0
concatenate_25[0][0]
dropout_28[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]			

dropout_29 (Dropout)	(None, 8, 8, 25)	0	conv2d_58[0][0]

concatenate_27 (Concatenate)	(None, 8, 8, 125)	0	
concatenate_26[0][0]			
dropout_29[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]			

dropout_30 (Dropout)	(None, 8, 8, 25)	0	conv2d_60[0][0]

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

dropout_30[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]

dropout_31 (Dropout) (None, 8, 8, 25) 0 conv2d_62[0][0]

concatenate_29 (Concatenate) (None, 8, 8, 175) 0
concatenate_28[0][0]
dropout_31[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]

-----

dropout_32 (Dropout)           (None, 8, 8, 25)      0          conv2d_64[0][0]

-----

concatenate_30 (Concatenate)   (None, 8, 8, 200)     0
concatenate_29[0][0]
dropout_32[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]

-----

dropout_33 (Dropout)           (None, 8, 8, 25)      0          conv2d_66[0][0]

-----

concatenate_31 (Concatenate)   (None, 8, 8, 225)     0
concatenate_30[0][0]
dropout_33[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]			

dropout_34 (Dropout)	(None, 8, 8, 25)	0	conv2d_68[0][0]

concatenate_32 (Concatenate)	(None, 8, 8, 250)	0	
concatenate_31[0][0]			
dropout_34[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]			
dropout_35 (Dropout)	(None, 8, 8, 25)	0	conv2d_70[0][0]
concatenate_33 (Concatenate)	(None, 8, 8, 275)	0	
concatenate_32[0][0]			
dropout_35[0][0]			
batch_normalization_70 (Batch Normalization)	(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 (Batch Normalization)	(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]			
dropout_36 (Dropout)	(None, 8, 8, 25)	0	conv2d_72[0][0]
concatenate_34 (Concatenate)	(None, 8, 8, 300)	0	
concatenate_33[0][0]			
dropout_36[0][0]			
batch_normalization_72 (Batch Normalization)	(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]			

dropout_37 (Dropout)	(None, 8, 8, 25)	0	conv2d_74[0][0]

concatenate_35 (Concatenate)	(None, 8, 8, 325)	0	
concatenate_34[0][0]			
dropout_37[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]			

dropout_38 (Dropout)	(None, 8, 8, 25)	0	conv2d_75[0][0]

average_pooling2d_2 (AveragePoo	(None, 4, 4, 25)	0	
dropout_38[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]

dropout_39 (Dropout) (None, 4, 4, 25) 0 conv2d_77[0][0]

concatenate_36 (Concatenate) (None, 4, 4, 50) 0
average_pooling2d_2[0][0]
dropout_39[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) activation_78[0][0]	(None, 4, 4, 25)	5625	

dropout_40 (Dropout)	(None, 4, 4, 25)	0	conv2d_79[0][0]

concatenate_37 (Concatenate) concatenate_36[0][0] dropout_40[0][0]	(None, 4, 4, 75)	0	

batch_normalization_79 (Batch Normalization) concatenate_37[0][0]	(None, 4, 4, 75)	300	

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

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

batch_normalization_80 (Batch Normalization) conv2d_80[0][0]	(None, 4, 4, 25)	100	conv2d_80[0][0]

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

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

dropout_41 (Dropout)	(None, 4, 4, 25)	0	conv2d_81[0][0]

concatenate_38 (Concatenate) concatenate_37[0][0] dropout_41[0][0]	(None, 4, 4, 100)	0	

batch_normalization_81 (Batch Normalization) concatenate_38[0][0]	(None, 4, 4, 100)	400	

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]			

dropout_42 (Dropout)	(None, 4, 4, 25)	0	conv2d_83[0][0]

concatenate_39 (Concatenate)	(None, 4, 4, 125)	0	
concatenate_38[0][0]			
dropout_42[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]			

dropout_43 (Dropout)	(None, 4, 4, 25)	0	conv2d_85[0][0]
concatenate_40 (Concatenate)	(None, 4, 4, 150)	0	
concatenate_39[0][0]			
dropout_43[0][0]			
batch_normalization_85 (Batch Normalization)	(None, 4, 4, 150)	600	
concatenate_40[0][0]			
activation_85 (Activation)	(None, 4, 4, 150)	0	
batch_normalization_85[0][0]			
conv2d_86 (Conv2D)	(None, 4, 4, 25)	3750	
activation_85[0][0]			
batch_normalization_86 (Batch Normalization)	(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]			
dropout_44 (Dropout)	(None, 4, 4, 25)	0	conv2d_87[0][0]
concatenate_41 (Concatenate)	(None, 4, 4, 175)	0	
concatenate_40[0][0]			
dropout_44[0][0]			
batch_normalization_87 (Batch Normalization)	(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]			
dropout_45 (Dropout)	(None, 4, 4, 25)	0	conv2d_89[0][0]
concatenate_42 (Concatenate)	(None, 4, 4, 200)	0	
concatenate_41[0][0]			
dropout_45[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]			
dropout_46 (Dropout)	(None, 4, 4, 25)	0	conv2d_91[0][0]

```

-----
concatenate_43 (Concatenate)      (None, 4, 4, 225)      0
concatenate_42[0][0]
dropout_46[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]
-----

dropout_47 (Dropout)                (None, 4, 4, 25)       0      conv2d_93[0][0]
-----

concatenate_44 (Concatenate)      (None, 4, 4, 250)      0
concatenate_43[0][0]
dropout_47[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]			
dropout_48 (Dropout)	(None, 4, 4, 25)	0	conv2d_95[0] [0]
concatenate_45 (Concatenate)	(None, 4, 4, 275)	0	
concatenate_44[0] [0]			
dropout_48[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]			
dropout_49 (Dropout)	(None, 4, 4, 25)	0	conv2d_97[0] [0]
concatenate_46 (Concatenate)	(None, 4, 4, 300)	0	

```

concatenate_45[0][0]
dropout_49[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]
-----
dropout_50 (Dropout)            (None, 4, 4, 25)      0      conv2d_99[0][0]
-----
concatenate_47 (Concatenate)    (None, 4, 4, 325)      0
concatenate_46[0][0]
dropout_50[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 (AveragePoo (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

```
-----  
-----
```

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

```
406
```

```
[ ]: 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 [=====] - 168s 163ms/step - loss: 1.9539 -
accuracy: 0.2869 - val_loss: 1.8601 - val_accuracy: 0.3451

Epoch 00001: val_accuracy improved from -inf to 0.34510, saving model to
model5_weights.best.hdf5

Epoch 2/100

781/781 [=====] - 125s 161ms/step - loss: 1.5744 -
accuracy: 0.4215 - val_loss: 2.7340 - val_accuracy: 0.2778

Epoch 00002: val_accuracy did not improve from 0.34510

Epoch 3/100

781/781 [=====] - 126s 161ms/step - loss: 1.3330 -
accuracy: 0.5172 - val_loss: 2.8528 - val_accuracy: 0.3612

Epoch 00003: val_accuracy improved from 0.34510 to 0.36120, saving model to
model5_weights.best.hdf5

Epoch 4/100

781/781 [=====] - 126s 161ms/step - loss: 1.1693 -
accuracy: 0.5798 - val_loss: 1.3349 - val_accuracy: 0.5871

Epoch 00004: val_accuracy improved from 0.36120 to 0.58710, saving model to

```

model5_weights.best.hdf5
Epoch 5/100
781/781 [=====] - 126s 161ms/step - loss: 1.0555 -
accuracy: 0.6220 - val_loss: 1.4948 - val_accuracy: 0.5794

Epoch 00005: val_accuracy did not improve from 0.58710
Epoch 6/100
781/781 [=====] - 125s 160ms/step - loss: 0.9628 -
accuracy: 0.6577 - val_loss: 3.3519 - val_accuracy: 0.3629

Epoch 00006: val_accuracy did not improve from 0.58710
Epoch 7/100
781/781 [=====] - 125s 160ms/step - loss: 0.8920 -
accuracy: 0.6870 - val_loss: 1.3352 - val_accuracy: 0.6218

Epoch 00007: val_accuracy improved from 0.58710 to 0.62180, saving model to
model5_weights.best.hdf5
Epoch 8/100
781/781 [=====] - 125s 160ms/step - loss: 0.8268 -
accuracy: 0.7100 - val_loss: 1.1063 - val_accuracy: 0.6978

Epoch 00008: val_accuracy improved from 0.62180 to 0.69780, saving model to
model5_weights.best.hdf5
Epoch 9/100
781/781 [=====] - 125s 160ms/step - loss: 0.7805 -
accuracy: 0.7263 - val_loss: 1.5754 - val_accuracy: 0.5854

Epoch 00009: val_accuracy did not improve from 0.69780
Epoch 10/100
781/781 [=====] - 125s 160ms/step - loss: 0.7375 -
accuracy: 0.7439 - val_loss: 1.3145 - val_accuracy: 0.6552

Epoch 00010: val_accuracy did not improve from 0.69780
Epoch 11/100
781/781 [=====] - 124s 159ms/step - loss: 0.7026 -
accuracy: 0.7547 - val_loss: 1.6132 - val_accuracy: 0.6050

Epoch 00011: val_accuracy did not improve from 0.69780
Epoch 12/100
781/781 [=====] - 124s 159ms/step - loss: 0.6798 -
accuracy: 0.7638 - val_loss: 0.6869 - val_accuracy: 0.7797

Epoch 00012: val_accuracy improved from 0.69780 to 0.77970, saving model to
model5_weights.best.hdf5
Epoch 13/100
781/781 [=====] - 125s 160ms/step - loss: 0.6541 -
accuracy: 0.7721 - val_loss: 1.0733 - val_accuracy: 0.6972

```

Epoch 00013: val_accuracy did not improve from 0.77970
Epoch 14/100
781/781 [=====] - 125s 160ms/step - loss: 0.6308 -
accuracy: 0.7798 - val_loss: 1.0946 - val_accuracy: 0.7073

Epoch 00014: val_accuracy did not improve from 0.77970
Epoch 15/100
781/781 [=====] - 124s 159ms/step - loss: 0.6116 -
accuracy: 0.7886 - val_loss: 0.6905 - val_accuracy: 0.7832

Epoch 00015: val_accuracy improved from 0.77970 to 0.78320, saving model to
model5_weights.best.hdf5
Epoch 16/100
781/781 [=====] - 124s 159ms/step - loss: 0.5951 -
accuracy: 0.7928 - val_loss: 1.1738 - val_accuracy: 0.6919

Epoch 00016: val_accuracy did not improve from 0.78320
Epoch 17/100
781/781 [=====] - 125s 160ms/step - loss: 0.5798 -
accuracy: 0.7982 - val_loss: 0.7496 - val_accuracy: 0.7758

Epoch 00017: val_accuracy did not improve from 0.78320
Epoch 18/100
781/781 [=====] - 125s 160ms/step - loss: 0.5659 -
accuracy: 0.8046 - val_loss: 0.6886 - val_accuracy: 0.7975

Epoch 00018: val_accuracy improved from 0.78320 to 0.79750, saving model to
model5_weights.best.hdf5
Epoch 19/100
781/781 [=====] - 125s 160ms/step - loss: 0.5497 -
accuracy: 0.8112 - val_loss: 0.8661 - val_accuracy: 0.7550

Epoch 00019: val_accuracy did not improve from 0.79750
Epoch 20/100
781/781 [=====] - 124s 159ms/step - loss: 0.5350 -
accuracy: 0.8157 - val_loss: 1.1197 - val_accuracy: 0.7082

Epoch 00020: val_accuracy did not improve from 0.79750
Epoch 21/100
781/781 [=====] - 125s 160ms/step - loss: 0.5278 -
accuracy: 0.8173 - val_loss: 0.6811 - val_accuracy: 0.7966

Epoch 00021: val_accuracy did not improve from 0.79750
Epoch 22/100
781/781 [=====] - 125s 159ms/step - loss: 0.5154 -
accuracy: 0.8207 - val_loss: 0.7040 - val_accuracy: 0.7911

Epoch 00022: val_accuracy did not improve from 0.79750

Epoch 23/100
781/781 [=====] - 125s 160ms/step - loss: 0.5016 - accuracy: 0.8256 - val_loss: 0.6738 - val_accuracy: 0.8055

Epoch 00023: val_accuracy improved from 0.79750 to 0.80550, saving model to model5_weights.best.hdf5

Epoch 24/100
781/781 [=====] - 124s 159ms/step - loss: 0.4982 - accuracy: 0.8278 - val_loss: 1.3099 - val_accuracy: 0.7262

Epoch 00024: val_accuracy did not improve from 0.80550

Epoch 25/100
781/781 [=====] - 125s 160ms/step - loss: 0.4934 - accuracy: 0.8297 - val_loss: 0.5844 - val_accuracy: 0.8245

Epoch 00025: val_accuracy improved from 0.80550 to 0.82450, saving model to model5_weights.best.hdf5

Epoch 26/100
781/781 [=====] - 125s 159ms/step - loss: 0.4094 - accuracy: 0.8584 - val_loss: 0.4688 - val_accuracy: 0.8598

Epoch 00026: val_accuracy improved from 0.82450 to 0.85980, saving model to model5_weights.best.hdf5

Epoch 27/100
781/781 [=====] - 124s 159ms/step - loss: 0.3893 - accuracy: 0.8632 - val_loss: 0.4260 - val_accuracy: 0.8694

Epoch 00027: val_accuracy improved from 0.85980 to 0.86940, saving model to model5_weights.best.hdf5

Epoch 28/100
781/781 [=====] - 125s 160ms/step - loss: 0.3821 - accuracy: 0.8683 - val_loss: 0.4430 - val_accuracy: 0.8650

Epoch 00028: val_accuracy did not improve from 0.86940

Epoch 29/100
781/781 [=====] - 124s 159ms/step - loss: 0.3752 - accuracy: 0.8703 - val_loss: 0.4483 - val_accuracy: 0.8640

Epoch 00029: val_accuracy did not improve from 0.86940

Epoch 30/100
781/781 [=====] - 125s 159ms/step - loss: 0.3675 - accuracy: 0.8719 - val_loss: 0.4330 - val_accuracy: 0.8685

Epoch 00030: val_accuracy did not improve from 0.86940

Epoch 31/100
781/781 [=====] - 124s 159ms/step - loss: 0.3609 - accuracy: 0.8746 - val_loss: 0.4285 - val_accuracy: 0.8714

Epoch 00031: val_accuracy improved from 0.86940 to 0.87140, saving model to model5_weights.best.hdf5

Epoch 32/100

781/781 [=====] - 124s 159ms/step - loss: 0.3572 - accuracy: 0.8765 - val_loss: 0.4048 - val_accuracy: 0.8776

Epoch 00032: val_accuracy improved from 0.87140 to 0.87760, saving model to model5_weights.best.hdf5

Epoch 33/100

781/781 [=====] - 124s 159ms/step - loss: 0.3538 - accuracy: 0.8764 - val_loss: 0.4468 - val_accuracy: 0.8666

Epoch 00033: val_accuracy did not improve from 0.87760

Epoch 34/100

781/781 [=====] - 124s 159ms/step - loss: 0.3451 - accuracy: 0.8806 - val_loss: 0.4183 - val_accuracy: 0.8751

Epoch 00034: val_accuracy did not improve from 0.87760

Epoch 35/100

781/781 [=====] - 125s 160ms/step - loss: 0.3463 - accuracy: 0.8809 - val_loss: 0.4134 - val_accuracy: 0.8743

Epoch 00035: val_accuracy did not improve from 0.87760

Epoch 36/100

781/781 [=====] - 125s 160ms/step - loss: 0.3373 - accuracy: 0.8829 - val_loss: 0.4375 - val_accuracy: 0.8720

Epoch 00036: val_accuracy did not improve from 0.87760

Epoch 37/100

781/781 [=====] - 125s 160ms/step - loss: 0.3428 - accuracy: 0.8805 - val_loss: 0.4091 - val_accuracy: 0.8754

Epoch 00037: val_accuracy did not improve from 0.87760

Epoch 38/100

781/781 [=====] - 125s 160ms/step - loss: 0.3376 - accuracy: 0.8825 - val_loss: 0.3819 - val_accuracy: 0.8815

Epoch 00038: val_accuracy improved from 0.87760 to 0.88150, saving model to model5_weights.best.hdf5

Epoch 39/100

781/781 [=====] - 125s 160ms/step - loss: 0.3375 - accuracy: 0.8821 - val_loss: 0.4721 - val_accuracy: 0.8637

Epoch 00039: val_accuracy did not improve from 0.88150

Epoch 40/100

781/781 [=====] - 125s 160ms/step - loss: 0.3339 - accuracy: 0.8854 - val_loss: 0.4030 - val_accuracy: 0.8788

Epoch 00040: val_accuracy did not improve from 0.88150
Epoch 41/100
781/781 [=====] - 125s 160ms/step - loss: 0.3334 -
accuracy: 0.8832 - val_loss: 0.4245 - val_accuracy: 0.8726

Epoch 00041: val_accuracy did not improve from 0.88150
Epoch 42/100
781/781 [=====] - 125s 160ms/step - loss: 0.3289 -
accuracy: 0.8873 - val_loss: 0.4051 - val_accuracy: 0.8773

Epoch 00042: val_accuracy did not improve from 0.88150
Epoch 43/100
781/781 [=====] - 125s 160ms/step - loss: 0.3271 -
accuracy: 0.8861 - val_loss: 0.4003 - val_accuracy: 0.8798

Epoch 00043: val_accuracy did not improve from 0.88150
Epoch 44/100
781/781 [=====] - 125s 160ms/step - loss: 0.3262 -
accuracy: 0.8862 - val_loss: 0.4186 - val_accuracy: 0.8755

Epoch 00044: val_accuracy did not improve from 0.88150
Epoch 45/100
781/781 [=====] - 124s 159ms/step - loss: 0.3253 -
accuracy: 0.8862 - val_loss: 0.4137 - val_accuracy: 0.8778

Epoch 00045: val_accuracy did not improve from 0.88150
Epoch 46/100
781/781 [=====] - 124s 159ms/step - loss: 0.3198 -
accuracy: 0.8879 - val_loss: 0.3992 - val_accuracy: 0.8838

Epoch 00046: val_accuracy improved from 0.88150 to 0.88380, saving model to
model5_weights.best.hdf5
Epoch 47/100
781/781 [=====] - 125s 160ms/step - loss: 0.3146 -
accuracy: 0.8895 - val_loss: 0.3991 - val_accuracy: 0.8800

Epoch 00047: val_accuracy did not improve from 0.88380
Epoch 48/100
781/781 [=====] - 125s 159ms/step - loss: 0.3127 -
accuracy: 0.8908 - val_loss: 0.3981 - val_accuracy: 0.8810

Epoch 00048: val_accuracy did not improve from 0.88380
Epoch 49/100
781/781 [=====] - 125s 160ms/step - loss: 0.3180 -
accuracy: 0.8898 - val_loss: 0.4189 - val_accuracy: 0.8791

Epoch 00049: val_accuracy did not improve from 0.88380
Epoch 50/100

781/781 [=====] - 125s 160ms/step - loss: 0.3189 -
accuracy: 0.8886 - val_loss: 0.4156 - val_accuracy: 0.8764

Epoch 00050: val_accuracy did not improve from 0.88380

Epoch 51/100

781/781 [=====] - 124s 159ms/step - loss: 0.3109 -
accuracy: 0.8916 - val_loss: 0.3908 - val_accuracy: 0.8856

Epoch 00051: val_accuracy improved from 0.88380 to 0.88560, saving model to
model5_weights.best.hdf5

Epoch 52/100

781/781 [=====] - 124s 159ms/step - loss: 0.3118 -
accuracy: 0.8902 - val_loss: 0.3922 - val_accuracy: 0.8845

Epoch 00052: val_accuracy did not improve from 0.88560

Epoch 53/100

781/781 [=====] - 124s 158ms/step - loss: 0.3063 -
accuracy: 0.8922 - val_loss: 0.3903 - val_accuracy: 0.8857

Epoch 00053: val_accuracy improved from 0.88560 to 0.88570, saving model to
model5_weights.best.hdf5

Epoch 54/100

781/781 [=====] - 124s 159ms/step - loss: 0.3006 -
accuracy: 0.8950 - val_loss: 0.3903 - val_accuracy: 0.8863

Epoch 00054: val_accuracy improved from 0.88570 to 0.88630, saving model to
model5_weights.best.hdf5

Epoch 55/100

781/781 [=====] - 124s 158ms/step - loss: 0.3004 -
accuracy: 0.8947 - val_loss: 0.3926 - val_accuracy: 0.8850

Epoch 00055: val_accuracy did not improve from 0.88630

Epoch 56/100

781/781 [=====] - 124s 159ms/step - loss: 0.3010 -
accuracy: 0.8934 - val_loss: 0.3814 - val_accuracy: 0.8881

Epoch 00056: val_accuracy improved from 0.88630 to 0.88810, saving model to
model5_weights.best.hdf5

Epoch 57/100

781/781 [=====] - 124s 159ms/step - loss: 0.2998 -
accuracy: 0.8943 - val_loss: 0.3942 - val_accuracy: 0.8854

Epoch 00057: val_accuracy did not improve from 0.88810

Epoch 58/100

781/781 [=====] - 124s 159ms/step - loss: 0.3010 -
accuracy: 0.8955 - val_loss: 0.3848 - val_accuracy: 0.8866

Epoch 00058: val_accuracy did not improve from 0.88810

Epoch 59/100
781/781 [=====] - 124s 159ms/step - loss: 0.3007 -
accuracy: 0.8941 - val_loss: 0.3823 - val_accuracy: 0.8879

Epoch 00059: val_accuracy did not improve from 0.88810
Epoch 60/100
781/781 [=====] - 124s 158ms/step - loss: 0.3023 -
accuracy: 0.8946 - val_loss: 0.3821 - val_accuracy: 0.8867

Epoch 00060: val_accuracy did not improve from 0.88810
Epoch 61/100
781/781 [=====] - 124s 159ms/step - loss: 0.2990 -
accuracy: 0.8958 - val_loss: 0.3833 - val_accuracy: 0.8874

Epoch 00061: val_accuracy did not improve from 0.88810
Epoch 62/100
781/781 [=====] - 125s 159ms/step - loss: 0.2999 -
accuracy: 0.8951 - val_loss: 0.3818 - val_accuracy: 0.8864

Epoch 00062: val_accuracy did not improve from 0.88810
Epoch 63/100
781/781 [=====] - 124s 159ms/step - loss: 0.3023 -
accuracy: 0.8949 - val_loss: 0.3909 - val_accuracy: 0.8843

Epoch 00063: val_accuracy did not improve from 0.88810
Epoch 64/100
781/781 [=====] - 124s 158ms/step - loss: 0.3020 -
accuracy: 0.8942 - val_loss: 0.3879 - val_accuracy: 0.8865

Epoch 00064: val_accuracy did not improve from 0.88810
Epoch 65/100
781/781 [=====] - 124s 159ms/step - loss: 0.2994 -
accuracy: 0.8946 - val_loss: 0.3842 - val_accuracy: 0.8877

Epoch 00065: val_accuracy did not improve from 0.88810
Epoch 66/100
781/781 [=====] - 124s 159ms/step - loss: 0.3017 -
accuracy: 0.8944 - val_loss: 0.3793 - val_accuracy: 0.8889

Epoch 00066: val_accuracy improved from 0.88810 to 0.88890, saving model to
model5_weights.best.hdf5
Epoch 67/100
781/781 [=====] - 124s 159ms/step - loss: 0.2956 -
accuracy: 0.8963 - val_loss: 0.3826 - val_accuracy: 0.8870

Epoch 00067: val_accuracy did not improve from 0.88890
Epoch 68/100
781/781 [=====] - 124s 159ms/step - loss: 0.3012 -

accuracy: 0.8942 - val_loss: 0.3844 - val_accuracy: 0.8867

Epoch 00068: val_accuracy did not improve from 0.88890

Epoch 69/100

781/781 [=====] - 124s 159ms/step - loss: 0.2973 -
accuracy: 0.8974 - val_loss: 0.3888 - val_accuracy: 0.8860

Epoch 00069: val_accuracy did not improve from 0.88890

Epoch 70/100

781/781 [=====] - 124s 159ms/step - loss: 0.2982 -
accuracy: 0.8953 - val_loss: 0.3884 - val_accuracy: 0.8866

Epoch 00070: val_accuracy did not improve from 0.88890

Epoch 71/100

781/781 [=====] - 124s 159ms/step - loss: 0.2997 -
accuracy: 0.8946 - val_loss: 0.3878 - val_accuracy: 0.8866

Epoch 00071: val_accuracy did not improve from 0.88890

Epoch 72/100

781/781 [=====] - 125s 160ms/step - loss: 0.2992 -
accuracy: 0.8974 - val_loss: 0.3867 - val_accuracy: 0.8876

Epoch 00072: val_accuracy did not improve from 0.88890

Epoch 73/100

781/781 [=====] - 124s 159ms/step - loss: 0.2951 -
accuracy: 0.8963 - val_loss: 0.3813 - val_accuracy: 0.8890

Epoch 00073: val_accuracy improved from 0.88890 to 0.88900, saving model to
model5_weights.best.hdf5

Epoch 74/100

781/781 [=====] - 125s 159ms/step - loss: 0.2981 -
accuracy: 0.8955 - val_loss: 0.3842 - val_accuracy: 0.8868

Epoch 00074: val_accuracy did not improve from 0.88900

Epoch 75/100

781/781 [=====] - 125s 159ms/step - loss: 0.2970 -
accuracy: 0.8969 - val_loss: 0.3855 - val_accuracy: 0.8867

Epoch 00075: val_accuracy did not improve from 0.88900

Epoch 76/100

781/781 [=====] - 125s 159ms/step - loss: 0.2939 -
accuracy: 0.8969 - val_loss: 0.3881 - val_accuracy: 0.8884

Epoch 00076: val_accuracy did not improve from 0.88900

Epoch 77/100

781/781 [=====] - 124s 159ms/step - loss: 0.2957 -
accuracy: 0.8960 - val_loss: 0.3855 - val_accuracy: 0.8890

Epoch 00077: val_accuracy did not improve from 0.88900
Epoch 78/100
781/781 [=====] - 124s 159ms/step - loss: 0.2948 -
accuracy: 0.8967 - val_loss: 0.3896 - val_accuracy: 0.8878

Epoch 00078: val_accuracy did not improve from 0.88900
Epoch 79/100
781/781 [=====] - 124s 159ms/step - loss: 0.2965 -
accuracy: 0.8962 - val_loss: 0.3808 - val_accuracy: 0.8890

Epoch 00079: val_accuracy did not improve from 0.88900
Epoch 80/100
781/781 [=====] - 124s 159ms/step - loss: 0.2952 -
accuracy: 0.8980 - val_loss: 0.3862 - val_accuracy: 0.8879

Epoch 00080: val_accuracy did not improve from 0.88900
Epoch 81/100
781/781 [=====] - 124s 159ms/step - loss: 0.2949 -
accuracy: 0.8969 - val_loss: 0.3879 - val_accuracy: 0.8863

Epoch 00081: val_accuracy did not improve from 0.88900
Epoch 82/100
781/781 [=====] - 124s 159ms/step - loss: 0.2986 -
accuracy: 0.8959 - val_loss: 0.3815 - val_accuracy: 0.8891

Epoch 00082: val_accuracy improved from 0.88900 to 0.88910, saving model to
model5_weights.best.hdf5
Epoch 83/100
781/781 [=====] - 125s 160ms/step - loss: 0.2974 -
accuracy: 0.8958 - val_loss: 0.3891 - val_accuracy: 0.8873

Epoch 00083: val_accuracy did not improve from 0.88910
Epoch 84/100
781/781 [=====] - 124s 159ms/step - loss: 0.2946 -
accuracy: 0.8971 - val_loss: 0.3838 - val_accuracy: 0.8889

Epoch 00084: val_accuracy did not improve from 0.88910
Epoch 85/100
781/781 [=====] - 124s 159ms/step - loss: 0.2971 -
accuracy: 0.8969 - val_loss: 0.3807 - val_accuracy: 0.8887

Epoch 00085: val_accuracy did not improve from 0.88910
Epoch 86/100
781/781 [=====] - 125s 159ms/step - loss: 0.2908 -
accuracy: 0.8973 - val_loss: 0.3842 - val_accuracy: 0.8889

Epoch 00086: val_accuracy did not improve from 0.88910
Epoch 87/100

781/781 [=====] - 124s 159ms/step - loss: 0.2960 -
accuracy: 0.8954 - val_loss: 0.3871 - val_accuracy: 0.8878

Epoch 00087: val_accuracy did not improve from 0.88910

Epoch 88/100

781/781 [=====] - 125s 160ms/step - loss: 0.2933 -
accuracy: 0.8969 - val_loss: 0.3809 - val_accuracy: 0.8880

Epoch 00088: val_accuracy did not improve from 0.88910

Epoch 89/100

781/781 [=====] - 124s 159ms/step - loss: 0.2929 -
accuracy: 0.8979 - val_loss: 0.3799 - val_accuracy: 0.8892

Epoch 00089: val_accuracy improved from 0.88910 to 0.88920, saving model to
model5_weights.best.hdf5

Epoch 90/100

781/781 [=====] - 124s 159ms/step - loss: 0.2910 -
accuracy: 0.8989 - val_loss: 0.3807 - val_accuracy: 0.8908

Epoch 00090: val_accuracy improved from 0.88920 to 0.89080, saving model to
model5_weights.best.hdf5

Epoch 91/100

781/781 [=====] - 124s 159ms/step - loss: 0.2920 -
accuracy: 0.8975 - val_loss: 0.3795 - val_accuracy: 0.8902

Epoch 00091: val_accuracy did not improve from 0.89080

Epoch 92/100

781/781 [=====] - 124s 159ms/step - loss: 0.2921 -
accuracy: 0.8981 - val_loss: 0.3800 - val_accuracy: 0.8899

Epoch 00092: val_accuracy did not improve from 0.89080

Epoch 93/100

781/781 [=====] - 125s 160ms/step - loss: 0.2974 -
accuracy: 0.8960 - val_loss: 0.3842 - val_accuracy: 0.8883

Epoch 00093: val_accuracy did not improve from 0.89080

Epoch 94/100

781/781 [=====] - 124s 159ms/step - loss: 0.2936 -
accuracy: 0.8975 - val_loss: 0.3880 - val_accuracy: 0.8876

Epoch 00094: val_accuracy did not improve from 0.89080

Epoch 95/100

781/781 [=====] - 124s 159ms/step - loss: 0.2912 -
accuracy: 0.8979 - val_loss: 0.3797 - val_accuracy: 0.8896

Epoch 00095: val_accuracy did not improve from 0.89080

Epoch 96/100

781/781 [=====] - 124s 159ms/step - loss: 0.2941 -

accuracy: 0.8977 - val_loss: 0.3795 - val_accuracy: 0.8900

Epoch 00096: val_accuracy did not improve from 0.89080

```
[ ]: <tensorflow.python.keras.callbacks.History at 0x7f175e69a550>
```

```
[ ]: import pandas as pd
import numpy as np
training_log = pd.read_csv('/content/training_model5.log')
training_log.head(100)
```

```
[ ]:      epoch  accuracy      loss      lr  val_accuracy  val_loss
0         0  0.286867  1.953920  0.0100         0.3451  1.860124
1         1  0.421540  1.574372  0.0100         0.2778  2.734015
2         2  0.517222  1.333009  0.0100         0.3612  2.852820
3         3  0.579842  1.169290  0.0100         0.5871  1.334862
4         4  0.622016  1.055502  0.0100         0.5794  1.494813
..      ...
91        91  0.898130  0.292132  0.0001         0.8899  0.380016
92        92  0.896007  0.297427  0.0001         0.8883  0.384248
93        93  0.897529  0.293597  0.0001         0.8876  0.387959
94        94  0.897889  0.291213  0.0001         0.8896  0.379663
95        95  0.897749  0.294055  0.0001         0.8900  0.379485
```

[96 rows x 6 columns]

```
[ ]: 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

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

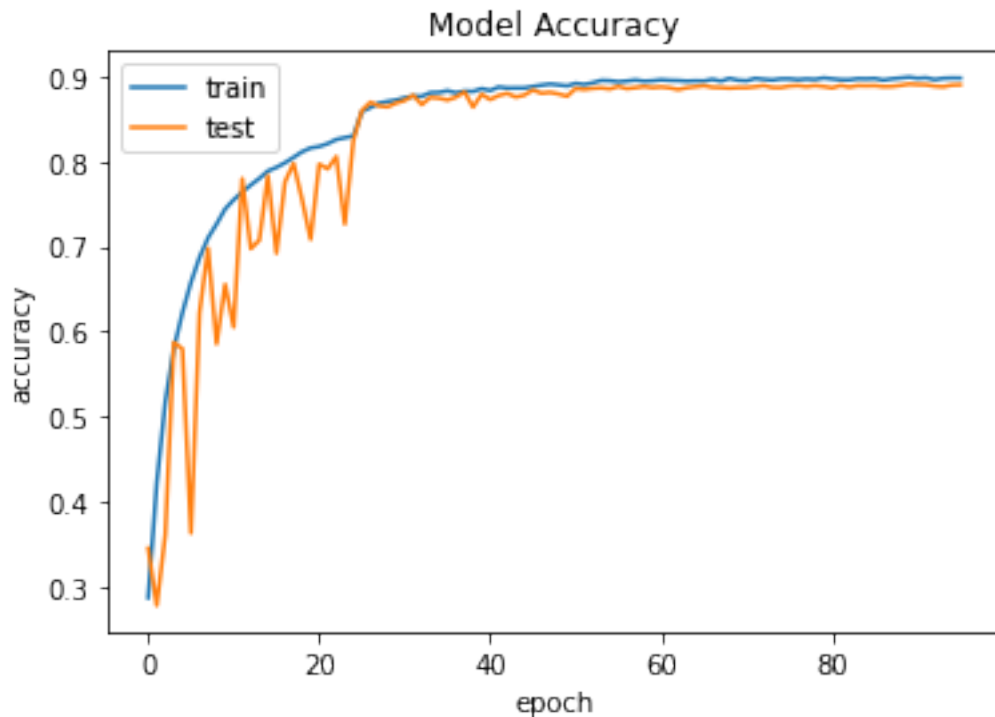
Test loss = 0.3807081878185272

Test accuracy = 0.8907999992370605

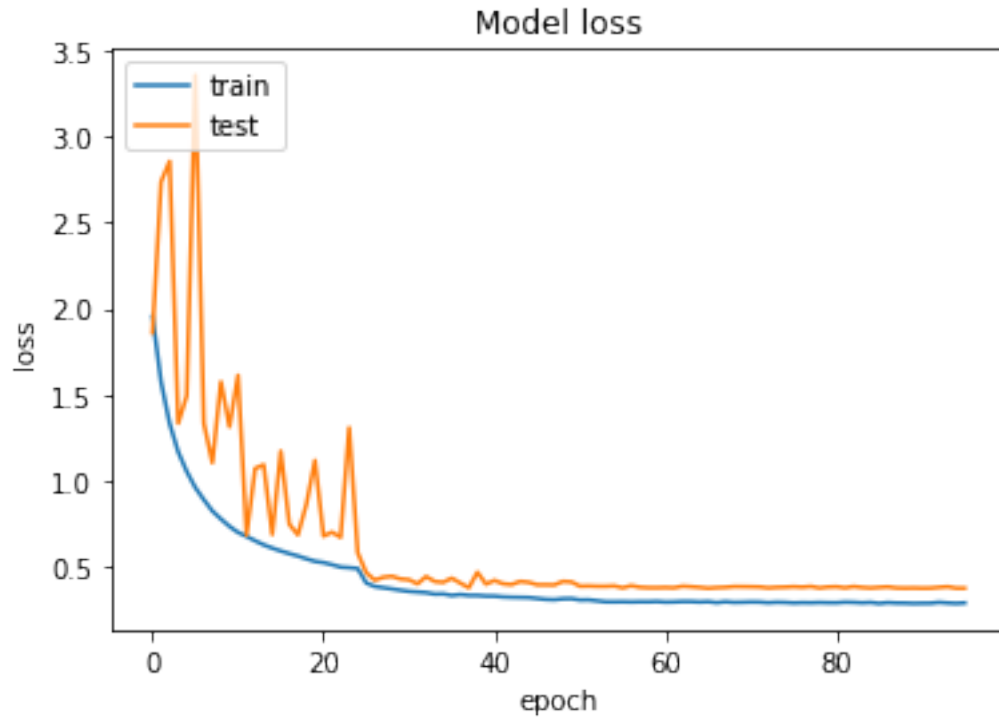
1.3.7 2.7 Plotting loss and accuracy of Model 5 above

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

```
[ ]: #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()
```



```
[ ]: #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

```
[6]: from prettytable import PrettyTable

x = PrettyTable()
x.field_names = ['Growth Rate', 'Compression', '# of Blocks', 'Test Accuracy']
x.add_row([24, 0.5, 12, 82.590])
x.add_row([32, 0.7, 12, 87.990])
x.add_row([36, 0.7, 12, 89.079])

print(x)
```

Growth Rate	Compression	# of Blocks	Test Accuracy
24	0.5	12	82.59
32	0.7	12	87.99
36	0.7	12	89.079

```
[7]: from prettytable import PrettyTable
```

```
x1 = PrettyTable()
x1.field_names = ['Growth Rate', 'Compression', '# of Blocks', 'Test Accuracy']
x1.add_row([24,0.5,12,82.590])
x1.add_row([32,0.7,12,87.990])
x1.add_row([36,0.7,12,89.079])

print(x1)
```

```
+-----+-----+-----+-----+
| Growth Rate | Compression | # of Blocks | Test Accuracy |
+-----+-----+-----+-----+
|      24      |      0.5      |      12      |      82.59      |
|      32      |      0.7      |      12      |      87.99      |
|      36      |      0.7      |      12      |      89.079     |
+-----+-----+-----+-----+
```

Summary:

1. I have used Keras callbacks to adjust the learning rate as per the performance of the model(ReduceLROnPlateau, LearningRate Sceduler).
2. Increase in Growth rate along with an increase with compression rate has led to improvement in test accuracy scores.

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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