image_classification

January 14, 2019

0.1 Project: Image Classification

In this project, I have classified images from the CIFAR-10 dataset (https://www.cs.toronto.edu/~kriz/cifar.html). The dataset consists of airplanes, dogs, cats, and other objects.

In this project I have used Keras a framework which supports convolutional neural network for classifying images form CIFAR-10 dataset. I have developed my onw model based on Keras and used it for training and testing.

I have used following steps in the project

- Download Keras
- Import all the modules
- Download the images from CIFAR-10 dataset
- Augment the images dataset by performing operations like rotation
- Define Keras model

Using TensorFlow backend.

• Train and test the dataset

I have run this on my laptop (Lenovo Legin Y720) which has 6 GB GPU.

0.2 Install Keras

0.3 Import all the modules

```
In [2]: import keras
from keras.datasets import cifar10
from keras.models import Model, Sequential
from keras.layers import Dense, Dropout, Flatten, Input, AveragePooling2D, merge, Actifrom keras.layers import Conv2D, MaxPooling2D, BatchNormalization
from keras.layers import Concatenate
from keras.optimizers import Adam
from keras.optimizers import SGD
```

0.4 Tensorflow configuration changes for GPU

0.5 Define Hyper parameters to be used

Following hyper parameters are used:

- batch_size Batch size
- num_classes Number of classes (10)
- epochs Number of epochs (250)
- 1 Number of layers
- num_filter Number of filters
- compression
- dropout_rate

```
In [4]: # Hyperparameters
    batch_size = 64
    num_classes = 10
    epochs = 250
    l = 22
    num_filter = 16
    compression = 0.5
    dropout_rate = 0.25
```

0.6 Download the dataset and perform one hot encoding

In this step, I have downloaded the cifar10 training and test data

Next, performed one-hot encoding on target labels train_y and test_y to get the resultant train_y, test_y which will be used for processing

0.7 Perform Image augmentation of images

- 1. Peform image augmentation by rotating 45 degrees for a randomly selected 1000 images and append the augmented images
- 2. Perform image augmentation by rotating 45 degrees for a randomly selected 1000 images and append the augmented images

```
In [6]: from keras.preprocessing.image import ImageDataGenerator
        import numpy as np
        # Rotate images by 45 degrees
        datagen1 = ImageDataGenerator(rotation_range=45)
        # fit parameters from data
        datagen1.fit(x_train)
        # Get 1000 augmented images and append to the training data
        for x_rotated_images_batch, y_rotated_images_batch in datagen1.flow(x_train, y_train,
            x_train = np.append(x_train, x_rotated_images_batch, axis=0)
            y_train = np.append(y_train, y_rotated_images_batch, axis=0)
            break
        # Horizontally flip image
        datagen2 = ImageDataGenerator(horizontal_flip=True)
        # fit parameters from data
        datagen2.fit(x_train)
        # Get 1000 augmented horizontally flipped images and append to the training data
        for x_flipped_images_batch, y_flipped_images_batch in datagen2.flow(x_train, y_train,
            x_train = np.append(x_train, x_flipped_images_batch, axis=0)
            y_train = np.append(y_train, y_flipped_images_batch, axis=0)
            break
```

0.8 Create building blocks for model

1. Create a method for add_denseblock of layers where each layer has following operations: perform Batch Normalization, Perform relu Activation Apply 3x3 convolution of a number

of filters num_filter*compression Add dropout All the layers are concated to form a dense block

- 2. Create a method add_transition which does the following:

 perform Batch Normalization, Perform relu Activation Apply 1x1 convolution of a number of filters num_filter*compression Add dropout Perform 2x2 Average Pooling
- 3. Create a method output_layer which does the following:

 perform Batch Normalization, Perform relu Activation Perform 2x2 Average Pooling Flatten

 Dense layer of 10 classes with softmax activation

```
In [7]: # Dense Block
        def add_denseblock(input, num_filter = 12, dropout_rate = 0.2):
            global compression
            temp = input
            for _ in range(1):
                BatchNorm = BatchNormalization()(temp)
                relu = Activation('relu')(BatchNorm)
                Conv2D_3_3 = Conv2D(int(num_filter*compression), (3,3), use_bias=False ,padding
                if dropout_rate>0:
                  Conv2D_3_3 = Dropout(dropout_rate)(Conv2D_3_3)
                concat = Concatenate(axis=-1)([temp,Conv2D_3_3])
                temp = concat
            return temp
In [8]: def add_transition(input, num_filter = 12, dropout_rate = 0.2):
            global compression
            BatchNorm = BatchNormalization()(input)
            relu = Activation('relu')(BatchNorm)
            Conv2D_BottleNeck = Conv2D(int(num_filter*compression), (1,1), use_bias=False ,pade
            if dropout_rate>0:
              Conv2D_BottleNeck = Dropout(dropout_rate)(Conv2D_BottleNeck)
            AvgPooling = AveragePooling2D(pool_size=(2,2))(Conv2D_BottleNeck)
            return AvgPooling
In [9]: def output_layer(input):
            global compression
            BatchNorm = BatchNormalization()(input)
            relu = Activation('relu')(BatchNorm)
            AvgPooling = AveragePooling2D(pool_size=(2,2))(relu)
            flat = Flatten()(AvgPooling)
            output = Dense(num_classes, activation='softmax')(flat)
            return output
```

0.9 Create Final Model

Apply building blocks of dense_block, transtion blocks and 3x3 convolution filters to create the final model

```
In [10]: input = Input(shape=(img height, img width, channel,))
         layer1_transition = Conv2D(num_filter, (3,3), use_bias=False ,padding='same')(input)
        layer2_block = add_denseblock(layer1_transition, num_filter, dropout_rate)
        layer2_transition = add_transition(layer2_block, num_filter, dropout_rate)
         skip_connection1 = layer2_transition
        layer3_block = add_denseblock(layer2_transition, num_filter, dropout_rate)
         layer3_transition = add_transition(layer3_block, num_filter, dropout_rate)
        layer4_block = add_denseblock(layer3_transition, num_filter, dropout_rate)
        layer4_transition = add_transition(layer4_block, num_filter, dropout_rate)
        layer5_skip_connection_block = add_denseblock(skip_connection1, num_filter, dropout_re
        layer5_skip_connection_block = Conv2D(int(num_filter*compression), (5,5), use_bias=Fal
        layer5 skip connection block = Conv2D(int(num filter*compression), (5,5), use bias=Fa
        layer5_skip_connection_block = Conv2D(int(num_filter*compression), (5,5), use_bias=Fa
        layer5_block = Concatenate(axis=-1)([layer5_skip_connection_block, layer4_transition]
        layer5_transition = add_transition(layer5_block, num_filter, dropout_rate)
        layer6_block = add_denseblock(layer5_transition, num_filter, dropout_rate)
         output = output_layer(layer6_block)
```

In	[11]:	model	= Model	<pre>(inputs=[input],</pre>	<pre>outputs=[output])</pre>
		model			

Layer (type)	Output	Shape	Param #	Connected to
input_1 (InputLayer)	(None,	32, 32, 3)) 0	
conv2d_1 (Conv2D)	(None,	32, 32, 16	6) 432	input_1[0][0]
batch_normalization_1 (BatchNor	(None,	32, 32, 16	6) 64	conv2d_1[0][0]
activation_1 (Activation)	(None,	32, 32, 16	6) 0	batch_normalization_1[0][0]
conv2d_2 (Conv2D)	(None,	32, 32, 8)) 1152	activation_1[0][0]
dropout_1 (Dropout)	(None,	32, 32, 8)) 0	conv2d_2[0][0]
concatenate_1 (Concatenate)	(None,	32, 32, 24	4) 0	conv2d_1[0][0] dropout_1[0][0]

batch_normalization_2 (BatchNor	(None,	32,	32,	24)	96	concatenate_1[0][0]
activation_2 (Activation)	(None,	32,	32,	24)	0	batch_normalization_2[0][0]
conv2d_3 (Conv2D)	(None,	32,	32,	8)	1728	activation_2[0][0]
dropout_2 (Dropout)	(None,	32,	32,	8)	0	conv2d_3[0][0]
concatenate_2 (Concatenate)	(None,	32,	32,	32)	0	concatenate_1[0][0] dropout_2[0][0]
batch_normalization_3 (BatchNor	(None,	32,	32,	32)	128	concatenate_2[0][0]
activation_3 (Activation)	(None,	32,	32,	32)	0	batch_normalization_3[0][0]
conv2d_4 (Conv2D)	(None,	32,	32,	8)	2304	activation_3[0][0]
dropout_3 (Dropout)	(None,	32,	32,	8)	0	conv2d_4[0][0]
concatenate_3 (Concatenate)	(None,	32,	32,	40)	0	concatenate_2[0][0] dropout_3[0][0]
batch_normalization_4 (BatchNor	(None,	32,	32,	40)	160	concatenate_3[0][0]
activation_4 (Activation)	(None,	32,	32,	40)	0	batch_normalization_4[0][0]
conv2d_5 (Conv2D)	(None,	32,	32,	8)	2880	activation_4[0][0]
dropout_4 (Dropout)	(None,	32,	32,	8)	0	conv2d_5[0][0]
concatenate_4 (Concatenate)	(None,	32,	32,	48)	0	concatenate_3[0][0] dropout_4[0][0]
batch_normalization_5 (BatchNor	(None,	32,	32,	48)	192	concatenate_4[0][0]
activation_5 (Activation)	(None,	32,	32,	48)	0	batch_normalization_5[0][0]
conv2d_6 (Conv2D)	(None,	32,	32,	8)	3456	activation_5[0][0]
dropout_5 (Dropout)	(None,	32,	32,	8)	0	conv2d_6[0][0]
concatenate_5 (Concatenate)	(None,	32,	32,	56)	0	concatenate_4[0][0] dropout_5[0][0]
batch_normalization_6 (BatchNor	(None,	32,	32,	56)	224	concatenate_5[0][0]
activation_6 (Activation)	(None,	32,	32,	56)	0	batch_normalization_6[0][0]

conv2d_7 (Conv2D)	(None,	32,	32,	8)	4032	activation_6[0][0]
dropout_6 (Dropout)	(None,	32,	32,	8)	0	conv2d_7[0][0]
concatenate_6 (Concatenate)	(None,	32,	32,	64)	0	concatenate_5[0][0] dropout_6[0][0]
batch_normalization_7 (BatchNor	(None,	32,	32,	64)	256 	concatenate_6[0][0]
activation_7 (Activation)	(None,	32,	32,	64)	0	batch_normalization_7[0][0]
conv2d_8 (Conv2D)	(None,	32,	32,	8)	4608	activation_7[0][0]
dropout_7 (Dropout)	(None,	32,	32,	8)	0	conv2d_8[0][0]
concatenate_7 (Concatenate)	(None,	32,	32,	72)	0	concatenate_6[0][0] dropout_7[0][0]
batch_normalization_8 (BatchNor	(None,	32,	32,	72)	288	concatenate_7[0][0]
activation_8 (Activation)	(None,	32,	32,	72)	0	batch_normalization_8[0][0]
conv2d_9 (Conv2D)	(None,	32,	32,	8)	5184	activation_8[0][0]
dropout_8 (Dropout)	(None,	32,	32,	8)	0	conv2d_9[0][0]
concatenate_8 (Concatenate)	(None,	32,	32,	80)	0	concatenate_7[0][0] dropout_8[0][0]
batch_normalization_9 (BatchNor	(None,	32,	32,	80)	320	concatenate_8[0][0]
activation_9 (Activation)	(None,	32,	32,	80)	0	batch_normalization_9[0][0]
conv2d_10 (Conv2D)	(None,	32,	32,	8)	5760	activation_9[0][0]
dropout_9 (Dropout)	(None,	32,	32,	8)	0	conv2d_10[0][0]
concatenate_9 (Concatenate)	(None,	32,	32,	88)	0	concatenate_8[0][0] dropout_9[0][0]
batch_normalization_10 (BatchNo	(None,	32,	32,	88)	352	concatenate_9[0][0]
activation_10 (Activation)	(None,	32,	32,	88)	0	batch_normalization_10[0][0]
conv2d_11 (Conv2D)	(None,	32,	32,	8)	6336	activation_10[0][0]
dropout_10 (Dropout)	(None,	32,	32,	8)	0	conv2d_11[0][0]

concatenate_10 (Concatenate)	(None,	32,	32,	96)	0	<pre>concatenate_9[0][0] dropout_10[0][0]</pre>
batch_normalization_11 (BatchNo	(None,	32,	32,	96)	384	concatenate_10[0][0]
activation_11 (Activation)	(None,	32,	32,	96)	0	batch_normalization_11[0][0]
conv2d_12 (Conv2D)	(None,	32,	32,	8)	6912	activation_11[0][0]
dropout_11 (Dropout)	(None,	32,	32,	8)	0	conv2d_12[0][0]
concatenate_11 (Concatenate)	(None,	32,	32,	104)	0	concatenate_10[0][0] dropout_11[0][0]
batch_normalization_12 (BatchNo	(None,	32,	32,	104)	416	concatenate_11[0][0]
activation_12 (Activation)	(None,	32,	32,	104)	0	batch_normalization_12[0][0]
conv2d_13 (Conv2D)	(None,	32,	32,	8)	7488	activation_12[0][0]
dropout_12 (Dropout)	(None,	32,	32,	8)	0	conv2d_13[0][0]
concatenate_12 (Concatenate)	(None,	32,	32,	112)	0	concatenate_11[0][0] dropout_12[0][0]
batch_normalization_13 (BatchNo	(None,	32,	32,	112)	448	concatenate_12[0][0]
activation_13 (Activation)	(None,	32,	32,	112)	0	batch_normalization_13[0][0]
conv2d_14 (Conv2D)	(None,	32,	32,	8)	8064	activation_13[0][0]
dropout_13 (Dropout)	(None,	32,	32,	8)	0	conv2d_14[0][0]
concatenate_13 (Concatenate)	(None,	32,	32,	120)	0	concatenate_12[0][0] dropout_13[0][0]
batch_normalization_14 (BatchNo	(None,	32,	32,	120)	480	concatenate_13[0][0]
activation_14 (Activation)	(None,	32,	32,	120)	0	batch_normalization_14[0][0]
conv2d_15 (Conv2D)	(None,	32,	32,	8)	8640	activation_14[0][0]
dropout_14 (Dropout)	(None,	32,	32,	8)	0	conv2d_15[0][0]
concatenate_14 (Concatenate)	(None,	32,	32,	128)	0	concatenate_13[0][0] dropout_14[0][0]
batch_normalization_15 (BatchNo	(None,	32,	32,	128)	512	concatenate_14[0][0]

activation_15 (Activation)	(None,	32,	32,	128)	0	batch_normalization_15[0][0]
conv2d_16 (Conv2D)	(None,	32,	32,	8)	9216	activation_15[0][0]
dropout_15 (Dropout)	(None,	32,	32,	8)	0	conv2d_16[0][0]
concatenate_15 (Concatenate)	(None,	32,	32,	136)	0	concatenate_14[0][0] dropout_15[0][0]
batch_normalization_16 (BatchNo	(None,	32,	32,	136)	544	concatenate_15[0][0]
activation_16 (Activation)	(None,	32,	32,	136)	0	batch_normalization_16[0][0]
conv2d_17 (Conv2D)	(None,	32,	32,	8)	9792	activation_16[0][0]
dropout_16 (Dropout)	(None,	32,	32,	8)	0	conv2d_17[0][0]
concatenate_16 (Concatenate)	(None,	32,	32,	144)	0	concatenate_15[0][0] dropout_16[0][0]
batch_normalization_17 (BatchNo	(None,	32,	32,	144)	576	concatenate_16[0][0]
activation_17 (Activation)	(None,	32,	32,	144)	0	batch_normalization_17[0][0]
conv2d_18 (Conv2D)	(None,	32,	32,	8)	10368	activation_17[0][0]
dropout_17 (Dropout)	(None,	32,	32,	8)	0	conv2d_18[0][0]
concatenate_17 (Concatenate)	(None,	32,	32,	152)	0	concatenate_16[0][0] dropout_17[0][0]
batch_normalization_18 (BatchNo	(None,	32,	32,	152)	608	concatenate_17[0][0]
activation_18 (Activation)	(None,	32,	32,	152)	0	batch_normalization_18[0][0]
conv2d_19 (Conv2D)	(None,	32,	32,	8)	10944	activation_18[0][0]
dropout_18 (Dropout)	(None,	32,	32,	8)	0	conv2d_19[0][0]
concatenate_18 (Concatenate)	(None,	32,	32,	160)	0	concatenate_17[0][0] dropout_18[0][0]
batch_normalization_19 (BatchNo	(None,	32,	32,	160)	640	concatenate_18[0][0]
activation_19 (Activation)	(None,	32,	32,	160)	0	batch_normalization_19[0][0]

dropout_19 (Dropout)	(None,	32,	32,	8)	0	conv2d_20[0][0]
concatenate_19 (Concatenate)	(None,	32,	32,	168)	0	concatenate_18[0][0] dropout_19[0][0]
batch_normalization_20 (BatchNo	(None,	32,	32,	168)	672	concatenate_19[0][0]
activation_20 (Activation)	(None,	32,	32,	168)	0	batch_normalization_20[0][0]
conv2d_21 (Conv2D)	(None,	32,	32,	8)	12096	activation_20[0][0]
dropout_20 (Dropout)	(None,	32,	32,	8)	0	conv2d_21[0][0]
concatenate_20 (Concatenate)	(None,	32,	32,	176)	0	concatenate_19[0][0] dropout_20[0][0]
batch_normalization_21 (BatchNo	(None,	32,	32,	176)	704	concatenate_20[0][0]
activation_21 (Activation)	(None,	32,	32,	176)	0	batch_normalization_21[0][0]
conv2d_22 (Conv2D)	(None,	32,	32,	8)	12672	activation_21[0][0]
dropout_21 (Dropout)	(None,	32,	32,	8)	0	conv2d_22[0][0]
concatenate_21 (Concatenate)	(None,	32,	32,	184)	0	concatenate_20[0][0] dropout_21[0][0]
batch_normalization_22 (BatchNo	(None,	32,	32,	184)	736	concatenate_21[0][0]
activation_22 (Activation)	(None,	32,	32,	184)	0	batch_normalization_22[0][0]
conv2d_23 (Conv2D)	(None,	32,	32,	8)	13248	activation_22[0][0]
dropout_22 (Dropout)	(None,	32,	32,	8)	0	conv2d_23[0][0]
concatenate_22 (Concatenate)	(None,	32,	32,	192)	0	concatenate_21[0][0] dropout_22[0][0]
batch_normalization_23 (BatchNo	(None,	32,	32,	192)	768	concatenate_22[0][0]
activation_23 (Activation)	(None,	32,	32,	192)	0	batch_normalization_23[0][0]
conv2d_24 (Conv2D)	(None,	32,	32,	8)	1536	activation_23[0][0]
dropout_23 (Dropout)	(None,	32,	32,	8)	0	conv2d_24[0][0]
average_pooling2d_1 (AveragePoo	(None,	16,	16,	8)	0	dropout_23[0][0]

batch_normalization_24 (BatchNo	(None,	16,	16,	8)	32	average_pooling2d_1[0][0]
activation_24 (Activation)	(None,	16,	16,	8)	0	batch_normalization_24[0][0]
conv2d_25 (Conv2D)	(None,	16,	16,	8)	576	activation_24[0][0]
dropout_24 (Dropout)	(None,	16,	16,	8)	0	conv2d_25[0][0]
concatenate_23 (Concatenate)	(None,	16,	16,	16)	0	average_pooling2d_1[0][0] dropout_24[0][0]
batch_normalization_25 (BatchNo	(None,	16,	16,	16)	64	concatenate_23[0][0]
activation_25 (Activation)	(None,	16,	16,	16)	0	batch_normalization_25[0][0]
conv2d_26 (Conv2D)	(None,	16,	16,	8)	1152	activation_25[0][0]
dropout_25 (Dropout)	(None,	16,	16,	8)	0	conv2d_26[0][0]
concatenate_24 (Concatenate)	(None,	16,	16,	24)	0	concatenate_23[0][0] dropout_25[0][0]
batch_normalization_26 (BatchNo	(None,	16,	16,	24)	96	concatenate_24[0][0]
activation_26 (Activation)	(None,	16,	16,	24)	0	batch_normalization_26[0][0]
conv2d_27 (Conv2D)	(None,	16,	16,	8)	1728	activation_26[0][0]
dropout_26 (Dropout)	(None,	16,	16,	8)	0	conv2d_27[0][0]
concatenate_25 (Concatenate)	(None,	16,	16,	32)	0	concatenate_24[0][0] dropout_26[0][0]
batch_normalization_27 (BatchNo	(None,	16,	16,	32)	128	concatenate_25[0][0]
activation_27 (Activation)	(None,	16,	16,	32)	0	batch_normalization_27[0][0]
conv2d_28 (Conv2D)	(None,	16,	16,	8)	2304	activation_27[0][0]
dropout_27 (Dropout)	(None,	16,	16,	8)	0	conv2d_28[0][0]
concatenate_26 (Concatenate)	(None,	16,	16,	40)	0	concatenate_25[0][0] dropout_27[0][0]
batch_normalization_28 (BatchNo	(None,	16,	16,	40)	160	concatenate_26[0][0]
activation_28 (Activation)	(None,	16,	16,	40)	0	batch_normalization_28[0][0]

conv2d_29 (Conv2D)	(None,	16,	16,	8)	2880	activation_28[0][0]
dropout_28 (Dropout)	(None,	16,	16,	8)	0	conv2d_29[0][0]
concatenate_27 (Concatenate)	(None,	16,	16,	48)	0	concatenate_26[0][0] dropout_28[0][0]
batch_normalization_29 (BatchNo	(None,	16,	16,	48)	192	concatenate_27[0][0]
activation_29 (Activation)	(None,	16,	16,	48)	0	batch_normalization_29[0][0]
conv2d_30 (Conv2D)	(None,	16,	16,	8)	3456	activation_29[0][0]
dropout_29 (Dropout)	(None,	16,	16,	8)	0	conv2d_30[0][0]
concatenate_28 (Concatenate)	(None,	16,	16,	56)	0	concatenate_27[0][0] dropout_29[0][0]
batch_normalization_30 (BatchNo	(None,	16,	16,	56)	224	concatenate_28[0][0]
activation_30 (Activation)	(None,	16,	16,	56)	0	batch_normalization_30[0][0]
conv2d_31 (Conv2D)	(None,	16,	16,	8)	4032	activation_30[0][0]
dropout_30 (Dropout)	(None,	16,	16,	8)	0	conv2d_31[0][0]
concatenate_29 (Concatenate)	(None,	16,	16,	64)	0	concatenate_28[0][0] dropout_30[0][0]
batch_normalization_31 (BatchNo	(None,	16,	16,	64)	256	concatenate_29[0][0]
activation_31 (Activation)	(None,	16,	16,	64)	0	batch_normalization_31[0][0]
conv2d_32 (Conv2D)	(None,	16,	16,	8)	4608	activation_31[0][0]
dropout_31 (Dropout)	(None,	16,	16,	8)	0	conv2d_32[0][0]
concatenate_30 (Concatenate)	(None,	16,	16,	72)	0	concatenate_29[0][0] dropout_31[0][0]
batch_normalization_32 (BatchNo	(None,	16,	16,	72)	288	concatenate_30[0][0]
activation_32 (Activation)	(None,	16,	16,	72)	0	batch_normalization_32[0][0]
conv2d_33 (Conv2D)	(None,	16,	16,	8)	5184	activation_32[0][0]
dropout_32 (Dropout)	(None,	16,	16,	8)	0	conv2d_33[0][0]

21 (0	(N	1.0	1.0			
concatenate_31 (Concatenate)	(None,	16,	16,	80)	0	concatenate_30[0][0] dropout_32[0][0]
batch_normalization_33 (BatchNo	(None,	16,	16,	80)	320	concatenate_31[0][0]
activation_33 (Activation)	(None,	16,	16,	80)	0	batch_normalization_33[0][0]
conv2d_34 (Conv2D)	(None,	16,	16,	8)	5760	activation_33[0][0]
dropout_33 (Dropout)	(None,	16,	16,	8)	0	conv2d_34[0][0]
concatenate_32 (Concatenate)	(None,	16,	16,	88)	0	concatenate_31[0][0] dropout_33[0][0]
batch_normalization_34 (BatchNo	(None,	16,	16,	88)	352	concatenate_32[0][0]
activation_34 (Activation)	(None,	16,	16,	88)	0	batch_normalization_34[0][0]
conv2d_35 (Conv2D)	(None,	16,	16,	8)	6336	activation_34[0][0]
dropout_34 (Dropout)	(None,	16,	16,	8)	0	conv2d_35[0][0]
concatenate_33 (Concatenate)	(None,	16,	16,	96)	0	concatenate_32[0][0] dropout_34[0][0]
batch_normalization_35 (BatchNo	(None,	16,	16,	96)	384	concatenate_33[0][0]
activation_35 (Activation)	(None,	16,	16,	96)	0	batch_normalization_35[0][0]
conv2d_36 (Conv2D)	(None,	16,	16,	8)	6912	activation_35[0][0]
dropout_35 (Dropout)	(None,	16,	16,	8)	0	conv2d_36[0][0]
concatenate_34 (Concatenate)	(None,	16,	16,	104)	0	concatenate_33[0][0] dropout_35[0][0]
batch_normalization_36 (BatchNo	(None,	16,	16,	104)	416	concatenate_34[0][0]
activation_36 (Activation)	(None,	16,	16,	104)	0	batch_normalization_36[0][0]
conv2d_37 (Conv2D)	(None,	16,	16,	8)	7488	activation_36[0][0]
dropout_36 (Dropout)	(None,	16,	16,	8)	0	conv2d_37[0][0]
concatenate_35 (Concatenate)	(None,	16,	16,	112)	0	concatenate_34[0][0] dropout_36[0][0]

batch_normalization_37 (BatchNo	(None,	16,	16,	112)	448	concatenate_35[0][0]
activation_37 (Activation)	(None,	16,	16,	112)	0	batch_normalization_37[0][0]
conv2d_38 (Conv2D)	(None,	16,	16,	8)	8064	activation_37[0][0]
dropout_37 (Dropout)	(None,	16,	16,	8)	0	conv2d_38[0][0]
concatenate_36 (Concatenate)	(None,	16,	16,	120)	0	concatenate_35[0][0] dropout_37[0][0]
batch_normalization_38 (BatchNo	(None,	16,	16,	120)	480	concatenate_36[0][0]
activation_38 (Activation)	(None,	16,	16,	120)	0	batch_normalization_38[0][0]
conv2d_39 (Conv2D)	(None,	16,	16,	8)	8640	activation_38[0][0]
dropout_38 (Dropout)	(None,	16,	16,	8)	0	conv2d_39[0][0]
concatenate_37 (Concatenate)	(None,	16,	16,	128)	0	concatenate_36[0][0] dropout_38[0][0]
batch_normalization_39 (BatchNo	(None,	16,	16,	128)	512	concatenate_37[0][0]
activation_39 (Activation)	(None,	16,	16,	128)	0	batch_normalization_39[0][0]
conv2d_40 (Conv2D)	(None,	16,	16,	8)	9216	activation_39[0][0]
dropout_39 (Dropout)	(None,	16,	16,	8)	0	conv2d_40[0][0]
concatenate_38 (Concatenate)	(None,	16,	16,	136)	0	concatenate_37[0][0] dropout_39[0][0]
batch_normalization_40 (BatchNo	(None,	16,	16,	136)	544	concatenate_38[0][0]
activation_40 (Activation)	(None,	16,	16,	136)	0	batch_normalization_40[0][0]
conv2d_41 (Conv2D)	(None,	16,	16,	8)	9792	activation_40[0][0]
dropout_40 (Dropout)	(None,	16,	16,	8)	0	conv2d_41[0][0]
concatenate_39 (Concatenate)	(None,	16,	16,	144)	0	concatenate_38[0][0] dropout_40[0][0]
batch_normalization_41 (BatchNo	(None,	16,	16,	144)	576	concatenate_39[0][0]
activation_41 (Activation)	(None,	16,	16,		0	

conv2d_42 (Conv2D)	(None,	16,	16,	8)	10368	activation_41[0][0]
dropout_41 (Dropout)	(None,	16,	16,	8)	0	conv2d_42[0][0]
concatenate_40 (Concatenate)	(None,	16,	16,	152)	0	concatenate_39[0][0] dropout_41[0][0]
batch_normalization_42 (BatchNo	(None,	16,	16,	152)	608	concatenate_40[0][0]
activation_42 (Activation)	(None,	16,	16,	152)	0	batch_normalization_42[0][0]
conv2d_43 (Conv2D)	(None,	16,	16,	8)	10944	activation_42[0][0]
dropout_42 (Dropout)	(None,	16,	16,	8)	0	conv2d_43[0][0]
<pre>concatenate_41 (Concatenate)</pre>	(None,	16,	16,	160)	0	concatenate_40[0][0] dropout_42[0][0]
batch_normalization_43 (BatchNo	(None,	16,	16,	160)	640	concatenate_41[0][0]
activation_43 (Activation)	(None,	16,	16,	160)	0	batch_normalization_43[0][0]
conv2d_44 (Conv2D)	(None,	16,	16,	8)	11520	activation_43[0][0]
dropout_43 (Dropout)	(None,	16,	16,	8)	0	conv2d_44[0][0]
concatenate_42 (Concatenate)	(None,	16,	16,	168)	0	concatenate_41[0][0] dropout_43[0][0]
batch_normalization_44 (BatchNo	(None,	16,	16,	168)	672	concatenate_42[0][0]
activation_44 (Activation)	(None,	16,	16,	168)	0	batch_normalization_44[0][0]
conv2d_45 (Conv2D)	(None,	16,	16,	8)	12096	activation_44[0][0]
dropout_44 (Dropout)	(None,	16,	16,	8)	0	conv2d_45[0][0]
concatenate_43 (Concatenate)	(None,	16,	16,	176)	0	concatenate_42[0][0] dropout_44[0][0]
batch_normalization_45 (BatchNo	(None,	16,	16,	176)	704	concatenate_43[0][0]
activation_45 (Activation)	(None,	16,	16,	176)	0	batch_normalization_45[0][0]
conv2d_46 (Conv2D)	(None,	16,	16,	8)	12672	activation_45[0][0]
dropout_45 (Dropout)	(None,	16,	16,	8)	0	conv2d_46[0][0]

concatenate_44 (Concatenate)	(None,	16, 16, 184)	0	concatenate_43[0][0] dropout_45[0][0]
batch_normalization_46 (BatchNo	(None,	16, 16, 184)	736	concatenate_44[0][0]
activation_46 (Activation)	(None,	16, 16, 184)	0	batch_normalization_46[0][0]
conv2d_47 (Conv2D)	(None,	16, 16, 8)	1472	activation_46[0][0]
dropout_46 (Dropout)	(None,	16, 16, 8)	0	conv2d_47[0][0]
average_pooling2d_2 (AveragePoo	(None,	8, 8, 8)	0	dropout_46[0][0]
batch_normalization_47 (BatchNo	(None,	8, 8, 8)	32	average_pooling2d_2[0][0]
activation_47 (Activation)	(None,	8, 8, 8)	0	batch_normalization_47[0][0]
batch_normalization_70 (BatchNo	(None,	16, 16, 8)	32	average_pooling2d_1[0][0]
conv2d_48 (Conv2D)	(None,	8, 8, 8)	576	activation_47[0][0]
activation_70 (Activation)	(None,	16, 16, 8)	0	batch_normalization_70[0][0]
dropout_47 (Dropout)	(None,	8, 8, 8)	0	conv2d_48[0][0]
conv2d_71 (Conv2D)	(None,	16, 16, 8)	576	activation_70[0][0]
concatenate_45 (Concatenate)	(None,	8, 8, 16)	0	average_pooling2d_2[0][0] dropout_47[0][0]
dropout_70 (Dropout)	(None,	16, 16, 8)	0	conv2d_71[0][0]
batch_normalization_48 (BatchNo	(None,	8, 8, 16)	64	concatenate_45[0][0]
concatenate_67 (Concatenate)	(None,	16, 16, 16)	0	average_pooling2d_1[0][0] dropout_70[0][0]
activation_48 (Activation)	(None,	8, 8, 16)	0	batch_normalization_48[0][0]
batch_normalization_71 (BatchNo	(None,	16, 16, 16)	64	concatenate_67[0][0]
conv2d_49 (Conv2D)	(None,	8, 8, 8)	1152	activation_48[0][0]
activation_71 (Activation)	(None,	16, 16, 16)	0	batch_normalization_71[0][0]
dropout_48 (Dropout)	(None,	8, 8, 8)	0	conv2d_49[0][0]
conv2d_72 (Conv2D)	(None,	16, 16, 8)	1152	activation_71[0][0]

concatenate_46 (Concatenate)	(None,	8, 8, 24)	0	concatenate_45[0][0] dropout_48[0][0]
dropout_71 (Dropout)	(None,	16, 16, 8)	0	conv2d_72[0][0]
batch_normalization_49 (BatchNo	(None,	8, 8, 24)	96	concatenate_46[0][0]
concatenate_68 (Concatenate)	(None,	16, 16, 24)	0	concatenate_67[0][0] dropout_71[0][0]
activation_49 (Activation)	(None,	8, 8, 24)	0	batch_normalization_49[0][0]
batch_normalization_72 (BatchNo	(None,	16, 16, 24)	96	concatenate_68[0][0]
conv2d_50 (Conv2D)	(None,	8, 8, 8)	1728	activation_49[0][0]
activation_72 (Activation)	(None,	16, 16, 24)	0	batch_normalization_72[0][0]
dropout_49 (Dropout)	(None,	8, 8, 8)	0	conv2d_50[0][0]
conv2d_73 (Conv2D)	(None,	16, 16, 8)	1728	activation_72[0][0]
concatenate_47 (Concatenate)	(None,	8, 8, 32)	0	concatenate_46[0][0] dropout_49[0][0]
dropout_72 (Dropout)	(None,	16, 16, 8)	0	conv2d_73[0][0]
batch_normalization_50 (BatchNo	(None,	8, 8, 32)	128	concatenate_47[0][0]
concatenate_69 (Concatenate)	(None,	16, 16, 32)	0	concatenate_68[0][0] dropout_72[0][0]
activation_50 (Activation)	(None,	8, 8, 32)	0	batch_normalization_50[0][0]
batch_normalization_73 (BatchNo	(None,	16, 16, 32)	128	concatenate_69[0][0]
conv2d_51 (Conv2D)	(None,	8, 8, 8)	2304	activation_50[0][0]
activation_73 (Activation)	(None,	16, 16, 32)	0	batch_normalization_73[0][0]
dropout_50 (Dropout)	(None,	8, 8, 8)	0	conv2d_51[0][0]
conv2d_74 (Conv2D)	(None,	16, 16, 8)	2304	activation_73[0][0]
concatenate_48 (Concatenate)	(None,	8, 8, 40)	0	concatenate_47[0][0] dropout_50[0][0]

dropout_73 (Dropout)	(None,	16, 16, 8)	0	conv2d_74[0][0]
batch_normalization_51 (BatchNo	(None,	8, 8, 40)	160	concatenate_48[0][0]
concatenate_70 (Concatenate)	(None,	16, 16, 40)	0	concatenate_69[0][0] dropout_73[0][0]
activation_51 (Activation)	(None,	8, 8, 40)	0	batch_normalization_51[0][0]
batch_normalization_74 (BatchNo	(None,	16, 16, 40)	160	concatenate_70[0][0]
conv2d_52 (Conv2D)	(None,	8, 8, 8)	2880	activation_51[0][0]
activation_74 (Activation)	(None,	16, 16, 40)	0	batch_normalization_74[0][0]
dropout_51 (Dropout)	(None,	8, 8, 8)	0	conv2d_52[0][0]
conv2d_75 (Conv2D)	(None,	16, 16, 8)	2880	activation_74[0][0]
concatenate_49 (Concatenate)	(None,	8, 8, 48)	0	concatenate_48[0][0] dropout_51[0][0]
dropout_74 (Dropout)	(None,	16, 16, 8)	0	conv2d_75[0][0]
batch_normalization_52 (BatchNo	(None,	8, 8, 48)	192	concatenate_49[0][0]
concatenate_71 (Concatenate)	(None,	16, 16, 48)	0	concatenate_70[0][0] dropout_74[0][0]
activation_52 (Activation)	(None,	8, 8, 48)	0	batch_normalization_52[0][0]
batch_normalization_75 (BatchNo	(None,	16, 16, 48)	192	concatenate_71[0][0]
conv2d_53 (Conv2D)	(None,	8, 8, 8)	3456	activation_52[0][0]
activation_75 (Activation)	(None,	16, 16, 48)	0	batch_normalization_75[0][0]
dropout_52 (Dropout)	(None,	8, 8, 8)	0	conv2d_53[0][0]
conv2d_76 (Conv2D)	(None,	16, 16, 8)	3456	activation_75[0][0]
concatenate_50 (Concatenate)	(None,	8, 8, 56)	0	concatenate_49[0][0] dropout_52[0][0]
dropout_75 (Dropout)	(None,	16, 16, 8)	0	conv2d_76[0][0]
batch_normalization_53 (BatchNo	(None,	8, 8, 56)	224	concatenate_50[0][0]

concatenate_72 (Concatenate)	(None,	16, 16, 56)	0	concatenate_71[0][0] dropout_75[0][0]
activation_53 (Activation)	(None,	8, 8, 56)	0	batch_normalization_53[0][0]
batch_normalization_76 (BatchNo	(None,	16, 16, 56)	224	concatenate_72[0][0]
conv2d_54 (Conv2D)	(None,	8, 8, 8)	4032	activation_53[0][0]
activation_76 (Activation)	(None,	16, 16, 56)	0	batch_normalization_76[0][0]
dropout_53 (Dropout)	(None,	8, 8, 8)	0	conv2d_54[0][0]
conv2d_77 (Conv2D)	(None,	16, 16, 8)	4032	activation_76[0][0]
concatenate_51 (Concatenate)	(None,	8, 8, 64)	0	concatenate_50[0][0] dropout_53[0][0]
dropout_76 (Dropout)	(None,	16, 16, 8)	0	conv2d_77[0][0]
batch_normalization_54 (BatchNo	(None,	8, 8, 64)	256	concatenate_51[0][0]
concatenate_73 (Concatenate)	(None,	16, 16, 64)	0	concatenate_72[0][0] dropout_76[0][0]
activation_54 (Activation)	(None,	8, 8, 64)	0	batch_normalization_54[0][0]
batch_normalization_77 (BatchNo	(None,	16, 16, 64)	256	concatenate_73[0][0]
conv2d_55 (Conv2D)	(None,	8, 8, 8)	4608	activation_54[0][0]
activation_77 (Activation)	(None,	16, 16, 64)	0	batch_normalization_77[0][0]
dropout_54 (Dropout)	(None,	8, 8, 8)	0	conv2d_55[0][0]
conv2d_78 (Conv2D)	(None,	16, 16, 8)	4608	activation_77[0][0]
concatenate_52 (Concatenate)	(None,	8, 8, 72)	0	concatenate_51[0][0] dropout_54[0][0]
dropout_77 (Dropout)	(None,	16, 16, 8)	0	conv2d_78[0][0]
batch_normalization_55 (BatchNo	(None,	8, 8, 72)	288	concatenate_52[0][0]
concatenate_74 (Concatenate)	(None,	16, 16, 72)	0	concatenate_73[0][0] dropout_77[0][0]
activation_55 (Activation)	(None,	8, 8, 72)	0	batch_normalization_55[0][0]

batch_normalization_78 (BatchNo	(None,	16, 16, 72)	288	concatenate_74[0][0]
conv2d_56 (Conv2D)	(None,	8, 8, 8)	5184	activation_55[0][0]
activation_78 (Activation)	(None,	16, 16, 72)	0	batch_normalization_78[0][0]
dropout_55 (Dropout)	(None,	8, 8, 8)	0	conv2d_56[0][0]
conv2d_79 (Conv2D)	(None,	16, 16, 8)	5184	activation_78[0][0]
concatenate_53 (Concatenate)	(None,	8, 8, 80)	0	concatenate_52[0][0] dropout_55[0][0]
dropout_78 (Dropout)	(None,	16, 16, 8)	0	conv2d_79[0][0]
batch_normalization_56 (BatchNo	(None,	8, 8, 80)	320	concatenate_53[0][0]
concatenate_75 (Concatenate)	(None,	16, 16, 80)	0	concatenate_74[0][0] dropout_78[0][0]
activation_56 (Activation)	(None,	8, 8, 80)	0	batch_normalization_56[0][0]
batch_normalization_79 (BatchNo	(None,	16, 16, 80)	320	concatenate_75[0][0]
conv2d_57 (Conv2D)	(None,	8, 8, 8)	5760	activation_56[0][0]
activation_79 (Activation)	(None,	16, 16, 80)	0	batch_normalization_79[0][0]
dropout_56 (Dropout)	(None,	8, 8, 8)	0	conv2d_57[0][0]
conv2d_80 (Conv2D)	(None,	16, 16, 8)	5760	activation_79[0][0]
concatenate_54 (Concatenate)	(None,	8, 8, 88)	0	concatenate_53[0][0] dropout_56[0][0]
dropout_79 (Dropout)	(None,	16, 16, 8)	0	conv2d_80[0][0]
batch_normalization_57 (BatchNo	(None,	8, 8, 88)	352	concatenate_54[0][0]
concatenate_76 (Concatenate)	(None,	16, 16, 88)	0	concatenate_75[0][0] dropout_79[0][0]
activation_57 (Activation)	(None,	8, 8, 88)	0	batch_normalization_57[0][0]
batch_normalization_80 (BatchNo	(None,	16, 16, 88)	352	concatenate_76[0][0]
conv2d_58 (Conv2D)	(None,	8, 8, 8)	6336	activation_57[0][0]

activation_80 (Activation)	(None,	16, 16, 88)	0	batch_normalization_80[0][0]
dropout_57 (Dropout)	(None,	8, 8, 8)	0	conv2d_58[0][0]
conv2d_81 (Conv2D)	(None,	16, 16, 8)	6336	activation_80[0][0]
concatenate_55 (Concatenate)	(None,	8, 8, 96)	0	concatenate_54[0][0] dropout_57[0][0]
dropout_80 (Dropout)	(None,	16, 16, 8)	0	conv2d_81[0][0]
batch_normalization_58 (BatchNo	(None,	8, 8, 96)	384	concatenate_55[0][0]
concatenate_77 (Concatenate)	(None,	16, 16, 96)	0	concatenate_76[0][0] dropout_80[0][0]
activation_58 (Activation)	(None,	8, 8, 96)	0	batch_normalization_58[0][0]
batch_normalization_81 (BatchNo	(None,	16, 16, 96)	384	concatenate_77[0][0]
conv2d_59 (Conv2D)	(None,	8, 8, 8)	6912	activation_58[0][0]
activation_81 (Activation)	(None,	16, 16, 96)	0	batch_normalization_81[0][0]
dropout_58 (Dropout)	(None,	8, 8, 8)	0	conv2d_59[0][0]
conv2d_82 (Conv2D)	(None,	16, 16, 8)	6912	activation_81[0][0]
concatenate_56 (Concatenate)	(None,	8, 8, 104)	0	concatenate_55[0][0] dropout_58[0][0]
dropout_81 (Dropout)	(None,	16, 16, 8)	0	conv2d_82[0][0]
batch_normalization_59 (BatchNo	(None,	8, 8, 104)	416	concatenate_56[0][0]
concatenate_78 (Concatenate)	(None,	16, 16, 104)	0	concatenate_77[0][0] dropout_81[0][0]
activation_59 (Activation)	(None,	8, 8, 104)	0	batch_normalization_59[0][0]
batch_normalization_82 (BatchNo	(None,	16, 16, 104)	416	concatenate_78[0][0]
conv2d_60 (Conv2D)	(None,	8, 8, 8)	7488	activation_59[0][0]
activation_82 (Activation)	(None,	16, 16, 104)	0	batch_normalization_82[0][0]
dropout_59 (Dropout)	(None,	8, 8, 8)	0	conv2d_60[0][0]

conv2d_83 (Conv2D)	(None,	16, 16, 8)	7488	activation_82[0][0]
concatenate_57 (Concatenate)	(None,	8, 8, 112)	0	concatenate_56[0][0] dropout_59[0][0]
dropout_82 (Dropout)	(None,	16, 16, 8)	0	conv2d_83[0][0]
batch_normalization_60 (BatchNo	(None,	8, 8, 112)	448	concatenate_57[0][0]
concatenate_79 (Concatenate)	(None,	16, 16, 112)	0	concatenate_78[0][0] dropout_82[0][0]
activation_60 (Activation)	(None,	8, 8, 112)	0	batch_normalization_60[0][0]
batch_normalization_83 (BatchNo	(None,	16, 16, 112)	448	concatenate_79[0][0]
conv2d_61 (Conv2D)	(None,	8, 8, 8)	8064	activation_60[0][0]
activation_83 (Activation)	(None,	16, 16, 112)	0	batch_normalization_83[0][0]
dropout_60 (Dropout)	(None,	8, 8, 8)	0	conv2d_61[0][0]
conv2d_84 (Conv2D)	(None,	16, 16, 8)	8064	activation_83[0][0]
concatenate_58 (Concatenate)	(None,	8, 8, 120)	0	concatenate_57[0][0] dropout_60[0][0]
dropout_83 (Dropout)	(None,	16, 16, 8)	0	conv2d_84[0][0]
batch_normalization_61 (BatchNo	(None,	8, 8, 120)	480	concatenate_58[0][0]
concatenate_80 (Concatenate)	(None,	16, 16, 120)	0	concatenate_79[0][0] dropout_83[0][0]
activation_61 (Activation)	(None,	8, 8, 120)	0	batch_normalization_61[0][0]
batch_normalization_84 (BatchNo	(None,	16, 16, 120)	480	concatenate_80[0][0]
conv2d_62 (Conv2D)	(None,	8, 8, 8)	8640	activation_61[0][0]
activation_84 (Activation)	(None,	16, 16, 120)	0	batch_normalization_84[0][0]
dropout_61 (Dropout)	(None,	8, 8, 8)	0	conv2d_62[0][0]
conv2d_85 (Conv2D)	(None,	16, 16, 8)	8640	activation_84[0][0]
concatenate_59 (Concatenate)	(None,	8, 8, 128)	0	concatenate_58[0][0]

				dropout_61[0][0]
dropout_84 (Dropout)	(None,	16, 16, 8)	0	conv2d_85[0][0]
batch_normalization_62 (BatchNo	(None,	8, 8, 128)	512	concatenate_59[0][0]
concatenate_81 (Concatenate)	(None,	16, 16, 128)	0	concatenate_80[0][0] dropout_84[0][0]
activation_62 (Activation)	(None,	8, 8, 128)	0	batch_normalization_62[0][0]
batch_normalization_85 (BatchNo	(None,	16, 16, 128)	512	concatenate_81[0][0]
conv2d_63 (Conv2D)	(None,	8, 8, 8)	9216	activation_62[0][0]
activation_85 (Activation)	(None,	16, 16, 128)	0	batch_normalization_85[0][0]
dropout_62 (Dropout)	(None,	8, 8, 8)	0	conv2d_63[0][0]
conv2d_86 (Conv2D)	(None,	16, 16, 8)	9216	activation_85[0][0]
concatenate_60 (Concatenate)	(None,	8, 8, 136)	0	concatenate_59[0][0] dropout_62[0][0]
dropout_85 (Dropout)	(None,	16, 16, 8)	0	conv2d_86[0][0]
batch_normalization_63 (BatchNo	(None,	8, 8, 136)	544	concatenate_60[0][0]
concatenate_82 (Concatenate)	(None,	16, 16, 136)	0	concatenate_81[0][0] dropout_85[0][0]
activation_63 (Activation)	(None,	8, 8, 136)	0	batch_normalization_63[0][0]
batch_normalization_86 (BatchNo	(None,	16, 16, 136)	544	concatenate_82[0][0]
conv2d_64 (Conv2D)	(None,	8, 8, 8)	9792	activation_63[0][0]
activation_86 (Activation)	(None,	16, 16, 136)	0	batch_normalization_86[0][0]
dropout_63 (Dropout)	(None,	8, 8, 8)	0	conv2d_64[0][0]
conv2d_87 (Conv2D)	(None,	16, 16, 8)	9792	activation_86[0][0]
concatenate_61 (Concatenate)	(None,	8, 8, 144)	0	concatenate_60[0][0] dropout_63[0][0]
dropout_86 (Dropout)	(None,	16, 16, 8)	0	conv2d_87[0][0]

batch_normalization_64 (BatchNo	(None,	8, 8, 144)	576	concatenate_61[0][0]
concatenate_83 (Concatenate)	(None,	16, 16, 144)	0	concatenate_82[0][0] dropout_86[0][0]
activation_64 (Activation)	(None,	8, 8, 144)	0	batch_normalization_64[0][0]
batch_normalization_87 (BatchNo	(None,	16, 16, 144)	576	concatenate_83[0][0]
conv2d_65 (Conv2D)	(None,	8, 8, 8)	10368	activation_64[0][0]
activation_87 (Activation)	(None,	16, 16, 144)	0	batch_normalization_87[0][0]
dropout_64 (Dropout)	(None,	8, 8, 8)	0	conv2d_65[0][0]
conv2d_88 (Conv2D)	(None,	16, 16, 8)	10368	activation_87[0][0]
concatenate_62 (Concatenate)	(None,	8, 8, 152)	0	concatenate_61[0][0] dropout_64[0][0]
dropout_87 (Dropout)	(None,	16, 16, 8)	0	conv2d_88[0][0]
batch_normalization_65 (BatchNo	(None,	8, 8, 152)	608	concatenate_62[0][0]
concatenate_84 (Concatenate)	(None,	16, 16, 152)	0	concatenate_83[0][0] dropout_87[0][0]
activation_65 (Activation)	(None,	8, 8, 152)	0	batch_normalization_65[0][0]
batch_normalization_88 (BatchNo	(None,	16, 16, 152)	608	concatenate_84[0][0]
conv2d_66 (Conv2D)	(None,	8, 8, 8)	10944	activation_65[0][0]
activation_88 (Activation)	(None,	16, 16, 152)	0	batch_normalization_88[0][0]
dropout_65 (Dropout)	(None,	8, 8, 8)	0	conv2d_66[0][0]
conv2d_89 (Conv2D)	(None,	16, 16, 8)	10944	activation_88[0][0]
concatenate_63 (Concatenate)	(None,	8, 8, 160)	0	concatenate_62[0][0] dropout_65[0][0]
dropout_88 (Dropout)	(None,	16, 16, 8)	0	conv2d_89[0][0]
batch_normalization_66 (BatchNo	(None,	8, 8, 160)	640	concatenate_63[0][0]
concatenate_85 (Concatenate)	(None,	16, 16, 160)	0	concatenate_84[0][0] dropout_88[0][0]

activation_66 (Activation)	(None,	8, 8, 160)	0	batch_normalization_66[0][0]
batch_normalization_89 (BatchNo	(None,	16, 16, 160)	640	concatenate_85[0][0]
conv2d_67 (Conv2D)	(None,	8, 8, 8)	11520	activation_66[0][0]
activation_89 (Activation)	(None,	16, 16, 160)	0	batch_normalization_89[0][0]
dropout_66 (Dropout)	(None,	8, 8, 8)	0	conv2d_67[0][0]
conv2d_90 (Conv2D)	(None,	16, 16, 8)	11520	activation_89[0][0]
concatenate_64 (Concatenate)	(None,	8, 8, 168)	0	concatenate_63[0][0] dropout_66[0][0]
dropout_89 (Dropout)	(None,	16, 16, 8)	0	conv2d_90[0][0]
batch_normalization_67 (BatchNo	(None,	8, 8, 168)	672	concatenate_64[0][0]
concatenate_86 (Concatenate)	(None,	16, 16, 168)	0	concatenate_85[0][0] dropout_89[0][0]
activation_67 (Activation)	(None,	8, 8, 168)	0	batch_normalization_67[0][0]
batch_normalization_90 (BatchNo	(None,	16, 16, 168)	672	concatenate_86[0][0]
conv2d_68 (Conv2D)	(None,	8, 8, 8)	12096	activation_67[0][0]
activation_90 (Activation)	(None,	16, 16, 168)	0	batch_normalization_90[0][0]
dropout_67 (Dropout)	(None,	8, 8, 8)	0	conv2d_68[0][0]
conv2d_91 (Conv2D)	(None,	16, 16, 8)	12096	activation_90[0][0]
concatenate_65 (Concatenate)	(None,	8, 8, 176)	0	concatenate_64[0][0] dropout_67[0][0]
dropout_90 (Dropout)	(None,	16, 16, 8)	0	conv2d_91[0][0]
batch_normalization_68 (BatchNo	(None,	8, 8, 176)	704	concatenate_65[0][0]
concatenate_87 (Concatenate)	(None,	16, 16, 176)	0	concatenate_86[0][0] dropout_90[0][0]
activation_68 (Activation)	(None,	8, 8, 176)	0	batch_normalization_68[0][0]
batch_normalization_91 (BatchNo	(None,	16, 16, 176)	704	concatenate_87[0][0]

conv2d_69 (Conv2D)	(None,	8, 8	3, 8)	12672	activation_68[0][0]
activation_91 (Activation)	(None,	16,	16, 176)	0	batch_normalization_91[0][0]
dropout_68 (Dropout)	(None,	8, 8	3, 8)	0	conv2d_69[0][0]
conv2d_92 (Conv2D)	(None,	16,	16, 8)	12672	activation_91[0][0]
concatenate_66 (Concatenate)	(None,	8, 8	3, 184)	0	concatenate_65[0][0] dropout_68[0][0]
dropout_91 (Dropout)	(None,	16,	16, 8)	0	conv2d_92[0][0]
batch_normalization_69 (BatchNo	(None,	8, 8	3, 184)	736	concatenate_66[0][0]
concatenate_88 (Concatenate)	(None,	16,	16, 184)	0	concatenate_87[0][0] dropout_91[0][0]
activation_69 (Activation)	(None,	8, 8	3, 184)	0	batch_normalization_69[0][0]
conv2d_93 (Conv2D)	(None,	12,	12, 8)	36800	concatenate_88[0][0]
conv2d_70 (Conv2D)	(None,	8, 8	3, 8)	1472	activation_69[0][0]
conv2d_94 (Conv2D)	(None,	8, 8	3, 8)	1600	conv2d_93[0][0]
dropout_69 (Dropout)	(None,	8, 8	3, 8)	0	conv2d_70[0][0]
conv2d_95 (Conv2D)	(None,	4, 4	, 8)	1600	conv2d_94[0][0]
average_pooling2d_3 (AveragePoo	(None,	4, 4	, 8)	0	dropout_69[0][0]
concatenate_89 (Concatenate)	(None,	4, 4	, 16)	0	conv2d_95[0][0] average_pooling2d_3[0][0]
batch_normalization_92 (BatchNo	(None,	4, 4	, 16)	64	concatenate_89[0][0]
activation_92 (Activation)	(None,	4, 4	, 16)	0	batch_normalization_92[0][0]
conv2d_96 (Conv2D)	(None,	4, 4	, 8)	128	activation_92[0][0]
dropout_92 (Dropout)	(None,	4, 4	, 8)	0	conv2d_96[0][0]
average_pooling2d_4 (AveragePoo	(None,	2, 2	2, 8)	0	dropout_92[0][0]
batch_normalization_93 (BatchNo	(None,	2, 2	2, 8)	32	average_pooling2d_4[0][0]

activation_93 (Activation)	(None,	2,	2,	8)	0	batch_normalization_93[0][0]
conv2d_97 (Conv2D)	(None,	2,	2,	8)	576	activation_93[0][0]
dropout_93 (Dropout)	(None,	2,	2,	8)	0	conv2d_97[0][0]
concatenate_90 (Concatenate)	(None,	2,	2,	16)	0	average_pooling2d_4[0][0] dropout_93[0][0]
batch_normalization_94 (BatchNo	(None,	2,	2,	16)	64	concatenate_90[0][0]
activation_94 (Activation)	(None,	2,	2,	16)	0	batch_normalization_94[0][0]
conv2d_98 (Conv2D)	(None,	2,	2,	8)	1152	activation_94[0][0]
dropout_94 (Dropout)	(None,	2,	2,	8)	0	conv2d_98[0][0]
concatenate_91 (Concatenate)	(None,	2,	2,	24)	0	concatenate_90[0][0] dropout_94[0][0]
batch_normalization_95 (BatchNo	(None,	2,	2,	24)	96	concatenate_91[0][0]
activation_95 (Activation)	(None,	2,	2,	24)	0	batch_normalization_95[0][0]
conv2d_99 (Conv2D)	(None,	2,	2,	8)	1728	activation_95[0][0]
dropout_95 (Dropout)	(None,	2,	2,	8)	0	conv2d_99[0][0]
concatenate_92 (Concatenate)	(None,	2,	2,	32)	0	concatenate_91[0][0] dropout_95[0][0]
batch_normalization_96 (BatchNo	(None,	2,	2,	32)	128	concatenate_92[0][0]
activation_96 (Activation)	(None,	2,	2,	32)	0	batch_normalization_96[0][0]
conv2d_100 (Conv2D)	(None,	2,	2,	8)	2304	activation_96[0][0]
dropout_96 (Dropout)	(None,	2,	2,	8)	0	conv2d_100[0][0]
concatenate_93 (Concatenate)	(None,	2,	2,	40)	0	concatenate_92[0][0] dropout_96[0][0]
batch_normalization_97 (BatchNo	(None,	2,	2,	40)	160	concatenate_93[0][0]
activation_97 (Activation)	(None,	2,	2,	40)	0	batch_normalization_97[0][0]
conv2d_101 (Conv2D)	(None,	2,	2,	8)	2880	activation_97[0][0]

dropout_97 (Dropout)	(None,	2,	2,	8)	0	conv2d_101[0][0]
concatenate_94 (Concatenate)	(None,	2,	2,	48)	0	concatenate_93[0][0] dropout_97[0][0]
batch_normalization_98 (BatchNo	(None,	2,	2,	48)	192	concatenate_94[0][0]
activation_98 (Activation)	(None,	2,	2,	48)	0	batch_normalization_98[0][0]
conv2d_102 (Conv2D)	(None,	2,	2,	8)	3456	activation_98[0][0]
dropout_98 (Dropout)	(None,	2,	2,	8)	0	conv2d_102[0][0]
concatenate_95 (Concatenate)	(None,	2,	2,	56)	0	concatenate_94[0][0] dropout_98[0][0]
batch_normalization_99 (BatchNo	(None,	2,	2,	56)	224	concatenate_95[0][0]
activation_99 (Activation)	(None,	2,	2,	56)	0	batch_normalization_99[0][0]
conv2d_103 (Conv2D)	(None,	2,	2,	8)	4032	activation_99[0][0]
dropout_99 (Dropout)	(None,	2,	2,	8)	0	conv2d_103[0][0]
concatenate_96 (Concatenate)	(None,	2,	2,	64)	0	concatenate_95[0][0] dropout_99[0][0]
batch_normalization_100 (BatchN	(None,	2,	2,	64)	256	concatenate_96[0][0]
activation_100 (Activation)	(None,	2,	2,	64)	0	batch_normalization_100[0][0]
conv2d_104 (Conv2D)	(None,	2,	2,	8)	4608	activation_100[0][0]
dropout_100 (Dropout)	(None,	2,	2,	8)	0	conv2d_104[0][0]
concatenate_97 (Concatenate)	(None,	2,	2,	72)	0	concatenate_96[0][0] dropout_100[0][0]
batch_normalization_101 (BatchN	(None,	2,	2,	72)	288	concatenate_97[0][0]
activation_101 (Activation)	(None,	2,	2,	72)		batch_normalization_101[0][0]
conv2d_105 (Conv2D)	(None,	2,	2,	8)		activation_101[0][0]
dropout_101 (Dropout)	(None,	2,	2,	8)	0	conv2d_105[0][0]
concatenate_98 (Concatenate)	(None,	2,	2,	80)	0	concatenate_97[0][0] dropout_101[0][0]

batch_normalization_102 (BatchN	 (None.	 2.	2.	 80)	320	
			,			
activation_102 (Activation)	(None,	2,	2,	80)	0	<pre>batch_normalization_102[0][0]</pre>
conv2d_106 (Conv2D)	(None,	2,	2,	8)	5760	activation_102[0][0]
dropout_102 (Dropout)	(None,	2,	2,	8)	0	conv2d_106[0][0]
concatenate_99 (Concatenate)	(None,	2,	2,	88)	0	concatenate_98[0][0] dropout_102[0][0]
batch_normalization_103 (BatchN	(None,	2,	2,	88)	352	concatenate_99[0][0]
activation_103 (Activation)	(None,	2,	2,	88)	0	batch_normalization_103[0][0]
conv2d_107 (Conv2D)	(None,	2,	2,	8)	6336	activation_103[0][0]
dropout_103 (Dropout)	(None,	2,	2,	8)	0	conv2d_107[0][0]
concatenate_100 (Concatenate)	(None,	2,	2,	96)	0	concatenate_99[0][0] dropout_103[0][0]
batch_normalization_104 (BatchN	(None,	2,	2,	96)	384	concatenate_100[0][0]
activation_104 (Activation)	(None,	2,	2,	96)	0	batch_normalization_104[0][0]
conv2d_108 (Conv2D)	(None,	2,	2,	8)	6912	activation_104[0][0]
dropout_104 (Dropout)	(None,	2,	2,	8)	0	conv2d_108[0][0]
concatenate_101 (Concatenate)	(None,	2,	2,	104)	0	concatenate_100[0][0] dropout_104[0][0]
batch_normalization_105 (BatchN	(None,	2,	2,	104)	416	concatenate_101[0][0]
activation_105 (Activation)	(None,	2,	2,	104)	0	batch_normalization_105[0][0]
conv2d_109 (Conv2D)	(None,	2,	2,	8)	7488	activation_105[0][0]
dropout_105 (Dropout)	(None,	2,	2,	8)	0	conv2d_109[0][0]
concatenate_102 (Concatenate)	(None,	2,	2,	112)	0	concatenate_101[0][0] dropout_105[0][0]
batch_normalization_106 (BatchN	(None,	2,	2,	112)	448	concatenate_102[0][0]
activation_106 (Activation)	(None,	2,	2,	112)	0	batch_normalization_106[0][0]

conv2d_110 (Conv2D)	(None,	2,	2,	8)	8064	activation_106[0][0]
dropout_106 (Dropout)	(None,	2,	2,	8)	0	conv2d_110[0][0]
concatenate_103 (Concatenate)	(None,	2,	2,	120)	0	concatenate_102[0][0] dropout_106[0][0]
batch_normalization_107 (BatchN	(None,	2,	2,	120)	480	concatenate_103[0][0]
activation_107 (Activation)	(None,	2,	2,	120)	0	batch_normalization_107[0][0]
conv2d_111 (Conv2D)	(None,	2,	2,	8)	8640	activation_107[0][0]
dropout_107 (Dropout)	(None,	2,	2,	8)	0	conv2d_111[0][0]
concatenate_104 (Concatenate)	(None,	2,	2,	128)	0	concatenate_103[0][0] dropout_107[0][0]
batch_normalization_108 (BatchN	(None,	2,	2,	128)	512	concatenate_104[0][0]
activation_108 (Activation)	(None,	2,	2,	128)	0	batch_normalization_108[0][0]
conv2d_112 (Conv2D)	(None,	2,	2,	8)	9216	activation_108[0][0]
dropout_108 (Dropout)	(None,	2,	2,	8)	0	conv2d_112[0][0]
concatenate_105 (Concatenate)	(None,	2,	2,	136)	0	concatenate_104[0][0] dropout_108[0][0]
batch_normalization_109 (BatchN	(None,	2,	2,	136)	544	concatenate_105[0][0]
activation_109 (Activation)	(None,	2,	2,	136)	0	batch_normalization_109[0][0]
conv2d_113 (Conv2D)	(None,	2,	2,	8)	9792	activation_109[0][0]
dropout_109 (Dropout)	(None,	2,	2,	8)	0	conv2d_113[0][0]
concatenate_106 (Concatenate)	(None,	2,	2,	144)	0	concatenate_105[0][0] dropout_109[0][0]
batch_normalization_110 (BatchN	(None,	2,	2,	144)	576	concatenate_106[0][0]
activation_110 (Activation)	(None,	2,	2,	144)	0	batch_normalization_110[0][0]
conv2d_114 (Conv2D)	(None,	2,	2,	8)	10368	activation_110[0][0]
dropout_110 (Dropout)	(None,	2,	2,	8)	0	conv2d_114[0][0]

concatenate_107 (Concatenate)	(None,	2,	2,	152)	0	concatenate_106[0][0] dropout_110[0][0]
batch_normalization_111 (BatchN	(None,	2,	2,	152)	608	concatenate_107[0][0]
activation_111 (Activation)	(None,	2,	2,	152)	0	batch_normalization_111[0][0]
conv2d_115 (Conv2D)	(None,	2,	2,	8)	10944	activation_111[0][0]
dropout_111 (Dropout)	(None,	2,	2,	8)	0	conv2d_115[0][0]
concatenate_108 (Concatenate)	(None,	2,	2,	160)	0	concatenate_107[0][0] dropout_111[0][0]
batch_normalization_112 (BatchN	(None,	2,	2,	160)	640	concatenate_108[0][0]
activation_112 (Activation)	(None,	2,	2,	160)	0	batch_normalization_112[0][0]
conv2d_116 (Conv2D)	(None,	2,	2,	8)	11520	activation_112[0][0]
dropout_112 (Dropout)	(None,	2,	2,	8)	0	conv2d_116[0][0]
concatenate_109 (Concatenate)	(None,	2,	2,	168)	0	concatenate_108[0][0] dropout_112[0][0]
batch_normalization_113 (BatchN	(None,	2,	2,	168)	672	concatenate_109[0][0]
activation_113 (Activation)	(None,	2,	2,	168)	0	batch_normalization_113[0][0]
conv2d_117 (Conv2D)	(None,	2,	2,	8)	12096	activation_113[0][0]
dropout_113 (Dropout)	(None,	2,	2,	8)	0	conv2d_117[0][0]
concatenate_110 (Concatenate)	(None,	2,	2,	176)	0	concatenate_109[0][0] dropout_113[0][0]
batch_normalization_114 (BatchN	(None,	2,	2,	176)	704	concatenate_110[0][0]
activation_114 (Activation)	(None,	2,	2,	176)	0	batch_normalization_114[0][0]
conv2d_118 (Conv2D)	(None,	2,	2,	8)	12672	activation_114[0][0]
dropout_114 (Dropout)	(None,	2,	2,	8)	0	conv2d_118[0][0]
concatenate_111 (Concatenate)	(None,	2,	2,	184)	0	concatenate_110[0][0] dropout_114[0][0]

batch_normalization_115 (BatchN	(None, 2, 2, 184)	736	concatenate_111[0][0]
activation_115 (Activation)	(None, 2, 2, 184)	0	batch_normalization_115[0][0]
average_pooling2d_5 (AveragePoo	(None, 1, 1, 184)	0	activation_115[0][0]
flatten_1 (Flatten)	(None, 184)	0	average_pooling2d_5[0][0]
dense_1 (Dense)	(None, 10)	1850	flatten_1[0][0]
Total params: 832,426			

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Total params: 832,426 Trainable params: 810,314 Non-trainable params: 22,112

0.10 Compile the model

0.11 Callbacks For Early Stopping and Saving the Best Model

```
In [13]: from keras.callbacks import Callback
         class EarlyStoppingByValidationAccuracy(Callback):
             def __init__(self, monitor='val_acc', value=1.0, verbose=0):
                 super(Callback, self).__init__()
                 self.monitor = monitor
                 self.value = value
                 self.verbose = verbose
             def on_epoch_end(self, epoch, logs={}):
                 accuracy_value= logs.get(self.monitor)
                 if accuracy_value >= self.value:
                     self.model.stop_training = True
                     if self.verbose == 1:
                         print("Epoch %d: Threshold for early stopping has reached" % (epoch +
In [14]: from keras.callbacks import ModelCheckpoint
         callbacks_list=[]
         model_save_path= "best_model-CIFAR10-monimoy-my_computer2.h5"
```

callbacks_list.append(EarlyStoppingByValidationAccuracy(monitor='val_acc', value=0.9callbacks_list.append(ModelCheckpoint(model_save_path, monitor='val_acc', verbose=1, save_path)

0.12 Fit the model on the train data and calculate metrics (Training Loss, Training Accuracy, Validation Losss, Validation Accuracy)

```
In [15]: model.fit(x_train, y_train,
                batch_size=batch_size,
                epochs=epochs,
                verbose=1.
                validation_data=(x_test, y_test),
                callbacks=callbacks list)
Train on 52000 samples, validate on 10000 samples
Epoch 1/250
Epoch 00001: val_acc improved from -inf to 0.20550, saving model to best_model-CIFAR10-monimoy
Epoch 2/250
Epoch 00002: val_acc improved from 0.20550 to 0.35920, saving model to best_model-CIFAR10-moning
Epoch 00003: val_acc improved from 0.35920 to 0.40760, saving model to best_model-CIFAR10-moning
Epoch 4/250
Epoch 00004: val acc did not improve from 0.40760
Epoch 5/250
Epoch 00005: val_acc improved from 0.40760 to 0.47370, saving model to best_model-CIFAR10-moning
Epoch 6/250
Epoch 00006: val_acc improved from 0.47370 to 0.48630, saving model to best_model-CIFAR10-moning
Epoch 7/250
Epoch 00007: val_acc improved from 0.48630 to 0.49210, saving model to best_model-CIFAR10-moning
Epoch 8/250
Epoch 00008: val_acc did not improve from 0.49210
Epoch 9/250
Epoch 00009: val_acc did not improve from 0.49210
Epoch 10/250
```

```
Epoch 00010: val_acc improved from 0.49210 to 0.49990, saving model to best_model-CIFAR10-moning
Epoch 11/250
Epoch 00011: val_acc did not improve from 0.49990
Epoch 12/250
Epoch 00012: val_acc did not improve from 0.49990
Epoch 13/250
Epoch 00013: val_acc improved from 0.49990 to 0.60330, saving model to best_model-CIFAR10-moning
Epoch 14/250
Epoch 00014: val_acc did not improve from 0.60330
Epoch 15/250
Epoch 00015: val_acc did not improve from 0.60330
Epoch 16/250
Epoch 00016: val_acc did not improve from 0.60330
Epoch 17/250
Epoch 00017: val_acc did not improve from 0.60330
Epoch 18/250
52000/52000 [============== ] - 310s 6ms/step - loss: 0.9599 - acc: 0.6580 - va
Epoch 00018: val_acc improved from 0.60330 to 0.65160, saving model to best_model-CIFAR10-moning
Epoch 19/250
Epoch 00019: val_acc did not improve from 0.65160
Epoch 20/250
Epoch 00020: val_acc did not improve from 0.65160
Epoch 21/250
Epoch 00021: val_acc improved from 0.65160 to 0.65300, saving model to best_model-CIFAR10-moning
Epoch 22/250
```

```
Epoch 00022: val_acc did not improve from 0.65300
Epoch 23/250
Epoch 00023: val_acc did not improve from 0.65300
Epoch 24/250
Epoch 00024: val_acc did not improve from 0.65300
Epoch 00025: val_acc improved from 0.65300 to 0.66890, saving model to best_model-CIFAR10-moning
Epoch 26/250
Epoch 00026: val_acc did not improve from 0.66890
Epoch 27/250
Epoch 00027: val_acc did not improve from 0.66890
Epoch 28/250
Epoch 00028: val_acc did not improve from 0.66890
Epoch 29/250
Epoch 00029: val_acc did not improve from 0.66890
Epoch 30/250
52000/52000 [============== ] - 311s 6ms/step - loss: 0.8165 - acc: 0.7087 - va
Epoch 00030: val_acc did not improve from 0.66890
Epoch 31/250
Epoch 00031: val_acc did not improve from 0.66890
Epoch 32/250
Epoch 00032: val_acc improved from 0.66890 to 0.68370, saving model to best_model-CIFAR10-moning
Epoch 33/250
Epoch 00033: val_acc improved from 0.68370 to 0.70340, saving model to best_model-CIFAR10-moning
```

Epoch 34/250

```
Epoch 00034: val_acc did not improve from 0.70340
Epoch 35/250
Epoch 00035: val_acc did not improve from 0.70340
Epoch 36/250
Epoch 00036: val_acc did not improve from 0.70340
Epoch 37/250
Epoch 00037: val_acc did not improve from 0.70340
Epoch 38/250
Epoch 00038: val_acc did not improve from 0.70340
Epoch 39/250
Epoch 00039: val_acc improved from 0.70340 to 0.72470, saving model to best_model-CIFAR10-moning
Epoch 40/250
Epoch 00040: val_acc did not improve from 0.72470
Epoch 41/250
Epoch 00041: val_acc did not improve from 0.72470
Epoch 42/250
52000/52000 [============== ] - 310s 6ms/step - loss: 0.7174 - acc: 0.7453 - va
Epoch 00042: val_acc did not improve from 0.72470
Epoch 43/250
Epoch 00043: val_acc did not improve from 0.72470
Epoch 44/250
Epoch 00044: val_acc did not improve from 0.72470
Epoch 45/250
Epoch 00045: val_acc did not improve from 0.72470
```

Epoch 46/250

```
Epoch 00046: val_acc improved from 0.72470 to 0.72540, saving model to best_model-CIFAR10-moning
Epoch 47/250
Epoch 00047: val_acc did not improve from 0.72540
Epoch 48/250
Epoch 00048: val_acc did not improve from 0.72540
Epoch 49/250
Epoch 00049: val_acc improved from 0.72540 to 0.73430, saving model to best_model-CIFAR10-moning
Epoch 50/250
Epoch 00050: val_acc did not improve from 0.73430
Epoch 51/250
Epoch 00051: val_acc did not improve from 0.73430
Epoch 52/250
Epoch 00052: val_acc did not improve from 0.73430
Epoch 53/250
Epoch 00053: val_acc improved from 0.73430 to 0.74760, saving model to best_model-CIFAR10-moning
Epoch 54/250
52000/52000 [============== ] - 310s 6ms/step - loss: 0.6439 - acc: 0.7734 - va
Epoch 00054: val_acc did not improve from 0.74760
Epoch 55/250
Epoch 00055: val_acc improved from 0.74760 to 0.75290, saving model to best_model-CIFAR10-moning
Epoch 56/250
Epoch 00056: val_acc did not improve from 0.75290
Epoch 57/250
Epoch 00057: val_acc did not improve from 0.75290
```

Epoch 58/250

```
Epoch 00058: val_acc improved from 0.75290 to 0.75320, saving model to best_model-CIFAR10-moning
Epoch 59/250
Epoch 00059: val_acc did not improve from 0.75320
Epoch 60/250
Epoch 00060: val_acc did not improve from 0.75320
Epoch 61/250
Epoch 00061: val_acc did not improve from 0.75320
Epoch 62/250
Epoch 00062: val_acc did not improve from 0.75320
Epoch 63/250
Epoch 00063: val_acc improved from 0.75320 to 0.76500, saving model to best_model-CIFAR10-moning
Epoch 64/250
Epoch 00064: val_acc did not improve from 0.76500
Epoch 65/250
Epoch 00065: val_acc did not improve from 0.76500
Epoch 66/250
52000/52000 [============== ] - 310s 6ms/step - loss: 0.5863 - acc: 0.7948 - va
Epoch 00066: val_acc improved from 0.76500 to 0.76610, saving model to best_model-CIFAR10-moning
Epoch 67/250
Epoch 00067: val_acc did not improve from 0.76610
Epoch 68/250
Epoch 00068: val_acc improved from 0.76610 to 0.77680, saving model to best_model-CIFAR10-moning
Epoch 69/250
Epoch 00069: val_acc did not improve from 0.77680
```

Epoch 70/250

```
52000/52000 [=============== ] - 310s 6ms/step - loss: 0.5640 - acc: 0.8064 - va
Epoch 00070: val_acc did not improve from 0.77680
Epoch 71/250
Epoch 00071: val_acc did not improve from 0.77680
Epoch 72/250
Epoch 00072: val_acc did not improve from 0.77680
Epoch 73/250
Epoch 00073: val_acc improved from 0.77680 to 0.79500, saving model to best_model-CIFAR10-moning
Epoch 74/250
Epoch 00074: val_acc did not improve from 0.79500
Epoch 75/250
Epoch 00075: val_acc did not improve from 0.79500
Epoch 76/250
Epoch 00076: val_acc did not improve from 0.79500
Epoch 77/250
Epoch 00077: val_acc did not improve from 0.79500
Epoch 78/250
52000/52000 [============== ] - 310s 6ms/step - loss: 0.5291 - acc: 0.8154 - va
Epoch 00078: val_acc did not improve from 0.79500
Epoch 79/250
Epoch 00079: val_acc did not improve from 0.79500
Epoch 80/250
Epoch 00080: val_acc did not improve from 0.79500
Epoch 81/250
Epoch 00081: val_acc did not improve from 0.79500
```

Epoch 82/250

```
Epoch 00082: val_acc did not improve from 0.79500
Epoch 00083: val_acc did not improve from 0.79500
Epoch 84/250
52000/52000 [=============== ] - 310s 6ms/step - loss: 0.5162 - acc: 0.8194 - va
Epoch 00084: val_acc did not improve from 0.79500
Epoch 85/250
Epoch 00085: val_acc did not improve from 0.79500
Epoch 86/250
Epoch 00086: val_acc did not improve from 0.79500
Epoch 87/250
Epoch 00087: val_acc did not improve from 0.79500
Epoch 88/250
Epoch 00088: val_acc did not improve from 0.79500
Epoch 89/250
Epoch 00089: val_acc did not improve from 0.79500
Epoch 90/250
52000/52000 [============== ] - 309s 6ms/step - loss: 0.4958 - acc: 0.8270 - va
Epoch 00090: val_acc did not improve from 0.79500
Epoch 91/250
Epoch 00091: val_acc did not improve from 0.79500
Epoch 92/250
52000/52000 [=============== ] - 310s 6ms/step - loss: 0.4809 - acc: 0.8324 - va
Epoch 00092: val_acc did not improve from 0.79500
Epoch 93/250
Epoch 00093: val_acc improved from 0.79500 to 0.80770, saving model to best_model-CIFAR10-moning
```

Epoch 94/250

```
Epoch 00094: val_acc did not improve from 0.80770
Epoch 00095: val_acc did not improve from 0.80770
Epoch 96/250
52000/52000 [=============== ] - 310s 6ms/step - loss: 0.4777 - acc: 0.8359 - va
Epoch 00096: val_acc did not improve from 0.80770
Epoch 00097: val_acc did not improve from 0.80770
Epoch 98/250
Epoch 00098: val_acc did not improve from 0.80770
Epoch 99/250
Epoch 00099: val_acc did not improve from 0.80770
Epoch 100/250
Epoch 00100: val_acc did not improve from 0.80770
Epoch 101/250
Epoch 00101: val_acc did not improve from 0.80770
Epoch 102/250
52000/52000 [============== ] - 310s 6ms/step - loss: 0.4626 - acc: 0.8388 - va
Epoch 00102: val_acc did not improve from 0.80770
Epoch 103/250
Epoch 00103: val_acc did not improve from 0.80770
Epoch 104/250
Epoch 00104: val_acc did not improve from 0.80770
Epoch 105/250
Epoch 00105: val_acc did not improve from 0.80770
```

Epoch 106/250

```
52000/52000 [============== ] - 313s 6ms/step - loss: 0.4494 - acc: 0.8440 - va
Epoch 00106: val_acc did not improve from 0.80770
Epoch 107/250
Epoch 00107: val_acc did not improve from 0.80770
Epoch 108/250
Epoch 00108: val_acc did not improve from 0.80770
Epoch 109/250
Epoch 00109: val_acc did not improve from 0.80770
Epoch 110/250
Epoch 00110: val_acc did not improve from 0.80770
Epoch 111/250
Epoch 00111: val_acc did not improve from 0.80770
Epoch 112/250
Epoch 00112: val_acc did not improve from 0.80770
Epoch 113/250
Epoch 00113: val_acc did not improve from 0.80770
Epoch 114/250
52000/52000 [============== ] - 313s 6ms/step - loss: 0.4377 - acc: 0.8504 - va
Epoch 00114: val_acc did not improve from 0.80770
Epoch 115/250
Epoch 00115: val_acc did not improve from 0.80770
Epoch 116/250
Epoch 00116: val_acc did not improve from 0.80770
Epoch 117/250
Epoch 00117: val_acc did not improve from 0.80770
```

Epoch 118/250

```
52000/52000 [=============== ] - 313s 6ms/step - loss: 0.4216 - acc: 0.8542 - va
Epoch 00118: val_acc did not improve from 0.80770
Epoch 119/250
Epoch 00119: val_acc did not improve from 0.80770
Epoch 120/250
Epoch 00120: val_acc did not improve from 0.80770
Epoch 121/250
Epoch 00121: val_acc did not improve from 0.80770
Epoch 122/250
Epoch 00122: val_acc did not improve from 0.80770
Epoch 123/250
Epoch 00123: val_acc did not improve from 0.80770
Epoch 124/250
Epoch 00124: val_acc did not improve from 0.80770
Epoch 125/250
Epoch 00125: val_acc did not improve from 0.80770
Epoch 126/250
52000/52000 [============== ] - 313s 6ms/step - loss: 0.4029 - acc: 0.8596 - va
Epoch 00126: val_acc improved from 0.80770 to 0.81100, saving model to best_model-CIFAR10-moning
Epoch 127/250
Epoch 00127: val_acc did not improve from 0.81100
Epoch 128/250
Epoch 00128: val_acc did not improve from 0.81100
Epoch 129/250
Epoch 00129: val_acc did not improve from 0.81100
```

Epoch 130/250

```
Epoch 00130: val_acc did not improve from 0.81100
Epoch 131/250
52000/52000 [============== ] - 311s 6ms/step - loss: 0.3961 - acc: 0.8618 - va
Epoch 00131: val_acc improved from 0.81100 to 0.81210, saving model to best_model-CIFAR10-moning
Epoch 132/250
Epoch 00132: val_acc did not improve from 0.81210
Epoch 133/250
Epoch 00133: val_acc did not improve from 0.81210
Epoch 134/250
Epoch 00134: val_acc improved from 0.81210 to 0.81610, saving model to best_model-CIFAR10-moning
Epoch 135/250
Epoch 00135: val_acc did not improve from 0.81610
Epoch 136/250
Epoch 00136: val_acc did not improve from 0.81610
Epoch 137/250
Epoch 00137: val_acc did not improve from 0.81610
Epoch 138/250
52000/52000 [============== ] - 312s 6ms/step - loss: 0.3822 - acc: 0.8679 - va
Epoch 00138: val_acc improved from 0.81610 to 0.82030, saving model to best_model-CIFAR10-moning
Epoch 139/250
Epoch 00139: val_acc did not improve from 0.82030
Epoch 140/250
Epoch 00140: val_acc did not improve from 0.82030
Epoch 141/250
Epoch 00141: val_acc did not improve from 0.82030
```

Epoch 142/250

```
Epoch 00142: val_acc did not improve from 0.82030
Epoch 143/250
Epoch 00143: val_acc did not improve from 0.82030
Epoch 144/250
Epoch 00144: val_acc improved from 0.82030 to 0.82210, saving model to best_model-CIFAR10-moning
Epoch 145/250
Epoch 00145: val_acc did not improve from 0.82210
Epoch 146/250
Epoch 00146: val_acc did not improve from 0.82210
Epoch 147/250
Epoch 00147: val_acc did not improve from 0.82210
Epoch 148/250
Epoch 00148: val_acc did not improve from 0.82210
Epoch 149/250
Epoch 00149: val_acc did not improve from 0.82210
Epoch 150/250
Epoch 00150: val_acc did not improve from 0.82210
Epoch 151/250
Epoch 00151: val_acc did not improve from 0.82210
Epoch 152/250
Epoch 00152: val_acc did not improve from 0.82210
Epoch 153/250
Epoch 00153: val_acc did not improve from 0.82210
```

Epoch 154/250

```
Epoch 00154: val_acc did not improve from 0.82210
Epoch 155/250
Epoch 00155: val_acc did not improve from 0.82210
Epoch 156/250
Epoch 00156: val_acc did not improve from 0.82210
Epoch 157/250
Epoch 00157: val_acc did not improve from 0.82210
Epoch 158/250
Epoch 00158: val_acc did not improve from 0.82210
Epoch 159/250
Epoch 00159: val_acc improved from 0.82210 to 0.82780, saving model to best_model-CIFAR10-moning
Epoch 160/250
Epoch 00160: val_acc did not improve from 0.82780
Epoch 161/250
Epoch 00161: val_acc did not improve from 0.82780
Epoch 162/250
52000/52000 [============== ] - 308s 6ms/step - loss: 0.3493 - acc: 0.8785 - va
Epoch 00162: val_acc did not improve from 0.82780
Epoch 163/250
Epoch 00163: val_acc did not improve from 0.82780
Epoch 164/250
Epoch 00164: val_acc did not improve from 0.82780
Epoch 165/250
Epoch 00165: val_acc did not improve from 0.82780
```

Epoch 166/250

```
Epoch 00166: val_acc did not improve from 0.82780
Epoch 167/250
Epoch 00167: val_acc did not improve from 0.82780
Epoch 168/250
Epoch 00168: val_acc did not improve from 0.82780
Epoch 169/250
Epoch 00169: val_acc did not improve from 0.82780
Epoch 170/250
Epoch 00170: val_acc did not improve from 0.82780
Epoch 171/250
Epoch 00171: val_acc did not improve from 0.82780
Epoch 172/250
Epoch 00172: val_acc did not improve from 0.82780
Epoch 173/250
Epoch 00173: val_acc did not improve from 0.82780
Epoch 174/250
52000/52000 [============== ] - 316s 6ms/step - loss: 0.3249 - acc: 0.8865 - va
Epoch 00174: val_acc did not improve from 0.82780
Epoch 175/250
Epoch 00175: val_acc did not improve from 0.82780
Epoch 176/250
Epoch 00176: val_acc did not improve from 0.82780
Epoch 177/250
Epoch 00177: val_acc did not improve from 0.82780
```

Epoch 178/250

```
Epoch 00178: val_acc did not improve from 0.82780
Epoch 179/250
Epoch 00179: val_acc did not improve from 0.82780
Epoch 180/250
Epoch 00180: val_acc did not improve from 0.82780
Epoch 181/250
Epoch 00181: val_acc did not improve from 0.82780
Epoch 182/250
Epoch 00182: val_acc did not improve from 0.82780
Epoch 183/250
Epoch 00183: val_acc did not improve from 0.82780
Epoch 184/250
Epoch 00184: val_acc did not improve from 0.82780
Epoch 185/250
Epoch 00185: val_acc did not improve from 0.82780
Epoch 186/250
52000/52000 [============== ] - 1421s 27ms/step - loss: 0.3084 - acc: 0.8923 -
Epoch 00186: val_acc did not improve from 0.82780
Epoch 187/250
Epoch 00187: val_acc did not improve from 0.82780
Epoch 188/250
Epoch 00188: val_acc did not improve from 0.82780
Epoch 189/250
Epoch 00189: val_acc did not improve from 0.82780
```

Epoch 190/250

```
Epoch 00190: val_acc improved from 0.82780 to 0.83050, saving model to best_model-CIFAR10-moning
Epoch 191/250
Epoch 00191: val_acc did not improve from 0.83050
Epoch 192/250
52000/52000 [=============== ] - 1422s 27ms/step - loss: 0.3052 - acc: 0.8934 -
Epoch 00192: val_acc did not improve from 0.83050
Epoch 193/250
Epoch 00193: val_acc did not improve from 0.83050
Epoch 194/250
Epoch 00194: val_acc improved from 0.83050 to 0.83240, saving model to best_model-CIFAR10-moning
Epoch 195/250
Epoch 00195: val_acc did not improve from 0.83240
Epoch 196/250
Epoch 00196: val_acc did not improve from 0.83240
Epoch 197/250
Epoch 00197: val_acc did not improve from 0.83240
Epoch 198/250
52000/52000 [============== ] - 1418s 27ms/step - loss: 0.2978 - acc: 0.8964 -
Epoch 00198: val_acc did not improve from 0.83240
Epoch 199/250
Epoch 00199: val_acc did not improve from 0.83240
Epoch 200/250
Epoch 00200: val_acc did not improve from 0.83240
Epoch 201/250
Epoch 00201: val_acc did not improve from 0.83240
```

Epoch 202/250

```
Epoch 00202: val_acc did not improve from 0.83240
Epoch 203/250
Epoch 00203: val_acc did not improve from 0.83240
Epoch 204/250
Epoch 00204: val_acc did not improve from 0.83240
Epoch 205/250
Epoch 00205: val_acc did not improve from 0.83240
Epoch 206/250
Epoch 00206: val_acc did not improve from 0.83240
Epoch 207/250
Epoch 00207: val_acc did not improve from 0.83240
Epoch 208/250
Epoch 00208: val_acc did not improve from 0.83240
Epoch 209/250
Epoch 00209: val_acc did not improve from 0.83240
Epoch 210/250
52000/52000 [============== ] - 1413s 27ms/step - loss: 0.2861 - acc: 0.8987 -
Epoch 00210: val_acc did not improve from 0.83240
Epoch 211/250
Epoch 00211: val_acc did not improve from 0.83240
Epoch 212/250
Epoch 00212: val_acc improved from 0.83240 to 0.83460, saving model to best_model-CIFAR10-moning
Epoch 213/250
Epoch 00213: val_acc did not improve from 0.83460
```

Epoch 214/250

```
Epoch 00214: val_acc did not improve from 0.83460
Epoch 215/250
52000/52000 [============== ] - 1411s 27ms/step - loss: 0.2817 - acc: 0.9014 -
Epoch 00215: val_acc did not improve from 0.83460
Epoch 216/250
Epoch 00216: val_acc did not improve from 0.83460
Epoch 217/250
Epoch 00217: val_acc improved from 0.83460 to 0.83800, saving model to best_model-CIFAR10-moning
Epoch 218/250
Epoch 00218: val_acc did not improve from 0.83800
Epoch 219/250
Epoch 00219: val_acc did not improve from 0.83800
Epoch 220/250
Epoch 00220: val_acc did not improve from 0.83800
Epoch 221/250
Epoch 00221: val_acc improved from 0.83800 to 0.83910, saving model to best_model-CIFAR10-moning
Epoch 222/250
52000/52000 [============== ] - 1394s 27ms/step - loss: 0.2774 - acc: 0.9022 -
Epoch 00222: val acc did not improve from 0.83910
Epoch 223/250
Epoch 00223: val_acc improved from 0.83910 to 0.84350, saving model to best_model-CIFAR10-moning
Epoch 224/250
Epoch 00224: val_acc did not improve from 0.84350
Epoch 225/250
Epoch 00225: val_acc did not improve from 0.84350
```

Epoch 226/250

```
Epoch 00226: val_acc did not improve from 0.84350
Epoch 227/250
Epoch 00227: val_acc did not improve from 0.84350
Epoch 228/250
Epoch 00228: val_acc did not improve from 0.84350
Epoch 229/250
Epoch 00229: val_acc did not improve from 0.84350
Epoch 230/250
Epoch 00230: val_acc did not improve from 0.84350
Epoch 231/250
Epoch 00231: val_acc did not improve from 0.84350
Epoch 232/250
Epoch 00232: val_acc did not improve from 0.84350
Epoch 233/250
Epoch 00233: val_acc did not improve from 0.84350
Epoch 234/250
Epoch 00234: val_acc improved from 0.84350 to 0.84530, saving model to best_model-CIFAR10-moning
Epoch 235/250
Epoch 00235: val_acc did not improve from 0.84530
Epoch 236/250
Epoch 00236: val_acc did not improve from 0.84530
Epoch 237/250
Epoch 00237: val_acc did not improve from 0.84530
```

Epoch 238/250

```
Epoch 00238: val_acc did not improve from 0.84530
Epoch 239/250
Epoch 00239: val_acc did not improve from 0.84530
Epoch 240/250
Epoch 00240: val_acc did not improve from 0.84530
Epoch 241/250
Epoch 00241: val_acc did not improve from 0.84530
Epoch 242/250
Epoch 00242: val_acc did not improve from 0.84530
Epoch 243/250
Epoch 00243: val_acc did not improve from 0.84530
Epoch 244/250
Epoch 00244: val_acc did not improve from 0.84530
Epoch 245/250
Epoch 00245: val_acc did not improve from 0.84530
Epoch 246/250
52000/52000 [============== ] - 1403s 27ms/step - loss: 0.2515 - acc: 0.9132 -
Epoch 00246: val_acc did not improve from 0.84530
Epoch 247/250
Epoch 00247: val_acc did not improve from 0.84530
Epoch 248/250
Epoch 00248: val_acc did not improve from 0.84530
Epoch 249/250
Epoch 00249: val_acc did not improve from 0.84530
```

Epoch 250/250

0.13 Conclusion:

On the CIFAR image dataset using Keras based Convolution Neural Network I have achieved testing accuracy of 84.51%