DENSE NET

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

```
In [1]:
```

```
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import cv2
```

1. Load dataset

```
In [2]:
```

```
(X_train, y_train), (X_test, y_test) = tf.keras.datasets.cifar10.load_data()
print(X_train.shape)
print(y_train.shape)
print(X_test.shape)
print(y_test.shape)

(50000, 32, 32, 3)
(50000, 1)
(10000, 32, 32, 3)
(10000, 1)
```

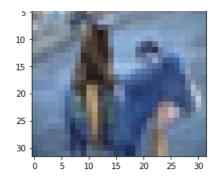
2. Look at some random images

```
In [3]:
```

```
random_num = np.random.randint(0, len(X_train))

plt.figure(figsize=(4,4))
plt.imshow(cv2.cvtColor(X_train[random_num], cv2.COLOR_BGR2RGB))
plt.title(y_train[random_num])
plt.show()

/home/ubuntu/anaconda3/envs/tensorflow2_p36/lib/python3.6/site-packages/matplotlib/text.py:1150: F
utureWarning: elementwise comparison failed; returning scalar instead, but in the future will perf
orm elementwise comparison
   if s != self._text:
```



3. Normalise the data

```
In [4]:
```

```
X_train = X_train/255.0
X_test = X_test/255.0
```

4. Reshaping the data

```
In [5]:
```

```
img_height = 32
img_width = 32
channels = 3
```

In [6]:

```
import tensorflow.keras.backend as K

if K.image_data_format() == 'channels_first':
    X_train = X_train.reshape(X_train.shape[0], channels, img_height, img_width).astype('float32')
    X_test = X_test.reshape(X_test.shape[0], channels, img_height, img_width).astype('float32')
    input_size = (channels, img_height, img_width)

else:
    X_train = X_train.reshape(X_train.shape[0], img_height, img_width, channels).astype('float32')
    X_test = X_test.reshape(X_test.shape[0], img_height, img_width, channels).astype('float32')
    input_size = (img_height, img_width, channels)
```

In [7]:

```
print(X_train.shape)
print(y_train.shape)
print(X_test.shape)
print(y_test.shape)
print(input_size)

(50000, 32, 32, 3)
(50000, 1)
(10000, 32, 32, 3)
(10000, 1)
(32, 32, 3)
```

5. Convert y to 10 classes

In [8]:

```
#convert y to 10 categories
y_train = tf.keras.utils.to_categorical(y_train, num_classes=10)
y_test = tf.keras.utils.to_categorical(y_test, num_classes=10)
print(y_train.shape)
print(y_test.shape)
```

```
(50000, 10)
(10000, 10)
```

6. Data Augmentation

In [9]:

7. Model- Architecture

- 1. Dense Block
- 2. Transition Block
- 3. Output layer ### Refer: https://www.pluralsight.com/guides/introduction-to-densenet-with-tensorflow

7.1 Dense Block

https://images.app.goo.gl/VDzxZGQUEL7kt5N58

```
In [ ]:
```

```
from tensorflow.keras import layers
from tensorflow.keras.models import Model
num_classes = 10
```

In [11]:

```
#BN-->ReLU-->Conv2D-->Dropout-->concat(input, output)-->(put in loop)
def denseblock(input, num filter, dropout rate):
                  global compression # to keep the growth rate of number of filters
                  temp = input
                  for in range(l):
                                     BatchNorm = layers.BatchNormalization()(temp)
                                     relu = layers.Activation('relu')(BatchNorm)
                                    \texttt{Conv2D\_3\_3} = \texttt{layers.Conv2D(int(num\_filter*compression), (3,3), use\_bias} = \textbf{False , padding='samble filter*compression'}, \textbf{(3,3), use\_bias} = \textbf{(3,3), use\_bias
e')(relu)
                                    if dropout rate>0:
                                                        Conv2D_3_3 = layers.Dropout(dropout_rate)(Conv2D_3_3)
                                     \#concat the input(temp) and output(conv2d 3 3) , in resnet we add but here we concat
                                     concat = layers.Concatenate(axis=-1)([temp,Conv2D 3 3])
                                     #change the concat as input
                                     temp = concat
                  return temp
```

7.2 Transition Block

- https://images.app.goo.gl/7ETwBQqHKwQwXPH26
- Conv 1x1 is the bottle neck layer here like in resnet

```
In [12]:
```

```
#BN-->relu-->conv2d(1x1) -->dropout-->avg_pool
def transition(input, num_filter, dropout_rate):
    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
```

7.3 Output layer

```
In [13]:
```

```
#BN-->relu-->avgpool-->flat-->softmax
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
```

7.4 Full Architecture

In [14]:

```
# Hyperparameters
1 = 12
num_filter = 12
compression = 0.5
dropout_rate = 0.2
num_classes = 10
```

In [15]:

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

#First dense and transition block
First_Block = denseblock(First_Conv2D, num_filter, dropout_rate)
First_Transition = transition (First_Block, num_filter, dropout_rate)

#Second dense and transition block
Second_Block = denseblock(First_Transition, num_filter, dropout_rate)

#Third dense and transition block
Third_Block = denseblock(Second_Transition, num_filter, dropout_rate)

#Third_Transition = transition(Third_Block, num_filter, dropout_rate)

#last dense and output block
Last_Block = denseblock(Third_Transition, num_filter, dropout_rate)
output = output_layer(Last_Block)
```

In [16]:

```
model = tf.keras.models.Model(inputs=[input], outputs=[output])
model.summary()
```

Model: "model"

input_1 (InputLayer)	[(None,				0	
conv2d (Conv2D)	(None,	32,	32,	12)	324	input_1[0][0]
batch_normalization (BatchNorma	(None,	32,	32,	12)	48	conv2d[0][0]
activation (Activation)	(None,	32,	32,	12)	0	batch_normalization[0][0]
conv2d_1 (Conv2D)	(None,	32,	32,	6)	648	activation[0][0]
dropout (Dropout)	(None,	32,	32,	6)	0	conv2d_1[0][0]
concatenate (Concatenate)	(None,	32,	32,	18)	0	conv2d[0][0] dropout[0][0]
batch_normalization_1 (BatchNor	(None,	32,	32,	18)	72	concatenate[0][0]
activation_1 (Activation)	(None,	32,	32,	18)	0	batch_normalization_1[0][0]
conv2d_2 (Conv2D)	(None,	32,	32,	6)	972	activation_1[0][0]
dropout_1 (Dropout)	(None,	32,	32,	6)	0	conv2d_2[0][0]
concatenate_1 (Concatenate)	(None,	32,	32,	24)	0	<pre>concatenate[0][0] dropout_1[0][0]</pre>
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,	6)	1296	activation_2[0][0]
dropout_2 (Dropout)	(None,	32,	32,	6)	0	conv2d_3[0][0]
concatenate_2 (Concatenate)	(None,	32,	32,	30)	0	concatenate_1[0][0] dropout_2[0][0]
batch_normalization_3 (BatchNor	(None,	32,	32,	30)	120	concatenate_2[0][0]
activation_3 (Activation)	(None,	32,	32,	30)	0	batch_normalization_3[0][0]
conv2d_4 (Conv2D)	(None,	32,	32,	6)	1620	activation_3[0][0]
dropout_3 (Dropout)	(None,	32,	32,	6)	0	conv2d_4[0][0]
concatenate_3 (Concatenate)	(None,	32,	32,	36)	0	concatenate_2[0][0] dropout_3[0][0]
batch_normalization_4 (BatchNor	(None,	32,	32,	36)	144	concatenate_3[0][0]
activation_4 (Activation)	(None,	32,	32,	36)	0	batch_normalization_4[0][0]
conv2d_5 (Conv2D)	(None,	32,	32,	6)	1944	activation_4[0][0]
dropout_4 (Dropout)	(None,	32,	32,	6)	0	conv2d_5[0][0]
concatenate_4 (Concatenate)	(None,	32,	32,	42)	0	concatenate_3[0][0] dropout_4[0][0]
batch_normalization_5 (BatchNor	(None,	32,	32,	42)	168	concatenate_4[0][0]
activation_5 (Activation)	(None,	32,	32,	42)	0	batch_normalization_5[0][0]
conv2d_6 (Conv2D)	(None,	32,	32,	6)	2268	activation_5[0][0]
dropout_5 (Dropout)	(None,	32,	32,	6)	0	conv2d_6[0][0]
concatenate_5 (Concatenate)	(None,	32,	32,	48)	0	concatenate_4[0][0] dropout_5[0][0]
batch_normalization_6 (BatchNor	(None,	32,	32,	48)	192	concatenate_5[0][0]
activation_6 (Activation)	(None,	32,	32,	48)	0	batch_normalization_6[0][0]
conv2d_7 (Conv2D)	(None,	32,	32,	6)	2592	activation_6[0][0]

dropout_6 (Dropout)	(None,	32,	32,	6)	0	conv2d_7[0][0]
concatenate_6 (Concatenate)	(None,	32,	32,	54)	0	concatenate_5[0][0] dropout_6[0][0]
batch_normalization_7 (BatchNor	(None,	32,	32,	54)	216	concatenate_6[0][0]
activation_7 (Activation)	(None,	32,	32,	54)	0	batch_normalization_7[0][0]
conv2d_8 (Conv2D)	(None,	32,	32,	6)	2916	activation_7[0][0]
dropout_7 (Dropout)	(None,	32,	32,	6)	0	conv2d_8[0][0]
concatenate_7 (Concatenate)	(None,	32,	32,	60)	0	concatenate_6[0][0] dropout_7[0][0]
batch_normalization_8 (BatchNor	(None,	32,	32,	60)	240	concatenate_7[0][0]
activation_8 (Activation)	(None,	32,	32,	60)	0	batch_normalization_8[0][0]
conv2d_9 (Conv2D)	(None,	32,	32,	6)	3240	activation_8[0][0]
dropout_8 (Dropout)	(None,	32,	32,	6)	0	conv2d_9[0][0]
concatenate_8 (Concatenate)	(None,	32,	32,	66)	0	concatenate_7[0][0] dropout_8[0][0]
batch_normalization_9 (BatchNor	(None,	32,	32,	66)	264	concatenate_8[0][0]
activation_9 (Activation)	(None,	32,	32,	66)	0	batch_normalization_9[0][0]
conv2d_10 (Conv2D)	(None,	32,	32,	6)	3564	activation_9[0][0]
dropout_9 (Dropout)	(None,	32,	32,	6)	0	conv2d_10[0][0]
concatenate_9 (Concatenate)	(None,	32,	32,	72)	0	concatenate_8[0][0] dropout_9[0][0]
batch_normalization_10 (BatchNo	(None,	32,	32,	72)	288	concatenate_9[0][0]
activation_10 (Activation)	(None,	32,	32,	72)	0	batch_normalization_10[0][0]
conv2d_11 (Conv2D)	(None,	32,	32,	6)	3888	activation_10[0][0]
dropout_10 (Dropout)	(None,	32,	32,	6)	0	conv2d_11[0][0]
concatenate_10 (Concatenate)	(None,	32,	32,	78)	0	concatenate_9[0][0] dropout_10[0][0]
batch_normalization_11 (BatchNo	(None,	32,	32,	78)	312	concatenate_10[0][0]
activation_11 (Activation)	(None,	32,	32,	78)	0	batch_normalization_11[0][0]
conv2d_12 (Conv2D)	(None,	32,	32,	6)	4212	activation_11[0][0]
dropout_11 (Dropout)	(None,	32,	32,	6)	0	conv2d_12[0][0]
concatenate_11 (Concatenate)	(None,	32,	32,	84)	0	concatenate_10[0][0] dropout_11[0][0]
batch_normalization_12 (BatchNo	(None,	32,	32,	84)	336	concatenate_11[0][0]
activation_12 (Activation)	(None,	32,	32,	84)	0	batch_normalization_12[0][0]
conv2d_13 (Conv2D)	(None,	32,	32,	6)	504	activation_12[0][0]
dropout_12 (Dropout)	(None,	32,	32,	6)	0	conv2d_13[0][0]
average_pooling2d (AveragePooli	(None,	16,	16,	6)	0	dropout_12[0][0]
batch_normalization_13 (BatchNo	(None,	16,	16,	6)	24	average_pooling2d[0][0]
activation_13 (Activation)	(None,	16,	16,	6)	0	batch_normalization_13[0][0]
conv2d_14 (Conv2D)	(None,	16,	16,	6)	324	activation_13[0][0]
dropout_13 (Dropout)	(None,	16,	16,	6)	0	conv2d_14[0][0]

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concatenate_12 (Concatenate)	(None,	16,	16,	12)	0	average_pooling2d[0][0] dropout_13[0][0]
batch_normalization_14 (BatchNo	(None,	16,	16,	12)	48	concatenate_12[0][0]
activation_14 (Activation)	(None,	16,	16,	12)	0	batch_normalization_14[0][0]
conv2d_15 (Conv2D)	(None,	16,	16,	6)	648	activation_14[0][0]
dropout_14 (Dropout)	(None,	16,	16,	6)	0	conv2d_15[0][0]
concatenate_13 (Concatenate)	(None,	16,	16,	18)	0	concatenate_12[0][0] dropout_14[0][0]
batch_normalization_15 (BatchNo	(None,	16,	16,	18)	72	concatenate_13[0][0]
activation_15 (Activation)	(None,	16,	16,	18)	0	batch_normalization_15[0][0]
conv2d_16 (Conv2D)	(None,	16,	16,	6)	972	activation_15[0][0]
dropout_15 (Dropout)	(None,	16,	16,	6)	0	conv2d_16[0][0]
concatenate_14 (Concatenate)	(None,	16,	16,	24)	0	concatenate_13[0][0] dropout_15[0][0]
batch_normalization_16 (BatchNo	(None,	16,	16,	24)	96	concatenate_14[0][0]
activation_16 (Activation)	(None,	16,	16,	24)	0	batch_normalization_16[0][0]
conv2d_17 (Conv2D)	(None,	16,	16,	6)	1296	activation_16[0][0]
dropout_16 (Dropout)	(None,	16,	16,	6)	0	conv2d_17[0][0]
concatenate_15 (Concatenate)	(None,	16,	16,	30)	0	concatenate_14[0][0] dropout_16[0][0]
batch_normalization_17 (BatchNo	(None,	16,	16,	30)	120	concatenate_15[0][0]
activation_17 (Activation)	(None,	16,	16,	30)	0	batch_normalization_17[0][0]
conv2d_18 (Conv2D)	(None,	16,	16,	6)	1620	activation_17[0][0]
dropout_17 (Dropout)	(None,	16,	16,	6)	0	conv2d_18[0][0]
concatenate_16 (Concatenate)	(None,	16,	16,	36)	0	concatenate_15[0][0] dropout_17[0][0]
batch_normalization_18 (BatchNo	(None,	16,	16,	36)	144	concatenate_16[0][0]
activation_18 (Activation)	(None,	16,	16,	36)	0	batch_normalization_18[0][0]
conv2d_19 (Conv2D)	(None,	16,	16,	6)	1944	activation_18[0][0]
dropout_18 (Dropout)	(None,	16,	16,	6)	0	conv2d_19[0][0]
concatenate_17 (Concatenate)	(None,	16,	16,	42)	0	<pre>concatenate_16[0][0] dropout_18[0][0]</pre>
batch_normalization_19 (BatchNo	(None,	16,	16,	42)	168	concatenate_17[0][0]
activation_19 (Activation)	(None,	16,	16,	42)	0	batch_normalization_19[0][0]
conv2d_20 (Conv2D)	(None,	16,	16,	6)	2268	activation_19[0][0]
dropout_19 (Dropout)	(None,	16,	16,	6)	0	conv2d_20[0][0]
concatenate_18 (Concatenate)	(None,	16,	16,	48)	0	concatenate_17[0][0] dropout_19[0][0]
batch_normalization_20 (BatchNo	(None,	16,	16,	48)	192	concatenate_18[0][0]
activation_20 (Activation)	(None,	16,	16,	48)	0	batch_normalization_20[0][0]
conv2d_21 (Conv2D)	(None,	16,	16,	6)	2592	activation_20[0][0]
dropout 20 (Dropout)	(None,	16,	16,	6)	0	conv2d 21[0][0]

concatenate_19 (Concatenate)	(None, 16, 16, 54)	0	concatenate_18[0][0] dropout_20[0][0]
batch_normalization_21 (BatchNo	(None, 16, 16, 54)	216	concatenate_19[0][0]
activation_21 (Activation)	(None, 16, 16, 54)	0	batch_normalization_21[0][0]
conv2d_22 (Conv2D)	(None, 16, 16, 6)	2916	activation_21[0][0]
dropout_21 (Dropout)	(None, 16, 16, 6)	0	conv2d_22[0][0]
concatenate_20 (Concatenate)	(None, 16, 16, 60)	0	concatenate_19[0][0] dropout_21[0][0]
batch_normalization_22 (BatchNo	(None, 16, 16, 60)	240	concatenate_20[0][0]
activation_22 (Activation)	(None, 16, 16, 60)	0	batch_normalization_22[0][0]
conv2d_23 (Conv2D)	(None, 16, 16, 6)	3240	activation_22[0][0]
dropout_22 (Dropout)	(None, 16, 16, 6)	0	conv2d_23[0][0]
concatenate_21 (Concatenate)	(None, 16, 16, 66)	0	concatenate_20[0][0] dropout_22[0][0]
batch_normalization_23 (BatchNo	(None, 16, 16, 66)	264	concatenate_21[0][0]
activation_23 (Activation)	(None, 16, 16, 66)	0	batch_normalization_23[0][0]
conv2d_24 (Conv2D)	(None, 16, 16, 6)	3564	activation_23[0][0]
dropout_23 (Dropout)	(None, 16, 16, 6)	0	conv2d_24[0][0]
concatenate_22 (Concatenate)	(None, 16, 16, 72)	0	concatenate_21[0][0] dropout_23[0][0]
batch_normalization_24 (BatchNo	(None, 16, 16, 72)	288	concatenate_22[0][0]
activation_24 (Activation)	(None, 16, 16, 72)	0	batch_normalization_24[0][0]
conv2d_25 (Conv2D)	(None, 16, 16, 6)	3888	activation_24[0][0]
dropout_24 (Dropout)	(None, 16, 16, 6)	0	conv2d_25[0][0]
concatenate_23 (Concatenate)	(None, 16, 16, 78)	0	concatenate_22[0][0] dropout_24[0][0]
batch_normalization_25 (BatchNo	(None, 16, 16, 78)	312	concatenate_23[0][0]
activation_25 (Activation)	(None, 16, 16, 78)	0	batch_normalization_25[0][0]
conv2d_26 (Conv2D)	(None, 16, 16, 6)	468	activation_25[0][0]
dropout_25 (Dropout)	(None, 16, 16, 6)	0	conv2d_26[0][0]
average_pooling2d_1 (AveragePoo	(None, 8, 8, 6)	0	dropout_25[0][0]
batch_normalization_26 (BatchNo	(None, 8, 8, 6)	24	average_pooling2d_1[0][0]
activation_26 (Activation)	(None, 8, 8, 6)	0	batch_normalization_26[0][0]
conv2d_27 (Conv2D)	(None, 8, 8, 6)	324	activation_26[0][0]
dropout_26 (Dropout)	(None, 8, 8, 6)	0	conv2d_27[0][0]
concatenate_24 (Concatenate)	(None, 8, 8, 12)	0	average_pooling2d_1[0][0] dropout_26[0][0]
batch_normalization_27 (BatchNo	(None, 8, 8, 12)	48	concatenate_24[0][0]
activation_27 (Activation)	(None, 8, 8, 12)	0	batch_normalization_27[0][0]
conv2d_28 (Conv2D)	(None, 8, 8, 6)	648	activation_27[0][0]
dropout_27 (Dropout)	(None, 8, 8, 6)	0	conv2d_28[0][0]

concatenate 25 (Concatenate)	(None,	8,	8,	18)	0	concatenate 24[0][0]
, , , , , , , , , , , , , , , , , ,	,	,	,	-,		dropout_27[0][0]
batch_normalization_28 (BatchNo	(None,	8,	8,	18)	72	concatenate_25[0][0]
activation_28 (Activation)	(None,	8,	8,	18)	0	batch_normalization_28[0][0]
conv2d_29 (Conv2D)	(None,	8,	8,	6)	972	activation_28[0][0]
dropout_28 (Dropout)	(None,	8,	8,	6)	0	conv2d_29[0][0]
concatenate_26 (Concatenate)	(None,	8,	8,	24)	0	concatenate_25[0][0] dropout_28[0][0]
batch_normalization_29 (BatchNo	(None,	8,	8,	24)	96	concatenate_26[0][0]
activation_29 (Activation)	(None,	8,	8,	24)	0	batch_normalization_29[0][0]
conv2d_30 (Conv2D)	(None,	8,	8,	6)	1296	activation_29[0][0]
dropout_29 (Dropout)	(None,	8,	8,	6)	0	conv2d_30[0][0]
concatenate_27 (Concatenate)	(None,	8,	8,	30)	0	concatenate_26[0][0] dropout_29[0][0]
batch_normalization_30 (BatchNo	(None,	8,	8,	30)	120	concatenate_27[0][0]
activation_30 (Activation)	(None,	8,	8,	30)	0	batch_normalization_30[0][0]
conv2d_31 (Conv2D)	(None,	8,	8,	6)	1620	activation_30[0][0]
dropout_30 (Dropout)	(None,	8,	8,	6)	0	conv2d_31[0][0]
concatenate_28 (Concatenate)	(None,	8,	8,	36)	0	concatenate_27[0][0] dropout_30[0][0]
batch_normalization_31 (BatchNo	(None,	8,	8,	36)	144	concatenate_28[0][0]
activation_31 (Activation)	(None,	8,	8,	36)	0	batch_normalization_31[0][0]
conv2d_32 (Conv2D)	(None,	8,	8,	6)	1944	activation_31[0][0]
dropout_31 (Dropout)	(None,	8,	8,	6)	0	conv2d_32[0][0]
concatenate_29 (Concatenate)	(None,	8,	8,	42)	0	concatenate_28[0][0] dropout_31[0][0]
batch_normalization_32 (BatchNo	(None,	8,	8,	42)	168	concatenate_29[0][0]
activation_32 (Activation)	(None,	8,	8,	42)	0	batch_normalization_32[0][0]
conv2d_33 (Conv2D)	(None,	8,	8,	6)	2268	activation_32[0][0]
dropout_32 (Dropout)	(None,	8,	8,	6)	0	conv2d_33[0][0]
concatenate_30 (Concatenate)	(None,	8,	8,	48)	0	concatenate_29[0][0] dropout_32[0][0]
batch_normalization_33 (BatchNo	(None,	8,	8,	48)	192	concatenate_30[0][0]
activation_33 (Activation)	(None,	8,	8,	48)	0	batch_normalization_33[0][0]
conv2d_34 (Conv2D)	(None,	8,	8,	6)	2592	activation_33[0][0]
dropout_33 (Dropout)	(None,	8,	8,	6)	0	conv2d_34[0][0]
concatenate_31 (Concatenate)	(None,	8,	8,	54)	0	concatenate_30[0][0] dropout_33[0][0]
batch_normalization_34 (BatchNo	(None,	8,	8,	54)	216	concatenate_31[0][0]
activation_34 (Activation)	(None,	8,	8,	54)	0	batch_normalization_34[0][0]
conv2d_35 (Conv2D)	(None,	8,	8,	6)	2916	activation_34[0][0]
dropout_34 (Dropout)	(None,	8,	8,	6)	0	conv2d_35[0][0]

(None,	8,	8,	60)	0	concatenate_31[0][0] dropout_34[0][0]
(None,	8,	8,	60)	240	concatenate_32[0][0]
(None,	8,	8,	60)	0	batch_normalization_35[0][0]
(None,	8,	8,	6)	3240	activation_35[0][0]
(None,	8,	8,	6)	0	conv2d_36[0][0]
(None,	8,	8,	66)	0	concatenate_32[0][0] dropout_35[0][0]
(None,	8,	8,	66)	264	concatenate_33[0][0]
(None,	8,	8,	66)	0	batch_normalization_36[0][0]
(None,	8,	8,	6)	3564	activation_36[0][0]
(None,	8,	8,	6)	0	conv2d_37[0][0]
(None,	8,	8,	72)	0	concatenate_33[0][0] dropout_36[0][0]
(None,	8,	8,	72)	288	concatenate_34[0][0]
(None,	8,	8,	72)	0	batch_normalization_37[0][0]
(None,	8,	8,	6)	3888	activation_37[0][0]
(None,	8,	8,	6)	0	conv2d_38[0][0]
(None,	8,	8,	78)	0	concatenate_34[0][0] dropout_37[0][0]
(None,	8,	8,	78)	312	concatenate_35[0][0]
(None,	8,	8,	78)	0	batch_normalization_38[0][0]
(None,	8,	8,	6)	468	activation_38[0][0]
(None,	8,	8,	6)	0	conv2d_39[0][0]
(None,	4,	4,	6)	0	dropout_38[0][0]
(None,	4,	4,	6)	24	average_pooling2d_2[0][0]
(None,	4,	4,	6)	0	batch_normalization_39[0][0]
(None,	4,	4,	6)	324	activation_39[0][0]
(None,	4,	4,	6)	0	conv2d_40[0][0]
(None,	4,	4,	12)	0	average_pooling2d_2[0][0] dropout_39[0][0]
(None,	4,	4,	12)	48	concatenate_36[0][0]
(None,	4,	4,	12)	0	batch_normalization_40[0][0]
(None,	4,	4,	6)	648	activation_40[0][0]
(None,	4,	4,	6)	0	conv2d_41[0][0]
(None,	4,	4,	18)	0	concatenate_36[0][0] dropout_40[0][0]
(None,	4,	4,	18)	72	concatenate_37[0][0]
(None,	4,	4,	18)	0	batch_normalization_41[0][0]
(None,	4,	4,	6)	972	activation_41[0][0]
(None,	4,	4,	6)	0	conv2d_42[0][0]
(None.	4.	4.	24)	0	concatenate 37[0][0]
	(None, (N	(None, 8, (None, 4, (None,	(None, 8, 8, 8, (None, 8, 8, (None, 8, 8, 8, (None, 4, 4, 4, 4, (None, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4,	(None, 8, 8, 60) (None, 8, 8, 60) (None, 8, 8, 60) (None, 8, 8, 6) (None, 8, 8, 6) (None, 8, 8, 66) (None, 8, 8, 66) (None, 8, 8, 6) (None, 8, 8, 6) (None, 8, 8, 72) (None, 8, 8, 72) (None, 8, 8, 72) (None, 8, 8, 72) (None, 8, 8, 73) (None, 8, 8, 78) (None, 8, 8, 78) (None, 8, 8, 78) (None, 8, 8, 6) (None, 8, 8, 6) (None, 8, 8, 6) (None, 4, 4, 6)	(None, 8, 8, 60) 240 (None, 8, 8, 60) 0 (None, 8, 8, 6) 3240 (None, 8, 8, 6) 0 (None, 8, 8, 6) 0 (None, 8, 8, 66) 0 (None, 8, 8, 66) 0 (None, 8, 8, 66) 0 (None, 8, 8, 6) 3564 (None, 8, 8, 6) 0 (None, 8, 8, 72) 0 (None, 8, 8, 73) 0 (None, 8, 8, 78) 0 (None, 8, 8, 6) 468 (None, 8, 8, 6) 0 (None, 4, 4, 6) 0

		-,	-,	,	·	dropout_41[0][0]
batch_normalization_42 (BatchNo	(None,	4,	4,	24)	96	concatenate_38[0][0]
activation_42 (Activation)	(None,	4,	4,	24)	0	batch_normalization_42[0][0]
conv2d_43 (Conv2D)	(None,	4,	4,	6)	1296	activation_42[0][0]
dropout_42 (Dropout)	(None,	4,	4,	6)	0	conv2d_43[0][0]
concatenate_39 (Concatenate)	(None,	4,	4,	30)	0	concatenate_38[0][0] dropout_42[0][0]
batch_normalization_43 (BatchNo	(None,	4,	4,	30)	120	concatenate_39[0][0]
activation_43 (Activation)	(None,	4,	4,	30)	0	batch_normalization_43[0][0]
conv2d_44 (Conv2D)	(None,	4,	4,	6)	1620	activation_43[0][0]
dropout_43 (Dropout)	(None,	4,	4,	6)	0	conv2d_44[0][0]
concatenate_40 (Concatenate)	(None,	4,	4,	36)	0	concatenate_39[0][0] dropout_43[0][0]
batch_normalization_44 (BatchNo	(None,	4,	4,	36)	144	concatenate_40[0][0]
activation_44 (Activation)	(None,	4,	4,	36)	0	batch_normalization_44[0][0]
conv2d_45 (Conv2D)	(None,	4,	4,	6)	1944	activation_44[0][0]
dropout_44 (Dropout)	(None,	4,	4,	6)	0	conv2d_45[0][0]
concatenate_41 (Concatenate)	(None,	4,	4,	42)	0	concatenate_40[0][0] dropout_44[0][0]
batch_normalization_45 (BatchNo	(None,	4,	4,	42)	168	concatenate_41[0][0]
activation 45 (Activation)	(None,	4,	4.	42)	0	batch normalization 45[0][0]
_			,	12/	0	paccii_iiofiiiaff2acfoii_45[0][0]
conv2d_46 (Conv2D)	(None,				2268	activation_45[0][0]
-		4,	4,	6)		
conv2d_46 (Conv2D)	(None,	4,	4,	6)	2268	activation_45[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout)	(None,	4,	4,	6) 6) 48)	2268	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate)	(None,	4,	4,	6) 6) 48)	0 0	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo	(None, (None, (None,	4,	4,	6) 6) 48) 48)	2268 0 0	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNoactivation_46 (Activation)	(None, (None, (None,	4, 4, 4, 4,	4, 4, 4, 4,	6) 6) 48) 48) 48)	2268 0 0 192	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D)	(None, (None, (None, (None, (None,	4, 4, 4, 4, 4,	4, 4, 4, 4, 4,	6) 6) 48) 48) 48) 6)	2268 0 0 192 0 2592	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D) dropout_46 (Dropout)	(None, (None, (None, (None, (None, (None, (None,	4, 4, 4, 4, 4,	4, 4, 4, 4, 4, 4,	6) 6) 48) 48) 48) 6) 6)	2268 0 0 192 0 2592	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0] conv2d_47[0][0] concatenate_42[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D) dropout_46 (Dropout) concatenate_43 (Concatenate)	(None, (None, (None, (None, (None, (None, (None,	4, 4, 4, 4, 4, 4,	4, 4, 4, 4, 4, 4,	6) 6) 48) 48) 48) 6) 6) 54)	2268 0 0 192 0 2592 0	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0] conv2d_47[0][0] concatenate_42[0][0] dropout_46[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D) dropout_46 (Dropout) concatenate_43 (Concatenate) batch_normalization_47 (BatchNo	(None, (None, (None, (None, (None, (None, (None, (None,	4, 4, 4, 4, 4, 4, 4,	4, 4, 4, 4, 4, 4, 4,	6) 6) 48) 48) 6) 6) 54) 54)	2268 0 0 192 0 2592 0 0	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0] conv2d_47[0][0] concatenate_42[0][0] concatenate_42[0][0] concatenate_43[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D) dropout_46 (Dropout) concatenate_43 (Concatenate) batch_normalization_47 (BatchNo activation_47 (Activation)	(None,	4, 4, 4, 4, 4, 4, 4,	4, 4, 4, 4, 4, 4, 4,	6) 6) 48) 48) 6) 6) 54) 54)	2268 0 0 192 0 2592 0 0 216	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0] conv2d_47[0][0] concatenate_42[0][0] concatenate_42[0][0] concatenate_43[0][0] batch_normalization_47[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D) dropout_46 (Dropout) concatenate_43 (Concatenate) batch_normalization_47 (BatchNo activation_47 (Activation) conv2d_48 (Conv2D)	(None,	4, 4, 4, 4, 4, 4, 4, 4,	4, 4, 4, 4, 4, 4, 4, 4, 4,	6) 48) 48) 48) 6) 6) 54) 54) 6)	2268 0 0 192 0 2592 0 0 216 0 2916	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0] conv2d_47[0][0] concatenate_42[0][0] concatenate_42[0][0] concatenate_43[0][0] concatenate_43[0][0] batch_normalization_47[0][0] activation_47[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D) dropout_46 (Dropout) concatenate_43 (Concatenate) batch_normalization_47 (BatchNo activation_47 (Activation) conv2d_48 (Conv2D) dropout_47 (Dropout)	(None,	4, 4, 4, 4, 4, 4, 4, 4,	4, 4, 4, 4, 4, 4, 4, 4,	6) 6) 48) 48) 6) 6) 54) 54) 6) 6) 6)	2268 0 0 192 0 2592 0 0 216 0 2916	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0] conv2d_47[0][0] concatenate_42[0][0] concatenate_42[0][0] concatenate_43[0][0] batch_normalization_47[0][0] activation_47[0][0] conv2d_48[0][0] conv2d_48[0][0] concatenate_43[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D) dropout_46 (Dropout) concatenate_43 (Concatenate) batch_normalization_47 (BatchNo activation_47 (Activation) conv2d_48 (Conv2D) dropout_47 (Dropout) concatenate_44 (Concatenate)	(None,	4, 4, 4, 4, 4, 4, 4, 4,	4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4,	6) 6) 48) 48) 6) 6) 54) 54) 6) 6) 60)	2268 0 0 192 0 2592 0 0 216 0 2916 0	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0] conv2d_47[0][0] concatenate_42[0][0] concatenate_42[0][0] concatenate_43[0][0] concatenate_43[0][0] batch_normalization_47[0][0] activation_47[0][0] conv2d_48[0][0] conv2d_48[0][0] concatenate_43[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D) dropout_46 (Dropout) concatenate_43 (Concatenate) batch_normalization_47 (BatchNo activation_47 (Activation) conv2d_48 (Conv2D) dropout_47 (Dropout) concatenate_44 (Concatenate) batch_normalization_48 (BatchNo	(None,	4, 4, 4, 4, 4, 4, 4, 4, 4,	4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4,	6) 48) 48) 48) 6) 6) 54) 54) 6) 6) 60)	2268 0 0 192 0 2592 0 0 216 0 2916 0	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0] conv2d_47[0][0] concatenate_42[0][0] concatenate_42[0][0] concatenate_43[0][0] concatenate_43[0][0] batch_normalization_47[0][0] activation_47[0][0] conv2d_48[0][0] concatenate_43[0][0] concatenate_43[0][0] concatenate_44[0][0]
conv2d_46 (Conv2D) dropout_45 (Dropout) concatenate_42 (Concatenate) batch_normalization_46 (BatchNo activation_46 (Activation) conv2d_47 (Conv2D) dropout_46 (Dropout) concatenate_43 (Concatenate) batch_normalization_47 (BatchNo activation_47 (Activation) conv2d_48 (Conv2D) dropout_47 (Dropout) concatenate_44 (Concatenate) batch_normalization_48 (BatchNo activation_48 (Activation)	(None,	4, 4, 4, 4, 4, 4, 4, 4, 4, 4,	4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4,	6) 48) 48) 48) 6) 6) 54) 54) 6) 6) 60) 60)	2268 0 0 192 0 2592 0 0 216 0 2916 0 0	activation_45[0][0] conv2d_46[0][0] concatenate_41[0][0] dropout_45[0][0] concatenate_42[0][0] batch_normalization_46[0][0] activation_46[0][0] conv2d_47[0][0] concatenate_42[0][0] concatenate_42[0][0] concatenate_43[0][0] concatenate_43[0][0] batch_normalization_47[0][0] activation_47[0][0] conv2d_48[0][0] concatenate_43[0][0] concatenate_43[0][0] concatenate_44[0][0] batch_normalization_48[0][0]

	(110110,	±1 ±	,	· · ,	V	dropout_48[0][0]
batch_normalization_49 (BatchNo	(None,	4, 4	,	66)	264	concatenate_45[0][0]
activation_49 (Activation)	(None,	4, 4	,	66)	0	batch_normalization_49[0][0]
conv2d_50 (Conv2D)	(None,	4, 4	,	6)	3564	activation_49[0][0]
dropout_49 (Dropout)	(None,	4, 4	,	6)	0	conv2d_50[0][0]
concatenate_46 (Concatenate)	(None,	4, 4	,	72)	0	concatenate_45[0][0] dropout_49[0][0]
batch_normalization_50 (BatchNo	(None,	4, 4	,	72)	288	concatenate_46[0][0]
activation_50 (Activation)	(None,	4, 4	,	72)	0	batch_normalization_50[0][0]
conv2d_51 (Conv2D)	(None,	4, 4	,	6)	3888	activation_50[0][0]
dropout_50 (Dropout)	(None,	4, 4	,	6)	0	conv2d_51[0][0]
concatenate_47 (Concatenate)	(None,	4, 4	,	78)	0	concatenate_46[0][0] dropout_50[0][0]
batch_normalization_51 (BatchNo	(None,	4, 4	,	78)	312	concatenate_47[0][0]
activation_51 (Activation)	(None,	4, 4	,	78)	0	batch_normalization_51[0][0]
average_pooling2d_3 (AveragePoo	(None,	2, 2	,	78)	0	activation_51[0][0]
flatten (Flatten)	(None,	312)			0	average_pooling2d_3[0][0]
dense (Dense)	(None,	10)			3130	flatten[0][0]

8. Model Compile

Trainable params: 114,394 Non-trainable params: 4,524

```
In [17]:
```

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

9. Callbacks

```
In [29]:
```

```
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
patience = 50
base_path = '/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/'
checkpoint_file_name = base_path + 'CIFAR' + '_{epoch:02d}-{val_loss:.2f}.hdf5'

model_checkpoint = ModelCheckpoint(checkpoint_file_name, monitor='val_loss', verbose=1, save_best_o
nly=True)
early_stop = EarlyStopping('val_loss', patience = patience)
reduce_LR = ReduceLROnPlateau(monitor='val_loss', factor=0.1, patience=int(patience/3))
```

```
In [30]:
```

```
callbacks = [model_checkpoint, early_stop, reduce_LR]
```

10. Train the model

```
In [22]:
```

```
epochs = 300
```

```
batch_size = 128
```

```
In [31]:
#https://keras.io/api/preprocessing/image/#flow-method
history = model.fit(data generator.flow(X train, y train, batch size),
            steps_per_epoch = int(len(X_train)/batch size),
            epochs = epochs,
            callbacks = callbacks,
            validation_data = (X_test, y_test), verbose=1)
WARNING:tensorflow:sample weight modes were coerced from
 ['...']
Train for 390 steps, validate on 10000 samples
Epoch 1/300
Epoch 00001: val loss improved from inf to 1.78309, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR 01-1.78.hdf5
oss: 1.7831 - val accuracy: 0.4376
Epoch 2/300
Epoch 00002: val loss improved from 1.78309 to 1.33771, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_02-1.34.hdf5
390/390 [============ ] - 28s 73ms/step - loss: 1.2729 - accuracy: 0.5363 - val 1
oss: 1.3377 - val accuracy: 0.5480
Epoch 3/300
Epoch 00003: val loss improved from 1.33771 to 1.29796, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR 03-1.30.hdf5
oss: 1.2980 - val_accuracy: 0.5614
Epoch 4/300
Epoch 00004: val loss did not improve from 1.29796
390/390 [=============== ] - 28s 72ms/step - loss: 1.1334 - accuracy: 0.5860 - val 1
oss: 1.3236 - val_accuracy: 0.5745
Epoch 5/300
Epoch 00005: val_loss did not improve from 1.29796
oss: 1.9929 - val accuracy: 0.4908
Epoch 6/300
Epoch 00006: val_loss improved from 1.29796 to 1.18068, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR 06-1.18.hdf5
390/390 [============ ] - 29s 74ms/step - loss: 1.0590 - accuracy: 0.6177 - val_1
oss: 1.1807 - val_accuracy: 0.6074
Epoch 7/300
Epoch 00007: val_loss did not improve from 1.18068
390/390 [============ ] - 28s 72ms/step - loss: 1.0251 - accuracy: 0.6277 - val 1
oss: 1.2659 - val accuracy: 0.5894
Epoch 8/300
Epoch 00008: val loss did not improve from 1.18068
oss: 1.1914 - val accuracy: 0.6099
Epoch 9/300
Epoch 00009: val loss did not improve from 1.18068
oss: 1.3551 - val accuracy: 0.5842
Epoch 10/300
Epoch 00010: val loss improved from 1.18068 to 1.10336, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_10-1.10.hdf5
390/390 [============ ] - 29s 73ms/step - loss: 0.9545 - accuracy: 0.6586 - val 1
oss: 1.1034 - val accuracy: 0.6427
Epoch 11/300
Epoch 00011: val loss improved from 1.10336 to 0.94139, saving model to
```

```
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR 11-0.94.hdf5
oss: 0.9414 - val accuracy: 0.6786
Epoch 12/300
Epoch 00012: val loss did not improve from 0.94139
390/390 [============= ] - 28s 72ms/step - loss: 0.9247 - accuracy: 0.6695 - val 1
oss: 1.0617 - val_accuracy: 0.6490
Epoch 13/300
Epoch 00013: val_loss did not improve from 0.94139
390/390 [=============== ] - 28s 73ms/step - loss: 0.9049 - accuracy: 0.6767 - val 1
oss: 1.1002 - val_accuracy: 0.6536
Epoch 14/300
Epoch 00014: val loss did not improve from 0.94139
390/390 [============ ] - 28s 72ms/step - loss: 0.8906 - accuracy: 0.6835 - val 1
oss: 1.0150 - val_accuracy: 0.6749
Epoch 15/300
Epoch 00015: val loss did not improve from 0.94139
390/390 [============ ] - 28s 72ms/step - loss: 0.8739 - accuracy: 0.6895 - val 1
oss: 1.1756 - val_accuracy: 0.6430
Epoch 16/300
Epoch 00016: val_loss did not improve from 0.94139
390/390 [============ ] - 28s 72ms/step - loss: 0.8600 - accuracy: 0.6927 - val 1
oss: 1.0592 - val accuracy: 0.6671
Epoch 17/300
Epoch 00017: val loss did not improve from 0.94139
390/390 [============ ] - 28s 72ms/step - loss: 0.8465 - accuracy: 0.6996 - val 1
oss: 1.2268 - val_accuracy: 0.6355
Epoch 18/300
Epoch 00018: val loss did not improve from 0.94139
oss: 1.2344 - val accuracy: 0.6350
Epoch 19/300
Epoch 00019: val loss did not improve from 0.94139
390/390 [============ ] - 28s 72ms/step - loss: 0.8188 - accuracy: 0.7083 - val 1
oss: 1.2781 - val_accuracy: 0.6406
Epoch 20/300
Epoch 00020: val loss did not improve from 0.94139
390/390 [============ ] - 28s 72ms/step - loss: 0.8106 - accuracy: 0.7139 - val 1
oss: 1.2194 - val accuracy: 0.6414
Epoch 21/300
Epoch 00021: val loss did not improve from 0.94139
oss: 1.2590 - val_accuracy: 0.6343
Epoch 22/300
Epoch 00022: val loss improved from 0.94139 to 0.87153, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_22-0.87.hdf5
390/390 [============ ] - 29s 73ms/step - loss: 0.7912 - accuracy: 0.7201 - val 1
oss: 0.8715 - val_accuracy: 0.7155
Epoch 23/300
Epoch 00023: val loss did not improve from 0.87153
oss: 1.0189 - val accuracy: 0.6918
Epoch 24/300
Epoch 00024: val_loss improved from 0.87153 to 0.81352, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_24-0.81.hdf5
390/390 [============ ] - 29s 74ms/step - loss: 0.7747 - accuracy: 0.7263 - val_1
oss: 0.8135 - val_accuracy: 0.7388
Epoch 25/300
Epoch 00025: val loss did not improve from 0.81352
390/390 [============ ] - 28s 72ms/step - loss: 0.7628 - accuracy: 0.7287 - val 1
oss: 0.8556 - val_accuracy: 0.7213
Epoch 26/300
```

```
Epoch 00026: val loss did not improve from 0.81352
oss: 0.8655 - val accuracy: 0.7198
Epoch 27/300
Epoch 00027: val loss did not improve from 0.81352
390/390 [============= ] - 28s 72ms/step - loss: 0.7546 - accuracy: 0.7343 - val 1
oss: 0.8424 - val_accuracy: 0.7317
Epoch 28/300
Epoch 00028: val loss did not improve from 0.81352
390/390 [========================== ] - 28s 72ms/step - loss: 0.7435 - accuracy: 0.7358 - val 1
oss: 0.8330 - val_accuracy: 0.7308
Epoch 29/300
Epoch 00029: val loss did not improve from 0.81352
390/390 [============ ] - 28s 72ms/step - loss: 0.7380 - accuracy: 0.7401 - val 1
oss: 0.8146 - val accuracy: 0.7381
Epoch 30/300
Epoch 00030: val loss did not improve from 0.81352
oss: 0.8778 - val_accuracy: 0.7308
Epoch 31/300
Epoch 00031: val_loss did not improve from 0.81352
390/390 [============ ] - 28s 72ms/step - loss: 0.7245 - accuracy: 0.7457 - val 1
oss: 0.9214 - val accuracy: 0.7228
Epoch 32/300
Epoch 00032: val loss did not improve from 0.81352
390/390 [============ ] - 28s 72ms/step - loss: 0.7204 - accuracy: 0.7467 - val 1
oss: 0.9153 - val accuracy: 0.7089
Epoch 33/300
Epoch 00033: val loss improved from 0.81352 to 0.77029, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR 33-0.77.hdf5
oss: 0.7703 - val accuracy: 0.7468
Epoch 34/300
Epoch 00034: val loss did not improve from 0.77029
390/390 [============= ] - 28s 72ms/step - loss: 0.7063 - accuracy: 0.7502 - val_1
oss: 1.1673 - val_accuracy: 0.6510
Epoch 35/300
Epoch 00035: val loss did not improve from 0.77029
390/390 [============= ] - 28s 73ms/step - loss: 0.7026 - accuracy: 0.7526 - val 1
oss: 0.8781 - val accuracy: 0.7302
Epoch 36/300
Epoch 00036: val loss did not improve from 0.77029
390/390 [============ ] - 28s 72ms/step - loss: 0.6992 - accuracy: 0.7558 - val_1
oss: 0.8242 - val_accuracy: 0.7402
Epoch 37/300
Epoch 00037: val_loss did not improve from 0.77029
390/390 [============ ] - 28s 72ms/step - loss: 0.6885 - accuracy: 0.7575 - val 1
oss: 1.0020 - val accuracy: 0.7077
Epoch 38/300
Epoch 00038: val loss did not improve from 0.77029
390/390 [============ ] - 28s 72ms/step - loss: 0.6877 - accuracy: 0.7586 - val 1
oss: 0.8783 - val accuracy: 0.7372
Epoch 39/300
Epoch 00039: val loss did not improve from 0.77029
oss: 1.0037 - val_accuracy: 0.7105
Epoch 40/300
Epoch 00040: val loss did not improve from 0.77029
390/390 [============ ] - 28s 72ms/step - loss: 0.6804 - accuracy: 0.7613 - val 1
oss: 0.9206 - val_accuracy: 0.7253
Epoch 41/300
Epoch 00041: val loss did not improve from 0.77029
```

```
390/390 [============ ] - 28s 72ms/step - loss: 0.6699 - accuracy: 0.7627 - val 1
oss: 0.8955 - val accuracy: 0.7375
Epoch 42/300
Epoch 00042: val loss did not improve from 0.77029
oss: 0.8193 - val accuracy: 0.7536
Epoch 43/300
Epoch 00043: val_loss did not improve from 0.77029
oss: 0.9809 - val accuracy: 0.7126
Epoch 44/300
Epoch 00044: val loss improved from 0.77029 to 0.71524, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_44-0.72.hdf5
oss: 0.7152 - val accuracy: 0.7723
Epoch 45/300
Epoch 00045: val loss did not improve from 0.71524
390/390 [============ ] - 28s 72ms/step - loss: 0.6592 - accuracy: 0.7678 - val 1
oss: 0.8049 - val accuracy: 0.7541
Epoch 46/300
Epoch 00046: val loss improved from 0.71524 to 0.67775, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_46-0.68.hdf5
390/390 [============ ] - 28s 73ms/step - loss: 0.6561 - accuracy: 0.7695 - val_1
oss: 0.6777 - val_accuracy: 0.7879
Epoch 47/300
Epoch 00047: val loss did not improve from 0.67775
oss: 0.8008 - val_accuracy: 0.7513
Epoch 48/300
Epoch 00048: val loss did not improve from 0.67775
oss: 0.7841 - val accuracy: 0.7539
Epoch 49/300
Epoch 00049: val loss did not improve from 0.67775
390/390 [============= ] - 28s 72ms/step - loss: 0.6425 - accuracy: 0.7738 - val 1
oss: 0.9057 - val_accuracy: 0.7353
Epoch 50/300
Epoch 00050: val_loss did not improve from 0.67775
390/390 [============ ] - 28s 72ms/step - loss: 0.6406 - accuracy: 0.7739 - val 1
oss: 0.8436 - val accuracy: 0.7444
Epoch 51/300
Epoch 00051: val loss did not improve from 0.67775
oss: 0.9483 - val_accuracy: 0.7206
Epoch 52/300
Epoch 00052: val loss did not improve from 0.67775
390/390 [============ ] - 28s 72ms/step - loss: 0.6318 - accuracy: 0.7779 - val 1
oss: 1.0205 - val accuracy: 0.7098
Epoch 53/300
Epoch 00053: val loss did not improve from 0.67775
390/390 [============ ] - 28s 71ms/step - loss: 0.6314 - accuracy: 0.7781 - val_1
oss: 0.7074 - val_accuracy: 0.7772
Epoch 54/300
Epoch 00054: val loss did not improve from 0.67775
390/390 [============ ] - 28s 72ms/step - loss: 0.6308 - accuracy: 0.7783 - val 1
oss: 0.8820 - val_accuracy: 0.7428
Epoch 55/300
Epoch 00055: val_loss did not improve from 0.67775
390/390 [============= ] - 28s 72ms/step - loss: 0.6281 - accuracy: 0.7794 - val_1
oss: 0.9102 - val_accuracy: 0.7383
Epoch 56/300
Epoch 00057: val loss did not improve from 0.67775
```

```
390/390 [============== ] - 28s 72ms/step - loss: 0.6223 - accuracy: 0.7821 - val 1
oss: 0.8572 - val accuracy: 0.7502
Epoch 58/300
Epoch 00058: val loss did not improve from 0.67775
390/390 [============ ] - 28s 72ms/step - loss: 0.6216 - accuracy: 0.7836 - val_1
oss: 0.7145 - val accuracy: 0.7728
Epoch 59/300
Epoch 00059: val loss did not improve from 0.67775
390/390 [============ ] - 28s 72ms/step - loss: 0.6186 - accuracy: 0.7838 - val 1
oss: 0.6832 - val_accuracy: 0.7870
Epoch 60/300
Epoch 00060: val loss did not improve from 0.67775
390/390 [============ ] - 28s 72ms/step - loss: 0.6118 - accuracy: 0.7861 - val 1
oss: 0.8445 - val_accuracy: 0.7423
Epoch 61/300
Epoch 00061: val loss did not improve from 0.67775
oss: 0.7096 - val accuracy: 0.7794
Epoch 62/300
Epoch 00062: val_loss did not improve from 0.67775
oss: 0.6784 - val_accuracy: 0.7823
Epoch 63/300
Epoch 00063: val loss improved from 0.67775 to 0.64425, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR 63-0.64.hdf5
oss: 0.6443 - val accuracy: 0.8006
Epoch 64/300
Epoch 00064: val loss improved from 0.64425 to 0.63837, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR 64-0.64.hdf5
390/390 [============ ] - 29s 75ms/step - loss: 0.5612 - accuracy: 0.8052 - val 1
oss: 0.6384 - val accuracy: 0.8023
Epoch 65/300
Epoch 00065: val loss improved from 0.63837 to 0.63668, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR 65-0.64.hdf5
390/390 [===================== ] - 29s 74ms/step - loss: 0.5651 - accuracy: 0.8026 - val 1
oss: 0.6367 - val_accuracy: 0.8006
Epoch 66/300
Epoch 00066: val loss improved from 0.63668 to 0.61334, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR 66-0.61.hdf5
oss: 0.6133 - val accuracy: 0.8083
Epoch 67/300
Epoch 00067: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5551 - accuracy: 0.8056 - val_1
oss: 0.6187 - val accuracy: 0.8103
Epoch 68/300
Epoch 00068: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5552 - accuracy: 0.8055 - val 1
oss: 0.6215 - val_accuracy: 0.8081
Epoch 69/300
Epoch 00069: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5519 - accuracy: 0.8079 - val 1
oss: 0.6524 - val accuracy: 0.8005
Epoch 70/300
Epoch 00070: val loss did not improve from 0.61334
oss: 0.6199 - val_accuracy: 0.8097
Epoch 71/300
Epoch 00071: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5509 - accuracy: 0.8086 - val 1
oss: 0.6481 - val accuracy: 0.8015
Epoch 72/300
```

```
Epoch 00072: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5508 - accuracy: 0.8065 - val 1
oss: 0.6337 - val_accuracy: 0.8084
Epoch 73/300
Epoch 00073: val loss did not improve from 0.61334
oss: 0.6421 - val accuracy: 0.8075
Epoch 74/300
Epoch 00074: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5547 - accuracy: 0.8057 - val 1
oss: 0.6373 - val_accuracy: 0.8060
Epoch 75/300
Epoch 00075: val loss did not improve from 0.61334
oss: 0.6167 - val accuracy: 0.8111
Epoch 76/300
Epoch 00076: val loss did not improve from 0.61334
390/390 [============ ] - 28s 73ms/step - loss: 0.5498 - accuracy: 0.8074 - val 1
oss: 0.6347 - val_accuracy: 0.8072
Epoch 77/300
Epoch 00077: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5492 - accuracy: 0.8062 - val 1
oss: 0.6422 - val_accuracy: 0.8064
Epoch 78/300
Epoch 00078: val loss did not improve from 0.61334
390/390 [========================== ] - 28s 72ms/step - loss: 0.5468 - accuracy: 0.8080 - val 1
oss: 0.6581 - val_accuracy: 0.8033
Epoch 79/300
Epoch 00079: val loss did not improve from 0.61334
oss: 0.6333 - val accuracy: 0.8066
Epoch 80/300
Epoch 00080: val loss did not improve from 0.61334
390/390 [============= ] - 28s 72ms/step - loss: 0.5442 - accuracy: 0.8100 - val 1
oss: 0.6199 - val_accuracy: 0.8100
Epoch 81/300
Epoch 00081: val loss did not improve from 0.61334
oss: 0.6548 - val accuracy: 0.8027
Epoch 82/300
Epoch 00082: val loss did not improve from 0.61334
390/390 [============= ] - 28s 72ms/step - loss: 0.5432 - accuracy: 0.8097 - val 1
oss: 0.6311 - val accuracy: 0.8073
Epoch 83/300
Epoch 00083: val loss did not improve from 0.61334
390/390 [============= ] - 28s 72ms/step - loss: 0.5410 - accuracy: 0.8101 - val 1
oss: 0.6273 - val_accuracy: 0.8099
Epoch 84/300
Epoch 00084: val loss did not improve from 0.61334
390/390 [========================= ] - 28s 72ms/step - loss: 0.5423 - accuracy: 0.8093 - val 1
oss: 0.6330 - val_accuracy: 0.8076
Epoch 85/300
Epoch 00085: val loss did not improve from 0.61334
oss: 0.6334 - val accuracy: 0.8071
Epoch 86/300
Epoch 00086: val_loss did not improve from 0.61334
oss: 0.6329 - val accuracy: 0.8076
Epoch 87/300
Epoch 00087: val loss did not improve from 0.61334
```

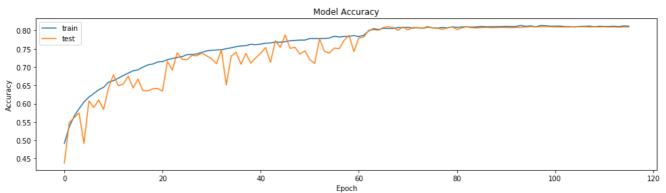
```
oss: 0.6290 - val_accuracy: 0.8083
Epoch 88/300
Epoch 00088: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5397 - accuracy: 0.8104 - val 1
oss: 0.6333 - val_accuracy: 0.8077
Epoch 89/300
Epoch 00089: val loss did not improve from 0.61334
oss: 0.6309 - val accuracy: 0.8078
Epoch 90/300
Epoch 00090: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5406 - accuracy: 0.8108 - val 1
oss: 0.6319 - val accuracy: 0.8083
Epoch 91/300
Epoch 00091: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5386 - accuracy: 0.8112 - val 1
oss: 0.6342 - val accuracy: 0.8079
Epoch 92/300
Epoch 00092: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5429 - accuracy: 0.8110 - val 1
oss: 0.6361 - val_accuracy: 0.8075
Epoch 93/300
Epoch 00093: val_loss did not improve from 0.61334
oss: 0.6344 - val_accuracy: 0.8078
Epoch 94/300
Epoch 00094: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5340 - accuracy: 0.8138 - val 1
oss: 0.6330 - val_accuracy: 0.8086
Epoch 95/300
Epoch 00095: val loss did not improve from 0.61334
oss: 0.6314 - val accuracy: 0.8095
Epoch 96/300
Epoch 00096: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5385 - accuracy: 0.8127 - val 1
oss: 0.6285 - val accuracy: 0.8104
Epoch 97/300
Epoch 00097: val loss did not improve from 0.61334
oss: 0.6297 - val accuracy: 0.8098
Epoch 98/300
Epoch 00098: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5332 - accuracy: 0.8136 - val_1
oss: 0.6319 - val_accuracy: 0.8098
Epoch 99/300
Epoch 00099: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5355 - accuracy: 0.8129 - val 1
oss: 0.6303 - val_accuracy: 0.8106
Epoch 100/300
Epoch 00100: val loss did not improve from 0.61334
oss: 0.6304 - val accuracy: 0.8101
Epoch 101/300
Epoch 00101: val loss did not improve from 0.61334
390/390 [============= ] - 28s 72ms/step - loss: 0.5375 - accuracy: 0.8117 - val 1
oss: 0.6314 - val_accuracy: 0.8097
Epoch 102/300
Epoch 00102: val_loss did not improve from 0.61334
390/390 [============= ] - 28s 72ms/step - loss: 0.5374 - accuracy: 0.8120 - val 1
```

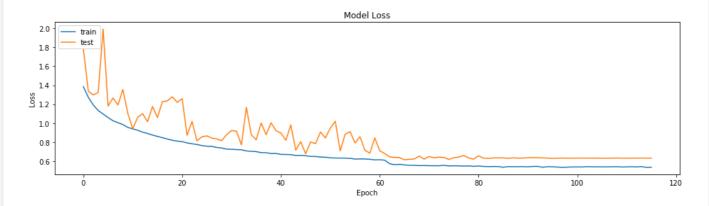
oss: 0.6313 - val accuracy: 0.8096

```
__----
Epoch 103/300
Epoch 00103: val loss did not improve from 0.61334
390/390 [============= ] - 28s 72ms/step - loss: 0.5402 - accuracy: 0.8104 - val 1
oss: 0.6311 - val_accuracy: 0.8095
Epoch 104/300
Epoch 00104: val loss did not improve from 0.61334
oss: 0.6305 - val accuracy: 0.8094
Epoch 105/300
Epoch 00105: val_loss did not improve from 0.61334
390/390 [============= ] - 28s 71ms/step - loss: 0.5388 - accuracy: 0.8093 - val 1
oss: 0.6312 - val_accuracy: 0.8102
Epoch 106/300
Epoch 00106: val_loss did not improve from 0.61334
390/390 [============= ] - 28s 72ms/step - loss: 0.5380 - accuracy: 0.8111 - val 1
oss: 0.6300 - val accuracy: 0.8101
Epoch 107/300
Epoch 00107: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5383 - accuracy: 0.8113 - val 1
oss: 0.6302 - val accuracy: 0.8101
Epoch 108/300
Epoch 00108: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5387 - accuracy: 0.8119 - val 1
oss: 0.6306 - val_accuracy: 0.8096
Epoch 109/300
Epoch 00109: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5400 - accuracy: 0.8101 - val 1
oss: 0.6319 - val accuracy: 0.8096
Epoch 110/300
Epoch 00110: val loss did not improve from 0.61334
oss: 0.6307 - val_accuracy: 0.8096
Epoch 111/300
Epoch 00111: val_loss did not improve from 0.61334
390/390 [============= ] - 28s 71ms/step - loss: 0.5382 - accuracy: 0.8111 - val_1
oss: 0.6308 - val accuracy: 0.8099
Epoch 112/300
Epoch 00112: val_loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5395 - accuracy: 0.8109 - val 1
oss: 0.6310 - val_accuracy: 0.8094
Epoch 113/300
Epoch 00113: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5376 - accuracy: 0.8116 - val 1
oss: 0.6309 - val_accuracy: 0.8096
Epoch 114/300
Epoch 00114: val loss did not improve from 0.61334
390/390 [=================== ] - 28s 72ms/step - loss: 0.5413 - accuracy: 0.8105 - val_1
oss: 0.6318 - val_accuracy: 0.8089
Epoch 115/300
Epoch 00115: val_loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5333 - accuracy: 0.8126 - val 1
oss: 0.6309 - val accuracy: 0.8096
Epoch 116/300
Epoch 00116: val loss did not improve from 0.61334
390/390 [============ ] - 28s 72ms/step - loss: 0.5351 - accuracy: 0.8117 - val 1
oss: 0.6307 - val accuracy: 0.8094
```

```
plt.plot(history, history['val_accuracy'])
plt.title("Model Accuracy")
plt.xlabel("Epoch")
plt.ylabel("Accuracy")
plt.legend(["train", "test"], loc="upper left")
plt.show()

# history plot for accuracy
plt.figure(figsize=(16,4))
plt.plot(history.history["loss"])
plt.plot(history.history["val_loss"])
plt.title("Model Loss")
plt.xlabel("Epoch")
plt.xlabel("Epoch")
plt.ylabel("Loss")
plt.legend(["train", "test"], loc="upper left")
plt.show()
```





In [38]:

```
model.metrics_names
```

Out[38]:

['loss', 'accuracy']

In [40]:

```
best_model_1 =
tf.keras.models.load_model('/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_66-0.61.hdf5')
scores = best_model_1.evaluate(X_test, y_test, verbose=1)
print(scores)
```

10000/10000 [============] - 3s 348us/sample - loss: 0.6133 - accuracy: 0.8083 [0.6133434212684631, 0.8083]

Summary:

• The model early stopped because the val loss is not reducing for 50 epochs. The best model is at epoch 66 and it gives the test accuracy of 0.8083. We can improve the accuracy further

11. By trying different learning rate

In [41]:

model_2 = tf.keras.models.load_model('/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_66-0
.61.hdf5')
model_2.summary()

Model	:	"model"
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Layer (type)	Output	Shape			Param #	Connected to
input_1 (InputLayer)	[(None,	32,	32 ,	3)]	0	
conv2d (Conv2D)	(None,	32, 3	2,	12)	324	input_1[0][0]
batch_normalization (BatchNorma	(None,	32, 3	2,	12)	48	conv2d[0][0]
activation (Activation)	(None,	32, 3	2,	12)	0	batch_normalization[0][0]
conv2d_1 (Conv2D)	(None,	32, 3	2,	6)	648	activation[0][0]
dropout (Dropout)	(None,	32, 3	2,	6)	0	conv2d_1[0][0]
concatenate (Concatenate)	(None,	32, 3	2,	18)	0	conv2d[0][0] dropout[0][0]
batch_normalization_1 (BatchNor	(None,	32, 3	2,	18)	72	concatenate[0][0]
activation_1 (Activation)	(None,	32, 3	2,	18)	0	batch_normalization_1[0][0]
conv2d_2 (Conv2D)	(None,	32, 3	2,	6)	972	activation_1[0][0]
dropout_1 (Dropout)	(None,	32, 3	2,	6)	0	conv2d_2[0][0]
concatenate_1 (Concatenate)	(None,	32, 3	2,	24)	0	<pre>concatenate[0][0] dropout_1[0][0]</pre>
batch_normalization_2 (BatchNor	(None,	32, 3	2,	24)	96	concatenate_1[0][0]
activation_2 (Activation)	(None,	32, 3	2,	24)	0	batch_normalization_2[0][0]
conv2d_3 (Conv2D)	(None,	32, 3	2,	6)	1296	activation_2[0][0]
dropout_2 (Dropout)	(None,	32, 3	2,	6)	0	conv2d_3[0][0]
concatenate_2 (Concatenate)	(None,	32, 3	2,	30)	0	<pre>concatenate_1[0][0] dropout_2[0][0]</pre>
batch_normalization_3 (BatchNor	(None,	32, 3	2,	30)	120	concatenate_2[0][0]
activation_3 (Activation)	(None,	32, 3	2,	30)	0	batch_normalization_3[0][0]
conv2d_4 (Conv2D)	(None,	32, 3	2,	6)	1620	activation_3[0][0]
dropout_3 (Dropout)	(None,	32, 3	2,	6)	0	conv2d_4[0][0]
concatenate_3 (Concatenate)	(None,	32, 3	2,	36)	0	concatenate_2[0][0] dropout_3[0][0]
batch_normalization_4 (BatchNor	(None,	32, 3	2,	36)	144	concatenate_3[0][0]
activation_4 (Activation)	(None,	32, 3	2,	36)	0	batch_normalization_4[0][0]
conv2d_5 (Conv2D)	(None,	32, 3	2,	6)	1944	activation_4[0][0]
dropout_4 (Dropout)	(None,	32, 3	2,	6)	0	conv2d_5[0][0]
concatenate_4 (Concatenate)	(None,	32, 3	2,	42)	0	concatenate_3[0][0] dropout_4[0][0]
batch_normalization_5 (BatchNor	(None,	32, 3	2,	42)	168	concatenate_4[0][0]
activation 5 (Activation)	(None.	32. 3	2	42)	0	batch normalization 5[0][0]

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conv2d_6 (Conv2D)	(None,	32,	32,	6)	2268	activation_5[0][0]
dropout_5 (Dropout)	(None,	32,	32,	6)	0	conv2d_6[0][0]
concatenate_5 (Concatenate)	(None,	32,	32,	48)	0	concatenate_4[0][0] dropout_5[0][0]
batch_normalization_6 (BatchNor	(None,	32,	32,	48)	192	concatenate_5[0][0]
activation_6 (Activation)	(None,	32,	32,	48)	0	batch_normalization_6[0][0]
conv2d_7 (Conv2D)	(None,	32,	32,	6)	2592	activation_6[0][0]
dropout_6 (Dropout)	(None,	32,	32,	6)	0	conv2d_7[0][0]
concatenate_6 (Concatenate)	(None,	32,	32,	54)	0	concatenate_5[0][0] dropout_6[0][0]
batch_normalization_7 (BatchNor	(None,	32,	32,	54)	216	concatenate_6[0][0]
activation_7 (Activation)	(None,	32,	32,	54)	0	batch_normalization_7[0][0]
conv2d_8 (Conv2D)	(None,	32,	32,	6)	2916	activation_7[0][0]
dropout_7 (Dropout)	(None,	32,	32,	6)	0	conv2d_8[0][0]
concatenate_7 (Concatenate)	(None,	32,	32,	60)	0	concatenate_6[0][0] dropout_7[0][0]
batch_normalization_8 (BatchNor	(None,	32,	32,	60)	240	concatenate_7[0][0]
activation_8 (Activation)	(None,	32,	32,	60)	0	batch_normalization_8[0][0]
conv2d_9 (Conv2D)	(None,	32,	32,	6)	3240	activation_8[0][0]
dropout_8 (Dropout)	(None,	32,	32,	6)	0	conv2d_9[0][0]
concatenate_8 (Concatenate)	(None,	32,	32,	66)	0	concatenate_7[0][0] dropout_8[0][0]
batch_normalization_9 (BatchNor	(None,	32,	32,	66)	264	concatenate_8[0][0]
activation_9 (Activation)	(None,	32,	32,	66)	0	batch_normalization_9[0][0]
conv2d_10 (Conv2D)	(None,	32,	32,	6)	3564	activation_9[0][0]
dropout_9 (Dropout)	(None,	32,	32,	6)	0	conv2d_10[0][0]
concatenate_9 (Concatenate)	(None,	32,	32,	72)	0	concatenate_8[0][0] dropout_9[0][0]
batch_normalization_10 (BatchNo	(None,	32,	32,	72)	288	concatenate_9[0][0]
activation_10 (Activation)	(None,	32,	32,	72)	0	batch_normalization_10[0][0]
conv2d_11 (Conv2D)	(None,	32,	32,	6)	3888	activation_10[0][0]
dropout_10 (Dropout)	(None,	32,	32,	6)	0	conv2d_11[0][0]
concatenate_10 (Concatenate)	(None,	32,	32,	78)	0	concatenate_9[0][0] dropout_10[0][0]
batch_normalization_11 (BatchNo	(None,	32,	32,	78)	312	concatenate_10[0][0]
activation_11 (Activation)	(None,	32,	32,	78)	0	batch_normalization_11[0][0]
conv2d_12 (Conv2D)	(None,	32,	32,	6)	4212	activation_11[0][0]
dropout_11 (Dropout)	(None,	32,	32,	6)	0	conv2d_12[0][0]
concatenate_11 (Concatenate)	(None,	32,	32,	84)	0	concatenate_10[0][0] dropout_11[0][0]
batch_normalization_12 (BatchNo	(None,	32,	32,	84)	336	concatenate_11[0][0]
activation 12 (Activation)	(None.	32.	32.	84)	n	hatch normalization 12[0][0]

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conv2d_13 (Conv2D)	(None,	32,	32,	6)	504	activation_12[0][0]
dropout_12 (Dropout)	(None,	32,	32,	6)	0	conv2d_13[0][0]
average_pooling2d (AveragePooli	(None,	16,	16,	6)	0	dropout_12[0][0]
batch_normalization_13 (BatchNo	(None,	16,	16,	6)	24	average_pooling2d[0][0]
activation_13 (Activation)	(None,	16,	16,	6)	0	batch_normalization_13[0][0]
conv2d_14 (Conv2D)	(None,	16,	16,	6)	324	activation_13[0][0]
dropout_13 (Dropout)	(None,	16,	16,	6)	0	conv2d_14[0][0]
concatenate_12 (Concatenate)	(None,	16,	16,	12)	0	average_pooling2d[0][0] dropout_13[0][0]
batch_normalization_14 (BatchNo	(None,	16,	16,	12)	48	concatenate_12[0][0]
activation_14 (Activation)	(None,	16,	16,	12)	0	batch_normalization_14[0][0]
conv2d_15 (Conv2D)	(None,	16,	16,	6)	648	activation_14[0][0]
dropout_14 (Dropout)	(None,	16,	16,	6)	0	conv2d_15[0][0]
concatenate_13 (Concatenate)	(None,	16,	16,	18)	0	concatenate_12[0][0] dropout_14[0][0]
batch_normalization_15 (BatchNo	(None,	16,	16,	18)	72	concatenate_13[0][0]
activation_15 (Activation)	(None,	16,	16,	18)	0	batch_normalization_15[0][0]
conv2d_16 (Conv2D)	(None,	16,	16,	6)	972	activation_15[0][0]
dropout_15 (Dropout)	(None,	16,	16,	6)	0	conv2d_16[0][0]
concatenate_14 (Concatenate)	(None,	16,	16,	24)	0	concatenate_13[0][0] dropout_15[0][0]
batch_normalization_16 (BatchNo	(None,	16,	16,	24)	96	concatenate_14[0][0]
activation_16 (Activation)	(None,	16,	16,	24)	0	batch_normalization_16[0][0]
conv2d_17 (Conv2D)	(None,	16,	16,	6)	1296	activation_16[0][0]
dropout_16 (Dropout)	(None,	16,	16,	6)	0	conv2d_17[0][0]
concatenate_15 (Concatenate)	(None,	16,	16,	30)	0	concatenate_14[0][0] dropout_16[0][0]
batch_normalization_17 (BatchNo	(None,	16,	16,	30)	120	concatenate_15[0][0]
activation_17 (Activation)	(None,	16,	16,	30)	0	batch_normalization_17[0][0]
conv2d_18 (Conv2D)	(None,	16,	16,	6)	1620	activation_17[0][0]
dropout_17 (Dropout)	(None,	16,	16,	6)	0	conv2d_18[0][0]
concatenate_16 (Concatenate)	(None,	16,	16,	36)	0	concatenate_15[0][0] dropout_17[0][0]
batch_normalization_18 (BatchNo	(None,	16,	16,	36)	144	concatenate_16[0][0]
activation_18 (Activation)	(None,	16,	16,	36)	0	batch_normalization_18[0][0]
conv2d_19 (Conv2D)	(None,	16,	16,	6)	1944	activation_18[0][0]
dropout_18 (Dropout)	(None,	16,	16,	6)	0	conv2d_19[0][0]
concatenate_17 (Concatenate)	(None,	16,	16,	42)	0	concatenate_16[0][0] dropout_18[0][0]
batch_normalization_19 (BatchNo	(None,	16,	16,	42)	168	concatenate_17[0][0]
activation_19 (Activation)	(None,	16,	16,	42)	0	batch_normalization_19[0][0]

conv2d_20 (Conv2D)	(None,	16,	16,	6)	2268	activation_19[0][0]
dropout_19 (Dropout)	(None,	16,	16,	6)	0	conv2d_20[0][0]
concatenate_18 (Concatenate)	(None,	16,	16,	48)	0	concatenate_17[0][0] dropout_19[0][0]
batch_normalization_20 (BatchNo	(None,	16,	16,	48)	192	concatenate_18[0][0]
activation_20 (Activation)	(None,	16,	16,	48)	0	batch_normalization_20[0][0]
conv2d_21 (Conv2D)	(None,	16,	16,	6)	2592	activation_20[0][0]
dropout_20 (Dropout)	(None,	16,	16,	6)	0	conv2d_21[0][0]
concatenate_19 (Concatenate)	(None,	16,	16,	54)	0	concatenate_18[0][0] dropout_20[0][0]
batch_normalization_21 (BatchNo	(None,	16,	16,	54)	216	concatenate_19[0][0]
activation_21 (Activation)	(None,	16,	16,	54)	0	batch_normalization_21[0][0]
conv2d_22 (Conv2D)	(None,	16,	16,	6)	2916	activation_21[0][0]
dropout_21 (Dropout)	(None,	16,	16,	6)	0	conv2d_22[0][0]
concatenate_20 (Concatenate)	(None,	16,	16,	60)	0	concatenate_19[0][0] dropout_21[0][0]
batch_normalization_22 (BatchNo	(None,	16,	16,	60)	240	concatenate_20[0][0]
activation_22 (Activation)	(None,	16,	16,	60)	0	batch_normalization_22[0][0]
conv2d_23 (Conv2D)	(None,	16,	16,	6)	3240	activation_22[0][0]
dropout_22 (Dropout)	(None,	16,	16,	6)	0	conv2d_23[0][0]
concatenate_21 (Concatenate)	(None,	16,	16,	66)	0	concatenate_20[0][0] dropout_22[0][0]
batch_normalization_23 (BatchNo	(None,	16,	16,	66)	264	concatenate_21[0][0]
activation_23 (Activation)	(None,	16,	16,	66)	0	batch_normalization_23[0][0]
conv2d_24 (Conv2D)	(None,	16,	16,	6)	3564	activation_23[0][0]
dropout_23 (Dropout)	(None,	16,	16,	6)	0	conv2d_24[0][0]
concatenate_22 (Concatenate)	(None,	16,	16,	72)	0	concatenate_21[0][0] dropout_23[0][0]
batch_normalization_24 (BatchNo	(None,	16,	16,	72)	288	concatenate_22[0][0]
activation_24 (Activation)	(None,	16,	16,	72)	0	batch_normalization_24[0][0]
conv2d_25 (Conv2D)	(None,	16,	16,	6)	3888	activation_24[0][0]
dropout_24 (Dropout)	(None,	16,	16,	6)	0	conv2d_25[0][0]
concatenate_23 (Concatenate)	(None,	16,	16,	78)	0	concatenate_22[0][0] dropout_24[0][0]
batch_normalization_25 (BatchNo	(None,	16,	16,	78)	312	concatenate_23[0][0]
activation_25 (Activation)	(None,	16,	16,	78)	0	batch_normalization_25[0][0]
conv2d_26 (Conv2D)	(None,	16,	16,	6)	468	activation_25[0][0]
dropout_25 (Dropout)	(None,	16,	16,	6)	0	conv2d_26[0][0]
average_pooling2d_1 (AveragePoo	(None,	8,	8, 6	)	0	dropout_25[0][0]
batch_normalization_26 (BatchNo	(None,	8,	8, 6	)	24	average_pooling2d_1[0][0]
activation_26 (Activation)	(None,	8,	8, 6	)	0	batch_normalization_26[0][0]
control 27 (Control)	/None	Ω	ρ ₆	١	321	activation 26[0][0]

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dropout_26 (Dropout)	(None,	8,	8,	6)	0	conv2d_27[0][0]
concatenate_24 (Concatenate)	(None,	8,	8,	12)	0	average_pooling2d_1[0][0] dropout_26[0][0]
batch_normalization_27 (BatchNo	(None,	8,	8,	12)	48	concatenate_24[0][0]
activation_27 (Activation)	(None,	8,	8,	12)	0	batch_normalization_27[0][0]
conv2d_28 (Conv2D)	(None,	8,	8,	6)	648	activation_27[0][0]
dropout_27 (Dropout)	(None,	8,	8,	6)	0	conv2d_28[0][0]
concatenate_25 (Concatenate)	(None,	8,	8,	18)	0	concatenate_24[0][0] dropout_27[0][0]
batch_normalization_28 (BatchNo	(None,	8,	8,	18)	72	concatenate_25[0][0]
activation_28 (Activation)	(None,	8,	8,	18)	0	batch_normalization_28[0][0]
conv2d_29 (Conv2D)	(None,	8,	8,	6)	972	activation_28[0][0]
dropout_28 (Dropout)	(None,	8,	8,	6)	0	conv2d_29[0][0]
concatenate_26 (Concatenate)	(None,	8,	8,	24)	0	concatenate_25[0][0] dropout_28[0][0]
batch_normalization_29 (BatchNo	(None,	8,	8,	24)	96	concatenate_26[0][0]
activation_29 (Activation)	(None,	8,	8,	24)	0	batch_normalization_29[0][0]
conv2d_30 (Conv2D)	(None,	8,	8,	6)	1296	activation_29[0][0]
dropout_29 (Dropout)	(None,	8,	8,	6)	0	conv2d_30[0][0]
concatenate_27 (Concatenate)	(None,	8,	8,	30)	0	concatenate_26[0][0] dropout_29[0][0]
batch_normalization_30 (BatchNo	(None,	8,	8,	30)	120	concatenate_27[0][0]
activation_30 (Activation)	(None,	8,	8,	30)	0	batch_normalization_30[0][0]
conv2d_31 (Conv2D)	(None,	8,	8,	6)	1620	activation_30[0][0]
dropout_30 (Dropout)	(None,	8,	8,	6)	0	conv2d_31[0][0]
concatenate_28 (Concatenate)	(None,	8,	8,	36)	0	concatenate_27[0][0] dropout_30[0][0]
batch_normalization_31 (BatchNo	(None,	8,	8,	36)	144	concatenate_28[0][0]
activation_31 (Activation)	(None,	8,	8,	36)	0	batch_normalization_31[0][0]
conv2d_32 (Conv2D)	(None,	8,	8,	6)	1944	activation_31[0][0]
dropout_31 (Dropout)	(None,	8,	8,	6)	0	conv2d_32[0][0]
concatenate_29 (Concatenate)	(None,	8,	8,	42)	0	concatenate_28[0][0] dropout_31[0][0]
batch_normalization_32 (BatchNo	(None,	8,	8,	42)	168	concatenate_29[0][0]
activation_32 (Activation)	(None,	8,	8,	42)	0	batch_normalization_32[0][0]
conv2d_33 (Conv2D)	(None,	8,	8,	6)	2268	activation_32[0][0]
dropout_32 (Dropout)	(None,	8,	8,	6)	0	conv2d_33[0][0]
concatenate_30 (Concatenate)	(None,	8,	8,	48)	0	concatenate_29[0][0] dropout_32[0][0]
batch_normalization_33 (BatchNo	(None,	8,	8,	48)	192	concatenate_30[0][0]
activation_33 (Activation)	(None,	8,	8,	48)	0	batch_normalization_33[0][0]
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dropout_33 (Dropout)	(None,	8,	8,	6)	0	conv2d_34[0][0]
concatenate_31 (Concatenate)	(None,	8,	8,	54)	0	concatenate_30[0][0] dropout_33[0][0]
batch_normalization_34 (BatchNo	(None,	8,	8,	54)	216	concatenate_31[0][0]
activation_34 (Activation)	(None,	8,	8,	54)	0	batch_normalization_34[0][0]
conv2d_35 (Conv2D)	(None,	8,	8,	6)	2916	activation_34[0][0]
dropout_34 (Dropout)	(None,	8,	8,	6)	0	conv2d_35[0][0]
concatenate_32 (Concatenate)	(None,	8,	8,	60)	0	concatenate_31[0][0] dropout_34[0][0]
batch_normalization_35 (BatchNo	(None,	8,	8,	60)	240	concatenate_32[0][0]
activation_35 (Activation)	(None,	8,	8,	60)	0	batch_normalization_35[0][0]
conv2d_36 (Conv2D)	(None,	8,	8,	6)	3240	activation_35[0][0]
dropout_35 (Dropout)	(None,	8,	8,	6)	0	conv2d_36[0][0]
concatenate_33 (Concatenate)	(None,	8,	8,	66)	0	concatenate_32[0][0] dropout_35[0][0]
oatch_normalization_36 (BatchNo	(None,	8,	8,	66)	264	concatenate_33[0][0]
activation_36 (Activation)	(None,	8,	8,	66)	0	batch_normalization_36[0][0]
conv2d_37 (Conv2D)	(None,	8,	8,	6)	3564	activation_36[0][0]
dropout_36 (Dropout)	(None,	8,	8,	6)	0	conv2d_37[0][0]
concatenate_34 (Concatenate)	(None,	8,	8,	72)	0	concatenate_33[0][0] dropout_36[0][0]
batch_normalization_37 (BatchNo	(None,	8,	8,	72)	288	concatenate_34[0][0]
activation_37 (Activation)	(None,	8,	8,	72)	0	batch_normalization_37[0][0]
conv2d_38 (Conv2D)	(None,	8,	8,	6)	3888	activation_37[0][0]
dropout_37 (Dropout)	(None,	8,	8,	6)	0	conv2d_38[0][0]
concatenate_35 (Concatenate)	(None,	8,	8,	78)	0	concatenate_34[0][0] dropout_37[0][0]
oatch_normalization_38 (BatchNo	(None,	8,	8,	78)	312	concatenate_35[0][0]
activation_38 (Activation)	(None,	8,	8,	78)	0	batch_normalization_38[0][0]
conv2d_39 (Conv2D)	(None,	8,	8,	6)	468	activation_38[0][0]
dropout_38 (Dropout)	(None,	8,	8,	6)	0	conv2d_39[0][0]
average_pooling2d_2 (AveragePoo	(None,	4,	4,	6)	0	dropout_38[0][0]
batch_normalization_39 (BatchNo	(None,	4,	4,	6)	24	average_pooling2d_2[0][0]
activation_39 (Activation)	(None,	4,	4,	6)	0	batch_normalization_39[0][0]
conv2d_40 (Conv2D)	(None,	4,	4,	6)	324	activation_39[0][0]
dropout_39 (Dropout)	(None,	4,	4,	6)	0	conv2d_40[0][0]
concatenate_36 (Concatenate)	(None,	4,	4,	12)	0	average_pooling2d_2[0][0] dropout_39[0][0]
batch_normalization_40 (BatchNo	(None,	4,	4,	12)	48	concatenate_36[0][0]
activation_40 (Activation)	(None,	4,	4,	12)	0	batch_normalization_40[0][0]
conv2d_41 (Conv2D)	(None,	4,	4,	6)	648	activation_40[0][0]

dropout_40 (Dropout)	(None,	4,	4,	6)	0	conv2d_41[0][0]
concatenate_37 (Concatenate)	(None,	4,	4,	18)	0	concatenate_36[0][0] dropout_40[0][0]
batch_normalization_41 (BatchNo	(None,	4,	4,	18)	72	concatenate_37[0][0]
activation_41 (Activation)	(None,	4,	4,	18)	0	batch_normalization_41[0][0]
conv2d_42 (Conv2D)	(None,	4,	4,	6)	972	activation_41[0][0]
dropout_41 (Dropout)	(None,	4,	4,	6)	0	conv2d_42[0][0]
concatenate_38 (Concatenate)	(None,	4,	4,	24)	0	concatenate_37[0][0] dropout_41[0][0]
<pre>batch_normalization_42 (BatchNo</pre>	(None,	4,	4,	24)	96	concatenate_38[0][0]
activation_42 (Activation)	(None,	4,	4,	24)	0	batch_normalization_42[0][0]
conv2d_43 (Conv2D)	(None,	4,	4,	6)	1296	activation_42[0][0]
dropout_42 (Dropout)	(None,	4,	4,	6)	0	conv2d_43[0][0]
concatenate_39 (Concatenate)	(None,	4,	4,	30)	0	concatenate_38[0][0] dropout_42[0][0]
<pre>batch_normalization_43 (BatchNo</pre>	(None,	4,	4,	30)	120	concatenate_39[0][0]
activation_43 (Activation)	(None,	4,	4,	30)	0	batch_normalization_43[0][0]
conv2d_44 (Conv2D)	(None,	4,	4,	6)	1620	activation_43[0][0]
dropout_43 (Dropout)	(None,	4,	4,	6)	0	conv2d_44[0][0]
concatenate_40 (Concatenate)	(None,	4,	4,	36)	0	concatenate_39[0][0] dropout_43[0][0]
<pre>batch_normalization_44 (BatchNo</pre>	(None,	4,	4,	36)	144	concatenate_40[0][0]
activation_44 (Activation)	(None,	4,	4,	36)	0	batch_normalization_44[0][0]
conv2d_45 (Conv2D)	(None,	4,	4,	6)	1944	activation_44[0][0]
dropout_44 (Dropout)	(None,	4,	4,	6)	0	conv2d_45[0][0]
concatenate_41 (Concatenate)	(None,	4,	4,	42)	0	concatenate_40[0][0] dropout_44[0][0]
<pre>batch_normalization_45 (BatchNo</pre>	(None,	4,	4,	42)	168	concatenate_41[0][0]
activation_45 (Activation)	(None,	4,	4,	42)	0	batch_normalization_45[0][0]
conv2d_46 (Conv2D)	(None,	4,	4,	6)	2268	activation_45[0][0]
dropout_45 (Dropout)	(None,	4,	4,	6)	0	conv2d_46[0][0]
concatenate_42 (Concatenate)	(None,	4,	4,	48)	0	concatenate_41[0][0] dropout_45[0][0]
batch_normalization_46 (BatchNo	(None,	4,	4,	48)	192	concatenate_42[0][0]
activation_46 (Activation)	(None,	4,	4,	48)	0	batch_normalization_46[0][0]
conv2d_47 (Conv2D)	(None,	4,	4,	6)	2592	activation_46[0][0]
dropout_46 (Dropout)	(None,	4,	4,	6)	0	conv2d_47[0][0]
concatenate_43 (Concatenate)	(None,	4,	4,	54)	0	concatenate_42[0][0] dropout_46[0][0]
batch_normalization_47 (BatchNo	(None,	4,	4,	54)	216	concatenate_43[0][0]
activation_47 (Activation)	(None,	4,	4,	54)	0	batch_normalization_47[0][0]
conv2d_48 (Conv2D)	(None,	4,	4,	6)	2916	activation_47[0][0]

dropout_47 (Dropout)	(None,	4, 4	1,	6)	0	conv2d_48[0][0]
concatenate_44 (Concatenate)	(None,	4, 4	1,	60)	0	concatenate_43[0][0] dropout_47[0][0]
batch_normalization_48 (BatchNo	(None,	4, 4	1,	60)	240	concatenate_44[0][0]
activation_48 (Activation)	(None,	4, 4	1,	60)	0	batch_normalization_48[0][0]
conv2d_49 (Conv2D)	(None,	4, 4	1,	6)	3240	activation_48[0][0]
dropout_48 (Dropout)	(None,	4, 4	1,	6)	0	conv2d_49[0][0]
concatenate_45 (Concatenate)	(None,	4, 4	1,	66)	0	concatenate_44[0][0] dropout_48[0][0]
batch_normalization_49 (BatchNo	(None,	4, 4	1,	66)	264	concatenate_45[0][0]
activation_49 (Activation)	(None,	4, 4	1,	66)	0	batch_normalization_49[0][0]
conv2d_50 (Conv2D)	(None,	4, 4	1,	6)	3564	activation_49[0][0]
dropout_49 (Dropout)	(None,	4, 4	1,	6)	0	conv2d_50[0][0]
concatenate_46 (Concatenate)	(None,	4, 4	1,	72)	0	concatenate_45[0][0] dropout_49[0][0]
batch_normalization_50 (BatchNo	(None,	4, 4	1,	72)	288	concatenate_46[0][0]
activation_50 (Activation)	(None,	4, 4	1,	72)	0	batch_normalization_50[0][0]
conv2d_51 (Conv2D)	(None,	4, 4	1,	6)	3888	activation_50[0][0]
dropout_50 (Dropout)	(None,	4, 4	1,	6)	0	conv2d_51[0][0]
concatenate_47 (Concatenate)	(None,	4, 4	1,	78)	0	concatenate_46[0][0] dropout_50[0][0]
batch_normalization_51 (BatchNo	(None,	4, 4	1,	78)	312	concatenate_47[0][0]
activation_51 (Activation)	(None,	4, 4	1,	78)	0	batch_normalization_51[0][0]
average_pooling2d_3 (AveragePoo	(None,	2, 2	2,	78)	0	activation_51[0][0]
flatten (Flatten)	(None,	312)			0	average_pooling2d_3[0][0]
dense (Dense)	(None,	10)			3130	flatten[0][0]

Total params: 118,918 Trainable params: 114,394 Non-trainable params: 4,524

### In [42]:

```
model_2.compile(optimizer=tf.keras.optimizers.Adam(learning_rate=0.0001),
loss='categorical_crossentropy', metrics=['accuracy'])
```

### In [43]:

```
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
patience = 50
base_path = '/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/'
checkpoint_file_name = base_path + 'CIFAR_model2' + '_{epoch:02d}-{val_accuracy:.2f}.hdf5'

model_checkpoint = ModelCheckpoint(checkpoint_file_name, monitor='val_accuracy', mode='max', verbos
e=1, save_best_only=True)
early_stop = EarlyStopping('val_accuracy', mode='max', patience = patience)
reduce_LR = ReduceLROnPlateau(monitor='val_accuracy', mode='max', factor=0.1, patience=int(patience
/3))
```

#### In [44]:

```
callbacks = [model_checkpoint, early_stop, reduce_LR]
```

#### In [45]:

```
epochs = 300
batch_size = 128
```

#### In [46]:

```
WARNING:tensorflow:sample_weight modes were coerced from
 ['...']
Train for 390 steps, validate on 10000 samples
Epoch 1/300
Epoch 00001: val accuracy improved from -inf to 0.80040, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model2 01-0.80.hdf5
oss: 0.6419 - val_accuracy: 0.8004
Epoch 2/300
Epoch 00002: val accuracy improved from 0.80040 to 0.80220, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model2_02-0.80.hdf5
oss: 0.6375 - val accuracy: 0.8022
Epoch 3/300
Epoch 00003: val accuracy did not improve from 0.80220
oss: 0.6571 - val accuracy: 0.7997
Epoch 4/300
Epoch 00004: val accuracy improved from 0.80220 to 0.81070, saving model to
/home/ubuntu/Project/my_data/CNN CIFAR/checkpoint/CIFAR model2 04-0.81.hdf5
oss: 0.6188 - val_accuracy: 0.8107
Epoch 5/300
Epoch 00005: val_accuracy did not improve from 0.81070
oss: 0.6297 - val accuracy: 0.8076
Epoch 6/300
Epoch 00006: val accuracy did not improve from 0.81070
oss: 0.6591 - val_accuracy: 0.8010
Epoch 7/300
Epoch 00007: val_accuracy did not improve from 0.81070
390/390 [============= ] - 28s 72ms/step - loss: 0.5492 - accuracy: 0.8074 - val 1
oss: 0.6238 - val_accuracy: 0.8076
Epoch 8/300
Epoch 00008: val accuracy did not improve from 0.81070
390/390 [============ ] - 28s 72ms/step - loss: 0.5467 - accuracy: 0.8092 - val 1
oss: 0.6312 - val_accuracy: 0.8067
Epoch 9/300
Epoch 00009: val accuracy did not improve from 0.81070
oss: 0.6622 - val_accuracy: 0.8012
Epoch 10/300
Epoch 00010: val_accuracy did not improve from 0.81070
oss. U 820 - Asl acchrach. U 8008
```

```
033. 0.0020
       var accuracy. U.UUU
Epoch 11/300
Epoch 00011: val_accuracy did not improve from 0.81070
oss: 0.6431 - val accuracy: 0.8018
Epoch 12/300
Epoch 00012: val accuracy did not improve from 0.81070
390/390 [============= ] - 28s 72ms/step - loss: 0.5409 - accuracy: 0.8104 - val 1
oss: 0.6439 - val accuracy: 0.8050
Epoch 13/300
Epoch 00013: val accuracy did not improve from 0.81070
390/390 [============ ] - 28s 72ms/step - loss: 0.5463 - accuracy: 0.8081 - val 1
oss: 0.6619 - val accuracy: 0.8008
Epoch 14/300
Epoch 00014: val_accuracy did not improve from 0.81070
390/390 [============ ] - 28s 72ms/step - loss: 0.5452 - accuracy: 0.8081 - val 1
oss: 0.6316 - val accuracy: 0.8095
Epoch 15/300
Epoch 00015: val accuracy improved from 0.81070 to 0.81330, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model2 15-0.81.hdf5
390/390 [============ ] - 29s 74ms/step - loss: 0.5492 - accuracy: 0.8097 - val 1
oss: 0.6198 - val accuracy: 0.8133
Epoch 16/300
Epoch 00016: val_accuracy did not improve from 0.81330
390/390 [============= ] - 28s 72ms/step - loss: 0.5433 - accuracy: 0.8091 - val 1
oss: 0.6484 - val_accuracy: 0.8048
Epoch 17/300
Epoch 00017: val_accuracy did not improve from 0.81330
390/390 [============= ] - 28s 72ms/step - loss: 0.5447 - accuracy: 0.8112 - val 1
oss: 0.6252 - val accuracy: 0.8105
Epoch 18/300
Epoch 00018: val accuracy did not improve from 0.81330
oss: 0.6236 - val accuracy: 0.8098
Epoch 19/300
Epoch 00019: val accuracy did not improve from 0.81330
390/390 [============= ] - 28s 72ms/step - loss: 0.5432 - accuracy: 0.8096 - val_1
oss: 0.6261 - val_accuracy: 0.8099
Epoch 20/300
Epoch 00020: val accuracy did not improve from 0.81330
390/390 [============= ] - 28s 72ms/step - loss: 0.5447 - accuracy: 0.8092 - val 1
oss: 0.6298 - val_accuracy: 0.8111
Epoch 21/300
Epoch 00021: val accuracy did not improve from 0.81330
390/390 [============ ] - 28s 72ms/step - loss: 0.5438 - accuracy: 0.8088 - val 1
oss: 0.6196 - val_accuracy: 0.8110
Epoch 22/300
Epoch 00022: val accuracy did not improve from 0.81330
oss: 0.6491 - val_accuracy: 0.8020
Epoch 23/300
Epoch 00023: val accuracy did not improve from 0.81330
390/390 [============ ] - 28s 72ms/step - loss: 0.5433 - accuracy: 0.8100 - val 1
oss: 0.6528 - val accuracy: 0.8067
Epoch 24/300
Epoch 00024: val accuracy did not improve from 0.81330
oss: 0.6227 - val accuracy: 0.8119
Epoch 25/300
Epoch 00025: val accuracy did not improve from 0.81330
oss: 0.6399 - val accuracy: 0.8086
```

Fnoch 26/300

```
EPUCII ZU/JUU
Epoch 00026: val_accuracy did not improve from 0.81330
oss: 0.6489 - val accuracy: 0.8045
Epoch 27/300
Epoch 00027: val accuracy did not improve from 0.81330
390/390 [============= ] - 28s 73ms/step - loss: 0.5461 - accuracy: 0.8083 - val 1
oss: 0.6226 - val accuracy: 0.8121
Epoch 28/300
Epoch 00028: val accuracy did not improve from 0.81330
oss: 0.6395 - val accuracy: 0.8060
Epoch 29/300
Epoch 00029: val accuracy did not improve from 0.81330
390/390 [============ ] - 28s 73ms/step - loss: 0.5368 - accuracy: 0.8132 - val_1
oss: 0.6366 - val_accuracy: 0.8097
Epoch 30/300
Epoch 00030: val accuracy did not improve from 0.81330
oss: 0.6457 - val_accuracy: 0.8061
Epoch 31/300
Epoch 00031: val accuracy did not improve from 0.81330
oss: 0.6383 - val_accuracy: 0.8075
Epoch 32/300
Epoch 00032: val_accuracy did not improve from 0.81330
oss: 0.6434 - val accuracy: 0.8048
Epoch 33/300
Epoch 00033: val accuracy did not improve from 0.81330
390/390 [============ ] - 27s 70ms/step - loss: 0.5378 - accuracy: 0.8125 - val 1
oss: 0.6457 - val accuracy: 0.8052
Epoch 34/300
Epoch 00034: val accuracy did not improve from 0.81330
390/390 [============ ] - 28s 71ms/step - loss: 0.5308 - accuracy: 0.8135 - val 1
oss: 0.6438 - val accuracy: 0.8054
Epoch 35/300
Epoch 00035: val accuracy did not improve from 0.81330
390/390 [============ ] - 27s 70ms/step - loss: 0.5360 - accuracy: 0.8118 - val 1
oss: 0.6379 - val_accuracy: 0.8071
Epoch 36/300
Epoch 00036: val accuracy did not improve from 0.81330
390/390 [============ ] - 27s 70ms/step - loss: 0.5380 - accuracy: 0.8111 - val 1
oss: 0.6425 - val accuracy: 0.8069
Epoch 37/300
Epoch 00037: val accuracy did not improve from 0.81330
oss: 0.6417 - val accuracy: 0.8068
Epoch 38/300
Epoch 00038: val_accuracy did not improve from 0.81330
390/390 [============== ] - 28s 71ms/step - loss: 0.5344 - accuracy: 0.8134 - val 1
oss: 0.6397 - val_accuracy: 0.8076
Epoch 39/300
Epoch 00039: val accuracy did not improve from 0.81330
oss: 0.6358 - val accuracy: 0.8082
Epoch 40/300
Epoch 00040: val accuracy did not improve from 0.81330
390/390 [============= ] - 28s 71ms/step - loss: 0.5308 - accuracy: 0.8141 - val 1
oss: 0.6359 - val accuracy: 0.8076
Epoch 41/300
```

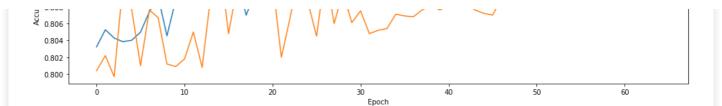
Enoch 00041. ... | 20012001 did not improve from 0 01220

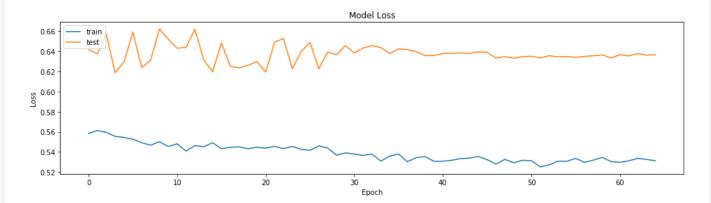
```
Epoch 00041: val_accuracy ata not improve from 0.01330
390/390 [============ ] - 28s 72ms/step - loss: 0.5308 - accuracy: 0.8154 - val 1
oss: 0.6379 - val_accuracy: 0.8082
Epoch 42/300
Epoch 00042: val_accuracy did not improve from 0.81330
390/390 [============= ] - 28s 72ms/step - loss: 0.5318 - accuracy: 0.8147 - val 1
oss: 0.6380 - val accuracy: 0.8082
Epoch 43/300
Epoch 00043: val accuracy did not improve from 0.81330
oss: 0.6386 - val accuracy: 0.8086
Epoch 44/300
Epoch 00044: val accuracy did not improve from 0.81330
390/390 [============= ] - 28s 72ms/step - loss: 0.5340 - accuracy: 0.8133 - val 1
oss: 0.6378 - val_accuracy: 0.8076
Epoch 45/300
Epoch 00045: val accuracy did not improve from 0.81330
390/390 [============= ] - 28s 72ms/step - loss: 0.5355 - accuracy: 0.8113 - val 1
oss: 0.6396 - val_accuracy: 0.8072
Epoch 46/300
Epoch 00046: val accuracy did not improve from 0.81330
390/390 [============ ] - 28s 73ms/step - loss: 0.5324 - accuracy: 0.8159 - val 1
oss: 0.6391 - val_accuracy: 0.8070
Epoch 47/300
Epoch 00047: val_accuracy did not improve from 0.81330
oss: 0.6334 - val accuracy: 0.8089
Epoch 48/300
Epoch 00048: val accuracy did not improve from 0.81330
390/390 [============ ] - 28s 72ms/step - loss: 0.5326 - accuracy: 0.8127 - val 1
oss: 0.6348 - val accuracy: 0.8083
Epoch 49/300
Epoch 00049: val accuracy did not improve from 0.81330
oss: 0.6332 - val accuracy: 0.8091
Epoch 50/300
Epoch 00050: val accuracy did not improve from 0.81330
oss: 0.6348 - val_accuracy: 0.8085
Epoch 51/300
Epoch 00051: val accuracy did not improve from 0.81330
oss: 0.6352 - val_accuracy: 0.8082
Epoch 52/300
Epoch 00052: val accuracy did not improve from 0.81330
oss: 0.6336 - val_accuracy: 0.8087
Epoch 53/300
Epoch 00053: val accuracy did not improve from 0.81330
oss: 0.6357 - val_accuracy: 0.8085
Epoch 54/300
Epoch 00054: val accuracy did not improve from 0.81330
390/390 [============ ] - 28s 72ms/step - loss: 0.5308 - accuracy: 0.8143 - val 1
oss: 0.6346 - val accuracy: 0.8083
Epoch 55/300
Epoch 00055: val accuracy did not improve from 0.81330
390/390 [============ ] - 28s 73ms/step - loss: 0.5308 - accuracy: 0.8147 - val 1
oss: 0.6348 - val accuracy: 0.8086
Epoch 56/300
Epoch 00056: val accuracy did not improve from 0.81330
390/390 [============ ] - 28s 72ms/step - loss: 0.5336 - accuracy: 0.8118 - val_1
```

--- 0 6240 ---1 ------ 0 0000

```
oss: U.6342 - Val accuracy: U.8092
Epoch 57/300
Epoch 00057: val_accuracy did not improve from 0.81330
oss: 0.6349 - val_accuracy: 0.8082
Epoch 58/300
Epoch 00058: val accuracy did not improve from 0.81330
oss: 0.6356 - val accuracy: 0.8085
Epoch 59/300
Epoch 00059: val accuracy did not improve from 0.81330
oss: 0.6365 - val accuracy: 0.8081
Epoch 60/300
Epoch 00060: val accuracy did not improve from 0.81330
390/390 [============== ] - 28s 72ms/step - loss: 0.5304 - accuracy: 0.8126 - val 1
oss: 0.6335 - val_accuracy: 0.8082
Epoch 61/300
Epoch 00061: val_accuracy did not improve from 0.81330
oss: 0.6367 - val_accuracy: 0.8081
Epoch 62/300
Epoch 00062: val accuracy did not improve from 0.81330
oss: 0.6355 - val_accuracy: 0.8080
Epoch 63/300
Epoch 00063: val_accuracy did not improve from 0.81330
390/390 [============= ] - 28s 72ms/step - loss: 0.5335 - accuracy: 0.8119 - val 1
oss: 0.6378 - val_accuracy: 0.8079
Epoch 64/300
Epoch 00064: val_accuracy did not improve from 0.81330
390/390 [=================== ] - 28s 73ms/step - loss: 0.5326 - accuracy: 0.8157 - val 1
oss: 0.6364 - val accuracy: 0.8078
Epoch 65/300
Epoch 00065: val accuracy did not improve from 0.81330
390/390 [============= ] - 28s 72ms/step - loss: 0.5312 - accuracy: 0.8159 - val 1
oss: 0.6366 - val accuracy: 0.8083
In [47]:
#history plot for accyracy
plt.figure(figsize=(16,4))
plt.plot(history 2.history['accuracy'])
plt.plot(history_2.history['val_accuracy'])
plt.title("Model Accuracy")
plt.xlabel("Epoch")
plt.ylabel("Accuracy")
plt.legend(["train", "test"], loc="upper left")
plt.show()
# history plot for accuracy
plt.figure(figsize=(16,4))
plt.plot(history 2.history["loss"])
plt.plot(history_2.history["val_loss"])
plt.title("Model Loss")
plt.xlabel("Epoch")
plt.ylabel("Loss")
plt.legend(["train", "test"], loc="upper left")
plt.show()
```







### **Summary:**

 Even after changing the learning rate the model accuracy is not improved. Next we can try is different convolution and different num of filters

#### In [65]:

### 12. Model-3

• Seperable convolution

[0.6198125976085663, 0.8133]

• num filters 32

### In [52]:

```
#BN-->ReLU-->Conv2D-->Dropout-->concat(input, output)-->(put in loop)
def denseblock(input, num filter, dropout rate = 0.2):
   global compression
                          # to keep the growth rate of number of filters
   temp = input
   for in range(l):
       BatchNorm = layers.BatchNormalization()(temp)
       relu = layers.Activation('relu')(BatchNorm)
       Conv2D 5 5= layers.SeparableConv2D(int(num filter*compression), (5,5), use bias=False, padd
ing='same') (relu)
        #Conv2D 3 3 = layers.Conv2D(int(num filter*compression), (3,3), use bias=False
,padding='same') (relu)
       if dropout rate>0:
           Conv2D 5 5 = layers.Dropout(dropout rate)(Conv2D 5 5)
        \#concat the input(temp) and output(conv2d 3 3) , in resnet we add but here we concat
        concat = layers.Concatenate(axis=-1)([temp,Conv2D_5_5])
        #change the concat as input
        temp = concat
   return temp
```

```
\#BN-->relu-->conv2d(1x1)-->dropout-->avg pool
def transition(input, num filter, 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)
    Conv2D BottleNeck = layers.SeparableConvolution2D(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
#BN-->relu-->avgpool-->flat-->softmax
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
# Hyperparameters
1 = 12
num_filter = 32
compression = 0.5
dropout_rate = 0.2
num classes = 10
input = layers.Input(shape=(input size))
First Conv2D = layers.Conv2D(num filter, (5,5), use bias=False ,padding='same')(input)
#First dense and transition block
First Block = denseblock(First Conv2D, num filter, dropout rate)
First Transition = transition(First_Block, num_filter, dropout_rate)
#Second dense and transition block
Second Block = denseblock(First Transition, num filter, dropout rate)
Second Transition = transition(Second Block, num filter, dropout rate)
#Third dense and transition block
Third_Block = denseblock(Second_Transition, num_filter, dropout_rate)
Third_Transition = transition(Third_Block, num_filter, dropout_rate)
#last dense and output block
Last Block = denseblock(Third Transition, num filter, dropout rate)
output = output layer(Last Block)
```

#### In [53]:

```
model_3 = tf.keras.models.Model(inputs=[input], outputs=[output])
model_3.summary()
```

Model: "model_2"

Layer (type)	Output Shape	Param #	Connected to
input_3 (InputLayer)	[(None, 32, 32,	3)] 0	
conv2d_53 (Conv2D)	(None, 32, 32,	32) 2400	input_3[0][0]
batch_normalization_104 (BatchN	(None, 32, 32,	32) 128	conv2d_53[0][0]
activation_104 (Activation)	(None, 32, 32,	32) 0	batch_normalization_104[0][0]
separable_conv2d_51 (SeparableC	(None, 32, 32,	16) 1312	activation_104[0][0]
dropout_102 (Dropout)	(None, 32, 32,	16) 0	separable_conv2d_51[0][0]
concatenate_96 (Concatenate)	(None, 32, 32,	48) 0	conv2d_53[0][0] dropout_102[0][0]
batch normalization 105 (BatchN	(None, 32, 32,	48) 192	concatenate 96[0][0]

activation_105 (Activation)	(None,	32,	32,	48)	0	batch_normalization_105[0][0]
separable_conv2d_52 (SeparableC	(None,	32,	32,	16)	1968	activation_105[0][0]
dropout_103 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_52[0][0]
concatenate_97 (Concatenate)	(None,	32,	32,	64)	0	concatenate_96[0][0] dropout_103[0][0]
batch_normalization_106 (BatchN	(None,	32,	32,	64)	256	concatenate_97[0][0]
activation_106 (Activation)	(None,	32,	32,	64)	0	batch_normalization_106[0][0]
separable_conv2d_53 (SeparableC	(None,	32,	32,	16)	2624	activation_106[0][0]
dropout_104 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_53[0][0]
concatenate_98 (Concatenate)	(None,	32,	32,	80)	0	concatenate_97[0][0] dropout_104[0][0]
batch_normalization_107 (BatchN	(None,	32,	32,	80)	320	concatenate_98[0][0]
activation_107 (Activation)	(None,	32,	32,	80)	0	batch_normalization_107[0][0]
separable_conv2d_54 (SeparableC	(None,	32,	32,	16)	3280	activation_107[0][0]
dropout_105 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_54[0][0]
concatenate_99 (Concatenate)	(None,	32,	32,	96)	0	concatenate_98[0][0] dropout_105[0][0]
batch_normalization_108 (BatchN	(None,	32,	32,	96)	384	concatenate_99[0][0]
activation_108 (Activation)	(None,	32,	32,	96)	0	batch_normalization_108[0][0]
separable_conv2d_55 (SeparableC	(None,	32,	32,	16)	3936	activation_108[0][0]
dropout_106 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_55[0][0]
concatenate_100 (Concatenate)	(None,	32,	32,	112)	0	concatenate_99[0][0] dropout_106[0][0]
batch_normalization_109 (BatchN	(None,	32,	32,	112)	448	concatenate_100[0][0]
activation_109 (Activation)	(None,	32,	32,	112)	0	batch_normalization_109[0][0]
separable_conv2d_56 (SeparableC	(None,	32,	32,	16)	4592	activation_109[0][0]
dropout_107 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_56[0][0]
concatenate_101 (Concatenate)	(None,	32,	32,	128)	0	concatenate_100[0][0] dropout_107[0][0]
batch_normalization_110 (BatchN	(None,	32,	32,	128)	512	concatenate_101[0][0]
activation_110 (Activation)	(None,	32,	32,	128)	0	batch_normalization_110[0][0]
separable_conv2d_57 (SeparableC	(None,	32,	32,	16)	5248	activation_110[0][0]
dropout_108 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_57[0][0]
concatenate_102 (Concatenate)	(None,	32,	32,	144)	0	concatenate_101[0][0] dropout_108[0][0]
batch_normalization_111 (BatchN	(None,	32,	32,	144)	576	concatenate_102[0][0]
activation_111 (Activation)	(None,	32,	32,	144)	0	batch_normalization_111[0][0]
separable_conv2d_58 (SeparableC	(None,	32,	32,	16)	5904	activation_111[0][0]
dropout_109 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_58[0][0]
concatenate_103 (Concatenate)	(None,	32,	32,	160)	0	concatenate_102[0][0] dropout_109[0][0]
batch_normalization_112 (BatchN	(None,	32,	32,	160)	640	concatenate_103[0][0]

activation_112 (Activation)	(None,	32,	32,	160)	0	batch_normalization_112[0][0]
separable_conv2d_59 (SeparableC	(None,	32,	32,	16)	6560	activation_112[0][0]
dropout_110 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_59[0][0]
concatenate_104 (Concatenate)	(None,	32,	32,	176)	0	concatenate_103[0][0] dropout_110[0][0]
oatch_normalization_113 (BatchN	(None,	32,	32,	176)	704	concatenate_104[0][0]
activation_113 (Activation)	(None,	32,	32,	176)	0	batch_normalization_113[0][0]
separable_conv2d_60 (SeparableC	(None,	32,	32,	16)	7216	activation_113[0][0]
dropout_111 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_60[0][0]
concatenate_105 (Concatenate)	(None,	32,	32,	192)	0	concatenate_104[0][0] dropout_111[0][0]
patch_normalization_114 (BatchN	(None,	32,	32,	192)	768	concatenate_105[0][0]
activation_114 (Activation)	(None,	32,	32,	192)	0	batch_normalization_114[0][0]
separable_conv2d_61 (SeparableC	(None,	32,	32,	16)	7872	activation_114[0][0]
dropout_112 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_61[0][0]
concatenate_106 (Concatenate)	(None,	32,	32,	208)	0	concatenate_105[0][0] dropout_112[0][0]
oatch_normalization_115 (BatchN	(None,	32,	32,	208)	832	concatenate_106[0][0]
activation_115 (Activation)	(None,	32,	32,	208)	0	batch_normalization_115[0][0]
separable_conv2d_62 (SeparableC	(None,	32,	32,	16)	8528	activation_115[0][0]
dropout_113 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_62[0][0]
concatenate_107 (Concatenate)	(None,	32,	32,	224)	0	concatenate_106[0][0] dropout_113[0][0]
patch_normalization_116 (BatchN	(None,	32,	32,	224)	896	concatenate_107[0][0]
activation_116 (Activation)	(None,	32,	32,	224)	0	batch_normalization_116[0][0]
separable_conv2d_63 (SeparableC	(None,	32,	32,	16)	3808	activation_116[0][0]
dropout_114 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_63[0][0]
average_pooling2d_8 (AveragePoo	(None,	16,	16,	16)	0	dropout_114[0][0]
oatch_normalization_117 (BatchN	(None,	16,	16,	16)	64	average_pooling2d_8[0][0]
activation_117 (Activation)	(None,	16,	16,	16)	0	batch_normalization_117[0][0]
separable_conv2d_64 (SeparableC	(None,	16,	16,	16)	656	activation_117[0][0]
dropout_115 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_64[0][0]
concatenate_108 (Concatenate)	(None,	16,	16,	32)	0	average_pooling2d_8[0][0] dropout_115[0][0]
oatch_normalization_118 (BatchN	(None,	16,	16,	32)	128	concatenate_108[0][0]
activation_118 (Activation)	(None,	16,	16,	32)	0	batch_normalization_118[0][0]
separable_conv2d_65 (SeparableC	(None,	16,	16,	16)	1312	activation_118[0][0]
dropout_116 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_65[0][0]
concatenate_109 (Concatenate)	(None,	16,	16,	48)	0	concatenate_108[0][0] dropout_116[0][0]
oatch_normalization_119 (BatchN	(None,	16,	16,	48)	192	concatenate 109[0][0]

activation_119 (Activation)	(None,	16,	16,	48)	0	batch_normalization_119[0][0]
separable_conv2d_66 (SeparableC	(None,	16,	16,	16)	1968	activation_119[0][0]
dropout_117 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_66[0][0]
concatenate_110 (Concatenate)	(None,	16,	16,	64)	0	concatenate_109[0][0] dropout_117[0][0]
batch_normalization_120 (BatchN	(None,	16,	16,	64)	256	concatenate_110[0][0]
activation_120 (Activation)	(None,	16,	16,	64)	0	batch_normalization_120[0][0]
separable_conv2d_67 (SeparableC	(None,	16,	16,	16)	2624	activation_120[0][0]
dropout_118 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_67[0][0]
concatenate_111 (Concatenate)	(None,	16,	16,	80)	0	concatenate_110[0][0] dropout_118[0][0]
batch_normalization_121 (BatchN	(None,	16,	16,	80)	320	concatenate_111[0][0]
activation_121 (Activation)	(None,	16,	16,	80)	0	batch_normalization_121[0][0]
separable_conv2d_68 (SeparableC	(None,	16,	16,	16)	3280	activation_121[0][0]
dropout_119 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_68[0][0]
concatenate_112 (Concatenate)	(None,	16,	16,	96)	0	concatenate_111[0][0] dropout_119[0][0]
batch_normalization_122 (BatchN	(None,	16,	16,	96)	384	concatenate_112[0][0]
activation_122 (Activation)	(None,	16,	16,	96)	0	batch_normalization_122[0][0]
separable_conv2d_69 (SeparableC	(None,	16,	16,	16)	3936	activation_122[0][0]
dropout_120 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_69[0][0]
concatenate_113 (Concatenate)	(None,	16,	16,	112)	0	concatenate_112[0][0] dropout_120[0][0]
batch_normalization_123 (BatchN	(None,	16,	16,	112)	448	concatenate_113[0][0]
activation_123 (Activation)	(None,	16,	16,	112)	0	batch_normalization_123[0][0]
separable_conv2d_70 (SeparableC	(None,	16,	16,	16)	4592	activation_123[0][0]
dropout_121 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_70[0][0]
concatenate_114 (Concatenate)	(None,	16,	16,	128)	0	concatenate_113[0][0] dropout_121[0][0]
batch_normalization_124 (BatchN	(None,	16,	16,	128)	512	concatenate_114[0][0]
activation_124 (Activation)	(None,	16,	16,	128)	0	batch_normalization_124[0][0]
separable_conv2d_71 (SeparableC	(None,	16,	16,	16)	5248	activation_124[0][0]
dropout_122 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_71[0][0]
concatenate_115 (Concatenate)	(None,	16,	16,	144)	0	concatenate_114[0][0] dropout_122[0][0]
batch_normalization_125 (BatchN	(None,	16,	16,	144)	576	concatenate_115[0][0]
activation_125 (Activation)	(None,	16,	16,	144)	0	batch_normalization_125[0][0]
separable_conv2d_72 (SeparableC	(None,	16,	16,	16)	5904	activation_125[0][0]
dropout_123 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_72[0][0]
concatenate_116 (Concatenate)	(None,	16,	16,	160)	0	concatenate_115[0][0] dropout_123[0][0]
batch_normalization_126 (BatchN	(None,	16,	16,	160)	640	concatenate_116[0][0]

activation_126 (Activation)	(None, 16, 16	, 160)	0	batch_normalization_126[0][0]
separable_conv2d_73 (SeparableC	(None, 16, 16	, 16)	6560	activation_126[0][0]
dropout_124 (Dropout)	(None, 16, 16	, 16)	0	separable_conv2d_73[0][0]
concatenate_117 (Concatenate)	(None, 16, 16	, 176)	0	concatenate_116[0][0] dropout_124[0][0]
batch_normalization_127 (BatchN	(None, 16, 16	, 176)	704	concatenate_117[0][0]
activation_127 (Activation)	(None, 16, 16	, 176)	0	batch_normalization_127[0][0]
separable_conv2d_74 (SeparableC	(None, 16, 16	, 16)	7216	activation_127[0][0]
dropout_125 (Dropout)	(None, 16, 16	, 16)	0	separable_conv2d_74[0][0]
concatenate_118 (Concatenate)	(None, 16, 16	, 192)	0	concatenate_117[0][0] dropout_125[0][0]
batch_normalization_128 (BatchN	(None, 16, 16	, 192)	768	concatenate_118[0][0]
activation_128 (Activation)	(None, 16, 16	, 192)	0	batch_normalization_128[0][0]
separable_conv2d_75 (SeparableC	(None, 16, 16	, 16)	7872	activation_128[0][0]
dropout_126 (Dropout)	(None, 16, 16	, 16)	0	separable_conv2d_75[0][0]
concatenate_119 (Concatenate)	(None, 16, 16	, 208)	0	concatenate_118[0][0] dropout_126[0][0]
batch_normalization_129 (BatchN	(None, 16, 16	, 208)	832	concatenate_119[0][0]
activation_129 (Activation)	(None, 16, 16	, 208)	0	batch_normalization_129[0][0]
separable_conv2d_76 (SeparableC	(None, 16, 16	, 16)	3536	activation_129[0][0]
dropout_127 (Dropout)	(None, 16, 16	, 16)	0	separable_conv2d_76[0][0]
average_pooling2d_9 (AveragePoo	(None, 8, 8,	16)	0	dropout_127[0][0]
batch_normalization_130 (BatchN	(None, 8, 8,	16)	64	average_pooling2d_9[0][0]
activation_130 (Activation)	(None, 8, 8,	16)	0	batch_normalization_130[0][0]
separable_conv2d_77 (SeparableC	(None, 8, 8,	16)	656	activation_130[0][0]
dropout_128 (Dropout)	(None, 8, 8,	16)	0	separable_conv2d_77[0][0]
concatenate_120 (Concatenate)	(None, 8, 8,	32)	0	average_pooling2d_9[0][0] dropout_128[0][0]
batch_normalization_131 (BatchN	(None, 8, 8,	32)	128	concatenate_120[0][0]
activation_131 (Activation)	(None, 8, 8,	32)	0	batch_normalization_131[0][0]
separable_conv2d_78 (SeparableC	(None, 8, 8,	16)	1312	activation_131[0][0]
dropout_129 (Dropout)	(None, 8, 8,	16)	0	separable_conv2d_78[0][0]
concatenate_121 (Concatenate)	(None, 8, 8,	48)	0	concatenate_120[0][0] dropout_129[0][0]
batch_normalization_132 (BatchN	(None, 8, 8,	48)	192	concatenate_121[0][0]
activation_132 (Activation)	(None, 8, 8,	48)	0	batch_normalization_132[0][0]
separable_conv2d_79 (SeparableC	(None, 8, 8,	16)	1968	activation_132[0][0]
dropout_130 (Dropout)	(None, 8, 8,	16)	0	separable_conv2d_79[0][0]
concatenate_122 (Concatenate)	(None, 8, 8,	64)	0	concatenate_121[0][0] dropout_130[0][0]
batch_normalization_133 (BatchN	(None, 8, 8,	64)	256	concatenate_122[0][0]
activation_133 (Activation)	(None, 8, 8,	64)	0	batch_normalization_133[0][0]

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separable_conv2d_80 (SeparableC	(None,	8,	8,	16)	2624	activation_133[0][0]
dropout_131 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_80[0][0]
concatenate_123 (Concatenate)	(None,	8,	8,	80)	0	concatenate_122[0][0] dropout_131[0][0]
batch_normalization_134 (BatchN	(None,	8,	8,	80)	320	concatenate_123[0][0]
activation_134 (Activation)	(None,	8,	8,	80)	0	batch_normalization_134[0][0]
separable_conv2d_81 (SeparableC	(None,	8,	8,	16)	3280	activation_134[0][0]
dropout_132 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_81[0][0]
concatenate_124 (Concatenate)	(None,	8,	8,	96)	0	concatenate_123[0][0] dropout_132[0][0]
batch_normalization_135 (BatchN	(None,	8,	8,	96)	384	concatenate_124[0][0]
activation_135 (Activation)	(None,	8,	8,	96)	0	batch_normalization_135[0][0]
separable_conv2d_82 (SeparableC	(None,	8,	8,	16)	3936	activation_135[0][0]
dropout_133 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_82[0][0]
concatenate_125 (Concatenate)	(None,	8,	8,	112)	0	concatenate_124[0][0] dropout_133[0][0]
batch_normalization_136 (BatchN	(None,	8,	8,	112)	448	concatenate_125[0][0]
activation_136 (Activation)	(None,	8,	8,	112)	0	batch_normalization_136[0][0]
separable_conv2d_83 (SeparableC	(None,	8,	8,	16)	4592	activation_136[0][0]
dropout_134 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_83[0][0]
concatenate_126 (Concatenate)	(None,	8,	8,	128)	0	concatenate_125[0][0] dropout_134[0][0]
batch_normalization_137 (BatchN	(None,	8,	8,	128)	512	concatenate_126[0][0]
activation_137 (Activation)	(None,	8,	8,	128)	0	batch_normalization_137[0][0]
separable_conv2d_84 (SeparableC	(None,	8,	8,	16)	5248	activation_137[0][0]
dropout_135 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_84[0][0]
concatenate_127 (Concatenate)	(None,	8,	8,	144)	0	concatenate_126[0][0] dropout_135[0][0]
batch_normalization_138 (BatchN	(None,	8,	8,	144)	576	concatenate_127[0][0]
activation_138 (Activation)	(None,	8,	8,	144)	0	batch_normalization_138[0][0]
separable_conv2d_85 (SeparableC	(None,	8,	8,	16)	5904	activation_138[0][0]
dropout_136 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_85[0][0]
concatenate_128 (Concatenate)	(None,	8,	8,	160)	0	concatenate_127[0][0] dropout_136[0][0]
batch_normalization_139 (BatchN	(None,	8,	8,	160)	640	concatenate_128[0][0]
activation_139 (Activation)	(None,	8,	8,	160)	0	batch_normalization_139[0][0]
separable_conv2d_86 (SeparableC	(None,	8,	8,	16)	6560	activation_139[0][0]
dropout_137 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_86[0][0]
concatenate_129 (Concatenate)	(None,	8,	8,	176)	0	concatenate_128[0][0] dropout_137[0][0]
batch_normalization_140 (BatchN	(None,	8,	8,	176)	704	concatenate_129[0][0]
activation 140 (Activation)	(None,	8,	8,	176)	0	batch normalization 140[0][0]

separable conv2d 87 (SeparableC (None, 8, 8, 16) 7216 activation 140[0][0] dropout_138 (Dropout) (None, 8, 8, 16) 0 separable_conv2d_87[0][0] concatenate 130 (Concatenate) (None, 8, 8, 192) concatenate_129[0][0] dropout_138[0][0] batch normalization 141 (BatchN (None, 8, 8, 192) 768 concatenate 130[0][0] activation 141 (Activation) (None, 8, 8, 192) 0 batch_normalization_141[0][0] separable conv2d 88 (SeparableC (None, 8, 8, 16) 7872 activation 141[0][0] dropout 139 (Dropout) (None, 8, 8, 16) 0 separable_conv2d_88[0][0] concatenate_130[0][0] concatenate 131 (Concatenate) (None, 8, 8, 208) dropout 139[0][0] batch normalization 142 (BatchN (None, 8, 8, 208) 832 concatenate 131[0][0] activation 142 (Activation) (None, 8, 8, 208) 0 batch normalization 142[0][0] separable conv2d 89 (SeparableC (None, 8, 8, 16) 3536 activation 142[0][0] dropout_140 (Dropout) 0 separable_conv2d_89[0][0] (None, 8, 8, 16) average pooling2d 10 (AveragePo (None, 4, 4, 16) 0 dropout 140[0][0] batch normalization 143 (BatchN (None, 4, 4, 16) average pooling2d 10[0][0] 64 activation 143 (Activation) (None, 4, 4, 16) 0 batch normalization 143[0][0] separable_conv2d_90 (SeparableC (None, 4, 4, 16) activation_143[0][0] 656 dropout 141 (Dropout) separable conv2d 90[0][0] (None, 4, 4, 16) 0 concatenate 132 (Concatenate) (None, 4, 4, 32) average_pooling2d 10[0][0] dropout 141[0][0] batch normalization 144 (BatchN (None, 4, 4, 32) 128 concatenate 132[0][0] activation 144 (Activation) (None, 4, 4, 32) batch normalization 144[0][0] separable_conv2d_91 (SeparableC (None, 4, 4, 16) 1312 activation_144[0][0] dropout 142 (Dropout) (None, 4, 4, 16) separable_conv2d_91[0][0] 0 concatenate 133 (Concatenate) (None, 4, 4, 48) concatenate 132[0][0] dropout 142[0][0] batch normalization 145 (BatchN (None, 4, 4, 48) 192 concatenate 133[0][0] activation 145 (Activation) 0 batch normalization 145[0][0] (None, 4, 4, 48) separable_conv2d_92 (SeparableC (None, 4, 4, 16) 1968 activation_145[0][0] dropout 143 (Dropout) (None, 4, 4, 16) 0 separable_conv2d_92[0][0] concatenate 134 (Concatenate) (None, 4, 4, 64) concatenate 133[0][0] dropout 143[0][0] batch normalization 146 (BatchN (None, 4, 4, 64) 256 concatenate 134[0][0] activation 146 (Activation) (None, 4, 4, 64) batch normalization 146[0][0] separable conv2d 93 (SeparableC (None, 4, 4, 16) 2624 activation 146[0][0] dropout 144 (Dropout) (None, 4, 4, 16) 0 separable conv2d 93[0][0] concatenate 135 (Concatenate) (None, 4, 4, 80) 0 concatenate 134[0][0] dropout_144[0][0] batch normalization 147 (BatchN (None, 4, 4, 80) 320 concatenate 135[0][0] activation 147 (Activation) (None, 4, 4, 80) 0 batch normalization 147[0][0]

separable_conv2d_94 (SeparableC	(None, 4	1,	4,	16)	3280	activation_147[0][0]
dropout_145 (Dropout)	(None,	1,	4,	16)	0	separable_conv2d_94[0][0]
concatenate_136 (Concatenate)	(None,	1,	4,	96)	0	concatenate_135[0][0] dropout_145[0][0]
batch_normalization_148 (BatchN	(None,	1,	4,	96)	384	concatenate_136[0][0]
activation_148 (Activation)	(None, 4	1,	4,	96)	0	batch_normalization_148[0][0]
separable_conv2d_95 (SeparableC	(None, 4	1,	4,	16)	3936	activation_148[0][0]
dropout_146 (Dropout)	(None,	1,	4,	16)	0	separable_conv2d_95[0][0]
concatenate_137 (Concatenate)	(None, 4	1,	4,	112)	0	concatenate_136[0][0] dropout_146[0][0]
batch_normalization_149 (BatchN	(None,	1,	4,	112)	448	concatenate_137[0][0]
activation_149 (Activation)	(None,	1,	4,	112)	0	batch_normalization_149[0][0]
separable_conv2d_96 (SeparableC	(None,	1,	4,	16)	4592	activation_149[0][0]
dropout_147 (Dropout)	(None,	1,	4,	16)	0	separable_conv2d_96[0][0]
concatenate_138 (Concatenate)	(None,	1,	4,	128)	0	concatenate_137[0][0] dropout_147[0][0]
batch_normalization_150 (BatchN	(None, 4	1,	4,	128)	512	concatenate_138[0][0]
activation_150 (Activation)	(None, 4	1,	4,	128)	0	batch_normalization_150[0][0]
separable_conv2d_97 (SeparableC	(None, 4	1,	4,	16)	5248	activation_150[0][0]
dropout_148 (Dropout)	(None, 4	1,	4,	16)	0	separable_conv2d_97[0][0]
concatenate_139 (Concatenate)	(None, 4	1,	4,	144)	0	concatenate_138[0][0] dropout_148[0][0]
batch_normalization_151 (BatchN	(None, 4	1,	4,	144)	576	concatenate_139[0][0]
activation_151 (Activation)	(None, 4	1,	4,	144)	0	batch_normalization_151[0][0]
separable_conv2d_98 (SeparableC	(None,	1,	4,	16)	5904	activation_151[0][0]
dropout_149 (Dropout)	(None,	1,	4,	16)	0	separable_conv2d_98[0][0]
concatenate_140 (Concatenate)	(None,	1,	4,	160)	0	concatenate_139[0][0] dropout_149[0][0]
batch_normalization_152 (BatchN	(None, 4	1,	4,	160)	640	concatenate_140[0][0]
activation_152 (Activation)	(None,	1,	4,	160)	0	batch_normalization_152[0][0]
separable_conv2d_99 (SeparableC	(None, 4	1,	4,	16)	6560	activation_152[0][0]
dropout_150 (Dropout)	(None,	1,	4,	16)	0	separable_conv2d_99[0][0]
concatenate_141 (Concatenate)	(None,	1,	4,	176)	0	concatenate_140[0][0] dropout_150[0][0]
batch_normalization_153 (BatchN	(None,	1,	4,	176)	704	concatenate_141[0][0]
activation_153 (Activation)	(None, 4	1,	4,	176)	0	batch_normalization_153[0][0]
separable_conv2d_100 (Separable	(None,	1,	4,	16)	7216	activation_153[0][0]
dropout_151 (Dropout)	(None,	1,	4,	16)	0	separable_conv2d_100[0][0]
concatenate_142 (Concatenate)	(None, 4	1,	4,	192)	0	concatenate_141[0][0] dropout_151[0][0]
batch_normalization_154 (BatchN	(None, 4	1,	4,	192)	768	concatenate_142[0][0]
activation_154 (Activation)	(None,	1,	4,	192)	0	batch_normalization_154[0][0]

separable_conv2d_101 (Separable	(None,	4, 4	, 16)	7872	activation_154[0][0]
dropout_152 (Dropout)	(None,	4, 4	, 16)	0	separable_conv2d_101[0][0]
concatenate_143 (Concatenate)	(None,	4, 4	, 208)	0	concatenate_142[0][0] dropout_152[0][0]
batch_normalization_155 (BatchN	(None,	4, 4	, 208)	832	concatenate_143[0][0]
activation_155 (Activation)	(None,	4, 4	, 208)	0	batch_normalization_155[0][0]
average_pooling2d_11 (AveragePo	(None,	2, 2	, 208)	0	activation_155[0][0]
flatten_2 (Flatten)	(None,	832)		0	average_pooling2d_11[0][0]
dense_2 (Dense)	(None,	10)		8330	flatten_2[0][0]

Total params: 258,282 Trainable params: 246,218 Non-trainable params: 12,064

#### In [54]:

```
model 3.compile(optimizer='adam', loss='categorical crossentropy', metrics=['accuracy'])
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
patience = 50
base_path = '/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/'
checkpoint file name = base path + 'CIFAR model3' + ' {epoch:02d}-{val accuracy:.2f}.hdf5'
model checkpoint = ModelCheckpoint (checkpoint file name, monitor='val accuracy', verbose=1, save be
early stop = EarlyStopping('val accuracy', mode='max', patience = patience)
reduce LR = ReduceLROnPlateau (monitor='val accuracy', mode='max', factor=0.1, patience=int(patience
/3), verbose=1)
callbacks = [model checkpoint, early stop, reduce LR]
epochs = 300
batch size = 256
#https://keras.io/api/preprocessing/image/#flow-method
history 3 = model_3.fit(data_generator.flow(X_train, y_train, batch_size),
                    steps_per_epoch = int(len(X_train)/batch size),
                    epochs = epochs,
                    callbacks = callbacks,
                    validation_data = (X_test, y_test), verbose=1)
```

WARNING:tensorflow:sample_weight modes were coerced from to ['...'] Train for 195 steps, validate on 10000 samples Epoch 1/300 Epoch 00001: val accuracy improved from -inf to 0.10000, saving model to /home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 01-0.10.hdf5 loss: 2.3508 - val_accuracy: 0.1000 Epoch 2/300 Epoch 00002: val accuracy improved from 0.10000 to 0.13040, saving model to /home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 02-0.13.hdf5 loss: 2.6361 - val_accuracy: 0.1304 Epoch 3/300 Epoch 00003: val_accuracy improved from 0.13040 to 0.54500, saving model to /home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model3_03-0.55.hdf5 loss: 1.3663 - val_accuracy: 0.5450 Epoch 4/300 

```
Epoch 00004: val accuracy did not improve from 0.54500
loss: 1.7844 - val accuracy: 0.5219
Epoch 5/300
Epoch 00005: val accuracy improved from 0.54500 to 0.59910, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 05-0.60.hdf5
loss: 1.2854 - val accuracy: 0.5991
Epoch 6/300
Epoch 00006: val accuracy did not improve from 0.59910
loss: 1.6606 - val_accuracy: 0.5682
Epoch 7/300
Epoch 00007: val accuracy improved from 0.59910 to 0.64520, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 07-0.65.hdf5
loss: 1.2518 - val accuracy: 0.6452
Epoch 8/300
Epoch 00008: val_accuracy improved from 0.64520 to 0.67210, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model3_08-0.67.hdf5
loss: 1.1228 - val_accuracy: 0.6721
Epoch 9/300
Epoch 00009: val accuracy did not improve from 0.67210
loss: 1.2116 - val accuracy: 0.6537
Epoch 10/300
Epoch 00010: val accuracy did not improve from 0.67210
loss: 1.6817 - val_accuracy: 0.5975
Epoch 11/300
Epoch 00011: val accuracy improved from 0.67210 to 0.71550, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 11-0.72.hdf5
loss: 0.9364 - val accuracy: 0.7155
Epoch 12/300
Epoch 00012: val accuracy did not improve from 0.71550
loss: 1.4434 - val accuracy: 0.6396
Epoch 13/300
Epoch 00013: val accuracy did not improve from 0.71550
loss: 1.2199 - val accuracy: 0.6821
Epoch 14/300
Epoch 00014: val accuracy did not improve from 0.71550
loss: 1.4715 - val_accuracy: 0.6352
Epoch 15/300
Epoch 00015: val_accuracy did not improve from 0.71550
loss: 1.2906 - val_accuracy: 0.6463
Epoch 16/300
Epoch 00016: val accuracy improved from 0.71550 to 0.72990, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 16-0.73.hdf5
loss: 0.9909 - val_accuracy: 0.7299
Epoch 17/300
Epoch 00017: val accuracy improved from 0.72990 to 0.75800, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 17-0.76.hdf5
loss: 0.8110 - val_accuracy: 0.7580
Epoch 18/300
Epoch 00018: val accuracy did not improve from 0.75800
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loss: 1.0143 - val accuracy: 0.7238
Epoch 19/300
Epoch 00019: val accuracy improved from 0.75800 to 0.78400, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model3_19-0.78.hdf5
loss: 0.7006 - val accuracy: 0.7840
Epoch 20/300
Epoch 00020: val_accuracy did not improve from 0.78400
loss: 0.8453 - val_accuracy: 0.7636
Epoch 21/300
Epoch 00021: val accuracy did not improve from 0.78400
loss: 1.2476 - val accuracy: 0.6963
Epoch 22/300
Epoch 00022: val accuracy did not improve from 0.78400
loss: 0.7055 - val accuracy: 0.7833
Epoch 23/300
Epoch 00023: val accuracy did not improve from 0.78400
loss: 0.9178 - val accuracy: 0.7420
Epoch 24/300
Epoch 00024: val accuracy did not improve from 0.78400
loss: 0.8734 - val_accuracy: 0.7555
Epoch 25/300
Epoch 00025: val accuracy did not improve from 0.78400
loss: 1.1021 - val accuracy: 0.7249
Epoch 26/300
Epoch 00026: val accuracy did not improve from 0.78400
loss: 0.7443 - val_accuracy: 0.7774
Epoch 27/300
Epoch 00027: val accuracy did not improve from 0.78400
loss: 0.8986 - val accuracy: 0.7506
Epoch 28/300
Epoch 00028: val accuracy did not improve from 0.78400
loss: 0.7706 - val_accuracy: 0.7776
Epoch 29/300
Epoch 00029: val accuracy improved from 0.78400 to 0.80120, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model3_29-0.80.hdf5
loss: 0.6982 - val accuracy: 0.8012
Epoch 30/300
Epoch 00030: val accuracy did not improve from 0.80120
loss: 1.0080 - val_accuracy: 0.7376
Epoch 31/300
Epoch 00031: val accuracy improved from 0.80120 to 0.81090, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 31-0.81.hdf5
loss: 0.6098 - val accuracy: 0.8109
Epoch 32/300
Epoch 00032: val_accuracy did not improve from 0.81090
loss: 0.7732 - val_accuracy: 0.7825
Epoch 33/300
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Epoch 00033: val_accuracy did not improve from 0.81090
loss: 0.8289 - val_accuracy: 0.7722
Epoch 34/300
Epoch 00034: val accuracy did not improve from 0.81090
loss: 0.7130 - val accuracy: 0.7963
Epoch 35/300
Epoch 00035: val accuracy did not improve from 0.81090
loss: 0.7658 - val_accuracy: 0.7869
Epoch 36/300
Epoch 00036: val accuracy did not improve from 0.81090
loss: 0.9025 - val_accuracy: 0.7674
Epoch 37/300
Epoch 00037: val accuracy did not improve from 0.81090
loss: 0.6681 - val accuracy: 0.8063
Epoch 38/300
Epoch 00038: val accuracy did not improve from 0.81090
loss: 0.8125 - val_accuracy: 0.7778
Epoch 39/300
Epoch 00039: val accuracy did not improve from 0.81090
loss: 0.9117 - val accuracy: 0.7663
Epoch 40/300
Epoch 00040: val accuracy improved from 0.81090 to 0.84530, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 40-0.85.hdf5
loss: 0.5169 - val accuracy: 0.8453
Epoch 41/300
Epoch 00041: val accuracy did not improve from 0.84530
loss: 0.7346 - val_accuracy: 0.7989
Epoch 42/300
Epoch 00042: val_accuracy did not improve from 0.84530
loss: 0.7958 - val accuracy: 0.7897
Epoch 43/300
Epoch 00043: val accuracy did not improve from 0.84530
loss: 0.7988 - val accuracy: 0.7866
Epoch 44/300
Epoch 00044: val accuracy did not improve from 0.84530
loss: 0.8270 - val_accuracy: 0.7836
Epoch 45/300
Epoch 00045: val accuracy did not improve from 0.84530
loss: 1.0867 - val_accuracy: 0.7382
Epoch 46/300
Epoch 00046: val accuracy did not improve from 0.84530
loss: 0.6583 - val_accuracy: 0.8115
Epoch 47/300
Epoch 00047: val_accuracy did not improve from 0.84530
loss: 0.6326 - val accuracy: 0.8270
Epoch 48/300
Epoch 00048: val accuracy did not improve from 0.84530
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loss: 0.7164 - val accuracy: 0.8055
Epoch 49/300
Epoch 00049: val accuracy did not improve from 0.84530
loss: 0.5649 - val accuracy: 0.8385
Epoch 50/300
Epoch 00050: val accuracy did not improve from 0.84530
loss: 0.6285 - val_accuracy: 0.8294
Epoch 51/300
Epoch 00051: val_accuracy did not improve from 0.84530
loss: 0.8583 - val accuracy: 0.7843
Epoch 52/300
Epoch 00052: val accuracy did not improve from 0.84530
loss: 0.7306 - val accuracy: 0.8084
Epoch 53/300
Epoch 00053: val accuracy did not improve from 0.84530
loss: 0.6683 - val_accuracy: 0.8223
Epoch 54/300
Epoch 00054: val accuracy did not improve from 0.84530
loss: 0.6907 - val_accuracy: 0.8068
Epoch 55/300
Epoch 00055: val accuracy did not improve from 0.84530
loss: 0.8694 - val_accuracy: 0.7840
Epoch 56/300
Epoch 00056: val accuracy did not improve from 0.84530
loss: 0.8278 - val_accuracy: 0.7906
Epoch 57/300
Epoch 00057: val_accuracy improved from 0.84530 to 0.85590, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 57-0.86.hdf5
loss: 0.5141 - val_accuracy: 0.8559
Epoch 58/300
Epoch 00058: val accuracy did not improve from 0.85590
loss: 0.5282 - val_accuracy: 0.8504
Epoch 59/300
Epoch 00059: val accuracy improved from 0.85590 to 0.85790, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 59-0.86.hdf5
loss: 0.5006 - val_accuracy: 0.8579
Epoch 60/300
Epoch 00060: val accuracy did not improve from 0.85790
loss: 0.5421 - val_accuracy: 0.8481
Epoch 61/300
Epoch 00061: val accuracy did not improve from 0.85790
loss: 0.5408 - val accuracy: 0.8508
Epoch 62/300
Epoch 00062: val_accuracy did not improve from 0.85790
loss: 0.5336 - val accuracy: 0.8522
Epoch 63/300
Epoch 00063: val accuracy improved from 0.85790 to 0.86340, saving model to
```

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/home/ubuntu/Project/my_data/CNN CIFAR/checkpoint/CIFAR model3 63-0.86.hdf5
loss: 0.4907 - val accuracy: 0.8634
Epoch 64/300
Epoch 00064: val accuracy did not improve from 0.86340
loss: 0.5027 - val accuracy: 0.8614
Epoch 65/300
Epoch 00065: val accuracy did not improve from 0.86340
loss: 0.5090 - val_accuracy: 0.8563
Epoch 66/300
Epoch 00066: val_accuracy did not improve from 0.86340
loss: 0.5374 - val_accuracy: 0.8527
Epoch 67/300
Epoch 00067: val accuracy did not improve from 0.86340
loss: 0.5612 - val accuracy: 0.8500
Epoch 68/300
Epoch 00068: val accuracy did not improve from 0.86340
loss: 0.5471 - val_accuracy: 0.8542
Epoch 69/300
Epoch 00069: val accuracy did not improve from 0.86340
loss: 0.4954 - val accuracy: 0.8631
Epoch 70/300
Epoch 00070: val accuracy improved from 0.86340 to 0.86810, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model3 70-0.87.hdf5
loss: 0.4829 - val accuracy: 0.8681
Epoch 71/300
Epoch 00071: val accuracy did not improve from 0.86810
loss: 0.4899 - val_accuracy: 0.8635
Epoch 72/300
Epoch 00072: val accuracy did not improve from 0.86810
loss: 0.5181 - val_accuracy: 0.8581
Epoch 73/300
Epoch 00073: val accuracy did not improve from 0.86810
loss: 0.5228 - val_accuracy: 0.8574
Epoch 74/300
Epoch 00074: val_accuracy did not improve from 0.86810
loss: 0.5139 - val_accuracy: 0.8611
Epoch 75/300
Epoch 00075: val_accuracy did not improve from 0.86810
loss: 0.5706 - val_accuracy: 0.8453
Epoch 76/300
Epoch 00076: val accuracy did not improve from 0.86810
loss: 0.5191 - val accuracy: 0.8594
Epoch 77/300
Epoch 00077: val_accuracy did not improve from 0.86810
loss: 0.4882 - val accuracy: 0.8658
Epoch 78/300
Epoch 00078: val accuracy did not improve from 0.86810
```

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loss: 0.5592 - val_accuracy: 0.8509
Epoch 79/300
Epoch 00079: val accuracy did not improve from 0.86810
loss: 0.5144 - val accuracy: 0.8605
Epoch 80/300
Epoch 00080: val accuracy did not improve from 0.86810
loss: 0.4895 - val accuracy: 0.8679
Epoch 81/300
Epoch 00081: val_accuracy did not improve from 0.86810
loss: 0.4984 - val accuracy: 0.8678
Epoch 82/300
Epoch 00082: val accuracy did not improve from 0.86810
loss: 0.5619 - val_accuracy: 0.8555
Epoch 83/300
Epoch 00083: val accuracy did not improve from 0.86810
loss: 0.5114 - val_accuracy: 0.8643
Epoch 84/300
Epoch 00084: val_accuracy did not improve from 0.86810
loss: 0.5042 - val accuracy: 0.8651
Epoch 85/300
Epoch 00085: val accuracy did not improve from 0.86810
loss: 0.5573 - val accuracy: 0.8531
Epoch 86/300
Epoch 00086: val accuracy did not improve from 0.86810
loss: 0.5333 - val accuracy: 0.8605
Epoch 87/300
Epoch 00087: val accuracy did not improve from 0.86810
loss: 0.5121 - val accuracy: 0.8645
Epoch 88/300
Epoch 00088: val accuracy did not improve from 0.86810
loss: 0.5187 - val accuracy: 0.8627
Epoch 89/300
Epoch 00089: val accuracy did not improve from 0.86810
loss: 0.5168 - val_accuracy: 0.8628
Epoch 90/300
Epoch 00090: val_accuracy did not improve from 0.86810
loss: 0.5122 - val accuracy: 0.8639
Epoch 91/300
Epoch 00091: val accuracy did not improve from 0.86810
loss: 0.5102 - val accuracy: 0.8649
Epoch 92/300
Epoch 00092: val accuracy did not improve from 0.86810
loss: 0.5104 - val accuracy: 0.8641
Epoch 93/300
Epoch 00093: val accuracy did not improve from 0.86810
```

loss: 0.5185 - val accuracy: 0.8630

```
Epoch 94/300
Epoch 00094: val_accuracy did not improve from 0.86810
loss: 0.5174 - val_accuracy: 0.8647
Epoch 95/300
Epoch 00095: val accuracy did not improve from 0.86810
loss: 0.5254 - val accuracy: 0.8620
Epoch 96/300
Epoch 00096: val accuracy did not improve from 0.86810
loss: 0.5124 - val accuracy: 0.8654
Epoch 97/300
Epoch 00097: val accuracy did not improve from 0.86810
loss: 0.5062 - val accuracy: 0.8655
Epoch 98/300
Epoch 00098: val accuracy did not improve from 0.86810
loss: 0.5155 - val_accuracy: 0.8645
Epoch 99/300
Epoch 00099: val_accuracy did not improve from 0.86810
loss: 0.5068 - val_accuracy: 0.8655
Epoch 100/300
Epoch 00100: val_accuracy did not improve from 0.86810
loss: 0.5099 - val_accuracy: 0.8650
Epoch 101/300
Epoch 00101: val accuracy did not improve from 0.86810
loss: 0.5109 - val accuracy: 0.8643
Epoch 102/300
Epoch 00102: val_accuracy did not improve from 0.86810
loss: 0.5141 - val_accuracy: 0.8636
Epoch 103/300
Epoch 00103: val accuracy did not improve from 0.86810
loss: 0.5122 - val_accuracy: 0.8637
Epoch 104/300
Epoch 00104: val accuracy did not improve from 0.86810
loss: 0.5127 - val accuracy: 0.8637
Epoch 105/300
Epoch 00105: val_accuracy did not improve from 0.86810
loss: 0.5129 - val_accuracy: 0.8636
Epoch 106/300
Epoch 00106: val_accuracy did not improve from 0.86810
loss: 0.5113 - val accuracy: 0.8641
Epoch 107/300
Epoch 00107: val accuracy did not improve from 0.86810
loss: 0.5136 - val_accuracy: 0.8635
Epoch 108/300
Epoch 00108: val_accuracy did not improve from 0.86810
loss: 0.5126 - val_accuracy: 0.8637
Epoch 109/300
```

```
Epoch 00109: val_accuracy did not improve from 0.86810
loss: 0.5131 - val accuracy: 0.8638
Epoch 110/300
Epoch 00110: val accuracy did not improve from 0.86810
loss: 0.5118 - val accuracy: 0.8638
Epoch 111/300
Epoch 00111: val accuracy did not improve from 0.86810
loss: 0.5127 - val accuracy: 0.8638
Epoch 112/300
Epoch 00112: val accuracy did not improve from 0.86810
loss: 0.5135 - val_accuracy: 0.8633
Epoch 113/300
Epoch 00113: val accuracy did not improve from 0.86810
loss: 0.5138 - val_accuracy: 0.8636
Epoch 114/300
Epoch 00114: val accuracy did not improve from 0.86810
loss: 0.5128 - val_accuracy: 0.8638
Epoch 115/300
Epoch 00115: val_accuracy did not improve from 0.86810
loss: 0.5123 - val_accuracy: 0.8640
Epoch 116/300
Epoch 00116: val accuracy did not improve from 0.86810
loss: 0.5144 - val accuracy: 0.8631
Epoch 117/300
Epoch 00117: val accuracy did not improve from 0.86810
loss: 0.5135 - val accuracy: 0.8640
Epoch 118/300
Epoch 00118: val accuracy did not improve from 0.86810
loss: 0.5135 - val_accuracy: 0.8638
Epoch 119/300
Epoch 00119: val accuracy did not improve from 0.86810
loss: 0.5127 - val_accuracy: 0.8642
Epoch 120/300
Epoch 00120: val accuracy did not improve from 0.86810
loss: 0.5135 - val accuracy: 0.8637
In [55]:
#history plot for accyracy
plt.figure(figsize=(16,4))
plt.plot(history_3.history['accuracy'])
plt.plot(history_3.history['val_accuracy'])
plt.title("Model Accuracy")
plt.xlabel("Epoch")
plt.ylabel("Accuracy")
plt.legend(["train", "test"], loc="upper left")
# history plot for accuracy
```

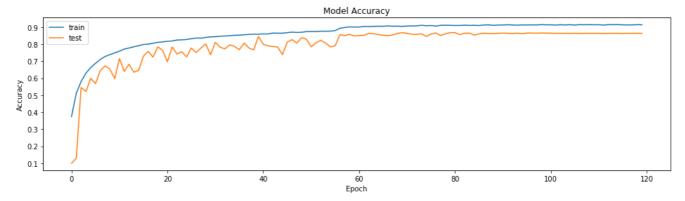
plt.figure(figsize=(16,4))

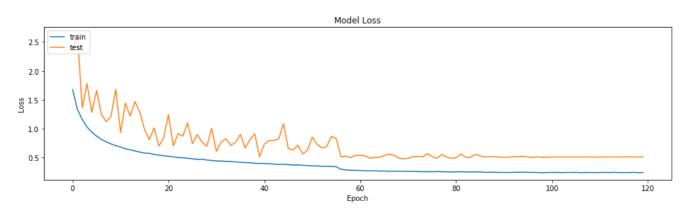
plt.plot(history_3.history["loss"])
plt.plot(history_3.history["val_loss"])
plt_title("Model_Loss")

----- VV

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```
plt.xlabel("Epoch")
plt.ylabel("Loss")
plt.legend(["train", "test"], loc="upper left")
plt.show()
```





# In [57]:

## Model -4

## In [58]:

```
#BN-->ReLU-->Conv2D-->Dropout-->concat(input, output)-->(put in loop)
def denseblock(input, num_filter, dropout_rate = 0.2):
                          # to keep the growth rate of number of filters
    global compression
    temp = input
for _ in range(l):
        BatchNorm = layers.BatchNormalization()(temp)
        relu = layers.Activation('relu')(BatchNorm)
        Conv2D 7 7= layers.SeparableConv2D(int(num filter*compression), (7,7), use bias=False, padd
ing='same') (relu)
        \#Conv2D 3 3 = layers.Conv2D(int(num filter*compression), (3,3), use bias=False
,padding='same') (relu)
        if dropout rate>0:
            Conv2D 7 7 = layers.Dropout(dropout rate)(Conv2D 7 7)
        #concat the input(temp) and output(conv2d 3 3) , in resnet we add but here we concat
        concat = layers.Concatenate(axis=-1)([temp,Conv2D_7_7])
        #change the concat as input
        temp = concat
```

```
return temp
\#BN-->relu-->conv2d(1x1)-->dropout-->avg pool
def transition(input, num_filter, 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)
    Conv2D BottleNeck = layers.SeparableConvolution2D(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
#BN-->relu-->avgpool-->flat-->softmax
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
# Hyperparameters
1 = 12
num filter = 32
compression = 0.5
dropout_rate = 0.2
num_classes = 10
input = layers.Input(shape=(input size))
First Conv2D = layers.Conv2D(num filter, (7,7), use bias=False ,padding='same')(input)
#First dense and transition block
First Block = denseblock(First Conv2D, num filter, dropout rate)
First Transition = transition(First Block, num filter, dropout rate)
#Second dense and transition block
Second Block = denseblock(First Transition, num filter, dropout rate)
Second Transition = transition(Second Block, num filter, dropout rate)
#Third dense and transition block
Third Block = denseblock(Second Transition, num filter, dropout rate)
Third_Transition = transition(Third_Block, num_filter, dropout_rate)
#last dense and output block
Last_Block = denseblock(Third_Transition, num_filter, dropout_rate)
output = output layer(Last Block)
```

## In [59]:

```
model_4 = tf.keras.models.Model(inputs=[input], outputs=[output])
model_4.summary()
```

Model: "model 3"

Layer (type)	Output Shape	Param #	Connected to
input_4 (InputLayer)	[(None, 32, 32, 3)]	0	
conv2d_54 (Conv2D)	(None, 32, 32, 32)	4704	input_4[0][0]
batch_normalization_156 (BatchN	(None, 32, 32, 32)	128	conv2d_54[0][0]
activation_156 (Activation)	(None, 32, 32, 32)	0	batch_normalization_156[0][0]
separable_conv2d_102 (Separable	(None, 32, 32, 16)	2080	activation_156[0][0]
dropout_153 (Dropout)	(None, 32, 32, 16)	0	separable_conv2d_102[0][0]
concatenate_144 (Concatenate)	(None, 32, 32, 48)	0	conv2d_54[0][0] dropout_153[0][0]

batch_normalization_157 (BatchN	(None,	32,	32,	48)	192	concatenate_144[0][0]
activation_157 (Activation)	(None,	32,	32,	48)	0	batch_normalization_157[0][0]
separable_conv2d_103 (Separable	(None,	32,	32,	16)	3120	activation_157[0][0]
dropout_154 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_103[0][0]
concatenate_145 (Concatenate)	(None,	32,	32,	64)	0	concatenate_144[0][0] dropout_154[0][0]
batch_normalization_158 (BatchN	(None,	32,	32,	64)	256	concatenate_145[0][0]
activation_158 (Activation)	(None,	32,	32,	64)	0	batch_normalization_158[0][0]
separable_conv2d_104 (Separable	(None,	32,	32,	16)	4160	activation_158[0][0]
dropout_155 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_104[0][0]
concatenate_146 (Concatenate)	(None,	32,	32,	80)	0	concatenate_145[0][0] dropout_155[0][0]
batch_normalization_159 (BatchN	(None,	32,	32,	80)	320	concatenate_146[0][0]
activation_159 (Activation)	(None,	32,	32,	80)	0	batch_normalization_159[0][0]
separable_conv2d_105 (Separable	(None,	32,	32,	16)	5200	activation_159[0][0]
dropout_156 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_105[0][0]
concatenate_147 (Concatenate)	(None,	32,	32,	96)	0	concatenate_146[0][0] dropout_156[0][0]
batch_normalization_160 (BatchN	(None,	32,	32,	96)	384	concatenate_147[0][0]
activation_160 (Activation)	(None,	32,	32,	96)	0	batch_normalization_160[0][0]
separable_conv2d_106 (Separable	(None,	32,	32,	16)	6240	activation_160[0][0]
dropout_157 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_106[0][0]
concatenate_148 (Concatenate)	(None,	32,	32,	112)	0	concatenate_147[0][0] dropout_157[0][0]
batch_normalization_161 (BatchN	(None,	32,	32,	112)	448	concatenate_148[0][0]
activation_161 (Activation)	(None,	32,	32,	112)	0	batch_normalization_161[0][0]
separable_conv2d_107 (Separable	(None,	32,	32,	16)	7280	activation_161[0][0]
dropout_158 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_107[0][0]
concatenate_149 (Concatenate)	(None,	32,	32,	128)	0	concatenate_148[0][0] dropout_158[0][0]
batch_normalization_162 (BatchN	(None,	32,	32,	128)	512	concatenate_149[0][0]
activation_162 (Activation)	(None,	32,	32,	128)	0	batch_normalization_162[0][0]
separable_conv2d_108 (Separable	(None,	32,	32,	16)	8320	activation_162[0][0]
dropout_159 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_108[0][0]
concatenate_150 (Concatenate)	(None,	32,	32,	144)	0	concatenate_149[0][0] dropout_159[0][0]
batch_normalization_163 (BatchN	(None,	32,	32,	144)	576	concatenate_150[0][0]
activation_163 (Activation)	(None,	32,	32,	144)	0	batch_normalization_163[0][0]
separable_conv2d_109 (Separable	(None,	32,	32,	16)	9360	activation_163[0][0]
dropout_160 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_109[0][0]
concatenate_151 (Concatenate)	(None,	32,	32,	160)	0	concatenate_150[0][0] dropout 160[0][0]

_

batch_normalization_164 (BatchN	(None,	32,	32,	160)	640	concatenate_151[0][0]
activation_164 (Activation)	(None,	32,	32,	160)	0	batch_normalization_164[0][0]
separable_conv2d_110 (Separable	(None,	32,	32,	16)	10400	activation_164[0][0]
dropout_161 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_110[0][0]
concatenate_152 (Concatenate)	(None,	32,	32,	176)	0	concatenate_151[0][0] dropout_161[0][0]
batch_normalization_165 (BatchN	(None,	32,	32,	176)	704	concatenate_152[0][0]
activation_165 (Activation)	(None,	32,	32,	176)	0	batch_normalization_165[0][0]
separable_conv2d_111 (Separable	(None,	32,	32,	16)	11440	activation_165[0][0]
dropout_162 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_111[0][0]
concatenate_153 (Concatenate)	(None,	32,	32,	192)	0	concatenate_152[0][0] dropout_162[0][0]
batch_normalization_166 (BatchN	(None,	32,	32,	192)	768	concatenate_153[0][0]
activation_166 (Activation)	(None,	32,	32,	192)	0	batch_normalization_166[0][0]
separable_conv2d_112 (Separable	(None,	32,	32,	16)	12480	activation_166[0][0]
dropout_163 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_112[0][0]
concatenate_154 (Concatenate)	(None,	32,	32,	208)	0	concatenate_153[0][0] dropout_163[0][0]
batch_normalization_167 (BatchN	(None,	32,	32,	208)	832	concatenate_154[0][0]
activation_167 (Activation)	(None,	32,	32,	208)	0	batch_normalization_167[0][0]
separable_conv2d_113 (Separable	(None,	32,	32,	16)	13520	activation_167[0][0]
dropout_164 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_113[0][0]
concatenate_155 (Concatenate)	(None,	32,	32,	224)	0	concatenate_154[0][0] dropout_164[0][0]
batch_normalization_168 (BatchN	(None,	32,	32,	224)	896	concatenate_155[0][0]
activation_168 (Activation)	(None,	32,	32,	224)	0	batch_normalization_168[0][0]
separable_conv2d_114 (Separable	(None,	32,	32,	16)	3808	activation_168[0][0]
dropout_165 (Dropout)	(None,	32,	32,	16)	0	separable_conv2d_114[0][0]
average_pooling2d_12 (AveragePo	(None,	16,	16,	16)	0	dropout_165[0][0]
batch_normalization_169 (BatchN	(None,	16,	16,	16)	64	average_pooling2d_12[0][0]
activation_169 (Activation)	(None,	16,	16,	16)	0	batch_normalization_169[0][0]
separable_conv2d_115 (Separable	(None,	16,	16,	16)	1040	activation_169[0][0]
dropout_166 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_115[0][0]
concatenate_156 (Concatenate)	(None,	16,	16,	32)	0	average_pooling2d_12[0][0] dropout_166[0][0]
batch_normalization_170 (BatchN	(None,	16,	16,	32)	128	concatenate_156[0][0]
activation_170 (Activation)	(None,	16,	16,	32)	0	batch_normalization_170[0][0]
separable_conv2d_116 (Separable	(None,	16,	16,	16)	2080	activation_170[0][0]
dropout_167 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_116[0][0]
concatenate_157 (Concatenate)	(None,	16,	16,	48)	0	concatenate_156[0][0] dropout_167[0][0]

batch_normalization_171 (BatchN	(None,	16,	16,	48)	192	concatenate_157[0][0]
activation_171 (Activation)	(None,	16,	16,	48)	0	batch_normalization_171[0][0]
separable_conv2d_117 (Separable	(None,	16,	16,	16)	3120	activation_171[0][0]
dropout_168 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_117[0][0]
concatenate_158 (Concatenate)	(None,	16,	16,	64)	0	concatenate_157[0][0] dropout_168[0][0]
oatch_normalization_172 (BatchN	(None,	16,	16,	64)	256	concatenate_158[0][0]
activation_172 (Activation)	(None,	16,	16,	64)	0	batch_normalization_172[0][0]
separable_conv2d_118 (Separable	(None,	16,	16,	16)	4160	activation_172[0][0]
dropout_169 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_118[0][0]
concatenate_159 (Concatenate)	(None,	16,	16,	80)	0	concatenate_158[0][0] dropout_169[0][0]
patch_normalization_173 (BatchN	(None,	16,	16,	80)	320	concatenate_159[0][0]
activation_173 (Activation)	(None,	16,	16,	80)	0	batch_normalization_173[0][0]
separable_conv2d_119 (Separable	(None,	16,	16,	16)	5200	activation_173[0][0]
dropout_170 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_119[0][0]
concatenate_160 (Concatenate)	(None,	16,	16,	96)	0	concatenate_159[0][0] dropout_170[0][0]
oatch_normalization_174 (BatchN	(None,	16,	16,	96)	384	concatenate_160[0][0]
activation_174 (Activation)	(None,	16,	16,	96)	0	batch_normalization_174[0][0]
separable_conv2d_120 (Separable	(None,	16,	16,	16)	6240	activation_174[0][0]
dropout_171 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_120[0][0]
concatenate_161 (Concatenate)	(None,	16,	16,	112)	0	concatenate_160[0][0] dropout_171[0][0]
oatch_normalization_175 (BatchN	(None,	16,	16,	112)	448	concatenate_161[0][0]
activation_175 (Activation)	(None,	16,	16,	112)	0	batch_normalization_175[0][0]
separable_conv2d_121 (Separable	(None,	16,	16,	16)	7280	activation_175[0][0]
dropout_172 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_121[0][0]
concatenate_162 (Concatenate)	(None,	16,	16,	128)	0	concatenate_161[0][0] dropout_172[0][0]
patch_normalization_176 (BatchN	(None,	16,	16,	128)	512	concatenate_162[0][0]
activation_176 (Activation)	(None,	16,	16,	128)	0	batch_normalization_176[0][0]
separable_conv2d_122 (Separable	(None,	16,	16,	16)	8320	activation_176[0][0]
dropout_173 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_122[0][0]
concatenate_163 (Concatenate)	(None,	16,	16,	144)	0	concatenate_162[0][0] dropout_173[0][0]
patch_normalization_177 (BatchN	(None,	16,	16,	144)	576	concatenate_163[0][0]
activation_177 (Activation)	(None,	16,	16,	144)	0	batch_normalization_177[0][0]
separable_conv2d_123 (Separable	(None,	16,	16,	16)	9360	activation_177[0][0]
dropout_174 (Dropout)	(None,	16,	16,	16)	0	separable_conv2d_123[0][0]
concatenate_164 (Concatenate)	(None,	16,	16,	160)	0	concatenate_163[0][0] dropout_174[0][0]

batch_normalization_178 (BatchN	(None, 16, 16, 160)	640	concatenate_164[0][0]
activation_178 (Activation)	(None, 16, 16, 160)	0	batch_normalization_178[0][0]
separable_conv2d_124 (Separable	(None, 16, 16, 16)	10400	activation_178[0][0]
dropout_175 (Dropout)	(None, 16, 16, 16)	0	separable_conv2d_124[0][0]
concatenate_165 (Concatenate)	(None, 16, 16, 176)	0	concatenate_164[0][0] dropout_175[0][0]
batch_normalization_179 (BatchN	(None, 16, 16, 176)	704	concatenate_165[0][0]
activation_179 (Activation)	(None, 16, 16, 176)	0	batch_normalization_179[0][0]
separable_conv2d_125 (Separable	(None, 16, 16, 16)	11440	activation_179[0][0]
dropout_176 (Dropout)	(None, 16, 16, 16)	0	separable_conv2d_125[0][0]
concatenate_166 (Concatenate)	(None, 16, 16, 192)	0	concatenate_165[0][0] dropout_176[0][0]
batch_normalization_180 (BatchN	(None, 16, 16, 192)	768	concatenate_166[0][0]
activation_180 (Activation)	(None, 16, 16, 192)	0	batch_normalization_180[0][0]
separable_conv2d_126 (Separable	(None, 16, 16, 16)	12480	activation_180[0][0]
dropout_177 (Dropout)	(None, 16, 16, 16)	0	separable_conv2d_126[0][0]
concatenate_167 (Concatenate)	(None, 16, 16, 208)	0	concatenate_166[0][0] dropout_177[0][0]
batch_normalization_181 (BatchN	(None, 16, 16, 208)	832	concatenate_167[0][0]
activation_181 (Activation)	(None, 16, 16, 208)	0	batch_normalization_181[0][0]
separable_conv2d_127 (Separable	(None, 16, 16, 16)	3536	activation_181[0][0]
dropout_178 (Dropout)	(None, 16, 16, 16)	0	separable_conv2d_127[0][0]
average_pooling2d_13 (AveragePo	(None, 8, 8, 16)	0	dropout_178[0][0]
batch_normalization_182 (BatchN	(None, 8, 8, 16)	64	average_pooling2d_13[0][0]
activation_182 (Activation)	(None, 8, 8, 16)	0	batch_normalization_182[0][0]
separable_conv2d_128 (Separable	(None, 8, 8, 16)	1040	activation_182[0][0]
dropout_179 (Dropout)	(None, 8, 8, 16)	0	separable_conv2d_128[0][0]
concatenate_168 (Concatenate)	(None, 8, 8, 32)	0	average_pooling2d_13[0][0] dropout_179[0][0]
batch_normalization_183 (BatchN	(None, 8, 8, 32)	128	concatenate_168[0][0]
activation_183 (Activation)	(None, 8, 8, 32)	0	batch_normalization_183[0][0]
activation_183 (Activation) separable_conv2d_129 (Separable		2080	batch_normalization_183[0][0] activation_183[0][0]
separable_conv2d_129 (Separable			
separable_conv2d_129 (Separable dropout_180 (Dropout)	(None, 8, 8, 16)	2080	activation_183[0][0]
separable_conv2d_129 (Separable dropout_180 (Dropout) concatenate_169 (Concatenate)	(None, 8, 8, 16) (None, 8, 8, 16) (None, 8, 8, 48)	2080	activation_183[0][0] separable_conv2d_129[0][0] concatenate_168[0][0]
separable_conv2d_129 (Separable dropout_180 (Dropout) concatenate_169 (Concatenate) batch_normalization_184 (BatchN	(None, 8, 8, 16) (None, 8, 8, 16) (None, 8, 8, 48)	2080	activation_183[0][0]  separable_conv2d_129[0][0]  concatenate_168[0][0] dropout_180[0][0]
_	(None, 8, 8, 16) (None, 8, 8, 16) (None, 8, 8, 48) (None, 8, 8, 48) (None, 8, 8, 48)	2080 0 0 192	activation_183[0][0]  separable_conv2d_129[0][0]  concatenate_168[0][0]  dropout_180[0][0]  concatenate_169[0][0]
separable_conv2d_129 (Separable dropout_180 (Dropout) concatenate_169 (Concatenate) batch_normalization_184 (BatchN activation_184 (Activation)	(None, 8, 8, 16) (None, 8, 8, 16) (None, 8, 8, 48) (None, 8, 8, 48) (None, 8, 8, 48)	2080 0 0 192	activation_183[0][0]  separable_conv2d_129[0][0]  concatenate_168[0][0]  dropout_180[0][0]  concatenate_169[0][0]  batch_normalization_184[0][0]
separable_conv2d_129 (Separable dropout_180 (Dropout)  concatenate_169 (Concatenate)  batch_normalization_184 (BatchNactivation_184 (Activation)  separable_conv2d_130 (Separable	(None, 8, 8, 16) (None, 8, 8, 16) (None, 8, 8, 48) (None, 8, 8, 48) (None, 8, 8, 48)	2080 0 0 192 0 3120	activation_183[0][0]  separable_conv2d_129[0][0]  concatenate_168[0][0]  dropout_180[0][0]  concatenate_169[0][0]  batch_normalization_184[0][0]  activation_184[0][0]

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activation_185 (Activation)	(None,	8,	8,	64)	0	batch_normalization_185[0][0]
separable_conv2d_131 (Separable	(None,	8,	8,	16)	4160	activation_185[0][0]
dropout_182 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_131[0][0]
concatenate_171 (Concatenate)	(None,	8,	8,	80)	0	concatenate_170[0][0] dropout_182[0][0]
batch_normalization_186 (BatchN	(None,	8,	8,	80)	320	concatenate_171[0][0]
activation_186 (Activation)	(None,	8,	8,	80)	0	batch_normalization_186[0][0]
separable_conv2d_132 (Separable	(None,	8,	8,	16)	5200	activation_186[0][0]
dropout_183 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_132[0][0]
concatenate_172 (Concatenate)	(None,	8,	8,	96)	0	concatenate_171[0][0] dropout_183[0][0]
batch_normalization_187 (BatchN	(None,	8,	8,	96)	384	concatenate_172[0][0]
activation_187 (Activation)	(None,	8,	8,	96)	0	batch_normalization_187[0][0]
separable_conv2d_133 (Separable	(None,	8,	8,	16)	6240	activation_187[0][0]
dropout_184 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_133[0][0]
concatenate_173 (Concatenate)	(None,	8,	8,	112)	0	concatenate_172[0][0] dropout_184[0][0]
batch_normalization_188 (BatchN	(None,	8,	8,	112)	448	concatenate_173[0][0]
activation_188 (Activation)	(None,	8,	8,	112)	0	batch_normalization_188[0][0]
separable_conv2d_134 (Separable	(None,	8,	8,	16)	7280	activation_188[0][0]
dropout_185 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_134[0][0]
concatenate_174 (Concatenate)	(None,	8,	8,	128)	0	concatenate_173[0][0] dropout_185[0][0]
batch_normalization_189 (BatchN	(None,	8,	8,	128)	512	concatenate_174[0][0]
activation_189 (Activation)	(None,	8,	8,	128)	0	batch_normalization_189[0][0]
separable_conv2d_135 (Separable	(None,	8,	8,	16)	8320	activation_189[0][0]
dropout_186 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_135[0][0]
concatenate_175 (Concatenate)	(None,	8,	8,	144)	0	concatenate_174[0][0] dropout_186[0][0]
batch_normalization_190 (BatchN	(None,	8,	8,	144)	576	concatenate_175[0][0]
activation_190 (Activation)	(None,	8,	8,	144)	0	batch_normalization_190[0][0]
separable_conv2d_136 (Separable	(None,	8,	8,	16)	9360	activation_190[0][0]
dropout_187 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_136[0][0]
concatenate_176 (Concatenate)	(None,	8,	8,	160)	0	concatenate_175[0][0] dropout_187[0][0]
batch_normalization_191 (BatchN	(None,	8,	8,	160)	640	concatenate_176[0][0]
activation_191 (Activation)	(None,	8,	8,	160)	0	batch_normalization_191[0][0]
separable_conv2d_137 (Separable	(None,	8,	8,	16)	10400	activation_191[0][0]
dropout_188 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_137[0][0]
concatenate_177 (Concatenate)	(None,	8,	8,	176)	0	concatenate_176[0][0] dropout_188[0][0]
hatch normalization 192 (BatchN	(None.	8.	8.	176)	704	concatenate 177[0][0]

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activation_192 (Activation)	(None,	8,	8,	176)	0	batch_normalization_192[0][0]
separable_conv2d_138 (Separable	(None,	8,	8,	16)	11440	activation_192[0][0]
dropout_189 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_138[0][0]
concatenate_178 (Concatenate)	(None,	8,	8,	192)	0	concatenate_177[0][0] dropout_189[0][0]
batch_normalization_193 (BatchN	(None,	8,	8,	192)	768	concatenate_178[0][0]
activation_193 (Activation)	(None,	8,	8,	192)	0	batch_normalization_193[0][0]
separable_conv2d_139 (Separable	(None,	8,	8,	16)	12480	activation_193[0][0]
dropout_190 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_139[0][0]
concatenate_179 (Concatenate)	(None,	8,	8,	208)	0	concatenate_178[0][0] dropout_190[0][0]
batch_normalization_194 (BatchN	(None,	8,	8,	208)	832	concatenate_179[0][0]
activation_194 (Activation)	(None,	8,	8,	208)	0	batch_normalization_194[0][0]
separable_conv2d_140 (Separable	(None,	8,	8,	16)	3536	activation_194[0][0]
dropout_191 (Dropout)	(None,	8,	8,	16)	0	separable_conv2d_140[0][0]
average_pooling2d_14 (AveragePo	(None,	4,	4,	16)	0	dropout_191[0][0]
batch_normalization_195 (BatchN	(None,	4,	4,	16)	64	average_pooling2d_14[0][0]
activation_195 (Activation)	(None,	4,	4,	16)	0	batch_normalization_195[0][0]
separable_conv2d_141 (Separable	(None,	4,	4,	16)	1040	activation_195[0][0]
dropout_192 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_141[0][0]
concatenate_180 (Concatenate)	(None,	4,	4,	32)	0	average_pooling2d_14[0][0] dropout_192[0][0]
batch_normalization_196 (BatchN	(None,	4,	4,	32)	128	concatenate_180[0][0]
activation_196 (Activation)	(None,	4,	4,	32)	0	batch_normalization_196[0][0]
separable_conv2d_142 (Separable	(None,	4,	4,	16)	2080	activation_196[0][0]
dropout_193 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_142[0][0]
concatenate_181 (Concatenate)	(None,	4,	4,	48)	0	concatenate_180[0][0] dropout_193[0][0]
batch_normalization_197 (BatchN	(None,	4,	4,	48)	192	concatenate_181[0][0]
activation_197 (Activation)	(None,	4,	4,	48)	0	batch_normalization_197[0][0]
separable_conv2d_143 (Separable	(None,	4,	4,	16)	3120	activation_197[0][0]
dropout_194 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_143[0][0]
concatenate_182 (Concatenate)	(None,	4,	4,	64)	0	concatenate_181[0][0] dropout_194[0][0]
batch_normalization_198 (BatchN	(None,	4,	4,	64)	256	concatenate_182[0][0]
activation_198 (Activation)	(None,	4,	4,	64)	0	batch_normalization_198[0][0]
separable_conv2d_144 (Separable	(None,	4,	4,	16)	4160	activation_198[0][0]
dropout_195 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_144[0][0]
concatenate_183 (Concatenate)	(None,	4,	4,	80)	0	concatenate_182[0][0] dropout_195[0][0]
batch_normalization_199 (BatchN	(None,	4,	4,	80)	320	concatenate_183[0][0]

activation_199 (Activation)	(None,	4,	4,	80)	0	batch_normalization_199[0][0]
separable_conv2d_145 (Separable	(None,	4,	4,	16)	5200	activation_199[0][0]
dropout_196 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_145[0][0]
concatenate_184 (Concatenate)	(None,	4,	4,	96)	0	concatenate_183[0][0] dropout_196[0][0]
batch_normalization_200 (BatchN	(None,	4,	4,	96)	384	concatenate_184[0][0]
activation_200 (Activation)	(None,	4,	4,	96)	0	batch_normalization_200[0][0]
separable_conv2d_146 (Separable	(None,	4,	4,	16)	6240	activation_200[0][0]
dropout_197 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_146[0][0]
concatenate_185 (Concatenate)	(None,	4,	4,	112)	0	concatenate_184[0][0] dropout_197[0][0]
batch_normalization_201 (BatchN	(None,	4,	4,	112)	448	concatenate_185[0][0]
activation_201 (Activation)	(None,	4,	4,	112)	0	batch_normalization_201[0][0]
separable_conv2d_147 (Separable	(None,	4,	4,	16)	7280	activation_201[0][0]
dropout_198 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_147[0][0]
concatenate_186 (Concatenate)	(None,	4,	4,	128)	0	concatenate_185[0][0] dropout_198[0][0]
batch_normalization_202 (BatchN	(None,	4,	4,	128)	512	concatenate_186[0][0]
activation_202 (Activation)	(None,	4,	4,	128)	0	batch_normalization_202[0][0]
separable_conv2d_148 (Separable	(None,	4,	4,	16)	8320	activation_202[0][0]
dropout_199 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_148[0][0]
concatenate_187 (Concatenate)	(None,	4,	4,	144)	0	concatenate_186[0][0] dropout_199[0][0]
batch_normalization_203 (BatchN	(None,	4,	4,	144)	576	concatenate_187[0][0]
activation_203 (Activation)	(None,	4,	4,	144)	0	batch_normalization_203[0][0]
separable_conv2d_149 (Separable	(None,	4,	4,	16)	9360	activation_203[0][0]
dropout_200 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_149[0][0]
concatenate_188 (Concatenate)	(None,	4,	4,	160)	0	concatenate_187[0][0] dropout_200[0][0]
batch_normalization_204 (BatchN	(None,	4,	4,	160)	640	concatenate_188[0][0]
activation_204 (Activation)	(None,	4,	4,	160)	0	batch_normalization_204[0][0]
separable_conv2d_150 (Separable	(None,	4,	4,	16)	10400	activation_204[0][0]
dropout_201 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_150[0][0]
concatenate_189 (Concatenate)	(None,	4,	4,	176)	0	concatenate_188[0][0] dropout_201[0][0]
batch_normalization_205 (BatchN	(None,	4,	4,	176)	704	concatenate_189[0][0]
activation_205 (Activation)	(None,	4,	4,	176)	0	batch_normalization_205[0][0]
separable_conv2d_151 (Separable	(None,	4,	4,	16)	11440	activation_205[0][0]
dropout_202 (Dropout)	(None,	4,	4,	16)	0	separable_conv2d_151[0][0]
concatenate_190 (Concatenate)	(None,	4,	4,	192)	0	concatenate_189[0][0] dropout_202[0][0]
batch normalization 206 (BatchN	/Nono		4 .	192)	768	concatenate 190[0][0]

(None,	4,	4,	192)	0	batch_normalization_206[0][0]
(None,	4,	4,	16)	12480	activation_206[0][0]
(None,	4,	4,	16)	0	separable_conv2d_152[0][0]
(None,	4,	4,	208)	0	concatenate_190[0][0] dropout_203[0][0]
(None,	4,	4,	208)	832	concatenate_191[0][0]
(None,	4,	4,	208)	0	batch_normalization_207[0][0]
(None,	2,	2,	208)	0	activation_207[0][0]
(None,	832	2)		0	average_pooling2d_15[0][0]
(None,	10)		8330	flatten_3[0][0]
	(None, (None, (None, (None, (None,	(None, 4, (None, 4, (None, 4, (None, 4, (None, 2, (None, 83.	(None, 4, 4, 4, (None, 4, 4, 4, (None, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4,	· · · · · · · · · · · · · · · · · · ·	(None, 4, 4, 16) 12480 (None, 4, 4, 16) 0 (None, 4, 4, 208) 0 (None, 4, 4, 208) 832 (None, 4, 4, 208) 0 (None, 2, 2, 208) 0 (None, 832) 0

Total params: 385,002 Trainable params: 372,938 Non-trainable params: 12,064

In [60]:

```
model 4.compile(optimizer='adam', loss='categorical crossentropy', metrics=['accuracy'])
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
patience = 50
base_path = '/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/'
checkpoint file name = base path + 'CIFAR model4' + ' {epoch:02d}-{val_accuracy:.2f}.hdf5'
model_checkpoint = ModelCheckpoint(checkpoint_file_name, monitor='val_accuracy', verbose=1, save_be
st only=True)
early stop = EarlyStopping('val accuracy', mode='max', patience = patience)
reduce LR = ReduceLROnPlateau (monitor='val accuracy', mode='max', factor=0.1, patience=int(patience
/3), verbose=1)
callbacks = [model checkpoint, early stop, reduce LR]
epochs = 300
batch size = 256
#https://keras.io/api/preprocessing/image/#flow-method
history_4 = model_4.fit(data_generator.flow(X_train, y_train, batch_size),
                    steps_per_epoch = int(len(X_train)/batch_size),
                    epochs = epochs,
                    callbacks = callbacks,
                    validation_data = (X_test, y_test), verbose=1)
```

WARNING:tensorflow:sample weight modes were coerced from to ['...'] Train for 195 steps, validate on 10000 samples Epoch 1/300 Epoch 00001: val accuracy improved from -inf to 0.10000, saving model to /home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model4 01-0.10.hdf5 loss: 2.3413 - val_accuracy: 0.1000 Epoch 2/300 Epoch 00002: val accuracy improved from 0.10000 to 0.19030, saving model to /home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model4 02-0.19.hdf5 loss: 2.2503 - val accuracy: 0.1903 Epoch 3/300 Epoch 00003: val accuracy improved from 0.19030 to 0.44100, saving model to /home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_03-0.44.hdf5 loss: 1.9132 - val accuracy: 0.4410

```
Epoch 4/300
Epoch 00004: val accuracy improved from 0.44100 to 0.48850, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model4 04-0.49.hdf5
loss: 2.1404 - val accuracy: 0.4885
Epoch 5/300
Epoch 00005: val accuracy improved from 0.48850 to 0.62000, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model4 05-0.62.hdf5
loss: 1.2004 - val_accuracy: 0.6200
Epoch 6/300
Epoch 00006: val accuracy did not improve from 0.62000
loss: 1.5936 - val accuracy: 0.5615
Epoch 7/300
Epoch 00007: val accuracy improved from 0.62000 to 0.65450, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_07-0.65.hdf5
loss: 1.1688 - val_accuracy: 0.6545
Epoch 8/300
Epoch 00008: val accuracy did not improve from 0.65450
loss: 1.9574 - val_accuracy: 0.5217
Epoch 9/300
Epoch 00009: val_accuracy improved from 0.65450 to 0.70890, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_09-0.71.hdf5
195/195 [============ ] - 59s 301ms/step - loss: 0.7668 - accuracy: 0.7304 - val
loss: 0.9380 - val_accuracy: 0.7089
Epoch 10/300
Epoch 00010: val accuracy did not improve from 0.70890
loss: 2.8126 - val accuracy: 0.4757
Epoch 11/300
Epoch 00011: val accuracy improved from 0.70890 to 0.73420, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_11-0.73.hdf5
loss: 0.8638 - val_accuracy: 0.7342
Epoch 12/300
Epoch 00012: val accuracy did not improve from 0.73420
loss: 1.4645 - val accuracy: 0.6306
Epoch 13/300
Epoch 00013: val accuracy did not improve from 0.73420
loss: 1.1560 - val_accuracy: 0.6718
Epoch 14/300
Epoch 00014: val accuracy did not improve from 0.73420
loss: 1.1602 - val_accuracy: 0.6892
Epoch 15/300
Epoch 00015: val accuracy did not improve from 0.73420
loss: 1.3772 - val_accuracy: 0.6478
Epoch 16/300
Epoch 00016: val accuracy did not improve from 0.73420
loss: 1.0045 - val_accuracy: 0.7176
Epoch 17/300
Epoch 00017: val_accuracy did not improve from 0.73420
loss: 1.4011 - val_accuracy: 0.6322
Epoch 18/300
```

```
Epoch 00018: val accuracy did not improve from 0.73420
loss: 1.0604 - val accuracy: 0.7087
Epoch 19/300
Epoch 00019: val accuracy improved from 0.73420 to 0.74620, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model4 19-0.75.hdf5
loss: 0.8668 - val accuracy: 0.7462
Epoch 20/300
Epoch 00020: val accuracy did not improve from 0.74620
loss: 0.8953 - val accuracy: 0.7378
Epoch 21/300
Epoch 00021: val accuracy did not improve from 0.74620
loss: 1.0524 - val_accuracy: 0.7157
Epoch 22/300
Epoch 00022: val accuracy did not improve from 0.74620
loss: 1.0577 - val accuracy: 0.7269
Epoch 23/300
Epoch 00023: val accuracy did not improve from 0.74620
loss: 1.1614 - val accuracy: 0.7007
Epoch 24/300
Epoch 00024: val accuracy improved from 0.74620 to 0.76850, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_24-0.77.hdf5
loss: 0.7614 - val accuracy: 0.7685
Epoch 25/300
Epoch 00025: val accuracy did not improve from 0.76850
loss: 0.8801 - val accuracy: 0.7589
Epoch 26/300
Epoch 00026: val_accuracy improved from 0.76850 to 0.79270, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_26-0.79.hdf5
loss: 0.7050 - val_accuracy: 0.7927
Epoch 27/300
Epoch 00027: val_accuracy did not improve from 0.79270
loss: 1.1469 - val accuracy: 0.7176
Epoch 28/300
Epoch 00028: val accuracy did not improve from 0.79270
loss: 1.2111 - val accuracy: 0.7168
Epoch 29/300
Epoch 00029: val_accuracy did not improve from 0.79270
loss: 0.7486 - val accuracy: 0.7850
Epoch 30/300
Epoch 00030: val accuracy improved from 0.79270 to 0.79300, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model4 30-0.79.hdf5
loss: 0.7231 - val accuracy: 0.7930
Epoch 31/300
Epoch 00031: val accuracy did not improve from 0.79300
loss: 0.8326 - val_accuracy: 0.7646
Epoch 32/300
Epoch 00032: val accuracy did not improve from 0.79300
```

loss: 1.0786 - val accuracy: 0.7239

```
Epoch 33/300
Epoch 00033: val accuracy improved from 0.79300 to 0.80390, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model4 33-0.80.hdf5
loss: 0.6970 - val_accuracy: 0.8039
Epoch 34/300
Epoch 00034: val accuracy did not improve from 0.80390
loss: 1.1122 - val_accuracy: 0.7356
Epoch 35/300
Epoch 00035: val_accuracy did not improve from 0.80390
loss: 0.9059 - val_accuracy: 0.7630
Epoch 36/300
Epoch 00036: val accuracy did not improve from 0.80390
loss: 1.1688 - val accuracy: 0.7184
Epoch 37/300
Epoch 00037: val accuracy improved from 0.80390 to 0.80720, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_37-0.81.hdf5
loss: 0.6776 - val_accuracy: 0.8072
Epoch 38/300
Epoch 00038: val_accuracy improved from 0.80720 to 0.81510, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_38-0.82.hdf5
loss: 0.6534 - val_accuracy: 0.8151
Epoch 39/300
Epoch 00039: val_accuracy did not improve from 0.81510
loss: 0.8656 - val_accuracy: 0.7855
Epoch 40/300
Epoch 00040: val accuracy did not improve from 0.81510
loss: 0.8044 - val_accuracy: 0.7890
Epoch 41/300
Epoch 00041: val_accuracy did not improve from 0.81510
loss: 1.1655 - val accuracy: 0.7376
Epoch 42/300
Epoch 00042: val accuracy did not improve from 0.81510
loss: 0.6769 - val accuracy: 0.8106
Epoch 43/300
Epoch 00043: val accuracy did not improve from 0.81510
loss: 0.8867 - val_accuracy: 0.7758
Epoch 44/300
Epoch 00044: val_accuracy did not improve from 0.81510
loss: 0.6781 - val_accuracy: 0.8115
Epoch 45/300
Epoch 00045: val accuracy did not improve from 0.81510
loss: 0.8694 - val_accuracy: 0.7861
Epoch 46/300
Epoch 00046: val accuracy did not improve from 0.81510
loss: 1.1198 - val_accuracy: 0.7407
Epoch 47/300
Epoch 00047: val_accuracy did not improve from 0.81510
```

```
1 JOB EJOMB, DUCP TODD: 0.0700 GCCGTGGT. 0.0000 VAT
loss: 0.7218 - val_accuracy: 0.8016
Epoch 48/300
Epoch 00048: val accuracy did not improve from 0.81510
loss: 0.7497 - val accuracy: 0.8040
Epoch 49/300
Epoch 00049: val accuracy improved from 0.81510 to 0.82350, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR model4 49-0.82.hdf5
loss: 0.6711 - val_accuracy: 0.8235
Epoch 50/300
Epoch 00050: val accuracy did not improve from 0.82350
loss: 0.7510 - val_accuracy: 0.8079
Epoch 51/300
Epoch 00051: val accuracy improved from 0.82350 to 0.83610, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_51-0.84.hdf5
loss: 0.5678 - val_accuracy: 0.8361
Epoch 52/300
Epoch 00052: val accuracy did not improve from 0.83610
loss: 0.7490 - val_accuracy: 0.8012
Epoch 53/300
Epoch 00053: val_accuracy did not improve from 0.83610
loss: 0.7325 - val accuracy: 0.8150
Epoch 54/300
Epoch 00054: val accuracy did not improve from 0.83610
loss: 0.6157 - val accuracy: 0.8282
Epoch 55/300
Epoch 00055: val accuracy did not improve from 0.83610
loss: 0.7196 - val accuracy: 0.8118
Epoch 56/300
Epoch 00056: val accuracy did not improve from 0.83610
loss: 0.9316 - val_accuracy: 0.7835
Epoch 57/300
Epoch 00057: val accuracy did not improve from 0.83610
loss: 1.0981 - val_accuracy: 0.7480
Epoch 58/300
Epoch 00058: val accuracy did not improve from 0.83610
loss: 0.7190 - val accuracy: 0.8167
Epoch 59/300
Epoch 00059: val_accuracy did not improve from 0.83610
loss: 0.8712 - val_accuracy: 0.7893
Epoch 60/300
Epoch 00060: val accuracy did not improve from 0.83610
loss: 0.8914 - val accuracy: 0.7858
Epoch 61/300
Epoch 00061: val accuracy did not improve from 0.83610
loss: 0.6456 - val_accuracy: 0.8250
Epoch 62/300
Epoch 00062: val accuracy did not improve from 0.83610
```

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                                       vа⊥
loss: 0.6300 - val_accuracy: 0.8296
Epoch 63/300
Epoch 00063: val_accuracy did not improve from 0.83610
loss: 0.6779 - val accuracy: 0.8112
Epoch 64/300
Epoch 00064: val accuracy did not improve from 0.83610
loss: 0.7649 - val accuracy: 0.8134
Epoch 65/300
Epoch 00065: val accuracy did not improve from 0.83610
loss: 0.9613 - val accuracy: 0.7889
Epoch 66/300
Epoch 00066: val accuracy did not improve from 0.83610
loss: 0.6972 - val accuracy: 0.8156
Epoch 67/300
Epoch 00067: val accuracy did not improve from 0.83610
Epoch 00067: ReduceLROnPlateau reducing learning rate to 0.00010000000474974513.
loss: 0.9667 - val accuracy: 0.7855
Epoch 68/300
Epoch 00068: val accuracy improved from 0.83610 to 0.86340, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_68-0.86.hdf5
loss: 0.4823 - val_accuracy: 0.8634
Epoch 69/300
Epoch 00069: val accuracy improved from 0.86340 to 0.86590, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model4 69-0.87.hdf5
loss: 0.4892 - val accuracy: 0.8659
Epoch 70/300
Epoch 00070: val accuracy did not improve from 0.86590
loss: 0.4795 - val_accuracy: 0.8637
Epoch 71/300
Epoch 00071: val_accuracy did not improve from 0.86590
loss: 0.5331 - val_accuracy: 0.8573
Epoch 72/300
Epoch 00072: val accuracy did not improve from 0.86590
loss: 0.5086 - val accuracy: 0.8637
Epoch 73/300
Epoch 00073: val accuracy did not improve from 0.86590
loss: 0.5293 - val_accuracy: 0.8612
Epoch 74/300
Epoch 00074: val accuracy did not improve from 0.86590
loss: 0.5277 - val accuracy: 0.8602
Epoch 75/300
Epoch 00075: val accuracy improved from 0.86590 to 0.86890, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model4 75-0.87.hdf5
loss: 0.4848 - val accuracy: 0.8689
Epoch 76/300
Epoch 00076: val accuracy did not improve from 0.86890
loss: 0.5205 - val accuracy: 0.8607
```

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Epoch 00077: val_accuracy did not improve from 0.86890
loss: 0.5112 - val_accuracy: 0.8644
Epoch 78/300
Epoch 00078: val accuracy improved from 0.86890 to 0.87080, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_78-0.87.hdf5
loss: 0.4802 - val accuracy: 0.8708
Epoch 79/300
Epoch 00079: val accuracy did not improve from 0.87080
loss: 0.5028 - val accuracy: 0.8660
Epoch 80/300
Epoch 00080: val_accuracy improved from 0.87080 to 0.87160, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_80-0.87.hdf5
loss: 0.4873 - val_accuracy: 0.8716
Epoch 81/300
Epoch 00081: val accuracy did not improve from 0.87160
loss: 0.5059 - val accuracy: 0.8672
Epoch 82/300
Epoch 00082: val accuracy did not improve from 0.87160
loss: 0.4989 - val_accuracy: 0.8686
Epoch 83/300
Epoch 00083: val_accuracy did not improve from 0.87160
loss: 0.5171 - val_accuracy: 0.8645
Epoch 84/300
Epoch 00084: val accuracy did not improve from 0.87160
loss: 0.4935 - val accuracy: 0.8711
Epoch 85/300
Epoch 00085: val accuracy did not improve from 0.87160
loss: 0.5030 - val_accuracy: 0.8680
Epoch 86/300
Epoch 00086: val_accuracy did not improve from 0.87160
loss: 0.5246 - val_accuracy: 0.8655
Epoch 87/300
Epoch 00087: val_accuracy did not improve from 0.87160
loss: 0.5086 - val accuracy: 0.8635
Epoch 88/300
Epoch 00088: val accuracy did not improve from 0.87160
loss: 0.5080 - val accuracy: 0.8695
Epoch 89/300
Epoch 00089: val_accuracy did not improve from 0.87160
loss: 0.5341 - val accuracy: 0.8611
Epoch 90/300
Epoch 00090: val accuracy did not improve from 0.87160
loss: 0.5030 - val accuracy: 0.8675
Epoch 91/300
Epoch 00091: val accuracy did not improve from 0.87160
loss: 0.5199 - val_accuracy: 0.8657
```

Enach 02/200

```
EPOCII 92/300
Epoch 00092: val_accuracy did not improve from 0.87160
loss: 0.5178 - val accuracy: 0.8633
Epoch 93/300
Epoch 00093: val accuracy did not improve from 0.87160
loss: 0.5139 - val_accuracy: 0.8652
Epoch 94/300
Epoch 00094: val accuracy did not improve from 0.87160
loss: 0.5276 - val accuracy: 0.8646
Epoch 95/300
Epoch 00095: val accuracy did not improve from 0.87160
loss: 0.5513 - val_accuracy: 0.8597
Epoch 96/300
Epoch 00096: val accuracy did not improve from 0.87160
Epoch 00096: ReduceLROnPlateau reducing learning rate to 1.0000000474974514e-05.
loss: 0.4776 - val accuracy: 0.8710
Epoch 97/300
Epoch 00097: val accuracy did not improve from 0.87160
loss: 0.5014 - val accuracy: 0.8693
Epoch 98/300
Epoch 00098: val accuracy did not improve from 0.87160
loss: 0.5089 - val_accuracy: 0.8675
Epoch 99/300
Epoch 00099: val accuracy did not improve from 0.87160
loss: 0.5057 - val_accuracy: 0.8687
Epoch 100/300
Epoch 00100: val_accuracy did not improve from 0.87160
loss: 0.5033 - val_accuracy: 0.8694
Epoch 101/300
Epoch 00101: val_accuracy did not improve from 0.87160
loss: 0.5015 - val_accuracy: 0.8692
Epoch 102/300
Epoch 00102: val accuracy did not improve from 0.87160
loss: 0.5028 - val accuracy: 0.8675
Epoch 103/300
Epoch 00103: val accuracy did not improve from 0.87160
loss: 0.4995 - val_accuracy: 0.8694
Epoch 104/300
Epoch 00104: val accuracy did not improve from 0.87160
loss: 0.5052 - val_accuracy: 0.8681
Epoch 105/300
Epoch 00105: val accuracy did not improve from 0.87160
loss: 0.4970 - val accuracy: 0.8693
Epoch 106/300
Epoch 00106: val_accuracy did not improve from 0.87160
loss: 0.5043 - val_accuracy: 0.8687
```

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Epocn IU//3UU
Epoch 00107: val accuracy did not improve from 0.87160
loss: 0.5058 - val accuracy: 0.8691
Epoch 108/300
Epoch 00108: val accuracy did not improve from 0.87160
loss: 0.5037 - val_accuracy: 0.8693
Epoch 109/300
Epoch 00109: val accuracy did not improve from 0.87160
loss: 0.5083 - val_accuracy: 0.8674
Epoch 110/300
Epoch 00110: val accuracy did not improve from 0.87160
loss: 0.5007 - val_accuracy: 0.8695
Epoch 111/300
Epoch 00111: val_accuracy did not improve from 0.87160
loss: 0.5015 - val accuracy: 0.8693
Epoch 112/300
Epoch 00112: val accuracy did not improve from 0.87160
Epoch 00112: ReduceLROnPlateau reducing learning rate to 1.0000000656873453e-06.
loss: 0.4992 - val accuracy: 0.8702
Epoch 113/300
Epoch 00113: val accuracy did not improve from 0.87160
loss: 0.5008 - val_accuracy: 0.8696
Epoch 114/300
Epoch 00114: val accuracy did not improve from 0.87160
loss: 0.5029 - val_accuracy: 0.8694
Epoch 115/300
Epoch 00115: val accuracy did not improve from 0.87160
loss: 0.5032 - val accuracy: 0.8692
Epoch 116/300
Epoch 00116: val accuracy did not improve from 0.87160
loss: 0.5033 - val accuracy: 0.8690
Epoch 117/300
Epoch 00117: val accuracy did not improve from 0.87160
loss: 0.5030 - val accuracy: 0.8694
Epoch 118/300
Epoch 00118: val accuracy did not improve from 0.87160
loss: 0.5023 - val_accuracy: 0.8689
Epoch 119/300
Epoch 00119: val accuracy did not improve from 0.87160
loss: 0.5032 - val_accuracy: 0.8689
Epoch 120/300
Epoch 00120: val accuracy did not improve from 0.87160
loss: 0.5026 - val accuracy: 0.8690
Epoch 121/300
Epoch 00121: val_accuracy did not improve from 0.87160
loss: 0.5038 - val accuracy: 0.8691
```

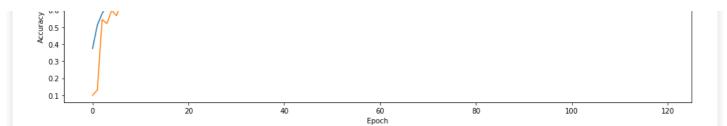
```
Epoch 122/300
Epoch 00122: val accuracy did not improve from 0.87160
loss: 0.5021 - val_accuracy: 0.8689
Epoch 123/300
Epoch 00123: val_accuracy did not improve from 0.87160
loss: 0.5026 - val_accuracy: 0.8691
Epoch 124/300
Epoch 00124: val_accuracy did not improve from 0.87160
loss: 0.5015 - val accuracy: 0.8693
Epoch 125/300
Epoch 00125: val accuracy did not improve from 0.87160
loss: 0.5024 - val_accuracy: 0.8689
Epoch 126/300
Epoch 00126: val_accuracy did not improve from 0.87160
loss: 0.5033 - val accuracy: 0.8689
Epoch 127/300
Epoch 00127: val accuracy did not improve from 0.87160
loss: 0.5031 - val accuracy: 0.8690
Epoch 128/300
Epoch 00128: val accuracy did not improve from 0.87160
Epoch 00128: ReduceLROnPlateau reducing learning rate to 1.0000001111620805e-07.
195/195 [============= ] - 58s 298ms/step - loss: 0.1997 - accuracy: 0.9286 - val_
loss: 0.5039 - val_accuracy: 0.8688
Epoch 129/300
Epoch 00129: val accuracy did not improve from 0.87160
loss: 0.5036 - val accuracy: 0.8690
Epoch 130/300
Epoch 00130: val accuracy did not improve from 0.87160
loss: 0.5010 - val accuracy: 0.8690
In [61]:
#history plot for accyracy
plt.figure(figsize=(16,4))
plt.plot(history_3.history['accuracy'])
plt.plot(history_3.history['val_accuracy'])
plt.title("Model Accuracy")
plt.xlabel("Epoch")
plt.ylabel("Accuracy")
plt.legend(["train", "test"], loc="upper left")
plt.show()
# history plot for accuracy
plt.figure(figsize=(16,4))
plt.plot(history_3.history["loss"])
plt.plot(history_3.history["val_loss"])
plt.title("Model Loss")
plt.xlabel("Epoch")
plt.ylabel("Loss")
plt.legend(["train", "test"], loc="upper left")
plt.show()
```

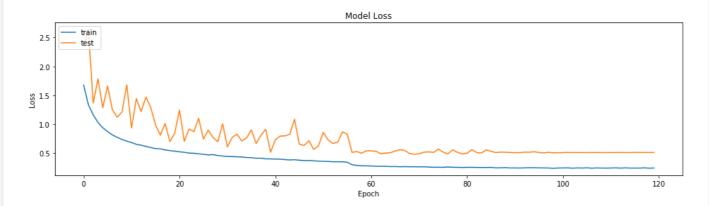
Model Accuracy

0.9

0.8

— train





In [63]:

```
best_model_4 =
    tf.keras.models.load_model('/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model4_80-0.87
    .hdf5')
    scores = best_model_4.evaluate(X_test, y_test, verbose=1)
    print(scores)
```

Model -5:

- dropout = 0.2
- compression = 0.3
- · using conv instead of fully connected layers

In [67]:

```
#BN-->ReLU-->Conv2D-->Dropout-->concat(input, output)-->(put in loop)
from tensorflow.keras import layers
from tensorflow.keras.models import Model
num classes = 10
def denseblock(input, num_filter, dropout_rate):
                         # to keep the growth rate of number of filters
    global compression
    temp = input
    for in range(l):
        BatchNorm = layers.BatchNormalization()(temp)
        relu = layers.Activation('relu')(BatchNorm)
        Conv2D 7 7= layers.SeparableConv2D(int(num filter*compression), (7,7), use bias=False, padd
ing='same') (relu)
        if dropout rate>0:
            Conv2D_7_7 = layers.Dropout(dropout_rate)(Conv2D_7_7)
        \#concat the input(temp) and output(conv2d 7 7) , in resnet we add but here we concat
        concat = layers.Concatenate(axis=-1)([temp,Conv2D 7 7])
        #change the concat as input
        temp = concat
    return temp
#BN-->relu-->conv2d(1x1)-->dropout-->avg pool
def transition(input, num filter, dropout rate):
    {\bf global} \ {\bf compression}
                lawara BatahNarmalization () (innut)
```

```
Batchworm = Tayers.Batchwormaltzation()(Imput)
    relu = layers.Activation('relu')(BatchNorm)
   Conv2D_BottleNeck = layers.SeparableConvolution2D(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
#BN-->relu-->avgpool-->flat-->softmax
def output_layer(input):
   global compression
   BatchNorm = layers.BatchNormalization()(input)
   relu = layers.Activation('relu')(BatchNorm)
   AvgPooling = layers.AveragePooling2D(pool_size=(2,2))(relu)
    conv op = layers.SeparableConvolution2D(num classes, (2,2), padding='valid') (AvgPooling)
   flat = layers.Flatten()(conv op)
    output = layers.Dense(num classes, activation='softmax')(flat)
    return output
# Hyperparameters
1 = 21
num filter = 48
compression = 0.3
dropout_rate = 0.2
num classes = 10
input = layers.Input(shape=(input size))
First Conv2D = layers.Conv2D(num filter, (7,7), use bias=False ,padding='same')(input)
#First dense and transition block
First Block = denseblock(First Conv2D, num filter, dropout rate)
First_Transition = transition(First_Block, num_filter, dropout_rate)
#Second dense and transition block
Second Block = denseblock(First Transition, num filter, dropout rate)
Second Transition = transition(Second Block, num filter, dropout rate)
#Third dense and transition block
Third Block = denseblock (Second Transition, num filter, dropout rate)
Third Transition = transition(Third Block, num filter, dropout rate)
#last dense and output block
Last Block = denseblock(Third Transition, num filter, dropout rate)
output = output_layer(Last_Block)
```

In [68]:

```
model_5 = tf.keras.models.Model(inputs=[input], outputs=[output])
model_5.summary()
```

Model: "model 20"

Layer (type)	Output Shape	Param #	Connected to
input_24 (InputLayer)	[(None, 32, 32, 3)]	0	
conv2d_32 (Conv2D)	(None, 32, 32, 48)	7056	input_24[0][0]
batch_normalization_2476 (Batch	(None, 32, 32, 48)	192	conv2d_32[0][0]
activation_2476 (Activation)	(None, 32, 32, 48)	0	batch_normalization_2476[0][0]
separable_conv2d_2463 (Separabl	(None, 32, 32, 14)	3024	activation_2476[0][0]
dropout_1339 (Dropout)	(None, 32, 32, 14)	0	separable_conv2d_2463[0][0]
concatenate_2384 (Concatenate)	(None, 32, 32, 62)	0	conv2d_32[0][0] dropout_1339[0][0]
batch_normalization_2477 (Batch	(None, 32, 32, 62)	248	concatenate_2384[0][0]
activation_2477 (Activation)	(None, 32, 32, 62)	0	batch_normalization_2477[0][0]
separable_conv2d_2464 (Separabl	(None, 32, 32, 14)	3906	activation_2477[0][0]

dropout_1340 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2464[0][0]
concatenate_2385 (Concatenate)	(None,	32,	32,	76)	0	concatenate_2384[0][0] dropout_1340[0][0]
oatch_normalization_2478 (Batch	(None,	32,	32,	76)	304	concatenate_2385[0][0]
activation_2478 (Activation)	(None,	32,	32,	76)	0	batch_normalization_2478[0][0]
separable_conv2d_2465 (Separabl	(None,	32,	32,	14)	4788	activation_2478[0][0]
dropout_1341 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2465[0][0]
concatenate_2386 (Concatenate)	(None,	32,	32,	90)	0	concatenate_2385[0][0] dropout_1341[0][0]
patch_normalization_2479 (Batch	(None,	32,	32,	90)	360	concatenate_2386[0][0]
activation_2479 (Activation)	(None,	32,	32,	90)	0	batch_normalization_2479[0][0]
separable_conv2d_2466 (Separabl	(None,	32,	32,	14)	5670	activation_2479[0][0]
dropout_1342 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2466[0][0]
concatenate_2387 (Concatenate)	(None,	32,	32,	104)	0	concatenate_2386[0][0] dropout_1342[0][0]
patch_normalization_2480 (Batch	(None,	32,	32,	104)	416	concatenate_2387[0][0]
activation_2480 (Activation)	(None,	32,	32,	104)	0	batch_normalization_2480[0][0]
separable_conv2d_2467 (Separabl	(None,	32,	32,	14)	6552	activation_2480[0][0]
dropout_1343 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2467[0][0]
concatenate_2388 (Concatenate)	(None,	32,	32,	118)	0	concatenate_2387[0][0] dropout_1343[0][0]
patch_normalization_2481 (Batch	(None,	32,	32,	118)	472	concatenate_2388[0][0]
activation_2481 (Activation)	(None,	32,	32,	118)	0	batch_normalization_2481[0][0]
separable_conv2d_2468 (Separabl	(None,	32,	32,	14)	7434	activation_2481[0][0]
dropout_1344 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2468[0][0]
concatenate_2389 (Concatenate)	(None,	32,	32,	132)	0	concatenate_2388[0][0] dropout_1344[0][0]
oatch_normalization_2482 (Batch	(None,	32,	32,	132)	528	concatenate_2389[0][0]
activation_2482 (Activation)	(None,	32,	32,	132)	0	batch_normalization_2482[0][0]
separable_conv2d_2469 (Separabl	(None,	32,	32,	14)	8316	activation_2482[0][0]
dropout_1345 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2469[0][0]
concatenate_2390 (Concatenate)	(None,	32,	32,	146)	0	concatenate_2389[0][0] dropout_1345[0][0]
patch_normalization_2483 (Batch	(None,	32,	32,	146)	584	concatenate_2390[0][0]
activation_2483 (Activation)	(None,	32,	32,	146)	0	batch_normalization_2483[0][0]
separable_conv2d_2470 (Separabl	(None,	32,	32,	14)	9198	activation_2483[0][0]
dropout_1346 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2470[0][0]
concatenate_2391 (Concatenate)	(None,	32,	32,	160)	0	concatenate_2390[0][0] dropout_1346[0][0]
patch_normalization_2484 (Batch	(None,	32,	32,	160)	640	concatenate_2391[0][0]
activation_2484 (Activation)	(None,	32,	32,	160)	0	batch_normalization_2484[0][0]
separable_conv2d_2471 (Separabl	(None,	32,	32,	14)	10080	activation_2484[0][0]

dropout_1347 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2471[0][0]
concatenate_2392 (Concatenate)	(None,	32,	32,	174)	0	concatenate_2391[0][0] dropout_1347[0][0]
oatch_normalization_2485 (Batch	(None,	32,	32,	174)	696	concatenate_2392[0][0]
activation_2485 (Activation)	(None,	32,	32,	174)	0	batch_normalization_2485[0][0]
separable_conv2d_2472 (Separabl	(None,	32,	32,	14)	10962	activation_2485[0][0]
dropout_1348 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2472[0][0]
concatenate_2393 (Concatenate)	(None,	32,	32,	188)	0	concatenate_2392[0][0] dropout_1348[0][0]
oatch_normalization_2486 (Batch	(None,	32,	32,	188)	752	concatenate_2393[0][0]
activation_2486 (Activation)	(None,	32,	32,	188)	0	batch_normalization_2486[0][0]
separable_conv2d_2473 (Separabl	(None,	32,	32,	14)	11844	activation_2486[0][0]
dropout_1349 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2473[0][0]
concatenate_2394 (Concatenate)	(None,	32,	32,	202)	0	concatenate_2393[0][0] dropout_1349[0][0]
patch_normalization_2487 (Batch	(None,	32,	32,	202)	808	concatenate_2394[0][0]
activation_2487 (Activation)	(None,	32,	32,	202)	0	batch_normalization_2487[0][0]
eparable_conv2d_2474 (Separabl	(None,	32,	32,	14)	12726	activation_2487[0][0]
Propout_1350 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2474[0][0]
concatenate_2395 (Concatenate)	(None,	32,	32,	216)	0	concatenate_2394[0][0] dropout_1350[0][0]
oatch_normalization_2488 (Batch	(None,	32,	32,	216)	864	concatenate_2395[0][0]
activation_2488 (Activation)	(None,	32,	32,	216)	0	batch_normalization_2488[0][0]
separable_conv2d_2475 (Separabl	(None,	32,	32,	14)	13608	activation_2488[0][0]
dropout_1351 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2475[0][0]
concatenate_2396 (Concatenate)	(None,	32,	32,	230)	0	concatenate_2395[0][0] dropout_1351[0][0]
patch_normalization_2489 (Batch	(None,	32,	32,	230)	920	concatenate_2396[0][0]
ctivation_2489 (Activation)	(None,	32,	32,	230)	0	batch_normalization_2489[0][0]
separable_conv2d_2476 (Separabl	(None,	32,	32,	14)	14490	activation_2489[0][0]
dropout_1352 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2476[0][0]
concatenate_2397 (Concatenate)	(None,	32,	32,	244)	0	concatenate_2396[0][0] dropout_1352[0][0]
oatch_normalization_2490 (Batch	(None,	32,	32,	244)	976	concatenate_2397[0][0]
activation_2490 (Activation)	(None,	32,	32,	244)	0	batch_normalization_2490[0][0]
separable_conv2d_2477 (Separabl	(None,	32,	32,	14)	15372	activation_2490[0][0]
dropout_1353 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2477[0][0]
concatenate_2398 (Concatenate)	(None,	32,	32,	258)	0	concatenate_2397[0][0] dropout_1353[0][0]
				2501	1032	2200[0][0]
patch_normalization_2491 (Batch	(None,	32,	32,	230)	1032	concatenate_2398[0][0]
patch_normalization_2491 (Batch_activation_2491 (Activation)	(None,				0	batch_normalization_2491[0][0]

dropout_1354 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2478[0][0]
concatenate_2399 (Concatenate)	(None,	32,	32,	272)	0	concatenate_2398[0][0] dropout_1354[0][0]
batch_normalization_2492 (Batch	(None,	32,	32,	272)	1088	concatenate_2399[0][0]
activation_2492 (Activation)	(None,	32,	32,	272)	0	batch_normalization_2492[0][0]
separable_conv2d_2479 (Separabl	(None,	32,	32,	14)	17136	activation_2492[0][0]
dropout_1355 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2479[0][0]
concatenate_2400 (Concatenate)	(None,	32,	32,	286)	0	concatenate_2399[0][0] dropout_1355[0][0]
batch_normalization_2493 (Batch	(None,	32,	32,	286)	1144	concatenate_2400[0][0]
activation_2493 (Activation)	(None,	32,	32,	286)	0	batch_normalization_2493[0][0]
separable_conv2d_2480 (Separabl	(None,	32,	32,	14)	18018	activation_2493[0][0]
dropout_1356 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2480[0][0]
concatenate_2401 (Concatenate)	(None,	32,	32,	300)	0	concatenate_2400[0][0] dropout_1356[0][0]
batch_normalization_2494 (Batch	(None,	32,	32,	300)	1200	concatenate_2401[0][0]
activation_2494 (Activation)	(None,	32,	32,	300)	0	batch_normalization_2494[0][0]
separable_conv2d_2481 (Separabl	(None,	32,	32,	14)	18900	activation_2494[0][0]
dropout_1357 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2481[0][0]
concatenate_2402 (Concatenate)	(None,	32,	32,	314)	0	concatenate_2401[0][0] dropout_1357[0][0]
batch_normalization_2495 (Batch	(None,	32,	32,	314)	1256	concatenate_2402[0][0]
activation_2495 (Activation)	(None,	32,	32,	314)	0	batch_normalization_2495[0][0]
separable_conv2d_2482 (Separabl	(None,	32,	32,	14)	19782	activation_2495[0][0]
dropout_1358 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2482[0][0]
concatenate_2403 (Concatenate)	(None,	32,	32,	328)	0	concatenate_2402[0][0] dropout_1358[0][0]
batch_normalization_2496 (Batch	(None,	32,	32,	328)	1312	concatenate_2403[0][0]
activation_2496 (Activation)	(None,	32,	32,	328)	0	batch_normalization_2496[0][0]
separable_conv2d_2483 (Separabl	(None,	32,	32,	14)	20664	activation_2496[0][0]
dropout_1359 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2483[0][0]
concatenate_2404 (Concatenate)	(None,	32,	32,	342)	0	concatenate_2403[0][0] dropout_1359[0][0]
batch_normalization_2497 (Batch	(None,	32,	32,	342)	1368	concatenate_2404[0][0]
activation_2497 (Activation)	(None,	32,	32,	342)	0	batch_normalization_2497[0][0]
separable_conv2d_2484 (Separabl	(None,	32,	32,	14)	5130	activation_2497[0][0]
dropout_1360 (Dropout)	(None,	32,	32,	14)	0	separable_conv2d_2484[0][0]
average_pooling2d_92 (AveragePo	(None,	16,	16,	14)	0	dropout_1360[0][0]
batch_normalization_2498 (Batch	(None,	16,	16,	14)	56	average_pooling2d_92[0][0]
activation_2498 (Activation)	(None,	16,	16,	14)	0	batch_normalization_2498[0][0]
separable_conv2d_2485 (Separabl	(None,	16,	16,	14)	882	activation_2498[0][0]
dronout 1361 (Dronout)	/None	16	16	141	n	senarable conv2d 2485[N][N]

aropoac_iooi (propoac)	(1MO11C' TO	, ₊ · ,	T = 1	v	seharante_comvsa_5100[0][0]
concatenate_2405 (Concatenate)	(None, 16	, 16,	28)	0	average_pooling2d_92[0][0] dropout_1361[0][0]
batch_normalization_2499 (Batch	(None, 16	, 16,	28)	112	concatenate_2405[0][0]
activation_2499 (Activation)	(None, 16	, 16,	28)	0	batch_normalization_2499[0][0]
separable_conv2d_2486 (Separabl	(None, 16	, 16,	14)	1764	activation_2499[0][0]
dropout_1362 (Dropout)	(None, 16	, 16,	14)	0	separable_conv2d_2486[0][0]
concatenate_2406 (Concatenate)	(None, 16	, 16,	42)	0	concatenate_2405[0][0] dropout_1362[0][0]
batch_normalization_2500 (Batch	(None, 16	, 16,	42)	168	concatenate_2406[0][0]
activation_2500 (Activation)	(None, 16	, 16,	42)	0	batch_normalization_2500[0][0]
separable_conv2d_2487 (Separabl	(None, 16	, 16,	14)	2646	activation_2500[0][0]
dropout_1363 (Dropout)	(None, 16	, 16,	14)	0	separable_conv2d_2487[0][0]
concatenate_2407 (Concatenate)	(None, 16	, 16,	56)	0	concatenate_2406[0][0] dropout_1363[0][0]
batch_normalization_2501 (Batch	(None, 16	, 16,	56)	224	concatenate_2407[0][0]
activation_2501 (Activation)	(None, 16	, 16,	56)	0	batch_normalization_2501[0][0]
separable_conv2d_2488 (Separabl	(None, 16	, 16,	14)	3528	activation_2501[0][0]
dropout_1364 (Dropout)	(None, 16	, 16,	14)	0	separable_conv2d_2488[0][0]
concatenate_2408 (Concatenate)	(None, 16	, 16,	70)	0	concatenate_2407[0][0] dropout_1364[0][0]
batch_normalization_2502 (Batch	(None, 16	, 16,	70)	280	concatenate_2408[0][0]
activation_2502 (Activation)	(None, 16	, 16,	70)	0	batch_normalization_2502[0][0]
separable_conv2d_2489 (Separabl	(None, 16	, 16,	14)	4410	activation_2502[0][0]
dropout_1365 (Dropout)	(None, 16	, 16,	14)	0	separable_conv2d_2489[0][0]
concatenate_2409 (Concatenate)	(None, 16	, 16,	84)	0	concatenate_2408[0][0] dropout_1365[0][0]
batch_normalization_2503 (Batch	(None, 16	, 16,	84)	336	concatenate_2409[0][0]
activation_2503 (Activation)	(None, 16	, 16,	84)	0	batch_normalization_2503[0][0]
separable_conv2d_2490 (Separabl	(None, 16	, 16,	14)	5292	activation_2503[0][0]
dropout_1366 (Dropout)	(None, 16	, 16,	14)	0	separable_conv2d_2490[0][0]
concatenate_2410 (Concatenate)	(None, 16	, 16,	98)	0	concatenate_2409[0][0] dropout_1366[0][0]
batch_normalization_2504 (Batch	(None, 16	, 16,	98)	392	concatenate_2410[0][0]
activation_2504 (Activation)	(None, 16	, 16,	98)	0	batch_normalization_2504[0][0]
separable_conv2d_2491 (Separabl	(None, 16	, 16,	14)	6174	activation_2504[0][0]
dropout_1367 (Dropout)	(None, 16	, 16,	14)	0	separable_conv2d_2491[0][0]
concatenate_2411 (Concatenate)	(None, 16	, 16,	112)	0	concatenate_2410[0][0] dropout_1367[0][0]
batch_normalization_2505 (Batch	(None, 16	, 16,	112)	448	concatenate_2411[0][0]
activation_2505 (Activation)	(None, 16	, 16,	112)	0	batch_normalization_2505[0][0]
separable_conv2d_2492 (Separabl	(None, 16	, 16,	14)	7056	activation_2505[0][0]
dropout 1368 (Propout)	/None 16	16	1 // \	Λ	canarahla control 2/02[0][0]

ατοροας τροο (πτοροας)	(14011 c ,	⊥∪,	⊥∪ ,	T.41	v	seharante_comvsd_s435[n][n]
concatenate_2412 (Concatenate)	(None,	16,	16,	126)	0	concatenate_2411[0][0] dropout_1368[0][0]
batch_normalization_2506 (Batch	(None,	16,	16,	126)	504	concatenate_2412[0][0]
activation_2506 (Activation)	(None,	16,	16,	126)	0	batch_normalization_2506[0][0]
separable_conv2d_2493 (Separabl	(None,	16,	16,	14)	7938	activation_2506[0][0]
dropout_1369 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2493[0][0]
concatenate_2413 (Concatenate)	(None,	16,	16,	140)	0	concatenate_2412[0][0] dropout_1369[0][0]
batch_normalization_2507 (Batch	(None,	16,	16,	140)	560	concatenate_2413[0][0]
activation_2507 (Activation)	(None,	16,	16,	140)	0	batch_normalization_2507[0][0]
separable_conv2d_2494 (Separabl	(None,	16,	16,	14)	8820	activation_2507[0][0]
dropout_1370 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2494[0][0]
concatenate_2414 (Concatenate)	(None,	16,	16,	154)	0	concatenate_2413[0][0] dropout_1370[0][0]
batch_normalization_2508 (Batch	(None,	16,	16,	154)	616	concatenate_2414[0][0]
activation_2508 (Activation)	(None,	16,	16,	154)	0	batch_normalization_2508[0][0]
separable_conv2d_2495 (Separabl	(None,	16,	16,	14)	9702	activation_2508[0][0]
dropout_1371 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2495[0][0]
concatenate_2415 (Concatenate)	(None,	16,	16,	168)	0	concatenate_2414[0][0] dropout_1371[0][0]
batch_normalization_2509 (Batch	(None,	16,	16,	168)	672	concatenate_2415[0][0]
activation_2509 (Activation)	(None,	16,	16,	168)	0	batch_normalization_2509[0][0]
separable_conv2d_2496 (Separabl	(None,	16,	16,	14)	10584	activation_2509[0][0]
dropout_1372 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2496[0][0]
concatenate_2416 (Concatenate)	(None,	16,	16,	182)	0	concatenate_2415[0][0] dropout_1372[0][0]
batch_normalization_2510 (Batch	(None,	16,	16,	182)	728	concatenate_2416[0][0]
activation_2510 (Activation)	(None,	16,	16,	182)	0	batch_normalization_2510[0][0]
separable_conv2d_2497 (Separabl	(None,	16,	16,	14)	11466	activation_2510[0][0]
dropout_1373 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2497[0][0]
concatenate_2417 (Concatenate)	(None,	16,	16,	196)	0	concatenate_2416[0][0] dropout_1373[0][0]
batch_normalization_2511 (Batch	(None,	16,	16,	196)	784	concatenate_2417[0][0]
activation_2511 (Activation)	(None,	16,	16,	196)	0	batch_normalization_2511[0][0]
separable_conv2d_2498 (Separabl	(None,	16,	16,	14)	12348	activation_2511[0][0]
dropout_1374 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2498[0][0]
concatenate_2418 (Concatenate)	(None,	16,	16,	210)	0	concatenate_2417[0][0] dropout_1374[0][0]
batch_normalization_2512 (Batch	(None,	16,	16,	210)	840	concatenate_2418[0][0]
activation_2512 (Activation)	(None,	16,	16,	210)	0	batch_normalization_2512[0][0]
separable_conv2d_2499 (Separabl	(None,	16,	16,	14)	13230	activation_2512[0][0]
dranaut 1275 (Dranaut)	/M^^^	1 6	1 4	1/11	^	00000010 000000 0140000101

arobour_13/3 (nrobour)	(None,	⊥0,	⊥♡,	14 <i>)</i>	U	sebarante_comvzd_z433[n][n]
concatenate_2419 (Concatenate)	(None,	16,	16,	224)	0	concatenate_2418[0][0] dropout_1375[0][0]
batch_normalization_2513 (Batch	(None,	16,	16,	224)	896	concatenate_2419[0][0]
activation_2513 (Activation)	(None,	16,	16,	224)	0	batch_normalization_2513[0][0]
separable_conv2d_2500 (Separabl	(None,	16,	16,	14)	14112	activation_2513[0][0]
dropout_1376 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2500[0][0]
concatenate_2420 (Concatenate)	(None,	16,	16,	238)	0	concatenate_2419[0][0] dropout_1376[0][0]
batch_normalization_2514 (Batch	(None,	16,	16,	238)	952	concatenate_2420[0][0]
activation_2514 (Activation)	(None,	16,	16,	238)	0	batch_normalization_2514[0][0]
separable_conv2d_2501 (Separabl	(None,	16,	16,	14)	14994	activation_2514[0][0]
dropout_1377 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2501[0][0]
concatenate_2421 (Concatenate)	(None,	16,	16,	252)	0	concatenate_2420[0][0] dropout_1377[0][0]
batch_normalization_2515 (Batch	(None,	16,	16,	252)	1008	concatenate_2421[0][0]
activation_2515 (Activation)	(None,	16,	16,	252)	0	batch_normalization_2515[0][0]
separable_conv2d_2502 (Separabl	(None,	16,	16,	14)	15876	activation_2515[0][0]
dropout_1378 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2502[0][0]
concatenate_2422 (Concatenate)	(None,	16,	16,	266)	0	concatenate_2421[0][0] dropout_1378[0][0]
batch_normalization_2516 (Batch	(None,	16,	16,	266)	1064	concatenate_2422[0][0]
activation_2516 (Activation)	(None,	16,	16,	266)	0	batch_normalization_2516[0][0]
separable_conv2d_2503 (Separabl	(None,	16,	16,	14)	16758	activation_2516[0][0]
dropout_1379 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2503[0][0]
concatenate_2423 (Concatenate)	(None,	16,	16,	280)	0	concatenate_2422[0][0] dropout_1379[0][0]
batch_normalization_2517 (Batch	(None,	16,	16,	280)	1120	concatenate_2423[0][0]
activation_2517 (Activation)	(None,	16,	16,	280)	0	batch_normalization_2517[0][0]
separable_conv2d_2504 (Separabl	(None,	16,	16,	14)	17640	activation_2517[0][0]
dropout_1380 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2504[0][0]
concatenate_2424 (Concatenate)	(None,	16,	16,	294)	0	concatenate_2423[0][0] dropout_1380[0][0]
batch_normalization_2518 (Batch	(None,	16,	16,	294)	1176	concatenate_2424[0][0]
activation_2518 (Activation)	(None,	16,	16,	294)	0	batch_normalization_2518[0][0]
separable_conv2d_2505 (Separabl	(None,	16,	16,	14)	18522	activation_2518[0][0]
dropout_1381 (Dropout)	(None,	16,	16,	14)	0	separable_conv2d_2505[0][0]
concatenate_2425 (Concatenate)	(None,	16,	16,	308)	0	concatenate_2424[0][0] dropout_1381[0][0]
batch_normalization_2519 (Batch	(None,	16,	16,	308)	1232	concatenate_2425[0][0]
activation_2519 (Activation)	(None,	16,	16,	308)	0	batch_normalization_2519[0][0]
separable_conv2d_2506 (Separabl	(None,	16,	16,	14)	4620	activation_2519[0][0]
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aropout_1382 (Dropout)	(None,	⊥ ,	Τ 6	o, 14)	U	separable_convzq_z5U0[U][U]
average_pooling2d_93 (AveragePo	(None,	8,	8,	14)	0	dropout_1382[0][0]
batch_normalization_2520 (Batch	(None,	8,	8,	14)	56	average_pooling2d_93[0][0]
activation_2520 (Activation)	(None,	8,	8,	14)	0	batch_normalization_2520[0][0]
separable_conv2d_2507 (Separabl	(None,	8,	8,	14)	882	activation_2520[0][0]
dropout_1383 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2507[0][0]
concatenate_2426 (Concatenate)	(None,	8,	8,	28)	0	average_pooling2d_93[0][0] dropout_1383[0][0]
oatch_normalization_2521 (Batch	(None,	8,	8,	28)	112	concatenate_2426[0][0]
activation_2521 (Activation)	(None,	8,	8,	28)	0	batch_normalization_2521[0][0]
separable_conv2d_2508 (Separabl	(None,	8,	8,	14)	1764	activation_2521[0][0]
dropout_1384 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2508[0][0]
concatenate_2427 (Concatenate)	(None,	8,	8,	42)	0	concatenate_2426[0][0] dropout_1384[0][0]
oatch_normalization_2522 (Batch	(None,	8,	8,	42)	168	concatenate_2427[0][0]
activation_2522 (Activation)	(None,	8,	8,	42)	0	batch_normalization_2522[0][0]
separable_conv2d_2509 (Separabl	(None,	8,	8,	14)	2646	activation_2522[0][0]
dropout_1385 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2509[0][0]
concatenate_2428 (Concatenate)	(None,	8,	8,	56)	0	concatenate_2427[0][0] dropout_1385[0][0]
patch_normalization_2523 (Batch	(None,	8,	8,	56)	224	concatenate_2428[0][0]
activation_2523 (Activation)	(None,	8,	8,	56)	0	batch_normalization_2523[0][0]
separable_conv2d_2510 (Separabl	(None,	8,	8,	14)	3528	activation_2523[0][0]
dropout_1386 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2510[0][0]
concatenate_2429 (Concatenate)	(None,	8,	8,	70)	0	concatenate_2428[0][0] dropout_1386[0][0]
oatch_normalization_2524 (Batch	(None,	8,	8,	70)	280	concatenate_2429[0][0]
activation_2524 (Activation)	(None,	8,	8,	70)	0	batch_normalization_2524[0][0]
separable_conv2d_2511 (Separabl	(None,	8,	8,	14)	4410	activation_2524[0][0]
dropout_1387 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2511[0][0]
concatenate_2430 (Concatenate)	(None,	8,	8,	84)	0	concatenate_2429[0][0] dropout_1387[0][0]
oatch_normalization_2525 (Batch	(None,	8,	8,	84)	336	concatenate_2430[0][0]
activation_2525 (Activation)	(None,	8,	8,	84)	0	batch_normalization_2525[0][0]
separable_conv2d_2512 (Separabl	(None,	8,	8,	14)	5292	activation_2525[0][0]
dropout_1388 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2512[0][0]
concatenate_2431 (Concatenate)	(None,	8,	8,	98)	0	concatenate_2430[0][0] dropout_1388[0][0]
patch_normalization_2526 (Batch	(None,	8,	8,	98)	392	concatenate_2431[0][0]
activation_2526 (Activation)	(None,	8,	8,	98)	0	batch_normalization_2526[0][0]
separable_conv2d_2513 (Separabl	(None,	8,	8,	14)	6174	activation_2526[0][0]
dropout_1389 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2513[0][0]

concatenate_2432 (Concatenate)	(None,	8,	8,	112)	0	concatenate_2431[0][0] dropout_1389[0][0]
batch_normalization_2527 (Batch	(None,	8,	8,	112)	448	concatenate_2432[0][0]
activation_2527 (Activation)	(None,	8,	8,	112)	0	batch_normalization_2527[0][0]
separable_conv2d_2514 (Separabl	(None,	8,	8,	14)	7056	activation_2527[0][0]
dropout_1390 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2514[0][0]
concatenate_2433 (Concatenate)	(None,	8,	8,	126)	0	concatenate_2432[0][0]
	/27			10.6)		dropout_1390[0][0]
batch_normalization_2528 (Batch					504	concatenate_2433[0][0]
activation_2528 (Activation)	(None,				0	batch_normalization_2528[0][0]
separable_conv2d_2515 (Separabl	(None,	8,	8,	14)	7938	activation_2528[0][0]
dropout_1391 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2515[0][0]
concatenate_2434 (Concatenate)	(None,	8,	8,	140)	0	concatenate_2433[0][0] dropout_1391[0][0]
batch_normalization_2529 (Batch	(None,	8,	8,	140)	560	concatenate_2434[0][0]
activation_2529 (Activation)	(None,	8,	8,	140)	0	batch_normalization_2529[0][0]
separable_conv2d_2516 (Separabl	(None,	8,	8,	14)	8820	activation_2529[0][0]
dropout_1392 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2516[0][0]
concatenate_2435 (Concatenate)	(None,	8,	8,	154)	0	concatenate_2434[0][0] dropout_1392[0][0]
batch_normalization_2530 (Batch	(None,	8,	8,	154)	616	concatenate_2435[0][0]
activation_2530 (Activation)	(None,	8,	8,	154)	0	batch_normalization_2530[0][0]
separable_conv2d_2517 (Separabl	(None,	8,	8,	14)	9702	activation_2530[0][0]
dropout_1393 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2517[0][0]
concatenate_2436 (Concatenate)	(None,	8,	8,	168)	0	concatenate_2435[0][0] dropout_1393[0][0]
batch_normalization_2531 (Batch	(None,	8,	8,	168)	672	concatenate_2436[0][0]
activation_2531 (Activation)	(None,	8,	8,	168)	0	batch_normalization_2531[0][0]
separable_conv2d_2518 (Separabl	(None,	8,	8,	14)	10584	activation_2531[0][0]
dropout_1394 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2518[0][0]
concatenate_2437 (Concatenate)	(None,	8,	8,	182)	0	concatenate_2436[0][0] dropout_1394[0][0]
batch_normalization_2532 (Batch	(None,	8,	8,	182)	728	concatenate_2437[0][0]
activation_2532 (Activation)	(None,	8,	8,	182)	0	batch_normalization_2532[0][0]
separable_conv2d_2519 (Separabl	(None,	8,	8,	14)	11466	activation_2532[0][0]
dropout_1395 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2519[0][0]
concatenate_2438 (Concatenate)	(None,	8,	8,	196)	0	concatenate_2437[0][0] dropout_1395[0][0]
batch_normalization_2533 (Batch	(None,	8,	8,	196)	784	concatenate_2438[0][0]
activation_2533 (Activation)	(None,	8,	8,	196)	0	batch_normalization_2533[0][0]
separable_conv2d_2520 (Separabl	(None,	8,	8,	14)	12348	activation_2533[0][0]
dropout_1396 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2520[0][0]

concatenate_2439 (Concatenate)						
	(None,	8,	8,	210)	0	concatenate_2438[0][0] dropout_1396[0][0]
batch_normalization_2534 (Batch	(None,	8,	8,	210)	840	concatenate_2439[0][0]
activation_2534 (Activation)	(None,	8,	8,	210)	0	batch_normalization_2534[0][0]
separable_conv2d_2521 (Separabl	(None,	8,	8,	14)	13230	activation_2534[0][0]
dropout_1397 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2521[0][0]
concatenate_2440 (Concatenate)	(None,	8,	8,	224)	0	concatenate_2439[0][0] dropout_1397[0][0]
batch_normalization_2535 (Batch	(None,	8,	8,	224)	896	concatenate_2440[0][0]
activation_2535 (Activation)	(None,	8,	8,	224)	0	batch_normalization_2535[0][0]
separable_conv2d_2522 (Separabl	(None,	8,	8,	14)	14112	activation_2535[0][0]
dropout_1398 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2522[0][0]
concatenate_2441 (Concatenate)	(None,	8,	8,	238)	0	concatenate_2440[0][0] dropout_1398[0][0]
batch_normalization_2536 (Batch	(None,	8,	8,	238)	952	concatenate_2441[0][0]
activation_2536 (Activation)	(None,	8,	8,	238)	0	batch_normalization_2536[0][0]
separable_conv2d_2523 (Separabl	(None,	8,	8,	14)	14994	activation_2536[0][0]
dropout_1399 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2523[0][0]
concatenate_2442 (Concatenate)	(None,	8,	8,	252)	0	concatenate_2441[0][0] dropout_1399[0][0]
batch_normalization_2537 (Batch	(None,	8,	8,	252)	1008	concatenate_2442[0][0]
activation_2537 (Activation)	(None,	8,	8,	252)	0	batch_normalization_2537[0][0]
separable_conv2d_2524 (Separabl	(None,	8,	8,	14)	15876	activation_2537[0][0]
dropout_1400 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2524[0][0]
concatenate_2443 (Concatenate)	(None,	8,	8,	266)	0	concatenate_2442[0][0] dropout_1400[0][0]
batch_normalization_2538 (Batch	(None,	8,	8,	266)	1064	concatenate_2443[0][0]
activation_2538 (Activation)	(None,	8,	8,	266)	0	batch_normalization_2538[0][0]
separable_conv2d_2525 (Separabl	(None,	8,	8,	14)	16758	activation_2538[0][0]
dropout_1401 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2525[0][0]
concatenate_2444 (Concatenate)	(None,	8,	8,	280)	0	concatenate_2443[0][0] dropout_1401[0][0]
batch_normalization_2539 (Batch	(None,	8,	8,	280)	1120	concatenate_2444[0][0]
activation_2539 (Activation)	(None,	8,	8,	280)	0	batch_normalization_2539[0][0]
separable_conv2d_2526 (Separabl	(None,	8,	8,	14)	17640	activation_2539[0][0]
dropout_1402 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2526[0][0]
concatenate_2445 (Concatenate)	(None,	8,	8,	294)	0	concatenate_2444[0][0] dropout_1402[0][0]
		Ω	8,	294)	1176	concatenate 2445[0][0]
batch_normalization_2540 (Batch	(None,	٠,	•			
batch_normalization_2540 (Batch_activation_2540 (Activation)	(None,			294)	0	batch_normalization_2540[0][0]
	(None,	8,	8,		0 18522	

concatenate_2446 (Concatenate)	(None,	8,	8,	308)	0	concatenate_2445[0][0] dropout_1403[0][0]
batch_normalization_2541 (Batch	(None,	8,	8,	308)	1232	concatenate_2446[0][0]
activation_2541 (Activation)	(None,	8,	8,	308)	0	batch_normalization_2541[0][0]
separable_conv2d_2528 (Separabl	(None,	8,	8,	14)	4620	activation_2541[0][0]
dropout_1404 (Dropout)	(None,	8,	8,	14)	0	separable_conv2d_2528[0][0]
average_pooling2d_94 (AveragePo	(None,	4,	4,	14)	0	dropout_1404[0][0]
batch_normalization_2542 (Batch	(None,	4,	4,	14)	56	average_pooling2d_94[0][0]
activation_2542 (Activation)	(None,	4,	4,	14)	0	batch_normalization_2542[0][0]
separable_conv2d_2529 (Separabl	(None,	4,	4,	14)	882	activation_2542[0][0]
dropout_1405 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2529[0][0]
concatenate_2447 (Concatenate)	(None,	4,	4,	28)	0	average_pooling2d_94[0][0] dropout_1405[0][0]
batch_normalization_2543 (Batch	(None,	4,	4,	28)	112	concatenate_2447[0][0]
activation_2543 (Activation)	(None,	4,	4,	28)	0	batch_normalization_2543[0][0]
separable_conv2d_2530 (Separabl	(None,	4,	4,	14)	1764	activation_2543[0][0]
dropout_1406 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2530[0][0]
concatenate_2448 (Concatenate)	(None,	4,	4,	42)	0	concatenate_2447[0][0] dropout_1406[0][0]
batch_normalization_2544 (Batch	(None,	4,	4,	42)	168	concatenate_2448[0][0]
activation_2544 (Activation)	(None,	4,	4,	42)	0	batch_normalization_2544[0][0]
separable_conv2d_2531 (Separabl	(None,	4,	4,	14)	2646	activation_2544[0][0]
dropout_1407 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2531[0][0]
concatenate_2449 (Concatenate)	(None,	4,	4,	56)	0	concatenate_2448[0][0] dropout_1407[0][0]
batch_normalization_2545 (Batch	(None,	4,	4,	56)	224	concatenate_2449[0][0]
activation_2545 (Activation)	(None,	4,	4,	56)	0	batch_normalization_2545[0][0]
separable_conv2d_2532 (Separabl	(None,	4,	4,	14)	3528	activation_2545[0][0]
dropout_1408 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2532[0][0]
concatenate_2450 (Concatenate)	(None,	4,	4,	70)	0	concatenate_2449[0][0] dropout_1408[0][0]
batch_normalization_2546 (Batch	(None,	4,	4,	70)	280	concatenate_2450[0][0]
activation_2546 (Activation)	(None,	4,	4,	70)	0	batch_normalization_2546[0][0]
separable_conv2d_2533 (Separabl	(None,	4,	4,	14)	4410	activation_2546[0][0]
dropout_1409 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2533[0][0]
concatenate_2451 (Concatenate)	(None,	4,	4,	84)	0	concatenate_2450[0][0] dropout_1409[0][0]
batch_normalization_2547 (Batch	(None,	4,	4,	84)	336	concatenate_2451[0][0]
activation_2547 (Activation)	(None,	4,	4,	84)	0	batch_normalization_2547[0][0]
separable_conv2d_2534 (Separabl	(None,	4,	4,	14)	5292	activation_2547[0][0]
dropout_1410 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2534[0][0]

concatenate_2452 (Concatenate)	(None, 4	ļ, ·	4,	98)	0	concatenate_2451[0][0] dropout_1410[0][0]
batch_normalization_2548 (Batch	(None, 4	,	4,	98)	392	concatenate_2452[0][0]
activation_2548 (Activation)	(None, 4	,	4,	98)	0	batch_normalization_2548[0][0]
separable_conv2d_2535 (Separabl	(None, 4	,	4,	14)	6174	activation_2548[0][0]
dropout_1411 (Dropout)	(None, 4	,	4,	14)	0	separable_conv2d_2535[0][0]
concatenate_2453 (Concatenate)	(None, 4	,	4,	112)	0	concatenate_2452[0][0] dropout_1411[0][0]
batch_normalization_2549 (Batch	(None, 4	,	4,	112)	448	concatenate_2453[0][0]
activation_2549 (Activation)	(None, 4	,	4,	112)	0	batch_normalization_2549[0][0]
separable_conv2d_2536 (Separabl	(None, 4	,	4,	14)	7056	activation_2549[0][0]
dropout_1412 (Dropout)	(None, 4	,	4,	14)	0	separable_conv2d_2536[0][0]
concatenate_2454 (Concatenate)	(None, 4	, ,	4,	126)	0	concatenate_2453[0][0] dropout_1412[0][0]
batch_normalization_2550 (Batch	(None, 4	, ,	4,	126)	504	concatenate_2454[0][0]
activation_2550 (Activation)	(None, 4	, ,	4,	126)	0	batch_normalization_2550[0][0]
separable_conv2d_2537 (Separabl	(None, 4	,	4,	14)	7938	activation_2550[0][0]
dropout_1413 (Dropout)	(None, 4	,	4,	14)	0	separable_conv2d_2537[0][0]
concatenate_2455 (Concatenate)	(None, 4	,	4,	140)	0	concatenate_2454[0][0] dropout_1413[0][0]
batch_normalization_2551 (Batch	(None, 4	,	4,	140)	560	concatenate_2455[0][0]
activation_2551 (Activation)	(None, 4	, ,	4,	140)	0	batch_normalization_2551[0][0]
separable_conv2d_2538 (Separabl	(None, 4	,	4,	14)	8820	activation_2551[0][0]
dropout_1414 (Dropout)	(None, 4	,	4,	14)	0	separable_conv2d_2538[0][0]
concatenate_2456 (Concatenate)	(None, 4	, ,	4,	154)	0	concatenate_2455[0][0] dropout_1414[0][0]
batch_normalization_2552 (Batch	(None, 4	,	4,	154)	616	concatenate_2456[0][0]
activation_2552 (Activation)	(None, 4	, ,	4,	154)	0	batch_normalization_2552[0][0]
separable_conv2d_2539 (Separabl	(None, 4	,	4,	14)	9702	activation_2552[0][0]
dropout_1415 (Dropout)	(None, 4	,	4,	14)	0	separable_conv2d_2539[0][0]
concatenate_2457 (Concatenate)	(None, 4	, ,	4,	168)	0	concatenate_2456[0][0] dropout_1415[0][0]
batch_normalization_2553 (Batch	(None, 4	,	4,	168)	672	concatenate_2457[0][0]
activation_2553 (Activation)	(None, 4	,	4,	168)	0	batch_normalization_2553[0][0]
separable_conv2d_2540 (Separabl	(None, 4	, ,	4,	14)	10584	activation_2553[0][0]
dropout_1416 (Dropout)	(None, 4	, ,	4,	14)	0	separable_conv2d_2540[0][0]
concatenate_2458 (Concatenate)	(None, 4	l,	4,	182)	0	concatenate_2457[0][0] dropout_1416[0][0]
batch_normalization_2554 (Batch	(None, 4	,	4,	182)	728	concatenate_2458[0][0]
activation_2554 (Activation)	(None, 4	,	4,	182)	0	batch_normalization_2554[0][0]
separable_conv2d_2541 (Separabl	(None, 4	,	4,	14)	11466	activation_2554[0][0]
dropout_1417 (Dropout)	(None, 4	,	4,	14)	0	separable_conv2d_2541[0][0]

concatenate_2459 (Concatenate)	(None,	4,	4,	196)	0	concatenate_2458[0][0] dropout_1417[0][0]
batch_normalization_2555 (Batch	(None,	4,	4,	196)	784	concatenate_2459[0][0]
activation_2555 (Activation)	(None,	4,	4,	196)	0	batch_normalization_2555[0][0]
separable_conv2d_2542 (Separabl	(None,	4,	4,	14)	12348	activation_2555[0][0]
dropout_1418 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2542[0][0]
concatenate_2460 (Concatenate)	(None,	4,	4,	210)	0	concatenate_2459[0][0] dropout_1418[0][0]
batch_normalization_2556 (Batch	(None,	4,	4,	210)	840	concatenate_2460[0][0]
activation_2556 (Activation)	(None,	4,	4,	210)	0	batch_normalization_2556[0][0]
separable_conv2d_2543 (Separabl	(None,	4,	4,	14)	13230	activation_2556[0][0]
dropout_1419 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2543[0][0]
concatenate_2461 (Concatenate)	(None,	4,	4,	224)	0	concatenate_2460[0][0] dropout_1419[0][0]
batch_normalization_2557 (Batch	(None,	4,	4,	224)	896	concatenate_2461[0][0]
activation_2557 (Activation)	(None,	4,	4,	224)	0	batch_normalization_2557[0][0]
separable_conv2d_2544 (Separabl	(None,	4,	4,	14)	14112	activation_2557[0][0]
dropout_1420 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2544[0][0]
concatenate_2462 (Concatenate)	(None,	4,	4,	238)	0	concatenate_2461[0][0] dropout_1420[0][0]
batch_normalization_2558 (Batch	(None,	4,	4,	238)	952	concatenate_2462[0][0]
activation_2558 (Activation)	(None,	4,	4,	238)	0	batch_normalization_2558[0][0]
separable_conv2d_2545 (Separabl	(None,	4,	4,	14)	14994	activation_2558[0][0]
dropout_1421 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2545[0][0]
concatenate_2463 (Concatenate)	(None,	4,	4,	252)	0	concatenate_2462[0][0] dropout_1421[0][0]
batch_normalization_2559 (Batch	(None,	4,	4,	252)	1008	concatenate_2463[0][0]
activation_2559 (Activation)	(None,	4,	4,	252)	0	batch_normalization_2559[0][0]
separable_conv2d_2546 (Separabl	(None,	4,	4,	14)	15876	activation_2559[0][0]
dropout_1422 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2546[0][0]
concatenate_2464 (Concatenate)	(None,	4,	4,	266)	0	concatenate_2463[0][0] dropout_1422[0][0]
batch_normalization_2560 (Batch	(None,	4,	4,	266)	1064	concatenate_2464[0][0]
activation_2560 (Activation)	(None,	4,	4,	266)	0	batch_normalization_2560[0][0]
separable_conv2d_2547 (Separabl	(None,	4,	4,	14)	16758	activation_2560[0][0]
dropout_1423 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2547[0][0]
concatenate_2465 (Concatenate)	(None,	4,	4,	280)	0	concatenate_2464[0][0] dropout_1423[0][0]
batch_normalization_2561 (Batch	(None,	4,	4,	280)	1120	concatenate_2465[0][0]
activation_2561 (Activation)	(None,	4,	4,	280)	0	batch_normalization_2561[0][0]
separable_conv2d_2548 (Separabl	(None,	4,	4,	14)	17640	activation_2561[0][0]
dropout_1424 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2548[0][0]

concatenate_2466 (Concatenate)	(None,	4,	4,	294)	0	concatenate_2465[0][0] dropout_1424[0][0]
batch_normalization_2562 (Batch	(None,	4,	4,	294)	1176	concatenate_2466[0][0]
activation_2562 (Activation)	(None,	4,	4,	294)	0	batch_normalization_2562[0][0]
separable_conv2d_2549 (Separabl	(None,	4,	4,	14)	18522	activation_2562[0][0]
dropout_1425 (Dropout)	(None,	4,	4,	14)	0	separable_conv2d_2549[0][0]
concatenate_2467 (Concatenate)	(None,	4,	4,	308)	0	concatenate_2466[0][0] dropout_1425[0][0]
batch_normalization_2563 (Batch	(None,	4,	4,	308)	1232	concatenate_2467[0][0]
activation_2563 (Activation)	(None,	4,	4,	308)	0	batch_normalization_2563[0][0]
average_pooling2d_95 (AveragePo	(None,	2,	2,	308)	0	activation_2563[0][0]
separable_conv2d_2550 (Separabl	(None,	1,	1,	10)	4322	average_pooling2d_95[0][0]
flatten_15 (Flatten)	(None,	10))		0	separable_conv2d_2550[0][0]
dense_17 (Dense)	(None,	10)			110	flatten_15[0][0]

Total params: 945,472 Trainable params: 915,640 Non-trainable params: 29,832

In [69]:

```
model 5.compile(optimizer='adam', loss='categorical crossentropy', metrics=['accuracy'])
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
patience = 50
base_path = '/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/'
checkpoint_file_name = base_path + 'CIFAR_model5' + '_{epoch:02d}-{val_accuracy:.2f}.hdf5'
model checkpoint = ModelCheckpoint (checkpoint file name, monitor='val accuracy', verbose=1, save be
st_only=True)
early_stop = EarlyStopping('val_accuracy', mode='max', patience = patience)
reduce LR = ReduceLROnPlateau (monitor='val accuracy', mode='max', factor=0.1, patience=int(patience
/3), verbose=1)
callbacks = [model checkpoint, early stop, reduce LR]
epochs = 300
batch_size = 128
#https://keras.io/api/preprocessing/image/#flow-method
history 5 = model_5.fit(data_generator.flow(X_train, y_train, batch_size),
                    steps_per_epoch = int(len(X_train)/batch_size),
                    epochs = epochs,
                    callbacks = callbacks,
                    validation_data = (X_test, y_test), verbose=1)
```

```
WARNING:tensorflow:sample weight modes were coerced from
 . . .
  to
 ['...']
Train for 390 steps, validate on 10000 samples
Epoch 1/300
Epoch 00001: val accuracy improved from -inf to 0.22800, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model5_01-0.23.hdf5
_loss: 2.3646 - val_accuracy: 0.2280
Epoch 2/300
Epoch 00002: val accuracy improved from 0.22800 to 0.30130, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_model5\_02-0.30.hdf5
390/390 [============ ] - 136s 349ms/step - loss: 1.3671 - accuracy: 0.4989 - val
_loss: 3.2475 - val_accuracy: 0.3013
```

```
Epoch 3/300
Epoch 00003: val accuracy improved from 0.30130 to 0.41590, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR model5 03-0.42.hdf5
_loss: 2.5937 - val_accuracy: 0.4159
Epoch 4/300
Epoch 00004: val accuracy improved from 0.41590 to 0.57930, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model5_04-0.58.hdf5
loss: 1.4217 - val accuracy: 0.5793
Epoch 5/300
Epoch 00005: val accuracy improved from 0.57930 to 0.61120, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 05-0.61.hdf5
390/390 [============= ] - 136s 349ms/step - loss: 0.9160 - accuracy: 0.6739 - val
loss: 1.3084 - val accuracy: 0.6112
Epoch 6/300
Epoch 00006: val_accuracy improved from 0.61120 to 0.62680, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model5_06-0.63.hdf5
390/390 [============ ] - 136s 349ms/step - loss: 0.8343 - accuracy: 0.7069 - val
loss: 1.3066 - val accuracy: 0.6268
Epoch 7/300
Epoch 00007: val accuracy did not improve from 0.62680
390/390 [============ ] - 135s 347ms/step - loss: 0.7776 - accuracy: 0.7296 - val
loss: 2.7267 - val_accuracy: 0.4788
Epoch 8/300
Epoch 00008: val_accuracy improved from 0.62680 to 0.65880, saving model to
/home/ubuntu/Project/my_data/CNN CIFAR/checkpoint/CIFAR model5 08-0.66.hdf5
390/390 [============ ] - 136s 350ms/step - loss: 0.7254 - accuracy: 0.7452 - val
_loss: 1.2443 - val_accuracy: 0.6588
Epoch 9/300
Epoch 00009: val accuracy improved from 0.65880 to 0.71540, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 09-0.72.hdf5
390/390 [============== ] - 136s 349ms/step - loss: 0.6860 - accuracy: 0.7621 - val
loss: 0.9944 - val accuracy: 0.7154
Epoch 10/300
Epoch 00010: val accuracy did not improve from 0.71540
390/390 [============ ] - 135s 346ms/step - loss: 0.6442 - accuracy: 0.7772 - val
loss: 2.1131 - val accuracy: 0.5154
Epoch 11/300
Epoch 00011: val accuracy improved from 0.71540 to 0.74050, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 11-0.74.hdf5
390/390 [============ ] - 136s 349ms/step - loss: 0.6253 - accuracy: 0.7823 - val
loss: 0.8934 - val accuracy: 0.7405
Epoch 12/300
Epoch 00012: val accuracy improved from 0.74050 to 0.74180, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model5_12-0.74.hdf5
390/390 [=========== ] - 136s 349ms/step - loss: 0.5945 - accuracy: 0.7925 - val
loss: 0.8817 - val_accuracy: 0.7418
Epoch 13/300
Epoch 00013: val accuracy did not improve from 0.74180
loss: 1.1154 - val accuracy: 0.7084
Epoch 14/300
Epoch 00014: val accuracy improved from 0.74180 to 0.77990, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_model5\_14-0.78.hdf5
390/390 [=========== ] - 136s 349ms/step - loss: 0.5532 - accuracy: 0.8072 - val
loss: 0.7306 - val accuracy: 0.7799
Epoch 15/300
Epoch 00015: val accuracy did not improve from 0.77990
390/390 [=============== ] - 135s 346ms/step - loss: 0.5359 - accuracy: 0.8134 - val
loss: 0.8631 - val accuracy: 0.7554
Epoch 16/300
Epoch 00016: val accuracy did not improve from 0.77990
```

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loss: 0.8329 - val accuracy: 0.7669
Epoch 17/300
Epoch 00017: val_accuracy did not improve from 0.77990
390/390 [============= ] - 135s 347ms/step - loss: 0.5050 - accuracy: 0.8245 - val
loss: 0.8937 - val_accuracy: 0.7489
Epoch 18/300
Epoch 00018: val_accuracy did not improve from 0.77990
390/390 [============= ] - 135s 346ms/step - loss: 0.4845 - accuracy: 0.8321 - val
loss: 0.7822 - val accuracy: 0.7752
Epoch 19/300
Epoch 00019: val accuracy did not improve from 0.77990
390/390 [============= ] - 135s 347ms/step - loss: 0.4740 - accuracy: 0.8350 - val
loss: 0.9793 - val accuracy: 0.7411
Epoch 20/300
Epoch 00020: val accuracy did not improve from 0.77990
390/390 [============ ] - 135s 347ms/step - loss: 0.4564 - accuracy: 0.8387 - val
loss: 1.3204 - val_accuracy: 0.6897
Epoch 21/300
Epoch 00021: val accuracy improved from 0.77990 to 0.78290, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 21-0.78.hdf5
390/390 [============ ] - 136s 350ms/step - loss: 0.4460 - accuracy: 0.8455 - val
_loss: 0.7568 - val_accuracy: 0.7829
Epoch 22/300
Epoch 00022: val accuracy did not improve from 0.78290
390/390 [=============== ] - 135s 347ms/step - loss: 0.4321 - accuracy: 0.8501 - val
loss: 0.9390 - val accuracy: 0.7539
Epoch 23/300
Epoch 00023: val accuracy improved from 0.78290 to 0.78860, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 23-0.79.hdf5
390/390 [============ ] - 136s 349ms/step - loss: 0.4254 - accuracy: 0.8521 - val
_loss: 0.7675 - val_accuracy: 0.7886
Epoch 24/300
Epoch 00024: val accuracy improved from 0.78860 to 0.79390, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model5_24-0.79.hdf5
390/390 [============ ] - 136s 349ms/step - loss: 0.4142 - accuracy: 0.8548 - val
loss: 0.7512 - val accuracy: 0.7939
Epoch 25/300
Epoch 00025: val accuracy improved from 0.79390 to 0.81740, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 25-0.82.hdf5
_loss: 0.6412 - val_accuracy: 0.8174
Epoch 26/300
Epoch 00026: val accuracy did not improve from 0.81740
390/390 [============ ] - 135s 346ms/step - loss: 0.3988 - accuracy: 0.8614 - val
loss: 1.0172 - val accuracy: 0.7459
Epoch 27/300
Epoch 00027: val accuracy did not improve from 0.81740
390/390 [============ ] - 135s 347ms/step - loss: 0.3837 - accuracy: 0.8658 - val
loss: 1.0873 - val accuracy: 0.7225
Epoch 28/300
Epoch 00028: val accuracy improved from 0.81740 to 0.81910, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 28-0.82.hdf5
loss: 0.6226 - val accuracy: 0.8191
Epoch 29/300
Epoch 00029: val accuracy did not improve from 0.81910
390/390 [============ ] - 135s 346ms/step - loss: 0.3719 - accuracy: 0.8699 - val
loss: 0.8904 - val accuracy: 0.7752
Epoch 30/300
Epoch 00030: val accuracy did not improve from 0.81910
390/390 [============ ] - 135s 347ms/step - loss: 0.3669 - accuracy: 0.8716 - val
loss: 0.7723 - val accuracy: 0.7989
```

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Epoch 31/300
Epoch 00031: val accuracy did not improve from 0.81910
loss: 0.6521 - val accuracy: 0.8153
Epoch 32/300
Epoch 00032: val accuracy did not improve from 0.81910
390/390 [============ ] - 135s 347ms/step - loss: 0.3500 - accuracy: 0.8779 - val
loss: 0.6707 - val accuracy: 0.8131
Epoch 33/300
Epoch 00033: val accuracy did not improve from 0.81910
loss: 0.7479 - val accuracy: 0.7991
Epoch 34/300
Epoch 00034: val accuracy did not improve from 0.81910
390/390 [============ ] - 135s 347ms/step - loss: 0.3367 - accuracy: 0.8807 - val
loss: 0.8278 - val accuracy: 0.7911
Epoch 35/300
Epoch 00035: val accuracy did not improve from 0.81910
390/390 [============= ] - 135s 347ms/step - loss: 0.3384 - accuracy: 0.8821 - val
_loss: 1.0773 - val_accuracy: 0.7542
Epoch 36/300
Epoch 00036: val accuracy did not improve from 0.81910
390/390 [============ ] - 135s 347ms/step - loss: 0.3267 - accuracy: 0.8842 - val
loss: 0.8458 - val accuracy: 0.7917
Epoch 37/300
Epoch 00037: val accuracy did not improve from 0.81910
390/390 [============= ] - 135s 346ms/step - loss: 0.3232 - accuracy: 0.8864 - val
loss: 1.1717 - val accuracy: 0.7484
Epoch 38/300
Epoch 00038: val accuracy did not improve from 0.81910
390/390 [============ ] - 135s 346ms/step - loss: 0.3151 - accuracy: 0.8904 - val
loss: 0.9669 - val accuracy: 0.7775
Epoch 39/300
Epoch 00039: val accuracy did not improve from 0.81910
390/390 [============ ] - 135s 347ms/step - loss: 0.3132 - accuracy: 0.8904 - val
loss: 0.9023 - val accuracy: 0.7860
Epoch 40/300
Epoch 00040: val accuracy improved from 0.81910 to 0.83030, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 40-0.83.hdf5
loss: 0.6351 - val accuracy: 0.8303
Epoch 41/300
Epoch 00041: val accuracy did not improve from 0.83030
390/390 [============ ] - 135s 346ms/step - loss: 0.3008 - accuracy: 0.8933 - val
_loss: 0.8391 - val_accuracy: 0.7941
Epoch 42/300
Epoch 00042: val accuracy improved from 0.83030 to 0.83730, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 42-0.84.hdf5
390/390 [=========== ] - 136s 349ms/step - loss: 0.2983 - accuracy: 0.8949 - val
loss: 0.6109 - val_accuracy: 0.8373
Epoch 43/300
Epoch 00043: val accuracy did not improve from 0.83730
_loss: 0.7076 - val_accuracy: 0.8122
Epoch 44/300
Epoch 00044: val accuracy did not improve from 0.83730
390/390 [============= ] - 135s 347ms/step - loss: 0.2867 - accuracy: 0.8987 - val
_loss: 0.9944 - val_accuracy: 0.7681
Epoch 45/300
Epoch 00045: val accuracy improved from 0.83730 to 0.83780, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 45-0.84.hdf5
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loss: 0.6084 - val accuracy: 0.8378
Epoch 46/300
Epoch 00046: val accuracy improved from 0.83780 to 0.84870, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 46-0.85.hdf5
390/390 [============ ] - 136s 349ms/step - loss: 0.2780 - accuracy: 0.9024 - val
loss: 0.5594 - val accuracy: 0.8487
Epoch 47/300
Epoch 00047: val accuracy did not improve from 0.84870
390/390 [================== ] - 135s 346ms/step - loss: 0.2781 - accuracy: 0.9028 - val
_loss: 0.8282 - val_accuracy: 0.8057
Epoch 48/300
Epoch 00048: val accuracy did not improve from 0.84870
390/390 [============ ] - 135s 346ms/step - loss: 0.2731 - accuracy: 0.9052 - val
loss: 0.6938 - val_accuracy: 0.8179
Epoch 49/300
Epoch 00049: val accuracy did not improve from 0.84870
loss: 0.6449 - val_accuracy: 0.8359
Epoch 50/300
Epoch 00050: val accuracy did not improve from 0.84870
390/390 [============ ] - 135s 346ms/step - loss: 0.2632 - accuracy: 0.9073 - val
loss: 0.8866 - val accuracy: 0.8046
Epoch 51/300
Epoch 00051: val accuracy did not improve from 0.84870
390/390 [============ ] - 135s 346ms/step - loss: 0.2622 - accuracy: 0.9066 - val
loss: 0.8214 - val accuracy: 0.8092
Epoch 52/300
Epoch 00052: val accuracy did not improve from 0.84870
390/390 [============ ] - 135s 347ms/step - loss: 0.2594 - accuracy: 0.9074 - val
loss: 0.8079 - val accuracy: 0.8077
Epoch 53/300
Epoch 00053: val accuracy did not improve from 0.84870
390/390 [=============== ] - 135s 347ms/step - loss: 0.2511 - accuracy: 0.9123 - val
loss: 0.9002 - val_accuracy: 0.7969
Epoch 54/300
Epoch 00054: val\_accuracy did not improve from 0.84870
loss: 0.6637 - val accuracy: 0.8340
Epoch 55/300
Epoch 00055: val accuracy did not improve from 0.84870
390/390 [============= ] - 136s 347ms/step - loss: 0.2464 - accuracy: 0.9133 - val
loss: 0.8131 - val_accuracy: 0.8097
Epoch 56/300
Epoch 00056: val_accuracy did not improve from 0.84870
390/390 [============= ] - 135s 347ms/step - loss: 0.2464 - accuracy: 0.9136 - val
loss: 0.6043 - val accuracy: 0.8413
Epoch 57/300
Epoch 00057: val accuracy did not improve from 0.84870
390/390 [============= ] - 135s 347ms/step - loss: 0.2449 - accuracy: 0.9135 - val
loss: 0.8142 - val accuracy: 0.8151
Epoch 58/300
Epoch 00058: val accuracy did not improve from 0.84870
390/390 [============ ] - 135s 347ms/step - loss: 0.2388 - accuracy: 0.9163 - val
loss: 0.6324 - val_accuracy: 0.8385
Epoch 59/300
Epoch 00059: val accuracy improved from 0.84870 to 0.84920, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 59-0.85.hdf5
390/390 [============ ] - 136s 348ms/step - loss: 0.2324 - accuracy: 0.9178 - val
_loss: 0.5738 - val_accuracy: 0.8492
Epoch 60/300
Epoch 00060: val accuracy did not improve from 0.84920
390/390 [============ ] - 135s 346ms/step - loss: 0.2325 - accuracy: 0.9179 - val
```

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loss: 0.7680 - val accuracy: 0.8197
Epoch 61/300
Epoch 00061: val accuracy did not improve from 0.84920
loss: 0.7218 - val accuracy: 0.8255
Epoch 62/300
Epoch 00062: val accuracy did not improve from 0.84920
loss: 0.7827 - val accuracy: 0.8194
Epoch 63/300
Epoch 00063: val accuracy improved from 0.84920 to 0.86040, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_model5\_63-0.86.hdf5
390/390 [============= ] - 136s 349ms/step - loss: 0.2217 - accuracy: 0.9221 - val
loss: 0.5561 - val accuracy: 0.8604
Epoch 64/300
Epoch 00064: val_accuracy did not improve from 0.86040
loss: 0.7433 - val accuracy: 0.8266
Epoch 65/300
Epoch 00065: val accuracy did not improve from 0.86040
loss: 0.6962 - val accuracy: 0.8417
Epoch 66/300
Epoch 00066: val accuracy did not improve from 0.86040
390/390 [================== ] - 135s 347ms/step - loss: 0.2181 - accuracy: 0.9217 - val
loss: 0.5255 - val accuracy: 0.8593
Epoch 67/300
Epoch 00067: val accuracy did not improve from 0.86040
390/390 [============ ] - 135s 347ms/step - loss: 0.2160 - accuracy: 0.9228 - val
loss: 0.8274 - val accuracy: 0.8146
Epoch 68/300
Epoch 00068: val_accuracy did not improve from 0.86040
390/390 [============ ] - 135s 346ms/step - loss: 0.2127 - accuracy: 0.9251 - val
loss: 0.8970 - val accuracy: 0.7974
Epoch 69/300
Epoch 00069: val accuracy did not improve from 0.86040
390/390 [============ ] - 135s 346ms/step - loss: 0.2094 - accuracy: 0.9258 - val
loss: 0.7490 - val accuracy: 0.8222
Epoch 70/300
Epoch 00070: val accuracy improved from 0.86040 to 0.86540, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR model5 70-0.87.hdf5
_loss: 0.5319 - val_accuracy: 0.8654
Epoch 71/300
Epoch 00071: val_accuracy did not improve from 0.86540
390/390 [============ ] - 135s 346ms/step - loss: 0.2055 - accuracy: 0.9276 - val
loss: 0.7056 - val accuracy: 0.8340
Epoch 72/300
Epoch 00072: val accuracy did not improve from 0.86540
loss: 0.6550 - val accuracy: 0.8428
Epoch 73/300
Epoch 00073: val accuracy did not improve from 0.86540
_loss: 0.7607 - val_accuracy: 0.8217
Epoch 74/300
Epoch 00074: val accuracy did not improve from 0.86540
390/390 [============= ] - 135s 346ms/step - loss: 0.1995 - accuracy: 0.9293 - val
_loss: 0.5658 - val_accuracy: 0.8615
Epoch 75/300
Epoch 00075: val_accuracy improved from 0.86540 to 0.86740, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 75-0.87.hdf5
```

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390/390 [============== ] - 136s 350ms/step - loss: 0.1975 - accuracy: 0.9299 - val
loss: 0.5567 - val accuracy: 0.8674
Epoch 76/300
Epoch 00076: val accuracy did not improve from 0.86740
390/390 [=================== ] - 135s 347ms/step - loss: 0.1906 - accuracy: 0.9325 - val
loss: 0.5900 - val accuracy: 0.8640
Epoch 77/300
Epoch 00077: val_accuracy did not improve from 0.86740
390/390 [=============== ] - 135s 346ms/step - loss: 0.1912 - accuracy: 0.9317 - val
loss: 0.7853 - val accuracy: 0.8285
Epoch 78/300
Epoch 00078: val_accuracy did not improve from 0.86740
loss: 0.8347 - val accuracy: 0.8214
Epoch 79/300
Epoch 00079: val accuracy did not improve from 0.86740
390/390 [============ ] - 135s 347ms/step - loss: 0.1904 - accuracy: 0.9321 - val
loss: 0.6842 - val accuracy: 0.8546
Epoch 80/300
Epoch 00080: val accuracy did not improve from 0.86740
390/390 [============ ] - 135s 346ms/step - loss: 0.1906 - accuracy: 0.9324 - val
_loss: 0.7611 - val_accuracy: 0.8312
Epoch 81/300
Epoch 00081: val accuracy did not improve from 0.86740
390/390 [============ ] - 135s 347ms/step - loss: 0.1863 - accuracy: 0.9337 - val
loss: 0.5795 - val accuracy: 0.8579
Epoch 82/300
Epoch 00082: val accuracy did not improve from 0.86740
loss: 0.6056 - val accuracy: 0.8547
Epoch 83/300
Epoch 00083: val accuracy did not improve from 0.86740
390/390 [============ ] - 135s 346ms/step - loss: 0.1780 - accuracy: 0.9364 - val
loss: 0.7099 - val accuracy: 0.8408
Epoch 84/300
Epoch 00084: val accuracy did not improve from 0.86740
loss: 0.8617 - val accuracy: 0.8185
Epoch 85/300
Epoch 00085: val accuracy did not improve from 0.86740
390/390 [============ ] - 135s 347ms/step - loss: 0.1784 - accuracy: 0.9380 - val
loss: 0.5710 - val accuracy: 0.8667
Epoch 86/300
Epoch 00086: val accuracy did not improve from 0.86740
390/390 [============ ] - 135s 346ms/step - loss: 0.1809 - accuracy: 0.9349 - val
_loss: 0.7086 - val_accuracy: 0.8446
Epoch 87/300
Epoch 00087: val accuracy improved from 0.86740 to 0.87590, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model5 87-0.88.hdf5
loss: 0.5157 - val accuracy: 0.8759
Epoch 88/300
Epoch 00088: val accuracy did not improve from 0.87590
390/390 [============== ] - 135s 347ms/step - loss: 0.1737 - accuracy: 0.9379 - val
loss: 0.6484 - val_accuracy: 0.8519
Epoch 89/300
Epoch 00089: val_accuracy did not improve from 0.87590
390/390 [============ ] - 135s 346ms/step - loss: 0.1695 - accuracy: 0.9401 - val
loss: 0.5067 - val_accuracy: 0.8738
Epoch 90/300
Epoch 00090: val_accuracy did not improve from 0.87590
390/390 [============ ] - 135s 346ms/step - loss: 0.1693 - accuracy: 0.9403 - val
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loss: 0.7201 - val accuracy: 0.8429
Epoch 91/300
Epoch 00091: val accuracy did not improve from 0.87590
390/390 [============ ] - 135s 347ms/step - loss: 0.1712 - accuracy: 0.9394 - val
loss: 0.5750 - val accuracy: 0.8631
Epoch 92/300
Epoch 00092: val accuracy did not improve from 0.87590
390/390 [============ ] - 135s 346ms/step - loss: 0.1711 - accuracy: 0.9390 - val
_loss: 0.6479 - val_accuracy: 0.8540
Epoch 93/300
Epoch 00093: val_accuracy did not improve from 0.87590
390/390 [============ ] - 135s 346ms/step - loss: 0.1674 - accuracy: 0.9398 - val
loss: 0.6256 - val accuracy: 0.8624
Epoch 94/300
Epoch 00094: val accuracy did not improve from 0.87590
loss: 0.6470 - val accuracy: 0.8616
Epoch 95/300
Epoch 00095: val accuracy did not improve from 0.87590
loss: 0.5145 - val accuracy: 0.8746
Epoch 96/300
Epoch 00096: val accuracy did not improve from 0.87590
loss: 0.7455 - val_accuracy: 0.8394
Epoch 97/300
Epoch 00097: val accuracy did not improve from 0.87590
390/390 [============ ] - 135s 347ms/step - loss: 0.1623 - accuracy: 0.9427 - val
loss: 0.5512 - val accuracy: 0.8722
Epoch 98/300
Epoch 00098: val accuracy did not improve from 0.87590
390/390 [============ ] - 135s 346ms/step - loss: 0.1613 - accuracy: 0.9421 - val
_loss: 0.7458 - val_accuracy: 0.8387
Epoch 99/300
Epoch 00099: val_accuracy did not improve from 0.87590
390/390 [============ ] - 135s 346ms/step - loss: 0.1567 - accuracy: 0.9436 - val
loss: 0.7160 - val accuracy: 0.8471
Epoch 100/300
Epoch 00100: val accuracy did not improve from 0.87590
loss: 0.8331 - val accuracy: 0.8279
Epoch 101/300
Epoch 00101: val accuracy did not improve from 0.87590
loss: 0.7311 - val accuracy: 0.8441
Epoch 102/300
Epoch 00102: val accuracy did not improve from 0.87590
390/390 [============ ] - 135s 346ms/step - loss: 0.1512 - accuracy: 0.9472 - val
loss: 0.6692 - val accuracy: 0.8583
Epoch 103/300
Epoch 00103: val_accuracy did not improve from 0.87590
Epoch 00103: ReduceLROnPlateau reducing learning rate to 0.00010000000474974513.
390/390 [============= ] - 135s 347ms/step - loss: 0.1543 - accuracy: 0.9452 - val
loss: 0.6255 - val accuracy: 0.8603
Epoch 104/300
Epoch 00104: val_accuracy improved from 0.87590 to 0.89080, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model5_104-0.89.hdf5
390/390 [============= ] - 136s 349ms/step - loss: 0.1180 - accuracy: 0.9588 - val
loss: 0.4858 - val_accuracy: 0.8908
Epoch 105/300
Epoch 00105: val accuracy did not improve from 0.89080
```

```
390/390 [============ ] - 135s 347ms/step - loss: 0.1008 - accuracy: 0.9637 - val
loss: 0.4950 - val accuracy: 0.8906
Epoch 106/300
Epoch 00106: val accuracy improved from 0.89080 to 0.89100, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_model5\_106-0.89.hdf5
loss: 0.5031 - val accuracy: 0.8910
Epoch 107/300
Epoch 00107: val accuracy did not improve from 0.89100
loss: 0.5264 - val accuracy: 0.8891
Epoch 108/300
Epoch 00108: val accuracy did not improve from 0.89100
loss: 0.5278 - val accuracy: 0.8897
Epoch 109/300
Epoch 00109: val accuracy did not improve from 0.89100
390/390 [============ ] - 135s 347ms/step - loss: 0.0853 - accuracy: 0.9699 - val
loss: 0.5183 - val accuracy: 0.8910
Epoch 110/300
Epoch 00110: val accuracy did not improve from 0.89100
390/390 [============ ] - 135s 347ms/step - loss: 0.0859 - accuracy: 0.9700 - val
_loss: 0.5395 - val_accuracy: 0.8886
Epoch 111/300
Epoch 00111: val accuracy did not improve from 0.89100
390/390 [============ ] - 135s 347ms/step - loss: 0.0829 - accuracy: 0.9702 - val
loss: 0.5458 - val accuracy: 0.8909
Epoch 112/300
Epoch 00112: val accuracy improved from 0.89100 to 0.89150, saving model to
/home/ubuntu/Project/my_data/CNN CIFAR/checkpoint/CIFAR model5 112-0.89.hdf5
390/390 [============ ] - 136s 350ms/step - loss: 0.0806 - accuracy: 0.9720 - val
loss: 0.5410 - val accuracy: 0.8915
Epoch 113/300
Epoch 00113: val accuracy did not improve from 0.89150
390/390 [=============== ] - 135s 347ms/step - loss: 0.0812 - accuracy: 0.9710 - val
loss: 0.5522 - val accuracy: 0.8909
Epoch 114/300
Epoch 00114: val accuracy improved from 0.89150 to 0.89180, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model5_114-0.89.hdf5
390/390 [============ ] - 136s 349ms/step - loss: 0.0790 - accuracy: 0.9719 - val
loss: 0.5481 - val accuracy: 0.8918
Epoch 115/300
Epoch 00115: val accuracy did not improve from 0.89180
loss: 0.5404 - val_accuracy: 0.8915
Epoch 116/300
Epoch 00116: val_accuracy did not improve from 0.89180
390/390 [============ ] - 135s 347ms/step - loss: 0.0780 - accuracy: 0.9724 - val
_loss: 0.5824 - val_accuracy: 0.8865
Epoch 117/300
Epoch 00117: val accuracy did not improve from 0.89180
390/390 [============= ] - 135s 347ms/step - loss: 0.0798 - accuracy: 0.9720 - val
loss: 0.5693 - val accuracy: 0.8873
Epoch 118/300
Epoch 00118: val accuracy did not improve from 0.89180
loss: 0.5688 - val_accuracy: 0.8904
Epoch 119/300
Epoch 00119: val accuracy did not improve from 0.89180
390/390 [============ ] - 135s 347ms/step - loss: 0.0734 - accuracy: 0.9745 - val
loss: 0.5809 - val_accuracy: 0.8889
Epoch 120/300
```

```
Epoch 00120: val_accuracy did not improve from 0.89180
loss: 0.5653 - val accuracy: 0.8902
Epoch 121/300
Epoch 00121: val accuracy did not improve from 0.89180
loss: 0.5992 - val accuracy: 0.8856
Epoch 122/300
Epoch 00122: val_accuracy did not improve from 0.89180
loss: 0.6007 - val accuracy: 0.8868
Epoch 123/300
Epoch 00123: val accuracy did not improve from 0.89180
390/390 [============ ] - 135s 347ms/step - loss: 0.0698 - accuracy: 0.9751 - val
loss: 0.6243 - val accuracy: 0.8824
Epoch 124/300
Epoch 00124: val accuracy did not improve from 0.89180
390/390 [============ ] - 135s 347ms/step - loss: 0.0710 - accuracy: 0.9747 - val
loss: 0.5675 - val accuracy: 0.8918
Epoch 125/300
Epoch 00125: val accuracy improved from 0.89180 to 0.89210, saving model to
/home/ubuntu/Project/my_data/CNN CIFAR/checkpoint/CIFAR model5 125-0.89.hdf5
390/390 [============ ] - 136s 349ms/step - loss: 0.0695 - accuracy: 0.9750 - val
_loss: 0.5825 - val_accuracy: 0.8921
Epoch 126/300
Epoch 00126: val accuracy did not improve from 0.89210
390/390 [============== ] - 136s 348ms/step - loss: 0.0691 - accuracy: 0.9760 - val
loss: 0.6159 - val accuracy: 0.8876
Epoch 127/300
Epoch 00127: val_accuracy did not improve from 0.89210
loss: 0.6079 - val accuracy: 0.8850
Epoch 128/300
Epoch 00128: val_accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0699 - accuracy: 0.9747 - val
_loss: 0.6289 - val_accuracy: 0.8841
Epoch 129/300
Epoch 00129: val accuracy did not improve from 0.89210
390/390 [============= ] - 135s 347ms/step - loss: 0.0680 - accuracy: 0.9754 - val
loss: 0.6260 - val accuracy: 0.8868
Epoch 130/300
Epoch 00130: val accuracy did not improve from 0.89210
390/390 [============= ] - 135s 347ms/step - loss: 0.0685 - accuracy: 0.9759 - val
loss: 0.6424 - val accuracy: 0.8821
Epoch 131/300
Epoch 00131: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0638 - accuracy: 0.9768 - val
loss: 0.6219 - val_accuracy: 0.8863
Epoch 132/300
Epoch 00132: val accuracy did not improve from 0.89210
390/390 [=============== ] - 135s 347ms/step - loss: 0.0664 - accuracy: 0.9763 - val
loss: 0.6491 - val_accuracy: 0.8841
Epoch 133/300
Epoch 00133: val accuracy did not improve from 0.89210
loss: 0.6434 - val accuracy: 0.8854
Epoch 134/300
Epoch 00134: val_accuracy did not improve from 0.89210
390/390 [============= ] - 135s 347ms/step - loss: 0.0651 - accuracy: 0.9771 - val
loss: 0.6236 - val accuracy: 0.8865
Epoch 135/300
Epoch 00135: val accuracy did not improve from 0.89210
```

```
390/390 [============ ] - 135s 347ms/step - loss: 0.0665 - accuracy: 0.9768 - val
_loss: 0.6119 - val_accuracy: 0.8880
Epoch 136/300
Epoch 00136: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0660 - accuracy: 0.9767 - val
loss: 0.6272 - val accuracy: 0.8867
Epoch 137/300
Epoch 00137: val accuracy did not improve from 0.89210
_loss: 0.6227 - val_accuracy: 0.8883
Epoch 138/300
Epoch 00138: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0663 - accuracy: 0.9762 - val
loss: 0.6828 - val accuracy: 0.8830
Epoch 139/300
Epoch 00139: val accuracy did not improve from 0.89210
loss: 0.6671 - val accuracy: 0.8811
Epoch 140/300
Epoch 00140: val accuracy did not improve from 0.89210
_loss: 0.6484 - val_accuracy: 0.8865
Epoch 141/300
Epoch 00141: val accuracy did not improve from 0.89210
Epoch 00141: ReduceLROnPlateau reducing learning rate to 1.0000000474974514e-05.
390/390 [============ ] - 135s 347ms/step - loss: 0.0651 - accuracy: 0.9770 - val
loss: 0.6447 - val accuracy: 0.8866
Epoch 142/300
Epoch 00142: val accuracy did not improve from 0.89210
390/390 [============= ] - 135s 347ms/step - loss: 0.0603 - accuracy: 0.9787 - val
loss: 0.6345 - val accuracy: 0.8887
Epoch 143/300
Epoch 00143: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0602 - accuracy: 0.9785 - val
loss: 0.6355 - val accuracy: 0.8880
Epoch 144/300
Epoch 00144: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0600 - accuracy: 0.9781 - val
loss: 0.6335 - val_accuracy: 0.8887
Epoch 145/300
Epoch 00145: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0586 - accuracy: 0.9793 - val
loss: 0.6340 - val_accuracy: 0.8878
Epoch 146/300
Epoch 00146: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0598 - accuracy: 0.9789 - val
loss: 0.6390 - val_accuracy: 0.8873
Epoch 147/300
Epoch 00147: val_accuracy did not improve from 0.89210
390/390 [=============== ] - 135s 347ms/step - loss: 0.0584 - accuracy: 0.9792 - val
loss: 0.6372 - val accuracy: 0.8878
Epoch 148/300
Epoch 00148: val accuracy did not improve from 0.89210
loss: 0.6324 - val accuracy: 0.8883
Epoch 149/300
Epoch 00149: val_accuracy did not improve from 0.89210
loss: 0.6336 - val accuracy: 0.8888
Epoch 150/300
Epoch 00150: val accuracy did not improve from 0.89210
```

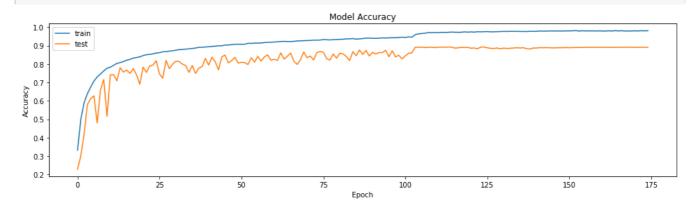
```
390/390 [============ ] - 135s 347ms/step - loss: 0.0561 - accuracy: 0.9796 - val
loss: 0.6279 - val_accuracy: 0.8890
Epoch 151/300
Epoch 00151: val_accuracy did not improve from 0.89210
loss: 0.6345 - val accuracy: 0.8883
Epoch 152/300
Epoch 00152: val accuracy did not improve from 0.89210
loss: 0.6357 - val accuracy: 0.8890
Epoch 153/300
Epoch 00153: val accuracy did not improve from 0.89210
390/390 [============= ] - 135s 347ms/step - loss: 0.0548 - accuracy: 0.9811 - val
loss: 0.6367 - val accuracy: 0.8892
Epoch 154/300
Epoch 00154: val accuracy did not improve from 0.89210
390/390 [============ ] - 136s 348ms/step - loss: 0.0597 - accuracy: 0.9786 - val
loss: 0.6346 - val accuracy: 0.8890
Epoch 155/300
Epoch 00155: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0566 - accuracy: 0.9803 - val
loss: 0.6294 - val_accuracy: 0.8892
Epoch 156/300
Epoch 00156: val_accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0591 - accuracy: 0.9792 - val
loss: 0.6295 - val accuracy: 0.8899
Epoch 157/300
Epoch 00157: val accuracy did not improve from 0.89210
Epoch 00157: ReduceLROnPlateau reducing learning rate to 1.0000000656873453e-06.
390/390 [============ ] - 135s 347ms/step - loss: 0.0569 - accuracy: 0.9795 - val
loss: 0.6296 - val accuracy: 0.8905
Epoch 158/300
Epoch 00158: val accuracy did not improve from 0.89210
loss: 0.6307 - val_accuracy: 0.8903
Epoch 159/300
Epoch 00159: val accuracy did not improve from 0.89210
390/390 [================== ] - 135s 347ms/step - loss: 0.0574 - accuracy: 0.9791 - val
loss: 0.6280 - val_accuracy: 0.8900
Epoch 160/300
Epoch 00160: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0569 - accuracy: 0.9796 - val
loss: 0.6276 - val accuracy: 0.8908
Epoch 161/300
Epoch 00161: val accuracy did not improve from 0.89210
390/390 [============== ] - 135s 347ms/step - loss: 0.0578 - accuracy: 0.9788 - val
loss: 0.6288 - val_accuracy: 0.8903
Epoch 162/300
Epoch 00162: val accuracy did not improve from 0.89210
390/390 [============== ] - 135s 347ms/step - loss: 0.0589 - accuracy: 0.9787 - val
loss: 0.6321 - val accuracy: 0.8902
Epoch 163/300
Epoch 00163: val accuracy did not improve from 0.89210
390/390 [=========== ] - 135s 347ms/step - loss: 0.0563 - accuracy: 0.9800 - val
loss: 0.6301 - val accuracy: 0.8898
Epoch 164/300
Epoch 00164: val accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0584 - accuracy: 0.9789 - val
loss: 0.6253 - val accuracy: 0.8903
Epoch 165/300
```

Epoch 00165: val accuracy did not improve from 0.89210

```
100. .ur_uccurac, uru .... rmprc.c r
loss: 0.6271 - val accuracy: 0.8904
Epoch 166/300
Epoch 00166: val accuracy did not improve from 0.89210
loss: 0.6305 - val accuracy: 0.8906
Epoch 167/300
Epoch 00167: val accuracy did not improve from 0.89210
390/390 [============= ] - 135s 347ms/step - loss: 0.0541 - accuracy: 0.9805 - val
loss: 0.6304 - val accuracy: 0.8903
Epoch 168/300
Epoch 00168: val accuracy did not improve from 0.89210
390/390 [=========== ] - 135s 347ms/step - loss: 0.0567 - accuracy: 0.9790 - val
loss: 0.6275 - val_accuracy: 0.8910
Epoch 169/300
Epoch 00169: val accuracy did not improve from 0.89210
390/390 [============== ] - 135s 347ms/step - loss: 0.0587 - accuracy: 0.9790 - val
loss: 0.6263 - val accuracy: 0.8912
Epoch 170/300
Epoch 00170: val accuracy did not improve from 0.89210
loss: 0.6336 - val_accuracy: 0.8901
Epoch 171/300
Epoch 00171: val_accuracy did not improve from 0.89210
390/390 [============== ] - 135s 347ms/step - loss: 0.0565 - accuracy: 0.9799 - val
loss: 0.6327 - val accuracy: 0.8901
Epoch 172/300
Epoch 00172: val accuracy did not improve from 0.89210
390/390 [================== ] - 135s 347ms/step - loss: 0.0595 - accuracy: 0.9792 - val
loss: 0.6267 - val accuracy: 0.8905
Epoch 173/300
Epoch 00173: val accuracy did not improve from 0.89210
Epoch 00173: ReduceLROnPlateau reducing learning rate to 1.0000001111620805e-07.
390/390 [============ ] - 135s 347ms/step - loss: 0.0556 - accuracy: 0.9805 - val
loss: 0.6276 - val_accuracy: 0.8906
Epoch 174/300
Epoch 00174: val_accuracy did not improve from 0.89210
390/390 [============ ] - 135s 347ms/step - loss: 0.0570 - accuracy: 0.9797 - val
loss: 0.6287 - val accuracy: 0.8908
Epoch 175/300
Epoch 00175: val accuracy did not improve from 0.89210
loss: 0.6282 - val accuracy: 0.8906
In [70]:
```

```
#history plot for accyracy
plt.figure(figsize=(16,4))
plt.plot(history_5.history['accuracy'])
plt.plot(history_5.history['val_accuracy'])
plt.title("Model Accuracy")
plt.xlabel("Epoch")
plt.ylabel("Accuracy")
plt.legend(["train", "test"], loc="upper left")
plt.show()
# history plot for accuracy
plt.figure(figsize=(16,4))
plt.plot(history 5.history["loss"])
plt.plot(history_5.history["val_loss"])
plt.title("Model Loss")
plt.xlabel("Epoch")
plt.ylabel("Loss")
plt.legend(["train", "test"], loc="upper left")
nlt show()
```





Model Loss 1.5 1.0 0.5 0.0 25 50 75 100 125 150 175 Epoch

In [71]:

```
best_model_5 =
tf.keras.models.load_model('/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model5_125-0.8
9.hdf5')
scores = best_model_5.evaluate(X_test, y_test, verbose=1)
print(scores)
```

6. Model -6

• Model-5 without dropouts

In [11]:

```
#BN-->ReLU-->Conv2D-->Dropout-->concat(input, output)-->(put in loop)
from tensorflow.keras import layers
from tensorflow.keras.models import Model
num classes = 10
def denseblock(input, num_filter, dropout_rate):
    global compression
                         # to keep the growth rate of number of filters
    temp = input
    for in range(l):
       BatchNorm = layers.BatchNormalization()(temp)
       relu = layers.Activation('relu')(BatchNorm)
       Conv2D_7_7= layers.SeparableConv2D(int(num_filter*compression), (7,7), use_bias=False, padd
ing='same')(relu)
       if dropout_rate>0:
            Conv2D 7 7 = layers.Dropout(dropout rate)(Conv2D 7 7)
       \#concat the input(temp) and output(conv2d_7_7) , in resnet we add but here we concat
       concat = layers.Concatenate(axis=-1)([temp,Conv2D 7 7])
       #change the concat as input
        temp = concat
```

```
return temp
#BN-->relu-->conv2d(1x1)-->dropout-->avg pool
def transition(input, num_filter, dropout_rate):
    global compression
   BatchNorm = layers.BatchNormalization()(input)
   relu = layers.Activation('relu')(BatchNorm)
   Conv2D BottleNeck = layers.SeparableConvolution2D(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
#BN-->relu-->avgpool-->flat-->softmax
def output layer(input):
    global compression
   BatchNorm = layers.BatchNormalization()(input)
   relu = layers.Activation('relu')(BatchNorm)
   AvgPooling = layers.AveragePooling2D(pool_size=(2,2))(relu)
    conv op = layers.SeparableConvolution2D(num classes, (2,2), padding='valid') (AvgPooling)
    flat = layers.Flatten()(conv_op)
   output = layers.Dense(num_classes, activation='softmax')(flat)
   return output
# Hyperparameters
1 = 21
num filter = 48
compression = 0.3
dropout rate = 0
num classes = 10
input = layers.Input(shape=(input size))
First_Conv2D = layers.Conv2D(num_filter, (7,7), use_bias=False ,padding='same')(input)
#First dense and transition block
First_Block = denseblock(First_Conv2D, num_filter, dropout_rate)
First Transition = transition(First Block, num filter, dropout rate)
#Second dense and transition block
Second Block = denseblock(First Transition, num filter, dropout rate)
Second_Transition = transition(Second_Block, num_filter, dropout_rate)
#Third dense and transition block
Third Block = denseblock(Second_Transition, num_filter, dropout_rate)
Third_Transition = transition(Third_Block, num filter, dropout rate)
#last dense and output block
Last Block = denseblock(Third Transition, num filter, dropout rate)
output = output layer(Last Block)
```

In [12]:

```
model_6 = tf.keras.models.Model(inputs=[input], outputs=[output])
model_6.summary()
```

Model: "model"

Layer (type)	Output	Sha	pe		Param #	Connected to
input_1 (InputLayer)	[(None	==== , 32	, 32	, 3)]	0	
conv2d (Conv2D)	(None,	32,	32,	48)	7056	input_1[0][0]
batch_normalization (BatchNorma	(None,	32,	32,	48)	192	conv2d[0][0]
activation (Activation)	(None,	32,	32,	48)	0	batch_normalization[0][0]
separable_conv2d (SeparableConv	(None,	32,	32,	14)	3024	activation[0][0]
concatenate (Concatenate)	(None,	32,	32,	62)	0	conv2d[0][0] separable_conv2d[0][0]
batch_normalization_1 (BatchNor	(None,	32,	32,	62)	248	concatenate[0][0]
activation 1 (Activation)	/ NT ~ ~ ~	27	27	621	Λ	hatch normalization 1[0][0]

activation_i (Activation)	(NOHe,	34,	34,	04)	U	patcu_normarrzacron_r[v][v]
separable_conv2d_1 (SeparableCo	(None,	32,	32,	14)	3906	activation_1[0][0]
concatenate_1 (Concatenate)	(None,	32,	32,	76)	0	<pre>concatenate[0][0] separable_conv2d_1[0][0]</pre>
batch_normalization_2 (BatchNor	(None,	32,	32,	76)	304	concatenate_1[0][0]
activation_2 (Activation)	(None,	32,	32,	76)	0	batch_normalization_2[0][0]
separable_conv2d_2 (SeparableCo	(None,	32,	32,	14)	4788	activation_2[0][0]
concatenate_2 (Concatenate)	(None,	32,	32,	90)	0	concatenate_1[0][0] separable_conv2d_2[0][0]
batch_normalization_3 (BatchNor	(None,	32,	32,	90)	360	concatenate_2[0][0]
activation_3 (Activation)	(None,	32,	32,	90)	0	batch_normalization_3[0][0]
separable_conv2d_3 (SeparableCo	(None,	32,	32,	14)	5670	activation_3[0][0]
concatenate_3 (Concatenate)	(None,	32,	32,	104)	0	<pre>concatenate_2[0][0] separable_conv2d_3[0][0]</pre>
batch_normalization_4 (BatchNor	(None,	32,	32,	104)	416	concatenate_3[0][0]
activation_4 (Activation)	(None,	32,	32,	104)	0	batch_normalization_4[0][0]
separable_conv2d_4 (SeparableCo	(None,	32,	32,	14)	6552	activation_4[0][0]
concatenate_4 (Concatenate)	(None,	32,	32,	118)	0	concatenate_3[0][0] separable_conv2d_4[0][0]
batch_normalization_5 (BatchNor	(None,	32,	32,	118)	472	concatenate_4[0][0]
activation_5 (Activation)	(None,	32,	32,	118)	0	batch_normalization_5[0][0]
separable_conv2d_5 (SeparableCo	(None,	32,	32,	14)	7434	activation_5[0][0]
concatenate_5 (Concatenate)	(None,	32,	32,	132)	0	concatenate_4[0][0] separable_conv2d_5[0][0]
batch_normalization_6 (BatchNor	(None,	32,	32,	132)	528	concatenate_5[0][0]
activation_6 (Activation)	(None,	32,	32,	132)	0	batch_normalization_6[0][0]
separable_conv2d_6 (SeparableCo	(None,	32,	32,	14)	8316	activation_6[0][0]
concatenate_6 (Concatenate)	(None,	32,	32,	146)	0	concatenate_5[0][0] separable_conv2d_6[0][0]
<pre>batch_normalization_7 (BatchNor</pre>	(None,	32,	32,	146)	584	concatenate_6[0][0]
activation_7 (Activation)	(None,	32,	32,	146)	0	batch_normalization_7[0][0]
separable_conv2d_7 (SeparableCo	(None,	32,	32,	14)	9198	activation_7[0][0]
concatenate_7 (Concatenate)	(None,	32,	32,	160)	0	concatenate_6[0][0] separable_conv2d_7[0][0]
batch_normalization_8 (BatchNor	(None,	32,	32,	160)	640	concatenate_7[0][0]
activation_8 (Activation)	(None,	32,	32,	160)	0	batch_normalization_8[0][0]
separable_conv2d_8 (SeparableCo	(None,	32,	32,	14)	10080	activation_8[0][0]
concatenate_8 (Concatenate)	(None,	32,	32,	174)	0	concatenate_7[0][0] separable_conv2d_8[0][0]
batch_normalization_9 (BatchNor	(None,	32,	32,	174)	696	concatenate_8[0][0]
activation_9 (Activation)	(None,	32,	32,	174)	0	batch_normalization_9[0][0]
separable_conv2d_9 (SeparableCo	(None,	32,	32,	14)	10962	activation_9[0][0]
concatenate_9 (Concatenate)	(None,	32,	32,	188)	0	concatenate_8[0][0]

batch_normalization_10 (BatchNo	(None,	32,	32,	188)	752	concatenate_9[0][0]
activation_10 (Activation)	(None,	32,	32,	188)	0	batch_normalization_10[0][0]
separable_conv2d_10 (SeparableC	(None,	32,	32,	14)	11844	activation_10[0][0]
concatenate_10 (Concatenate)	(None,	32,	32,	202)	0	concatenate_9[0][0] separable_conv2d_10[0][0]
batch_normalization_11 (BatchNo	(None,	32,	32,	202)	808	concatenate_10[0][0]
activation_11 (Activation)	(None,	32,	32,	202)	0	batch_normalization_11[0][0]
separable_conv2d_11 (SeparableC	(None,	32,	32,	14)	12726	activation_11[0][0]
concatenate_11 (Concatenate)	(None,	32,	32,	216)	0	concatenate_10[0][0] separable_conv2d_11[0][0]
batch_normalization_12 (BatchNo	(None,	32,	32,	216)	864	concatenate_11[0][0]
activation_12 (Activation)	(None,	32,	32,	216)	0	batch_normalization_12[0][0]
separable_conv2d_12 (SeparableC	(None,	32,	32,	14)	13608	activation_12[0][0]
concatenate_12 (Concatenate)	(None,	32,	32,	230)	0	concatenate_11[0][0] separable_conv2d_12[0][0]
batch_normalization_13 (BatchNo	(None,	32,	32,	230)	920	concatenate_12[0][0]
activation_13 (Activation)	(None,	32,	32,	230)	0	batch_normalization_13[0][0]
separable_conv2d_13 (SeparableC	(None,	32,	32,	14)	14490	activation_13[0][0]
concatenate_13 (Concatenate)	(None,	32,	32,	244)	0	concatenate_12[0][0] separable_conv2d_13[0][0]
<pre>batch_normalization_14 (BatchNo</pre>	(None,	32,	32,	244)	976	concatenate_13[0][0]
activation_14 (Activation)	(None,	32,	32,	244)	0	batch_normalization_14[0][0]
separable_conv2d_14 (SeparableC	(None,	32,	32,	14)	15372	activation_14[0][0]
concatenate_14 (Concatenate)	(None,	32,	32,	258)	0	concatenate_13[0][0] separable_conv2d_14[0][0]
batch_normalization_15 (BatchNo	(None,	32,	32,	258)	1032	concatenate_14[0][0]
activation_15 (Activation)	(None,	32,	32,	258)	0	batch_normalization_15[0][0]
separable_conv2d_15 (SeparableC	(None,	32,	32,	14)	16254	activation_15[0][0]
concatenate_15 (Concatenate)	(None,	32,	32,	272)	0	concatenate_14[0][0] separable_conv2d_15[0][0]
batch_normalization_16 (BatchNo	(None,	32,	32,	272)	1088	concatenate_15[0][0]
activation_16 (Activation)	(None,	32,	32,	272)	0	batch_normalization_16[0][0]
separable_conv2d_16 (SeparableC	(None,	32,	32,	14)	17136	activation_16[0][0]
concatenate_16 (Concatenate)	(None,	32,	32,	286)	0	concatenate_15[0][0] separable_conv2d_16[0][0]
batch_normalization_17 (BatchNo	(None,	32,	32,	286)	1144	concatenate_16[0][0]
activation_17 (Activation)	(None,	32,	32,	286)	0	batch_normalization_17[0][0]
separable_conv2d_17 (SeparableC	(None,	32,	32,	14)	18018	activation_17[0][0]
concatenate_17 (Concatenate)	(None,	32,	32,	300)	0	concatenate_16[0][0] separable_conv2d_17[0][0]
batch_normalization_18 (BatchNo	(None,	32,	32,	300)	1200	concatenate_17[0][0]
activation_18 (Activation)	(None,	32,	32,	300)	0	batch_normalization_18[0][0]

separable_conv2d_18 (SeparableC	(None,	32,	32,	14)	18900	activation_18[0][0]
concatenate_18 (Concatenate)	(None,	32,	32,	314)	0	concatenate_17[0][0] separable_conv2d_18[0][0]
oatch_normalization_19 (BatchNo	(None,	32,	32,	314)	1256	concatenate_18[0][0]
activation_19 (Activation)	(None,	32,	32,	314)	0	batch_normalization_19[0][0]
separable_conv2d_19 (SeparableC	(None,	32,	32,	14)	19782	activation_19[0][0]
concatenate_19 (Concatenate)	(None,	32,	32,	328)	0	concatenate_18[0][0] separable_conv2d_19[0][0]
batch_normalization_20 (BatchNo	(None,	32,	32,	328)	1312	concatenate_19[0][0]
activation_20 (Activation)	(None,	32,	32,	328)	0	batch_normalization_20[0][0]
separable_conv2d_20 (SeparableC	(None,	32,	32,	14)	20664	activation_20[0][0]
concatenate_20 (Concatenate)	(None,	32,	32,	342)	0	concatenate_19[0][0] separable_conv2d_20[0][0]
batch_normalization_21 (BatchNo	(None,	32,	32,	342)	1368	concatenate_20[0][0]
activation_21 (Activation)	(None,	32,	32,	342)	0	batch_normalization_21[0][0]
separable_conv2d_21 (SeparableC	(None,	32,	32,	14)	5130	activation_21[0][0]
average_pooling2d (AveragePooli	(None,	16,	16,	14)	0	separable_conv2d_21[0][0]
batch_normalization_22 (BatchNo	(None,	16,	16,	14)	56	average_pooling2d[0][0]
activation_22 (Activation)	(None,	16,	16,	14)	0	batch_normalization_22[0][0]
separable_conv2d_22 (SeparableC	(None,	16,	16,	14)	882	activation_22[0][0]
concatenate_21 (Concatenate)	(None,	16,	16,	28)	0	average_pooling2d[0][0] separable_conv2d_22[0][0]
batch_normalization_23 (BatchNo	(None,	16,	16,	28)	112	concatenate_21[0][0]
activation_23 (Activation)	(None,	16,	16,	28)	0	batch_normalization_23[0][0]
separable_conv2d_23 (SeparableC	(None,	16,	16,	14)	1764	activation_23[0][0]
concatenate_22 (Concatenate)	(None,	16,	16,	42)	0	concatenate_21[0][0] separable_conv2d_23[0][0]
batch_normalization_24 (BatchNo	(None,	16,	16,	42)	168	concatenate_22[0][0]
activation_24 (Activation)	(None,	16,	16,	42)	0	batch_normalization_24[0][0]
separable_conv2d_24 (SeparableC	(None,	16,	16,	14)	2646	activation_24[0][0]
concatenate_23 (Concatenate)	(None,	16,	16,	56)	0	concatenate_22[0][0] separable_conv2d_24[0][0]
batch_normalization_25 (BatchNo	(None,	16,	16,	56)	224	concatenate_23[0][0]
activation_25 (Activation)	(None,	16,	16,	56)	0	batch_normalization_25[0][0]
separable_conv2d_25 (SeparableC	(None,	16,	16,	14)	3528	activation_25[0][0]
concatenate_24 (Concatenate)	(None,	16,	16,	70)	0	concatenate_23[0][0] separable_conv2d_25[0][0]
batch_normalization_26 (BatchNo	(None,	16,	16,	70)	280	concatenate_24[0][0]
activation_26 (Activation)	(None,	16,	16,	70)	0	batch_normalization_26[0][0]
separable_conv2d_26 (SeparableC	(None,	16,	16,	14)	4410	activation_26[0][0]
concatenate_25 (Concatenate)	(None,	16,	16,	84)	0	concatenate_24[0][0] separable_conv2d_26[0][0]
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batch_normalization_27 (BatchNo	(None,	16,	16,	84)	336	concatenate_25[0][0]
activation_27 (Activation)	(None,	16,	16,	84)	0	batch_normalization_27[0][0]
separable_conv2d_27 (SeparableC	(None,	16,	16,	14)	5292	activation_27[0][0]
concatenate_26 (Concatenate)	(None,	16,	16,	98)	0	concatenate_25[0][0] separable_conv2d_27[0][0]
batch_normalization_28 (BatchNo	(None,	16,	16,	98)	392	concatenate_26[0][0]
activation_28 (Activation)	(None,	16,	16,	98)	0	batch_normalization_28[0][0]
separable_conv2d_28 (SeparableC	(None,	16,	16,	14)	6174	activation_28[0][0]
concatenate_27 (Concatenate)	(None,	16,	16,	112)	0	concatenate_26[0][0] separable_conv2d_28[0][0]
batch_normalization_29 (BatchNo	(None,	16,	16,	112)	448	concatenate_27[0][0]
activation_29 (Activation)	(None,	16,	16,	112)	0	batch_normalization_29[0][0]
separable_conv2d_29 (SeparableC	(None,	16,	16,	14)	7056	activation_29[0][0]
concatenate_28 (Concatenate)	(None,	16,	16,	126)	0	concatenate_27[0][0] separable_conv2d_29[0][0]
batch_normalization_30 (BatchNo	(None,	16,	16,	126)	504	concatenate_28[0][0]
activation_30 (Activation)	(None,	16,	16,	126)	0	batch_normalization_30[0][0]
separable_conv2d_30 (SeparableC	(None,	16,	16,	14)	7938	activation_30[0][0]
concatenate_29 (Concatenate)	(None,	16,	16,	140)	0	concatenate_28[0][0] separable_conv2d_30[0][0]
batch_normalization_31 (BatchNo	(None,	16,	16,	140)	560	concatenate_29[0][0]
activation_31 (Activation)	(None,	16,	16,	140)	0	batch_normalization_31[0][0]
separable_conv2d_31 (SeparableC	(None,	16,	16,	14)	8820	activation_31[0][0]
concatenate_30 (Concatenate)	(None,	16,	16,	154)	0	concatenate_29[0][0] separable_conv2d_31[0][0]
batch_normalization_32 (BatchNo	(None,	16,	16,	154)	616	concatenate_30[0][0]
activation_32 (Activation)	(None,	16,	16,	154)	0	batch_normalization_32[0][0]
separable_conv2d_32 (SeparableC	(None,	16,	16,	14)	9702	activation_32[0][0]
concatenate_31 (Concatenate)	(None,	16,	16,	168)	0	concatenate_30[0][0] separable_conv2d_32[0][0]
batch_normalization_33 (BatchNo	(None,	16,	16,	168)	672	concatenate_31[0][0]
activation_33 (Activation)	(None,	16,	16,	168)	0	batch_normalization_33[0][0]
separable_conv2d_33 (SeparableC	(None,	16,	16,	14)	10584	activation_33[0][0]
concatenate_32 (Concatenate)	(None,	16,	16,	182)	0	concatenate_31[0][0] separable_conv2d_33[0][0]
batch_normalization_34 (BatchNo	(None,	16,	16,	182)	728	concatenate_32[0][0]
activation_34 (Activation)	(None,	16,	16,	182)	0	batch_normalization_34[0][0]
separable_conv2d_34 (SeparableC	(None,	16,	16,	14)	11466	activation_34[0][0]
concatenate_33 (Concatenate)	(None,	16,	16,	196)	0	concatenate_32[0][0] separable_conv2d_34[0][0]
batch_normalization_35 (BatchNo	(None,	16,	16,	196)	784	concatenate_33[0][0]
activation_35 (Activation)	(None,	16,	16,	196)	0	batch_normalization_35[0][0]
separable conv2d 35 (SeparableC	(None,	16,	16,	14)	12348	activation_35[0][0]

concatenate_34 (Concatenate)	(None,	16,	16,	210)	0	concatenate_33[0][0] separable_conv2d_35[0][0]
batch_normalization_36 (BatchNo	(None,	16,	16,	210)	840	concatenate_34[0][0]
activation_36 (Activation)	(None,	16,	16,	210)	0	batch_normalization_36[0][0]
separable_conv2d_36 (SeparableC	(None,	16,	16,	14)	13230	activation_36[0][0]
concatenate_35 (Concatenate)	(None,	16,	16,	224)	0	concatenate_34[0][0] separable_conv2d_36[0][0]
batch_normalization_37 (BatchNo	(None,	16,	16,	224)	896	concatenate_35[0][0]
activation_37 (Activation)	(None,	16,	16,	224)	0	batch_normalization_37[0][0]
separable_conv2d_37 (SeparableC	(None,	16,	16,	14)	14112	activation_37[0][0]
concatenate_36 (Concatenate)	(None,	16,	16,	238)	0	concatenate_35[0][0] separable_conv2d_37[0][0]
batch_normalization_38 (BatchNo	(None,	16,	16,	238)	952	concatenate_36[0][0]
activation_38 (Activation)	(None,	16,	16,	238)	0	batch_normalization_38[0][0]
separable_conv2d_38 (SeparableC	(None,	16,	16,	14)	14994	activation_38[0][0]
concatenate_37 (Concatenate)	(None,	16,	16,	252)	0	concatenate_36[0][0] separable_conv2d_38[0][0]
batch_normalization_39 (BatchNo	(None,	16,	16,	252)	1008	concatenate_37[0][0]
activation_39 (Activation)	(None,	16,	16,	252)	0	batch_normalization_39[0][0]
separable_conv2d_39 (SeparableC	(None,	16,	16,	14)	15876	activation_39[0][0]
concatenate_38 (Concatenate)	(None,	16,	16,	266)	0	concatenate_37[0][0] separable_conv2d_39[0][0]
batch_normalization_40 (BatchNo	(None,	16,	16,	266)	1064	concatenate_38[0][0]
activation_40 (Activation)	(None,	16,	16,	266)	0	batch_normalization_40[0][0]
separable_conv2d_40 (SeparableC	(None,	16,	16,	14)	16758	activation_40[0][0]
concatenate_39 (Concatenate)	(None,	16,	16,	280)	0	concatenate_38[0][0] separable_conv2d_40[0][0]
batch_normalization_41 (BatchNo	(None,	16,	16,	280)	1120	concatenate_39[0][0]
activation_41 (Activation)	(None,	16,	16,	280)	0	batch_normalization_41[0][0]
separable_conv2d_41 (SeparableC	(None,	16,	16,	14)	17640	activation_41[0][0]
concatenate_40 (Concatenate)	(None,	16,	16,	294)	0	concatenate_39[0][0] separable_conv2d_41[0][0]
batch_normalization_42 (BatchNo	(None,	16,	16,	294)	1176	concatenate_40[0][0]
activation_42 (Activation)	(None,	16,	16,	294)	0	batch_normalization_42[0][0]
separable_conv2d_42 (SeparableC	(None,	16,	16,	14)	18522	activation_42[0][0]
concatenate_41 (Concatenate)	(None,	16,	16,	308)	0	concatenate_40[0][0] separable_conv2d_42[0][0]
batch_normalization_43 (BatchNo	(None,	16,	16,	308)	1232	concatenate_41[0][0]
activation_43 (Activation)	(None,	16,	16,	308)	0	batch_normalization_43[0][0]
separable_conv2d_43 (SeparableC	(None,	16,	16,	14)	4620	activation_43[0][0]
average_pooling2d_1 (AveragePoo	(None,	8,	8, 1	4)	0	separable_conv2d_43[0][0]
batch_normalization_44 (BatchNo	(None,	8,	8, 1	4)	56	average_pooling2d_1[0][0]

activation_44 (Activation)	(None,	8,	8,	14)	0	batch_normalization_44[0][0]
separable_conv2d_44 (SeparableC	(None,	8,	8,	14)	882	activation_44[0][0]
concatenate_42 (Concatenate)	(None,	8,	8,	28)	0	average_pooling2d_1[0][0] separable_conv2d_44[0][0]
batch_normalization_45 (BatchNo	(None,	8,	8,	28)	112	concatenate_42[0][0]
activation_45 (Activation)	(None,	8,	8,	28)	0	batch_normalization_45[0][0]
separable_conv2d_45 (SeparableC	(None,	8,	8,	14)	1764	activation_45[0][0]
concatenate_43 (Concatenate)	(None,	8,	8,	42)	0	concatenate_42[0][0] separable_conv2d_45[0][0]
batch_normalization_46 (BatchNo	(None,	8,	8,	42)	168	concatenate_43[0][0]
activation_46 (Activation)	(None,	8,	8,	42)	0	batch_normalization_46[0][0]
separable_conv2d_46 (SeparableC	(None,	8,	8,	14)	2646	activation_46[0][0]
concatenate_44 (Concatenate)	(None,	8,	8,	56)	0	concatenate_43[0][0] separable_conv2d_46[0][0]
batch_normalization_47 (BatchNo	(None,	8,	8,	56)	224	concatenate_44[0][0]
activation_47 (Activation)	(None,	8,	8,	56)	0	batch_normalization_47[0][0]
separable_conv2d_47 (SeparableC	(None,	8,	8,	14)	3528	activation_47[0][0]
concatenate_45 (Concatenate)	(None,	8,	8,	70)	0	concatenate_44[0][0] separable_conv2d_47[0][0]
batch_normalization_48 (BatchNo	(None,	8,	8,	70)	280	concatenate_45[0][0]
activation_48 (Activation)	(None,	8,	8,	70)	0	batch_normalization_48[0][0]
separable_conv2d_48 (SeparableC	(None,	8,	8,	14)	4410	activation_48[0][0]
concatenate_46 (Concatenate)	(None,	8,	8,	84)	0	concatenate_45[0][0] separable_conv2d_48[0][0]
batch_normalization_49 (BatchNo	(None,	8,	8,	84)	336	concatenate_46[0][0]
activation_49 (Activation)	(None,	8,	8,	84)	0	batch_normalization_49[0][0]
separable_conv2d_49 (SeparableC	(None,	8,	8,	14)	5292	activation_49[0][0]
concatenate_47 (Concatenate)	(None,	8,	8,	98)	0	concatenate_46[0][0] separable_conv2d_49[0][0]
batch_normalization_50 (BatchNo	(None,	8,	8,	98)	392	concatenate_47[0][0]
activation_50 (Activation)	(None,	8,	8,	98)	0	batch_normalization_50[0][0]
separable_conv2d_50 (SeparableC	(None,	8,	8,	14)	6174	activation_50[0][0]
concatenate_48 (Concatenate)	(None,	8,	8,	112)	0	concatenate_47[0][0] separable_conv2d_50[0][0]
batch_normalization_51 (BatchNo	(None,	8,	8,	112)	448	concatenate_48[0][0]
activation_51 (Activation)	(None,	8,	8,	112)	0	batch_normalization_51[0][0]
separable_conv2d_51 (SeparableC	(None,	8,	8,	14)	7056	activation_51[0][0]
concatenate_49 (Concatenate)	(None,	8,	8,	126)	0	concatenate_48[0][0] separable_conv2d_51[0][0]
batch_normalization_52 (BatchNo	(None,	8,	8,	126)	504	concatenate_49[0][0]
activation_52 (Activation)	(None,	8,	8,	126)	0	batch_normalization_52[0][0]
separable_conv2d_52 (SeparableC	(None,	8,	8,	14)	7938	activation_52[0][0]
concatenate_50 (Concatenate)	(None,	8,	8,	140)	0	concatenate_49[0][0]

batch_normalization_53 (BatchNo	(None,	8,	8,	140)	560	concatenate_50[0][0]
activation_53 (Activation)	(None,	8,	8,	140)	0	batch_normalization_53[0][0]
separable_conv2d_53 (SeparableC	(None,	8,	8,	14)	8820	activation_53[0][0]
concatenate_51 (Concatenate)	(None,	8,	8,	154)	0	concatenate_50[0][0] separable_conv2d_53[0][0]
batch_normalization_54 (BatchNo	(None,	8,	8,	154)	616	concatenate_51[0][0]
activation_54 (Activation)	(None,	8,	8,	154)	0	batch_normalization_54[0][0]
separable_conv2d_54 (SeparableC	(None,	8,	8,	14)	9702	activation_54[0][0]
concatenate_52 (Concatenate)	(None,	8,	8,	168)	0	concatenate_51[0][0] separable_conv2d_54[0][0]
batch_normalization_55 (BatchNo	(None,	8,	8,	168)	672	concatenate_52[0][0]
activation_55 (Activation)	(None,	8,	8,	168)	0	batch_normalization_55[0][0]
separable_conv2d_55 (SeparableC	(None,	8,	8,	14)	10584	activation_55[0][0]
concatenate_53 (Concatenate)	(None,	8,	8,	182)	0	concatenate_52[0][0] separable_conv2d_55[0][0]
batch_normalization_56 (BatchNo	(None,	8,	8,	182)	728	concatenate_53[0][0]
activation_56 (Activation)	(None,	8,	8,	182)	0	batch_normalization_56[0][0]
separable_conv2d_56 (SeparableC	(None,	8,	8,	14)	11466	activation_56[0][0]
concatenate_54 (Concatenate)	(None,	8,	8,	196)	0	concatenate_53[0][0] separable_conv2d_56[0][0]
batch_normalization_57 (BatchNo	(None,	8,	8,	196)	784	concatenate_54[0][0]
activation_57 (Activation)	(None,	8,	8,	196)	0	batch_normalization_57[0][0]
separable_conv2d_57 (SeparableC	(None,	8,	8,	14)	12348	activation_57[0][0]
concatenate_55 (Concatenate)	(None,	8,	8,	210)	0	concatenate_54[0][0] separable_conv2d_57[0][0]
batch_normalization_58 (BatchNo	(None,	8,	8,	210)	840	concatenate_55[0][0]
activation_58 (Activation)	(None,	8,	8,	210)	0	batch_normalization_58[0][0]
separable_conv2d_58 (SeparableC	(None,	8,	8,	14)	13230	activation_58[0][0]
concatenate_56 (Concatenate)	(None,	8,	8,	224)	0	concatenate_55[0][0] separable_conv2d_58[0][0]
<pre>batch_normalization_59 (BatchNo</pre>	(None,	8,	8,	224)	896	concatenate_56[0][0]
activation_59 (Activation)	(None,	8,	8,	224)	0	batch_normalization_59[0][0]
separable_conv2d_59 (SeparableC	(None,	8,	8,	14)	14112	activation_59[0][0]
concatenate_57 (Concatenate)	(None,	8,	8,	238)	0	concatenate_56[0][0] separable_conv2d_59[0][0]
batch_normalization_60 (BatchNo	(None,	8,	8,	238)	952	concatenate_57[0][0]
activation_60 (Activation)	(None,	8,	8,	238)	0	batch_normalization_60[0][0]
separable_conv2d_60 (SeparableC	(None,	8,	8,	14)	14994	activation_60[0][0]
concatenate_58 (Concatenate)	(None,	8,	8,	252)	0	concatenate_57[0][0] separable_conv2d_60[0][0]
batch_normalization_61 (BatchNo	(None,	8,	8,	252)	1008	concatenate_58[0][0]

separable_conv2d_61 (SeparableC	(None,	8,	8,	14)	15876	activation_61[0][0]
concatenate_59 (Concatenate)	(None,	8,	8,	266)	0	concatenate_58[0][0] separable_conv2d_61[0][0]
batch_normalization_62 (BatchNo	(None,	8,	8,	266)	1064	concatenate_59[0][0]
activation_62 (Activation)	(None,	8,	8,	266)	0	batch_normalization_62[0][0]
separable_conv2d_62 (SeparableC	(None,	8,	8,	14)	16758	activation_62[0][0]
concatenate_60 (Concatenate)	(None,	8,	8,	280)	0	concatenate_59[0][0] separable_conv2d_62[0][0]
batch_normalization_63 (BatchNo	(None,	8,	8,	280)	1120	concatenate_60[0][0]
activation_63 (Activation)	(None,	8,	8,	280)	0	batch_normalization_63[0][0]
separable_conv2d_63 (SeparableC	(None,	8,	8,	14)	17640	activation_63[0][0]
concatenate_61 (Concatenate)	(None,	8,	8,	294)	0	concatenate_60[0][0] separable_conv2d_63[0][0]
batch_normalization_64 (BatchNo	(None,	8,	8,	294)	1176	concatenate_61[0][0]
activation_64 (Activation)	(None,	8,	8,	294)	0	batch_normalization_64[0][0]
separable_conv2d_64 (SeparableC	(None,	8,	8,	14)	18522	activation_64[0][0]
concatenate_62 (Concatenate)	(None,	8,	8,	308)	0	concatenate_61[0][0] separable_conv2d_64[0][0]
batch_normalization_65 (BatchNo	(None,	8,	8,	308)	1232	concatenate_62[0][0]
activation_65 (Activation)	(None,	8,	8,	308)	0	batch_normalization_65[0][0]
separable_conv2d_65 (SeparableC	(None,	8,	8,	14)	4620	activation_65[0][0]
average_pooling2d_2 (AveragePoo	(None,	4,	4,	14)	0	separable_conv2d_65[0][0]
batch_normalization_66 (BatchNo	(None,	4,	4,	14)	56	average_pooling2d_2[0][0]
activation_66 (Activation)	(None,	4,	4,	14)	0	batch_normalization_66[0][0]
separable_conv2d_66 (SeparableC	(None,	4,	4,	14)	882	activation_66[0][0]
concatenate_63 (Concatenate)	(None,	4,	4,	28)	0	average_pooling2d_2[0][0] separable_conv2d_66[0][0]
batch_normalization_67 (BatchNo	(None,	4,	4,	28)	112	concatenate_63[0][0]
activation_67 (Activation)	(None,	4,	4,	28)	0	batch_normalization_67[0][0]
separable_conv2d_67 (SeparableC	(None,	4,	4,	14)	1764	activation_67[0][0]
concatenate_64 (Concatenate)	(None,	4,	4,	42)	0	concatenate_63[0][0] separable_conv2d_67[0][0]
batch_normalization_68 (BatchNo	(None,	4,	4,	42)	168	concatenate_64[0][0]
activation_68 (Activation)	(None,	4,	4,	42)	0	batch_normalization_68[0][0]
separable_conv2d_68 (SeparableC	(None,	4,	4,	14)	2646	activation_68[0][0]
concatenate_65 (Concatenate)	(None,	4,	4,	56)	0	concatenate_64[0][0] separable_conv2d_68[0][0]
batch_normalization_69 (BatchNo	(None,	4,	4,	56)	224	concatenate_65[0][0]
activation_69 (Activation)	(None,	4,	4,	56)	0	batch_normalization_69[0][0]
separable_conv2d_69 (SeparableC	(None,	4,	4,	14)	3528	activation_69[0][0]
concatenate_66 (Concatenate)	(None,	4,	4,	70)	0	concatenate_65[0][0] separable_conv2d_69[0][0]

batch_normalization_70 (BatchNo	(None,	4,	4,	70)	280	concatenate_66[0][0]
activation_70 (Activation)	(None,	4,	4,	70)	0	batch_normalization_70[0][0]
separable_conv2d_70 (SeparableC	(None,	4,	4,	14)	4410	activation_70[0][0]
concatenate_67 (Concatenate)	(None,	4,	4,	84)	0	concatenate_66[0][0] separable_conv2d_70[0][0]
batch_normalization_71 (BatchNo	(None,	4,	4,	84)	336	concatenate_67[0][0]
activation_71 (Activation)	(None,	4,	4,	84)	0	batch_normalization_71[0][0]
separable_conv2d_71 (SeparableC	(None,	4,	4,	14)	5292	activation_71[0][0]
concatenate_68 (Concatenate)	(None,	4,	4,	98)	0	concatenate_67[0][0] separable_conv2d_71[0][0]
batch_normalization_72 (BatchNo	(None,	4,	4,	98)	392	concatenate_68[0][0]
activation_72 (Activation)	(None,	4,	4,	98)	0	batch_normalization_72[0][0]
separable_conv2d_72 (SeparableC	(None,	4,	4,	14)	6174	activation_72[0][0]
concatenate_69 (Concatenate)	(None,	4,	4,	112)	0	concatenate_68[0][0] separable_conv2d_72[0][0]
batch_normalization_73 (BatchNo	(None,	4,	4,	112)	448	concatenate_69[0][0]
activation_73 (Activation)	(None,	4,	4,	112)	0	batch_normalization_73[0][0]
separable_conv2d_73 (SeparableC	(None,	4,	4,	14)	7056	activation_73[0][0]
concatenate_70 (Concatenate)	(None,	4,	4,	126)	0	concatenate_69[0][0] separable_conv2d_73[0][0]
batch_normalization_74 (BatchNo	(None,	4,	4,	126)	504	concatenate_70[0][0]
activation_74 (Activation)	(None,	4,	4,	126)	0	batch_normalization_74[0][0]
separable_conv2d_74 (SeparableC	(None,	4,	4,	14)	7938	activation_74[0][0]
concatenate_71 (Concatenate)	(None,	4,	4,	140)	0	concatenate_70[0][0] separable_conv2d_74[0][0]
batch_normalization_75 (BatchNo	(None,	4,	4,	140)	560	concatenate_71[0][0]
activation_75 (Activation)	(None,	4,	4,	140)	0	batch_normalization_75[0][0]
separable_conv2d_75 (SeparableC	(None,	4,	4,	14)	8820	activation_75[0][0]
concatenate_72 (Concatenate)	(None,	4,	4,	154)	0	concatenate_71[0][0] separable_conv2d_75[0][0]
batch_normalization_76 (BatchNo	(None,	4,	4,	154)	616	concatenate_72[0][0]
activation_76 (Activation)	(None,	4,	4,	154)	0	batch_normalization_76[0][0]
separable_conv2d_76 (SeparableC	(None,	4,	4,	14)	9702	activation_76[0][0]
concatenate_73 (Concatenate)	(None,	4,	4,	168)	0	concatenate_72[0][0] separable_conv2d_76[0][0]
batch_normalization_77 (BatchNo	(None,	4,	4,	168)	672	concatenate_73[0][0]
activation_77 (Activation)	(None,	4,	4,	168)	0	batch_normalization_77[0][0]
separable_conv2d_77 (SeparableC	(None,	4,	4,	14)	10584	activation_77[0][0]
concatenate_74 (Concatenate)	(None,	4,	4,	182)	0	concatenate_73[0][0] separable_conv2d_77[0][0]
batch_normalization_78 (BatchNo	(None,	4,	4,	182)	728	concatenate_74[0][0]
activation_78 (Activation)	(None,	4,	4,	182)	0	batch_normalization_78[0][0]
separable_conv2d_78 (SeparableC	(None,	4,	4,	14)	11466	activation_78[0][0]

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concatenate_75 (Concatenate)	(None, 4	, 4	,	196)	0	concatenate_74[0][0] separable_conv2d_78[0][0]
batch_normalization_79 (BatchNo	(None, 4	, 4	,	196)	784	concatenate_75[0][0]
activation_79 (Activation)	(None, 4	, 4	,	196)	0	batch_normalization_79[0][0]
separable_conv2d_79 (SeparableC	(None, 4	, 4	,	14)	12348	activation_79[0][0]
concatenate_76 (Concatenate)	(None, 4	, 4	,	210)	0	concatenate_75[0][0] separable_conv2d_79[0][0]
batch_normalization_80 (BatchNo	(None, 4	, 4	,	210)	840	concatenate_76[0][0]
activation_80 (Activation)	(None, 4	, 4	,	210)	0	batch_normalization_80[0][0]
separable_conv2d_80 (SeparableC	(None, 4	, 4	,	14)	13230	activation_80[0][0]
concatenate_77 (Concatenate)	(None, 4	, 4	,	224)	0	concatenate_76[0][0] separable_conv2d_80[0][0]
batch_normalization_81 (BatchNo	(None, 4	, 4	,	224)	896	concatenate_77[0][0]
activation_81 (Activation)	(None, 4	, 4	,	224)	0	batch_normalization_81[0][0]
separable_conv2d_81 (SeparableC	(None, 4	, 4	,	14)	14112	activation_81[0][0]
concatenate_78 (Concatenate)	(None, 4	, 4	,	238)	0	concatenate_77[0][0] separable_conv2d_81[0][0]
batch_normalization_82 (BatchNo	(None, 4	, 4	,	238)	952	concatenate_78[0][0]
activation_82 (Activation)	(None, 4	, 4	,	238)	0	batch_normalization_82[0][0]
separable_conv2d_82 (SeparableC	(None, 4	, 4	,	14)	14994	activation_82[0][0]
concatenate_79 (Concatenate)	(None, 4	, 4	,	252)	0	concatenate_78[0][0] separable_conv2d_82[0][0]
batch_normalization_83 (BatchNo	(None, 4	, 4	,	252)	1008	concatenate_79[0][0]
activation_83 (Activation)	(None, 4	, 4	,	252)	0	batch_normalization_83[0][0]
separable_conv2d_83 (SeparableC	(None, 4	, 4	,	14)	15876	activation_83[0][0]
concatenate_80 (Concatenate)	(None, 4	, 4	,	266)	0	concatenate_79[0][0] separable_conv2d_83[0][0]
batch_normalization_84 (BatchNo	(None, 4	, 4	,	266)	1064	concatenate_80[0][0]
activation_84 (Activation)	(None, 4	, 4	,	266)	0	batch_normalization_84[0][0]
separable_conv2d_84 (SeparableC	(None, 4	, 4	,	14)	16758	activation_84[0][0]
concatenate_81 (Concatenate)	(None, 4	, 4	,	280)	0	concatenate_80[0][0] separable_conv2d_84[0][0]
batch_normalization_85 (BatchNo	(None, 4	, 4	,	280)	1120	concatenate_81[0][0]
activation_85 (Activation)	(None, 4	, 4	,	280)	0	batch_normalization_85[0][0]
separable_conv2d_85 (SeparableC	(None, 4	, 4	,	14)	17640	activation_85[0][0]
concatenate_82 (Concatenate)	(None, 4	, 4	,	294)	0	concatenate_81[0][0] separable_conv2d_85[0][0]
batch_normalization_86 (BatchNo	(None, 4	, 4	,	294)	1176	concatenate_82[0][0]
activation_86 (Activation)	(None, 4	, 4	,	294)	0	batch_normalization_86[0][0]
separable_conv2d_86 (SeparableC	(None, 4	, 4	,	14)	18522	activation_86[0][0]
concatenate_83 (Concatenate)	(None, 4	, 4	,	308)	0	concatenate_82[0][0] separable_conv2d_86[0][0]
batch normalization 87 (BatchNo	(None, 4	, 4	,	308)	1232	concatenate 83[0][0]

activation 87 (Activation) batch normalization 87[0][0] (None, 4, 4, 308) average pooling2d 3 (AveragePoo (None, 2, 2, 308) activation 87[0][0] 0 separable conv2d 87 (SeparableC (None, 1, 1, 10) 4322 average pooling2d 3[0][0] flatten (Flatten) 0 separable conv2d 87[0][0] (None, 10) dense (Dense) (None, 10) 110 flatten[0][0] ______

Total params: 945,472 Trainable params: 915,640 Non-trainable params: 29,832

In [13]:

```
model 6.compile(optimizer='adam', loss='categorical crossentropy', metrics=['accuracy'])
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
patience = 50
base_path = '/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/'
checkpoint_file_name = base_path + 'CIFAR_model6' + '_{epoch:02d}-{val_accuracy:.2f}.hdf5'
model checkpoint = ModelCheckpoint (checkpoint file name, monitor='val accuracy', verbose=1, save be
early_stop = EarlyStopping('val_accuracy', mode='max', patience = patience)
reduce LR = ReduceLROnPlateau (monitor='val accuracy', mode='max', factor=0.1, patience=int(patience
/3), verbose=1)
callbacks = [model checkpoint, early stop, reduce LR]
epochs = 300
batch size = 128
#https://keras.io/api/preprocessing/image/#flow-method
history_6 = model_6.fit(data_generator.flow(X_train, y_train, batch_size),
                    steps per epoch = int(len(X train)/batch size),
                    epochs = epochs,
                    callbacks = callbacks,
                    validation_data = (X_test, y_test), verbose=1)
```

```
WARNING:tensorflow:sample weight modes were coerced from
  to
 ['...']
Train for 390 steps, validate on 10000 samples
Epoch 1/300
Epoch 00001: val accuracy improved from -inf to 0.28830, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model6 01-0.29.hdf5
390/390 [================== ] - 161s 413ms/step - loss: 1.6753 - accuracy: 0.3734 - val
loss: 1.9707 - val accuracy: 0.2883
Epoch 2/300
Epoch 00002: val accuracy improved from 0.28830 to 0.50130, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model6_02-0.50.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 1.2799 - accuracy: 0.5369 - val
loss: 1.4081 - val_accuracy: 0.5013
Epoch 3/300
Epoch 00003: val accuracy improved from 0.50130 to 0.55070, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model6_03-0.55.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 1.0746 - accuracy: 0.6183 - val
_loss: 1.3930 - val_accuracy: 0.5507
Epoch 4/300
Epoch 00004: val_accuracy improved from 0.55070 to 0.56570, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model6_04-0.57.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 0.9169 - accuracy: 0.6762 - val
loss: 1.3800 - val accuracy: 0.5657
Epoch 5/300
Epoch 00005: val accuracy improved from 0.56570 to 0.67600, saving model to
```

```
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR model6 05-0.68.hdf5
390/390 [=================== ] - 138s 354ms/step - loss: 0.8097 - accuracy: 0.7159 - val
loss: 0.9822 - val accuracy: 0.6760
Epoch 6/300
Epoch 00006: val accuracy did not improve from 0.67600
390/390 [============ ] - 137s 351ms/step - loss: 0.7262 - accuracy: 0.7463 - val
_loss: 1.1215 - val_accuracy: 0.6348
Epoch 7/300
Epoch 00007: val accuracy improved from 0.67600 to 0.73430, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model6_07-0.73.hdf5
390/390 [============= ] - 138s 354ms/step - loss: 0.6701 - accuracy: 0.7649 - val
loss: 0.7764 - val accuracy: 0.7343
Epoch 8/300
Epoch 00008: val accuracy improved from 0.73430 to 0.77720, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR model6 08-0.78.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 0.6186 - accuracy: 0.7825 - val
loss: 0.6698 - val accuracy: 0.7772
Epoch 9/300
Epoch 00009: val_accuracy did not improve from 0.77720
390/390 [============ ] - 137s 351ms/step - loss: 0.5798 - accuracy: 0.7984 - val
loss: 0.8203 - val accuracy: 0.7292
Epoch 10/300
Epoch 00010: val accuracy did not improve from 0.77720
390/390 [============ ] - 137s 352ms/step - loss: 0.5450 - accuracy: 0.8105 - val
loss: 0.7036 - val accuracy: 0.7676
Epoch 11/300
Epoch 00011: val accuracy improved from 0.77720 to 0.79050, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_model6\_11-0.79.hdf5
loss: 0.6237 - val accuracy: 0.7905
Epoch 12/300
Epoch 00012: val_accuracy did not improve from 0.79050
390/390 [============ ] - 137s 351ms/step - loss: 0.4872 - accuracy: 0.8315 - val
_loss: 0.6534 - val_accuracy: 0.7857
Epoch 13/300
Epoch 00013: val accuracy improved from 0.79050 to 0.80100, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_model6\_13-0.80.hdf5
loss: 0.6053 - val accuracy: 0.8010
Epoch 14/300
Epoch 00014: val accuracy did not improve from 0.80100
390/390 [============= ] - 137s 351ms/step - loss: 0.4516 - accuracy: 0.8422 - val
loss: 0.6065 - val_accuracy: 0.8000
Epoch 15/300
Epoch 00015: val_accuracy did not improve from 0.80100
390/390 [============= ] - 137s 351ms/step - loss: 0.4318 - accuracy: 0.8495 - val
loss: 0.8728 - val accuracy: 0.7389
Epoch 16/300
Epoch 00016: val accuracy did not improve from 0.80100
390/390 [============= ] - 137s 352ms/step - loss: 0.4149 - accuracy: 0.8558 - val
loss: 0.6928 - val accuracy: 0.7833
Epoch 17/300
Epoch 00017: val accuracy did not improve from 0.80100
390/390 [============ ] - 137s 351ms/step - loss: 0.3961 - accuracy: 0.8628 - val
loss: 0.9070 - val_accuracy: 0.7273
Epoch 18/300
Epoch 00018: val accuracy improved from 0.80100 to 0.80490, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model6 18-0.80.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 0.3868 - accuracy: 0.8667 - val
_loss: 0.6350 - val_accuracy: 0.8049
Epoch 19/300
Epoch 00019: val accuracy improved from 0.80490 to 0.81600, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model6 19-0.82.hdf5
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390/390 [============ ] - 138s 355ms/step - loss: 0.3677 - accuracy: 0.8714 - val
loss: 0.5698 - val accuracy: 0.8160
Epoch 20/300
Epoch 00020: val accuracy improved from 0.81600 to 0.82040, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model6 20-0.82.hdf5
loss: 0.5517 - val_accuracy: 0.8204
Epoch 21/300
Epoch 00021: val_accuracy did not improve from 0.82040
390/390 [=============== ] - 137s 351ms/step - loss: 0.3474 - accuracy: 0.8784 - val
loss: 0.6536 - val_accuracy: 0.7993
Epoch 22/300
Epoch 00022: val accuracy did not improve from 0.82040
390/390 [================== ] - 137s 352ms/step - loss: 0.3370 - accuracy: 0.8830 - val
loss: 0.6179 - val accuracy: 0.8077
Epoch 23/300
Epoch 00023: val accuracy did not improve from 0.82040
390/390 [============ ] - 137s 351ms/step - loss: 0.3240 - accuracy: 0.8877 - val
loss: 0.7175 - val_accuracy: 0.7860
Epoch 24/300
Epoch 00024: val_accuracy did not improve from 0.82040
390/390 [============== ] - 137s 351ms/step - loss: 0.3189 - accuracy: 0.8880 - val
loss: 0.5838 - val accuracy: 0.8171
Epoch 25/300
Epoch 00025: val accuracy improved from 0.82040 to 0.82790, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_model6\_25-0.83.hdf5
390/390 [============== ] - 138s 354ms/step - loss: 0.3098 - accuracy: 0.8921 - val
loss: 0.5647 - val accuracy: 0.8279
Epoch 26/300
Epoch 00026: val accuracy did not improve from 0.82790
390/390 [=========== ] - 137s 351ms/step - loss: 0.2965 - accuracy: 0.8961 - val
loss: 0.5734 - val accuracy: 0.8218
Epoch 27/300
Epoch 00027: val accuracy did not improve from 0.82790
390/390 [================== ] - 137s 351ms/step - loss: 0.2893 - accuracy: 0.8979 - val
loss: 0.5820 - val accuracy: 0.8230
Epoch 28/300
Epoch 00028: val accuracy did not improve from 0.82790
390/390 [============= ] - 137s 352ms/step - loss: 0.2770 - accuracy: 0.9021 - val
loss: 0.6923 - val accuracy: 0.8024
Epoch 29/300
Epoch 00029: val accuracy did not improve from 0.82790
390/390 [============ ] - 137s 351ms/step - loss: 0.2754 - accuracy: 0.9049 - val
loss: 0.6549 - val accuracy: 0.8176
Epoch 30/300
Epoch 00030: val accuracy improved from 0.82790 to 0.83870, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model6 30-0.84.hdf5
390/390 [=========== ] - 138s 354ms/step - loss: 0.2654 - accuracy: 0.9073 - val
_loss: 0.5280 - val_accuracy: 0.8387
Epoch 31/300
Epoch 00031: val accuracy did not improve from 0.83870
390/390 [============ ] - 137s 352ms/step - loss: 0.2660 - accuracy: 0.9066 - val
loss: 0.5413 - val accuracy: 0.8364
Epoch 32/300
Epoch 00032: val accuracy did not improve from 0.83870
390/390 [============ ] - 137s 351ms/step - loss: 0.2543 - accuracy: 0.9106 - val
_loss: 0.5786 - val_accuracy: 0.8306
Epoch 33/300
Epoch 00033: val_accuracy improved from 0.83870 to 0.84930, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model6 33-0.85.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 0.2453 - accuracy: 0.9141 - val
loss: 0.5121 - val accuracy: 0.8493
```

Epoch 34/300

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Epoch 00034: val accuracy did not improve from 0.84930
390/390 [============= ] - 137s 352ms/step - loss: 0.2445 - accuracy: 0.9136 - val
loss: 0.6663 - val accuracy: 0.8154
Epoch 35/300
Epoch 00035: val accuracy did not improve from 0.84930
loss: 0.7122 - val accuracy: 0.8040
Epoch 36/300
Epoch 00036: val accuracy did not improve from 0.84930
390/390 [============ ] - 137s 351ms/step - loss: 0.2322 - accuracy: 0.9188 - val
loss: 0.5439 - val_accuracy: 0.8392
Epoch 37/300
Epoch 00037: val accuracy did not improve from 0.84930
390/390 [============ ] - 137s 352ms/step - loss: 0.2263 - accuracy: 0.9202 - val
loss: 0.7712 - val accuracy: 0.7727
Epoch 38/300
Epoch 00038: val accuracy did not improve from 0.84930
390/390 [============ ] - 137s 351ms/step - loss: 0.2228 - accuracy: 0.9225 - val
loss: 0.5727 - val_accuracy: 0.8386
Epoch 39/300
Epoch 00039: val_accuracy did not improve from 0.84930
390/390 [============ ] - 137s 351ms/step - loss: 0.2155 - accuracy: 0.9225 - val
loss: 0.5577 - val accuracy: 0.8404
Epoch 40/300
Epoch 00040: val accuracy did not improve from 0.84930
390/390 [============ ] - 137s 352ms/step - loss: 0.2102 - accuracy: 0.9266 - val
loss: 0.4954 - val accuracy: 0.8489
Epoch 41/300
Epoch 00041: val accuracy did not improve from 0.84930
loss: 0.6793 - val accuracy: 0.8150
Epoch 42/300
Epoch 00042: val accuracy did not improve from 0.84930
loss: 0.6001 - val accuracy: 0.8385
Epoch 43/300
Epoch 00043: val accuracy did not improve from 0.84930
390/390 [============= ] - 137s 352ms/step - loss: 0.1981 - accuracy: 0.9293 - val
loss: 0.5893 - val accuracy: 0.8382
Epoch 44/300
Epoch 00044: val accuracy did not improve from 0.84930
390/390 [============ ] - 137s 351ms/step - loss: 0.1928 - accuracy: 0.9331 - val
loss: 0.5809 - val_accuracy: 0.8305
Epoch 45/300
Epoch 00045: val accuracy did not improve from 0.84930
390/390 [============ ] - 137s 351ms/step - loss: 0.1903 - accuracy: 0.9325 - val
loss: 0.5722 - val accuracy: 0.8367
Epoch 46/300
Epoch 00046: val accuracy did not improve from 0.84930
390/390 [============ ] - 137s 352ms/step - loss: 0.1849 - accuracy: 0.9349 - val
loss: 0.6191 - val accuracy: 0.8349
Epoch 47/300
Epoch 00047: val accuracy did not improve from 0.84930
loss: 0.5609 - val_accuracy: 0.8484
Epoch 48/300
Epoch 00048: val accuracy did not improve from 0.84930
_loss: 0.5538 - val_accuracy: 0.8435
Epoch 49/300
Epoch 00049: val accuracy improved from 0.84930 to 0.86030, saving model to
```

```
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model6 49-0.86.hdf5
390/390 [================== ] - 138s 354ms/step - loss: 0.1705 - accuracy: 0.9397 - val
loss: 0.5008 - val accuracy: 0.8603
Epoch 50/300
Epoch 00050: val accuracy did not improve from 0.86030
390/390 [============ ] - 137s 351ms/step - loss: 0.1728 - accuracy: 0.9391 - val
loss: 0.5259 - val accuracy: 0.8520
Epoch 51/300
Epoch 00051: val_accuracy did not improve from 0.86030
390/390 [============ ] - 137s 351ms/step - loss: 0.1672 - accuracy: 0.9415 - val
loss: 0.5809 - val accuracy: 0.8392
Epoch 52/300
Epoch 00052: val accuracy did not improve from 0.86030
390/390 [============ ] - 137s 352ms/step - loss: 0.1684 - accuracy: 0.9409 - val
loss: 0.6328 - val accuracy: 0.8346
Epoch 53/300
Epoch 00053: val accuracy improved from 0.86030 to 0.86160, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model6_53-0.86.hdf5
loss: 0.5077 - val accuracy: 0.8616
Epoch 54/300
Epoch 00054: val_accuracy did not improve from 0.86160
390/390 [============ ] - 137s 351ms/step - loss: 0.1585 - accuracy: 0.9440 - val
loss: 0.5327 - val accuracy: 0.8547
Epoch 55/300
Epoch 00055: val accuracy improved from 0.86160 to 0.86380, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model6_55-0.86.hdf5
390/390 [=========== ] - 138s 354ms/step - loss: 0.1549 - accuracy: 0.9452 - val
loss: 0.5082 - val_accuracy: 0.8638
Epoch 56/300
Epoch 00056: val accuracy improved from 0.86380 to 0.87290, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_model6\_56-0.87.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 0.1560 - accuracy: 0.9444 - val
_loss: 0.4334 - val_accuracy: 0.8729
Epoch 57/300
Epoch 00057: val_accuracy did not improve from 0.87290
390/390 [============ ] - 137s 351ms/step - loss: 0.1515 - accuracy: 0.9462 - val
loss: 0.4645 - val accuracy: 0.8719
Epoch 58/300
Epoch 00058: val accuracy did not improve from 0.87290
loss: 0.5121 - val accuracy: 0.8672
Epoch 59/300
Epoch 00059: val accuracy did not improve from 0.87290
loss: 0.5502 - val accuracy: 0.8542
Epoch 60/300
Epoch 00060: val accuracy did not improve from 0.87290
390/390 [============ ] - 137s 351ms/step - loss: 0.1431 - accuracy: 0.9486 - val
loss: 0.5808 - val accuracy: 0.8520
Epoch 61/300
Epoch 00061: val accuracy did not improve from 0.87290
390/390 [=============== ] - 137s 351ms/step - loss: 0.1423 - accuracy: 0.9504 - val
loss: 0.5301 - val accuracy: 0.8596
Epoch 62/300
Epoch 00062: val accuracy did not improve from 0.87290
390/390 [============ ] - 137s 351ms/step - loss: 0.1401 - accuracy: 0.9504 - val
loss: 0.6384 - val_accuracy: 0.8413
Epoch 63/300
Epoch 00063: val_accuracy did not improve from 0.87290
390/390 [============== ] - 137s 351ms/step - loss: 0.1373 - accuracy: 0.9509 - val
loss: 0.5965 - val accuracy: 0.8532
Epoch 64/300
```

```
Epoch 00064: val accuracy did not improve from 0.87290
390/390 [============= ] - 137s 351ms/step - loss: 0.1316 - accuracy: 0.9535 - val
loss: 0.5519 - val accuracy: 0.8591
Epoch 65/300
Epoch 00065: val accuracy did not improve from 0.87290
loss: 0.6794 - val_accuracy: 0.8387
Epoch 66/300
Epoch 00066: val_accuracy did not improve from 0.87290
390/390 [=================== ] - 137s 351ms/step - loss: 0.1321 - accuracy: 0.9532 - val
loss: 0.5336 - val_accuracy: 0.8595
Epoch 67/300
Epoch 00067: val_accuracy did not improve from 0.87290
loss: 0.5444 - val accuracy: 0.8577
Epoch 68/300
Epoch 00068: val accuracy did not improve from 0.87290
loss: 0.5496 - val accuracy: 0.8597
Epoch 69/300
Epoch 00069: val_accuracy did not improve from 0.87290
loss: 0.5733 - val accuracy: 0.8591
Epoch 70/300
Epoch 00070: val accuracy did not improve from 0.87290
390/390 [=============== ] - 137s 352ms/step - loss: 0.1250 - accuracy: 0.9554 - val
loss: 0.5491 - val accuracy: 0.8535
Epoch 71/300
Epoch 00071: val accuracy did not improve from 0.87290
390/390 [============ ] - 137s 351ms/step - loss: 0.1243 - accuracy: 0.9560 - val
loss: 0.5506 - val accuracy: 0.8616
Epoch 72/300
Epoch 00072: val accuracy did not improve from 0.87290
Epoch 00072: ReduceLROnPlateau reducing learning rate to 0.00010000000474974513.
390/390 [============ ] - 137s 351ms/step - loss: 0.1221 - accuracy: 0.9564 - val
loss: 0.7693 - val accuracy: 0.8225
Epoch 73/300
Epoch 00073: val accuracy improved from 0.87290 to 0.89580, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model6_73-0.90.hdf5
loss: 0.4004 - val accuracy: 0.8958
Epoch 74/300
Epoch 00074: val_accuracy did not improve from 0.89580
390/390 [============ ] - 137s 351ms/step - loss: 0.0614 - accuracy: 0.9795 - val
loss: 0.4125 - val_accuracy: 0.8946
Epoch 75/300
Epoch 00075: val_accuracy did not improve from 0.89580
loss: 0.4108 - val accuracy: 0.8951
Epoch 76/300
Epoch 00076: val accuracy improved from 0.89580 to 0.89610, saving model to
/home/ubuntu/Project/my data/CNN CIFAR/checkpoint/CIFAR model6 76-0.90.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 0.0528 - accuracy: 0.9819 - val
loss: 0.4177 - val accuracy: 0.8961
Epoch 77/300
Epoch 00077: val_accuracy did not improve from 0.89610
390/390 [=================== ] - 137s 351ms/step - loss: 0.0473 - accuracy: 0.9841 - val
loss: 0.4338 - val accuracy: 0.8952
Epoch 78/300
Epoch 00078: val accuracy did not improve from 0.89610
390/390 [=============== ] - 137s 351ms/step - loss: 0.0447 - accuracy: 0.9851 - val
```

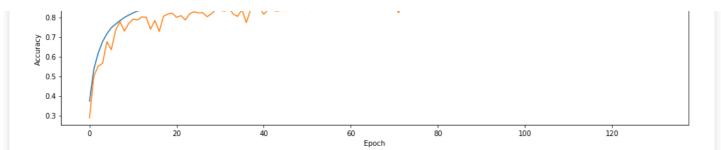
```
loss: 0.4628 - val accuracy: 0.8919
Epoch 79/300
Epoch 00079: val accuracy did not improve from 0.89610
loss: 0.4731 - val accuracy: 0.8919
Epoch 80/300
Epoch 00080: val accuracy did not improve from 0.89610
390/390 [============= ] - 137s 351ms/step - loss: 0.0397 - accuracy: 0.9865 - val
loss: 0.4634 - val_accuracy: 0.8931
Epoch 81/300
Epoch 00081: val_accuracy improved from 0.89610 to 0.89920, saving model to
/home/ubuntu/Project/my\_data/CNN\_CIFAR/checkpoint/CIFAR\_model6\_81-0.90.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 0.0375 - accuracy: 0.9872 - val
loss: 0.4493 - val accuracy: 0.8992
Epoch 82/300
Epoch 00082: val accuracy improved from 0.89920 to 0.89990, saving model to
/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model6_82-0.90.hdf5
390/390 [============ ] - 138s 354ms/step - loss: 0.0364 - accuracy: 0.9880 - val
loss: 0.4682 - val accuracy: 0.8999
Epoch 83/300
Epoch 00083: val_accuracy did not improve from 0.89990
loss: 0.4960 - val accuracy: 0.8919
Epoch 84/300
Epoch 00084: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0371 - accuracy: 0.9872 - val
loss: 0.5114 - val accuracy: 0.8920
Epoch 85/300
Epoch 00085: val accuracy did not improve from 0.89990
loss: 0.4935 - val accuracy: 0.8941
Epoch 86/300
Epoch 00086: val accuracy did not improve from 0.89990
_loss: 0.5134 - val_accuracy: 0.8921
Epoch 87/300
Epoch 00087: val accuracy did not improve from 0.89990
390/390 [================== ] - 137s 351ms/step - loss: 0.0338 - accuracy: 0.9881 - val
loss: 0.4842 - val_accuracy: 0.8960
Epoch 88/300
Epoch 00088: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 352ms/step - loss: 0.0344 - accuracy: 0.9883 - val
loss: 0.5214 - val_accuracy: 0.8917
Epoch 89/300
Epoch 00089: val_accuracy did not improve from 0.89990
loss: 0.5089 - val accuracy: 0.8964
Epoch 90/300
Epoch 00090: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0305 - accuracy: 0.9897 - val
loss: 0.5021 - val accuracy: 0.8962
Epoch 91/300
Epoch 00091: val_accuracy did not improve from 0.89990
loss: 0.5550 - val accuracy: 0.8916
Epoch 92/300
Epoch 00092: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0298 - accuracy: 0.9898 - val
loss: 0.5071 - val_accuracy: 0.8972
Epoch 93/300
Epoch 00093: val accuracy did not improve from 0.89990
```

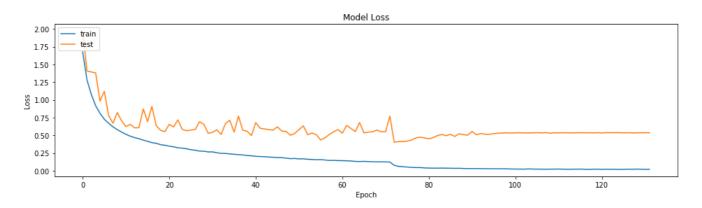
390/390 [============] - 137s 351ms/step - loss: 0.0297 - accuracy: 0.9901 - val

```
loss: 0.5258 - val accuracy: 0.8939
Epoch 94/300
Epoch 00094: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 352ms/step - loss: 0.0286 - accuracy: 0.9903 - val
loss: 0.5146 - val accuracy: 0.8958
Epoch 95/300
Epoch 00095: val accuracy did not improve from 0.89990
390/390 [============= ] - 137s 351ms/step - loss: 0.0277 - accuracy: 0.9909 - val
loss: 0.5141 - val accuracy: 0.8984
Epoch 96/300
Epoch 00096: val accuracy did not improve from 0.89990
390/390 [============== ] - 137s 351ms/step - loss: 0.0276 - accuracy: 0.9907 - val
loss: 0.5237 - val_accuracy: 0.8962
Epoch 97/300
Epoch 00097: val accuracy did not improve from 0.89990
390/390 [=============== ] - 137s 352ms/step - loss: 0.0282 - accuracy: 0.9899 - val
loss: 0.5292 - val accuracy: 0.8979
Epoch 98/300
Epoch 00098: val accuracy did not improve from 0.89990
Epoch 00098: ReduceLROnPlateau reducing learning rate to 1.0000000474974514e-05.
390/390 [=========== ] - 137s 351ms/step - loss: 0.0269 - accuracy: 0.9909 - val
loss: 0.5317 - val accuracy: 0.8951
Epoch 99/300
Epoch 00099: val_accuracy did not improve from 0.89990
loss: 0.5343 - val_accuracy: 0.8962
Epoch 100/300
Epoch 00100: val accuracy did not improve from 0.89990
loss: 0.5313 - val accuracy: 0.8969
Epoch 101/300
Epoch 00101: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0236 - accuracy: 0.9914 - val
_loss: 0.5344 - val_accuracy: 0.8961
Epoch 102/300
Epoch 00102: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0240 - accuracy: 0.9918 - val
loss: 0.5361 - val accuracy: 0.8971
Epoch 103/300
Epoch 00103: val_accuracy did not improve from 0.89990
loss: 0.5332 - val accuracy: 0.8971
Epoch 104/300
Epoch 00104: val_accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0259 - accuracy: 0.9914 - val
loss: 0.5336 - val accuracy: 0.8972
Epoch 105/300
Epoch 00105: val accuracy did not improve from 0.89990
loss: 0.5345 - val accuracy: 0.8969
Epoch 106/300
Epoch 00106: val accuracy did not improve from 0.89990
390/390 [============== ] - 137s 351ms/step - loss: 0.0225 - accuracy: 0.9918 - val
loss: 0.5367 - val accuracy: 0.8972
Epoch 107/300
Epoch 00107: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0216 - accuracy: 0.9931 - val
_loss: 0.5334 - val_accuracy: 0.8979
Epoch 108/300
Epoch 00108: val accuracy did not improve from 0.89990
```

```
loss: 0.5375 - val accuracy: 0.8961
Epoch 109/300
Epoch 00109: val accuracy did not improve from 0.89990
loss: 0.5292 - val accuracy: 0.8986
Epoch 110/300
Epoch 00110: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0221 - accuracy: 0.9925 - val
loss: 0.5354 - val accuracy: 0.8982
Epoch 111/300
Epoch 00111: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0236 - accuracy: 0.9918 - val
_loss: 0.5348 - val_accuracy: 0.8974
Epoch 112/300
Epoch 00112: val_accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0218 - accuracy: 0.9926 - val
loss: 0.5355 - val accuracy: 0.8981
Epoch 113/300
Epoch 00113: val accuracy did not improve from 0.89990
loss: 0.5347 - val accuracy: 0.8971
Epoch 114/300
Epoch 00114: val accuracy did not improve from 0.89990
Epoch 00114: ReduceLROnPlateau reducing learning rate to 1.0000000656873453e-06.
loss: 0.5340 - val accuracy: 0.8975
Epoch 115/300
Epoch 00115: val accuracy did not improve from 0.89990
390/390 [============== ] - 137s 351ms/step - loss: 0.0223 - accuracy: 0.9926 - val
loss: 0.5359 - val_accuracy: 0.8975
Epoch 116/300
Epoch 00116: val_accuracy did not improve from 0.89990
390/390 [============ ] - 137s 352ms/step - loss: 0.0229 - accuracy: 0.9923 - val
loss: 0.5363 - val_accuracy: 0.8971
Epoch 117/300
Epoch 00117: val_accuracy did not improve from 0.89990
390/390 [============== ] - 137s 351ms/step - loss: 0.0217 - accuracy: 0.9928 - val
loss: 0.5350 - val accuracy: 0.8975
Epoch 118/300
Epoch 00118: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0194 - accuracy: 0.9937 - val
loss: 0.5358 - val accuracy: 0.8971
Epoch 119/300
Epoch 00119: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 352ms/step - loss: 0.0216 - accuracy: 0.9925 - val
_loss: 0.5338 - val_accuracy: 0.8978
Epoch 120/300
Epoch 00120: val accuracy did not improve from 0.89990
loss: 0.5363 - val accuracy: 0.8974
Epoch 121/300
Epoch 00121: val accuracy did not improve from 0.89990
390/390 [============== ] - 137s 351ms/step - loss: 0.0201 - accuracy: 0.9932 - val
loss: 0.5330 - val_accuracy: 0.8979
Epoch 122/300
Epoch 00122: val_accuracy did not improve from 0.89990
_loss: 0.5376 - val_accuracy: 0.8965
Epoch 123/300
Epoch 00123: val_accuracy did not improve from 0.89990
```

```
loss: 0.5361 - val accuracy: 0.8972
Epoch 124/300
Epoch 00124: val accuracy did not improve from 0.89990
390/390 [============= ] - 137s 351ms/step - loss: 0.0208 - accuracy: 0.9931 - val
loss: 0.5376 - val_accuracy: 0.8967
Epoch 125/300
Epoch 00125: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0202 - accuracy: 0.9933 - val
loss: 0.5372 - val_accuracy: 0.8971
Epoch 126/300
Epoch 00126: val_accuracy did not improve from 0.89990
loss: 0.5341 - val accuracy: 0.8977
Epoch 127/300
Epoch 00127: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0222 - accuracy: 0.9923 - val
loss: 0.5363 - val accuracy: 0.8966
Epoch 128/300
Epoch 00128: val_accuracy did not improve from 0.89990
loss: 0.5334 - val accuracy: 0.8975
Epoch 129/300
Epoch 00129: val accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0230 - accuracy: 0.9923 - val
loss: 0.5349 - val accuracy: 0.8973
Epoch 130/300
Epoch 00130: val accuracy did not improve from 0.89990
Epoch 00130: ReduceLROnPlateau reducing learning rate to 1.0000001111620805e-07.
390/390 [=============== ] - 137s 351ms/step - loss: 0.0214 - accuracy: 0.9925 - val
loss: 0.5356 - val accuracy: 0.8976
Epoch 131/300
Epoch 00131: val accuracy did not improve from 0.89990
390/390 [============== ] - 137s 352ms/step - loss: 0.0201 - accuracy: 0.9932 - val
loss: 0.5365 - val accuracy: 0.8970
Epoch 132/300
Epoch 00132: val_accuracy did not improve from 0.89990
390/390 [============ ] - 137s 351ms/step - loss: 0.0217 - accuracy: 0.9921 - val
loss: 0.5356 - val accuracy: 0.8976
In [14]:
#history plot for accyracy
plt.figure(figsize=(16,4))
plt.plot(history 6.history['accuracy'])
plt.plot(history 6.history['val accuracy'])
plt.title("Model Accuracy")
plt.xlabel("Epoch")
plt.ylabel("Accuracy")
plt.legend(["train", "test"], loc="upper left")
plt.show()
# history plot for accuracy
plt.figure(figsize=(16,4))
plt.plot(history 6.history["loss"])
plt.plot(history_6.history["val_loss"])
plt.title("Model Loss")
plt.xlabel("Epoch")
plt.ylabel("Loss")
plt.legend(["train", "test"], loc="upper left")
plt.show()
```





In [15]:

```
best_model_6 =
    tf.keras.models.load_model('/home/ubuntu/Project/my_data/CNN_CIFAR/checkpoint/CIFAR_model6_82-0.90
    .hdf5')
    scores = best_model_6.evaluate(X_test, y_test, verbose=1)
    print(scores)
```

10000/10000 [=============] - 9s 854us/sample - loss: 0.4682 - accuracy: 0.8999 [0.46815555404126646, 0.8999]

Summary:

In [16]:

```
from prettytable import PrettyTable

x = PrettyTable()

x.field_names = ["Model", "Conv", 'final_layer', "kernel_size", 'dropout', 'No of parameters', "Tes t Accuracy"]

x.add_row(['model_1','Conv2D','dense', '3x3', '118,918', '0.2','80.83%'])

x.add_row(['model_2','Conv2D','dense', '3x3', '118,918', '0.2','81.33%'])

x.add_row(['model_2','Conv2D','dense', '3x3', '118,918', '0.2','81.33%'])

x.add_row(['model_3','SeparableConvolution2D', 'dense', '5x5', '0.2','258,282', '86.81%'])

x.add_row(['model_4','SeparableConvolution2D', 'dense', '7x7', '0.2','385,002', '87.16%'])

x.add_row(['model_5','SeparableConvolution2D', 'Conv2D', '7x7', '0.2','945,472', '89.21%'])

print(x)
```

++		+		+-		+-		+		+	
+ Model Accuracy	Conv	:	final_layer		kernel_size	1	dropout	N	o of parameters		Test
+											
\mid model_1 \mid	Conv2D	1	dense		3x3		118,918		0.2		3
.83%											
model_2	Conv2D		dense		3x3		118,918	1	0.2		3
.33%											
	SeparableConvolution2D		dense		5x5		0.2		258 , 282		8
.81%											
model_4	SeparableConvolution2D		dense		7x7		0.2		385,002		8
.16%											
model 5	SeparableConvolution2D		Conv2D		7x7		0.2	1	945 , 472		

• After tried, final layer as Conv2D instead of flatten layer without dropout i have achieved test accuracy of 89.999%