CNN_ Cifar10

With image augmentation

codes from below source are used in this casesetudy

https://machinelearningmastery.com/how-to-develop-a-cnn-from-scratch-for-cifar-10-photo-classification/
https://machinelearningmastery.com/how-to-configure-image-data-augmentation-when-training-deep-learning-neural-networks/
https://machinelearningmastery.com/how-to-stop-training-deep-neural-networks-at-the-right-time-using-early-stopping/
https://github.com/moritzhambach/Image-Augmentation-in-Keras-CIFAR-10/blob/master/CNN%20with%20Image%20Augmentation%20(CIFAR10).jpynb

Customised Dense Net with Image Augmentation

https://github.com/ludwikbukowski/mysupernet/blob/master/densenetcustom.py

```
In [0]:
```

```
%tensorflow_version 1.x
```

TensorFlow 1.x selected.

```
In [0]:
```

```
import os
import numpy as np
from keras import backend as K
from keras.datasets import cifar10
from keras.models import Model, Sequential
from keras import models, layers
from keras.layers import Dense, Dropout, Flatten, Input, AveragePooling2D, merge, Activation
from keras.layers import Conv2D, MaxPooling2D, BatchNormalization, DepthwiseConv2D
from keras.layers import Concatenate
from keras.models import load model
from keras.optimizers import SGD, Adam, RMSprop
from keras.preprocessing.image import ImageDataGenerator
from keras.callbacks import ReduceLROnPlateau, ModelCheckpoint, EarlyStopping,
LearningRateScheduler, CSVLogger
from keras.callbacks import Callback
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
import keras
import tensorflow as tf
from keras import backend as k
Using TensorFlow backend.
```

In [0]:

```
%time
datagen = ImageDataGenerator(
    rotation_range=20,
    width_shift_range=0.125,
    height_shift_range=0.125,
    horizontal_flip=True,
    fill_mode='nearest',
    zoom_range=0.10
)
```

```
CPU times: user 3 \mu s, sys: 0 ns, total: 3 \mu s Wall time: 5.96 \mu s
```

```
In [0]:
```

```
num_classes = 10
```

In [0]:

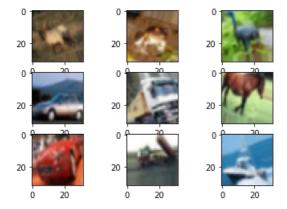
```
# Load CIFAR10 Data
(X_train, y_train), (X_test, y_test) = tf.keras.datasets.cifar10.load_data()
img_height, img_width, channel = X_train.shape[1], X_train.shape[2], X_train.shape[3]

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

In [0]:

```
for X_batch, y_batch in datagen.flow(X_train[:9], y_train[:9], batch_size=9):
    for i in range(0, 9):
        plt.subplot(330 + 1 + i)

        plt.imshow(X_batch[i].astype('uint8'), cmap=plt.get_cmap('prism'))
        plt.show()
        break
```



In [0]:

```
# Dense Block
def denseblock(input, num_filter = 12):
   global compression
    temp = input
    for in range(l):
        BatchNorm = layers.BatchNormalization()(temp)
        relu = layers.Activation('relu')(BatchNorm)
        Conv2D 3 3 = layers.Conv2D(int(num filter*compression), (3,3), use bias=False ,padding='sam
e') (relu)
        #if dropout rate>0:
            #Conv2D 3 3 = layers.Dropout(dropout rate)(Conv2D 3 3)
        concat = layers.Concatenate(axis=-1)([temp,Conv2D 3 3])
        temp = concat
    return temp
## transition Blosck
def transition(input, num_filter = 12):
    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 ava
#output layer
def output_layer(input):
    global compression
    BatchNorm = layers.BatchNormalization()(input)
    relu = layers.Activation('relu')(BatchNorm)
    AvgPooling = layers.AveragePooling2D(pool size=(2,2))(relu)
    #flat = layers.Flatten()(AvgPooling)
    #output = layers.Dense(num classes, activation='softmax')(flat)
    #replcaing Dense layer by conv layer
    #http://cs231n.github.io/convolutional-networks/#convert
    conv_layer = layers.Conv2D(num_classes, (1,1), use_bias=False ,padding='same') (AvgPooling)
    last = layers.GlobalMaxPooling2D()(conv layer)
#https://www.researchgate.net/post/Differences between Global Max Pooling and Global Average pooling
    output = layers.Activation('softmax')(last)
    return output
4
                                                                                                ....▶
In [0]:
num filter = 24
batch_size = 128
\#epochs = 100
compression = 1.041
dropout rate = 0.2
1 = 9
input = layers.Input(shape=(img height, img width, channel,))
First_Conv2D = layers.Conv2D(num_filter, (3,3), activation = 'relu', kernel initializer = 'he norma
l', use bias=False , padding='same') (input)
First_Block = denseblock(First_Conv2D, num_filter)
First Transition = transition(First Block, num filter)
Second Block = denseblock(First Transition, num filter)
Second Transition = transition(Second Block, num filter)
Third Block = denseblock(Second Transition, num filter)
Third Transition = transition(Third Block, num filter)
Last Block = denseblock(Third Transition, num filter)
output = output layer(Last Block)
WARNING:tensorflow:From /tensorflow-
1.15.2/python3.6/tensorflow core/python/ops/resource variable ops.py:1630: calling
BaseResourceVariable.__init__ (from tensorflow.python.ops.resource_variable_ops) with constraint i
s deprecated and will be removed in a future version.
Instructions for updating:
If using Keras pass * constraint arguments to layers.
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-
packages/keras/backend/tensorflow backend.py:4074: The name tf.nn.avg pool is deprecated. Please u
se tf.nn.avg_pool2d instead.
In [0]:
model = Model(inputs=[input], outputs=[output])
model.summary()
Model: "model 1"
```

Layer (type)	Output	Sha	 ре		Param #	Connected to
input_1 (InputLayer)	(None,	32,	32,	3)	0	
conv2d_1 (Conv2D)	(None,	32,	32,	24)	648	input_1[0][0]
batch_normalization_1 (BatchNor	(None,	32,	32,	24)	96	conv2d_1[0][0]
activation_1 (Activation)	(None,	32,	32,	24)	0	batch_normalization_1[0][0]
conv2d_2 (Conv2D)	(None,	32,	32,	24)	5184	activation_1[0][0]

concatenate_1 (Concatenate)	(None,	32,	32,	48)	0	conv2d_1[0][0] conv2d_2[0][0]
batch_normalization_2 (BatchNor	(None,	32,	32,	48)	192	concatenate_1[0][0]
activation_2 (Activation)	(None,	32,	32,	48)	0	batch_normalization_2[0][0]
conv2d_3 (Conv2D)	(None,	32,	32,	24)	10368	activation_2[0][0]
concatenate_2 (Concatenate)	(None,	32,	32,	72)	0	concatenate_1[0][0] conv2d_3[0][0]
batch_normalization_3 (BatchNor	(None,	32,	32,	72)	288	concatenate_2[0][0]
activation_3 (Activation)	(None,	32,	32,	72)	0	batch_normalization_3[0][0]
conv2d_4 (Conv2D)	(None,	32,	32,	24)	15552	activation_3[0][0]
concatenate_3 (Concatenate)	(None,	32,	32,	96)	0	concatenate_2[0][0] conv2d_4[0][0]
batch_normalization_4 (BatchNor	(None,	32,	32,	96)	384	concatenate_3[0][0]
activation_4 (Activation)	(None,	32,	32,	96)	0	batch_normalization_4[0][0]
conv2d_5 (Conv2D)	(None,	32,	32,	24)	20736	activation_4[0][0]
concatenate_4 (Concatenate)	(None,	32,	32,	120)	0	concatenate_3[0][0] conv2d_5[0][0]
batch_normalization_5 (BatchNor	(None,	32,	32,	120)	480	concatenate_4[0][0]
activation_5 (Activation)	(None,	32,	32,	120)	0	batch_normalization_5[0][0]
conv2d_6 (Conv2D)	(None,	32,	32,	24)	25920	activation_5[0][0]
concatenate_5 (Concatenate)	(None,	32,	32,	144)	0	concatenate_4[0][0] conv2d_6[0][0]
batch_normalization_6 (BatchNor	(None,	32,	32,	144)	576	concatenate_5[0][0]
activation_6 (Activation)	(None,	32,	32,	144)	0	batch_normalization_6[0][0]
conv2d_7 (Conv2D)	(None,	32,	32,	24)	31104	activation_6[0][0]
concatenate_6 (Concatenate)	(None,	32,	32,	168)	0	concatenate_5[0][0] conv2d_7[0][0]
batch_normalization_7 (BatchNor	(None,	32,	32,	168)	672	concatenate_6[0][0]
activation_7 (Activation)	(None,	32,	32,	168)	0	batch_normalization_7[0][0]
conv2d_8 (Conv2D)	(None,	32,	32,	24)	36288	activation_7[0][0]
concatenate_7 (Concatenate)	(None,	32,	32,	192)	0	concatenate_6[0][0] conv2d_8[0][0]
batch_normalization_8 (BatchNor	(None,	32,	32,	192)	768	concatenate_7[0][0]
activation_8 (Activation)	(None,	32,	32,	192)	0	batch_normalization_8[0][0]
conv2d_9 (Conv2D)	(None,	32,	32,	24)	41472	activation_8[0][0]
concatenate_8 (Concatenate)	(None,	32,	32,	216)	0	concatenate_7[0][0] conv2d_9[0][0]
batch_normalization_9 (BatchNor	(None,	32,	32,	216)	864	concatenate_8[0][0]
activation_9 (Activation)	(None,	32,	32,	216)	0	batch_normalization_9[0][0]
conv2d_10 (Conv2D)	(None,	32,	32,	24)	46656	activation_9[0][0]
concatenate_9 (Concatenate)	(None,	32,	32,	240)	0	concatenate_8[0][0] conv2d_10[0][0]
batch_normalization_10 (BatchNo	(None,	32,	32,	240)	960	concatenate_9[0][0]

activation_10 (Activation)	(None,	32,	32,	240)	0	batch_normalization_10[0][0]
conv2d_11 (Conv2D)	(None,	32,	32,	24)	5760	activation_10[0][0]
average_pooling2d_1 (AveragePoo	(None,	16,	16,	24)	0	conv2d_11[0][0]
oatch_normalization_11 (BatchNo	(None,	16,	16,	24)	96	average_pooling2d_1[0][0]
activation_11 (Activation)	(None,	16,	16,	24)	0	batch_normalization_11[0][0]
conv2d_12 (Conv2D)	(None,	16,	16,	24)	5184	activation_11[0][0]
concatenate_10 (Concatenate)	(None,	16,	16,	48)	0	average_pooling2d_1[0][0] conv2d_12[0][0]
oatch_normalization_12 (BatchNo	(None,	16,	16,	48)	192	concatenate_10[0][0]
activation_12 (Activation)	(None,	16,	16,	48)	0	batch_normalization_12[0][0]
conv2d_13 (Conv2D)	(None,	16,	16,	24)	10368	activation_12[0][0]
concatenate_11 (Concatenate)	(None,	16,	16,	72)	0	concatenate_10[0][0] conv2d_13[0][0]
oatch_normalization_13 (BatchNo	(None,	16,	16,	72)	288	concatenate_11[0][0]
activation_13 (Activation)	(None,	16,	16,	72)	0	batch_normalization_13[0][0]
conv2d_14 (Conv2D)	(None,	16,	16,	24)	15552	activation_13[0][0]
concatenate_12 (Concatenate)	(None,	16,	16,	96)	0	concatenate_11[0][0] conv2d_14[0][0]
oatch_normalization_14 (BatchNo	(None,	16,	16,	96)	384	concatenate_12[0][0]
activation_14 (Activation)	(None,	16,	16,	96)	0	batch_normalization_14[0][0]
conv2d_15 (Conv2D)	(None,	16,	16,	24)	20736	activation_14[0][0]
concatenate_13 (Concatenate)	(None,	16,	16,	120)	0	concatenate_12[0][0] conv2d_15[0][0]
oatch_normalization_15 (BatchNo	(None,	16,	16,	120)	480	concatenate_13[0][0]
activation_15 (Activation)	(None,	16,	16,	120)	0	batch_normalization_15[0][0]
conv2d_16 (Conv2D)	(None,	16,	16,	24)	25920	activation_15[0][0]
concatenate_14 (Concatenate)	(None,	16,	16,	144)	0	concatenate_13[0][0] conv2d_16[0][0]
oatch_normalization_16 (BatchNo	(None,	16,	16,	144)	576	concatenate_14[0][0]
activation_16 (Activation)	(None,	16,	16,	144)	0	batch_normalization_16[0][0]
conv2d_17 (Conv2D)	(None,	16,	16,	24)	31104	activation_16[0][0]
concatenate_15 (Concatenate)	(None,	16,	16,	168)	0	concatenate_14[0][0] conv2d_17[0][0]
oatch_normalization_17 (BatchNo	(None,	16,	16,	168)	672	concatenate_15[0][0]
activation_17 (Activation)	(None,	16,	16,	168)	0	batch_normalization_17[0][0]
conv2d_18 (Conv2D)	(None,	16,	16,	24)	36288	activation_17[0][0]
concatenate_16 (Concatenate)	(None,	16,	16,	192)	0	concatenate_15[0][0] conv2d_18[0][0]
oatch_normalization_18 (BatchNo	(None,	16,	16,	192)	768	concatenate_16[0][0]
activation_18 (Activation)	(None,	16,	16,	192)	0	batch_normalization_18[0][0]
conv2d_19 (Conv2D)	(None,	16,	16,	24)	41472	activation_18[0][0]
concatenate 17 (Concatenate)	(None,	16,	16,	216)	0	concatenate 16[0][0]

conv2d_19[0][0]

batch_normalization_19 (BatchNo	(None, 16, 16, 216)	864	concatenate_17[0][0]
activation_19 (Activation)	(None, 16, 16, 216)	0	batch_normalization_19[0][0]
conv2d_20 (Conv2D)	(None, 16, 16, 24)	46656	activation_19[0][0]
concatenate_18 (Concatenate)	(None, 16, 16, 240)	0	concatenate_17[0][0] conv2d_20[0][0]
batch_normalization_20 (BatchNo	(None, 16, 16, 240)	960	concatenate_18[0][0]
activation_20 (Activation)	(None, 16, 16, 240)	0	batch_normalization_20[0][0]
conv2d_21 (Conv2D)	(None, 16, 16, 24)	5760	activation_20[0][0]
average_pooling2d_2 (AveragePoo	(None, 8, 8, 24)	0	conv2d_21[0][0]
batch_normalization_21 (BatchNo	(None, 8, 8, 24)	96	average_pooling2d_2[0][0]
activation_21 (Activation)	(None, 8, 8, 24)	0	batch_normalization_21[0][0]
conv2d_22 (Conv2D)	(None, 8, 8, 24)	5184	activation_21[0][0]
concatenate_19 (Concatenate)	(None, 8, 8, 48)	0	average_pooling2d_2[0][0] conv2d_22[0][0]
batch_normalization_22 (BatchNo	(None, 8, 8, 48)	192	concatenate_19[0][0]
activation_22 (Activation)	(None, 8, 8, 48)	0	batch_normalization_22[0][0]
conv2d_23 (Conv2D)	(None, 8, 8, 24)	10368	activation_22[0][0]
concatenate_20 (Concatenate)	(None, 8, 8, 72)	0	concatenate_19[0][0] conv2d_23[0][0]
batch_normalization_23 (BatchNo	(None, 8, 8, 72)	288	concatenate_20[0][0]
activation_23 (Activation)	(None, 8, 8, 72)	0	batch_normalization_23[0][0]
conv2d_24 (Conv2D)	(None, 8, 8, 24)	15552	activation_23[0][0]
concatenate_21 (Concatenate)	(None, 8, 8, 96)	0	concatenate_20[0][0] conv2d_24[0][0]
batch_normalization_24 (BatchNo	(None, 8, 8, 96)	384	concatenate_21[0][0]
activation_24 (Activation)	(None, 8, 8, 96)	0	batch_normalization_24[0][0]
conv2d_25 (Conv2D)	(None, 8, 8, 24)	20736	activation_24[0][0]
concatenate_22 (Concatenate)	(None, 8, 8, 120)	0	concatenate_21[0][0] conv2d_25[0][0]
batch_normalization_25 (BatchNo	(None, 8, 8, 120)	480	concatenate_22[0][0]
activation_25 (Activation)	(None, 8, 8, 120)	0	batch_normalization_25[0][0]
conv2d_26 (Conv2D)	(None, 8, 8, 24)	25920	activation_25[0][0]
concatenate_23 (Concatenate)	(None, 8, 8, 144)	0	concatenate_22[0][0] conv2d_26[0][0]
batch_normalization_26 (BatchNo	(None, 8, 8, 144)	576	concatenate_23[0][0]
activation_26 (Activation)	(None, 8, 8, 144)	0	batch_normalization_26[0][0]
conv2d_27 (Conv2D)	(None, 8, 8, 24)	31104	activation_26[0][0]
concatenate_24 (Concatenate)	(None, 8, 8, 168)	0	concatenate_23[0][0] conv2d_27[0][0]
batch_normalization_27 (BatchNo	(None, 8, 8, 168)	672	concatenate_24[0][0]
activation_27 (Activation)	(None, 8, 8, 168)	0	batch_normalization_27[0][0]

conv2d_28 (Conv2D)	(None,	8,	8,	24)	36288	activation_27[0][0]
concatenate_25 (Concatenate)	(None,	8,	8,	192)	0	concatenate_24[0][0] conv2d_28[0][0]
batch_normalization_28 (BatchNo	(None,	8,	8,	192)	768	concatenate_25[0][0]
activation_28 (Activation)	(None,	8,	8,	192)	0	batch_normalization_28[0][0]
conv2d_29 (Conv2D)	(None,	8,	8,	24)	41472	activation_28[0][0]
concatenate_26 (Concatenate)	(None,	8,	8,	216)	0	concatenate_25[0][0] conv2d_29[0][0]
batch_normalization_29 (BatchNo	(None,	8,	8,	216)	864	concatenate_26[0][0]
activation_29 (Activation)	(None,	8,	8,	216)	0	batch_normalization_29[0][0]
conv2d_30 (Conv2D)	(None,	8,	8,	24)	46656	activation_29[0][0]
concatenate_27 (Concatenate)	(None,	8,	8,	240)	0	concatenate_26[0][0] conv2d_30[0][0]
batch_normalization_30 (BatchNo	(None,	8,	8,	240)	960	concatenate_27[0][0]
activation_30 (Activation)	(None,	8,	8,	240)	0	batch_normalization_30[0][0]
conv2d_31 (Conv2D)	(None,	8,	8,	24)	5760	activation_30[0][0]
average_pooling2d_3 (AveragePoo	(None,	4,	4,	24)	0	conv2d_31[0][0]
oatch_normalization_31 (BatchNo	(None,	4,	4,	24)	96	average_pooling2d_3[0][0]
activation_31 (Activation)	(None,	4,	4,	24)	0	batch_normalization_31[0][0]
conv2d_32 (Conv2D)	(None,	4,	4,	24)	5184	activation_31[0][0]
concatenate_28 (Concatenate)	(None,	4,	4,	48)	0	average_pooling2d_3[0][0] conv2d_32[0][0]
batch_normalization_32 (BatchNo	(None,	4,	4,	48)	192	concatenate_28[0][0]
activation_32 (Activation)	(None,	4,	4,	48)	0	batch_normalization_32[0][0]
conv2d_33 (Conv2D)	(None,	4,	4,	24)	10368	activation_32[0][0]
concatenate_29 (Concatenate)	(None,	4,	4,	72)	0	concatenate_28[0][0] conv2d_33[0][0]
batch_normalization_33 (BatchNo	(None,	4,	4,	72)	288	concatenate_29[0][0]
activation_33 (Activation)	(None,	4,	4,	72)	0	batch_normalization_33[0][0]
conv2d_34 (Conv2D)	(None,	4,	4,	24)	15552	activation_33[0][0]
concatenate_30 (Concatenate)	(None,	4,	4,	96)	0	concatenate_29[0][0] conv2d_34[0][0]
batch_normalization_34 (BatchNo	(None,	4,	4,	96)	384	concatenate_30[0][0]
activation_34 (Activation)	(None,	4,	4,	96)	0	batch_normalization_34[0][0]
conv2d_35 (Conv2D)	(None,	4,	4,	24)	20736	activation_34[0][0]
concatenate_31 (Concatenate)	(None,	4,	4,	120)	0	concatenate_30[0][0] conv2d_35[0][0]
batch_normalization_35 (BatchNo	(None,	4,	4,	120)	480	concatenate_31[0][0]
activation_35 (Activation)	(None,	4,	4,	120)	0	batch_normalization_35[0][0]
conv2d_36 (Conv2D)	(None,	4,	4,	24)	25920	activation_35[0][0]
concatenate_32 (Concatenate)	(None,	4,	4,	144)	0	concatenate_31[0][0] conv2d_36[0][0]
batch normalization 36 (BatchNo	(None,	4,	4,	144)	576	concatenate 32[0][0]

activation_36 (Activation)	(None,	4,	4,	144)	0	batch_normalization_36[0][0]
conv2d_37 (Conv2D)	(None,	4,	4,	24)	31104	activation_36[0][0]
concatenate_33 (Concatenate)	(None,	4,	4,	168)	0	concatenate_32[0][0] conv2d_37[0][0]
batch_normalization_37 (BatchNo	(None,	4,	4,	168)	672	concatenate_33[0][0]
activation_37 (Activation)	(None,	4,	4,	168)	0	batch_normalization_37[0][0]
conv2d_38 (Conv2D)	(None,	4,	4,	24)	36288	activation_37[0][0]
concatenate_34 (Concatenate)	(None,	4,	4,	192)	0	concatenate_33[0][0] conv2d_38[0][0]
batch_normalization_38 (BatchNo	(None,	4,	4,	192)	768	concatenate_34[0][0]
activation_38 (Activation)	(None,	4,	4,	192)	0	batch_normalization_38[0][0]
conv2d_39 (Conv2D)	(None,	4,	4,	24)	41472	activation_38[0][0]
concatenate_35 (Concatenate)	(None,	4,	4,	216)	0	concatenate_34[0][0] conv2d_39[0][0]
oatch_normalization_39 (BatchNo	(None,	4,	4,	216)	864	concatenate_35[0][0]
activation_39 (Activation)	(None,	4,	4,	216)	0	batch_normalization_39[0][0]
conv2d_40 (Conv2D)	(None,	4,	4,	24)	46656	activation_39[0][0]
concatenate_36 (Concatenate)	(None,	4,	4,	240)	0	concatenate_35[0][0] conv2d_40[0][0]
oatch_normalization_40 (BatchNo	(None,	4,	4,	240)	960	concatenate_36[0][0]
activation_40 (Activation)	(None,	4,	4,	240)	0	batch_normalization_40[0][0]
average_pooling2d_4 (AveragePoo	(None,	2,	2,	240)	0	activation_40[0][0]
conv2d_41 (Conv2D)	(None,	2,	2,	10)	2400	average_pooling2d_4[0][0]
global_max_pooling2d_1 (GlobalM	(None,	10)		0	conv2d_41[0][0]
activation_41 (Activation)	(None,	10)		0	global_max_pooling2d_1[0][0]
		===	===		=========	

Total params: 974,568 Trainable params: 964,008 Non-trainable params: 10,560

In [0]:

```
reduce_lr = ReduceLROnPlateau(monitor = 'val_accuracy', factor = 0.1, patience = 5, min_lr = 0.0000
01)
early_stop = EarlyStopping(monitor = "val_loss", patience = 10)
```

In [0]:

```
def decay_fn(epoch, lr):
    if epoch < 50:
        return 0.001
    elif epoch >= 50 and epoch < 75:
        return 0.0001
    else:
        return 0.00001

lr_scheduler = LearningRateScheduler(decay_fn)

csv_logger = CSVLogger('training.log')</pre>
```

```
111 [V].
```

```
filepath = "{epoch:03d}-{val_accuracy:.3f}.hdf5"
model_chkpt = ModelCheckpoint(filepath, monitor = 'val_accuracy', save_best_only=True, verbose = 1)
```

In [0]:

Model: "model 1"

Layer (type) ====================================	Output		-		Param #	Connected to
input_1 (InputLayer)	(None,				0	
conv2d_1 (Conv2D)	(None,	32,	32,	24)	648	input_1[0][0]
batch_normalization_1 (BatchNor	(None,	32,	32,	24)	96	conv2d_1[0][0]
activation_1 (Activation)	(None,	32,	32,	24)	0	batch_normalization_1[0][0]
conv2d_2 (Conv2D)	(None,	32,	32,	24)	5184	activation_1[0][0]
concatenate_1 (Concatenate)	(None,	32,	32,	48)	0	conv2d_1[0][0] conv2d_2[0][0]
batch_normalization_2 (BatchNor	(None,	32,	32,	48)	192	concatenate_1[0][0]
activation_2 (Activation)	(None,	32,	32,	48)	0	batch_normalization_2[0][0]
conv2d_3 (Conv2D)	(None,	32,	32,	24)	10368	activation_2[0][0]
concatenate_2 (Concatenate)	(None,	32,	32,	72)	0	concatenate_1[0][0] conv2d_3[0][0]
batch_normalization_3 (BatchNor	(None,	32,	32,	72)	288	concatenate_2[0][0]
activation_3 (Activation)	(None,	32,	32,	72)	0	batch_normalization_3[0][0]
conv2d_4 (Conv2D)	(None,	32,	32,	24)	15552	activation_3[0][0]
concatenate_3 (Concatenate)	(None,	32,	32,	96)	0	concatenate_2[0][0] conv2d_4[0][0]
batch_normalization_4 (BatchNor	(None,	32,	32,	96)	384	concatenate_3[0][0]
activation_4 (Activation)	(None,	32,	32,	96)	0	batch_normalization_4[0][0]
conv2d_5 (Conv2D)	(None,	32,	32,	24)	20736	activation_4[0][0]
concatenate_4 (Concatenate)	(None,	32,	32,	120)	0	concatenate_3[0][0] conv2d_5[0][0]
batch_normalization_5 (BatchNor	(None,	32,	32,	120)	480	concatenate_4[0][0]
activation_5 (Activation)	(None,	32,	32,	120)	0	batch_normalization_5[0][0]
conv2d_6 (Conv2D)	(None,	32,	32,	24)	25920	activation_5[0][0]
concatenate_5 (Concatenate)	(None,	32,	32,	144)	0	concatenate_4[0][0] conv2d_6[0][0]
batch_normalization_6 (BatchNor	(None,	32,	32,	144)	576	concatenate_5[0][0]
activation_6 (Activation)	(None,	32,	32,	144)	0	batch_normalization_6[0][0]
conv2d_7 (Conv2D)	(None,	32,	32,	24)	31104	activation_6[0][0]
concatenate_6 (Concatenate)	(None,	32,	32,	168)	0	concatenate_5[0][0] conv2d_7[0][0]
batch_normalization_7 (BatchNor	(None,	32,	32,	168)	672	concatenate_6[0][0]

activation_7 (Activation)	(None,	32,	32,	168)	0	batch_normalization_7[0][0]
conv2d_8 (Conv2D)	(None,	32,	32,	24)	36288	activation_7[0][0]
concatenate_7 (Concatenate)	(None,	32,	32,	192)	0	concatenate_6[0][0] conv2d_8[0][0]
batch_normalization_8 (BatchNor	(None,	32,	32,	192)	768	concatenate_7[0][0]
activation_8 (Activation)	(None,	32,	32,	192)	0	batch_normalization_8[0][0]
conv2d_9 (Conv2D)	(None,	32,	32,	24)	41472	activation_8[0][0]
concatenate_8 (Concatenate)	(None,	32,	32,	216)	0	concatenate_7[0][0] conv2d_9[0][0]
batch_normalization_9 (BatchNor	(None,	32,	32,	216)	864	concatenate_8[0][0]
activation_9 (Activation)	(None,	32,	32,	216)	0	batch_normalization_9[0][0]
conv2d_10 (Conv2D)	(None,	32,	32,	24)	46656	activation_9[0][0]
concatenate_9 (Concatenate)	(None,	32,	32,	240)	0	concatenate_8[0][0] conv2d_10[0][0]
batch_normalization_10 (BatchNo	(None,	32,	32,	240)	960	concatenate_9[0][0]
activation_10 (Activation)	(None,	32,	32,	240)	0	batch_normalization_10[0][0]
conv2d_11 (Conv2D)	(None,	32,	32,	24)	5760	activation_10[0][0]
average_pooling2d_1 (AveragePoo	(None,	16,	16,	24)	0	conv2d_11[0][0]
batch_normalization_11 (BatchNo	(None,	16,	16,	24)	96	average_pooling2d_1[0][0]
activation_11 (Activation)	(None,	16,	16,	24)	0	batch_normalization_11[0][0]
conv2d_12 (Conv2D)	(None,	16,	16,	24)	5184	activation_11[0][0]
concatenate_10 (Concatenate)	(None,	16,	16,	48)	0	average_pooling2d_1[0][0] conv2d_12[0][0]
batch_normalization_12 (BatchNo	(None,	16,	16,	48)	192	concatenate_10[0][0]
activation_12 (Activation)	(None,	16,	16,	48)	0	batch_normalization_12[0][0]
conv2d_13 (Conv2D)	(None,	16,	16,	24)	10368	activation_12[0][0]
concatenate_11 (Concatenate)	(None,	16,	16,	72)	0	concatenate_10[0][0] conv2d_13[0][0]
batch_normalization_13 (BatchNo	(None,	16,	16,	72)	288	concatenate_11[0][0]
activation_13 (Activation)	(None,	16,	16,	72)	0	batch_normalization_13[0][0]
conv2d_14 (Conv2D)	(None,	16,	16,	24)	15552	activation_13[0][0]
concatenate_12 (Concatenate)	(None,	16,	16,	96)	0	concatenate_11[0][0] conv2d_14[0][0]
batch_normalization_14 (BatchNo	(None,	16,	16,	96)	384	concatenate_12[0][0]
activation_14 (Activation)	(None,	16,	16,	96)	0	batch_normalization_14[0][0]
conv2d_15 (Conv2D)	(None,	16,	16,	24)	20736	activation_14[0][0]
concatenate_13 (Concatenate)	(None,	16,	16,	120)	0	concatenate_12[0][0] conv2d_15[0][0]
batch_normalization_15 (BatchNo	(None,	16,	16,	120)	480	concatenate_13[0][0]
activation_15 (Activation)	(None,	16,	16,	120)	0	batch_normalization_15[0][0]
conv2d_16 (Conv2D)	(None,	16,	16,	24)	25920	activation_15[0][0]
concatenate_14 (Concatenate)	(None,	16,	16,	144)	0	concatenate_13[0][0] conv2d_16[0][0]

batch_normalization_16 (BatchNo	(None, 16, 16,	144) 576	concatenate_14[0][0]
activation_16 (Activation)	(None, 16, 16,	144) 0	batch_normalization_16[0][0]
conv2d_17 (Conv2D)	(None, 16, 16,	24) 31104	activation_16[0][0]
concatenate_15 (Concatenate)	(None, 16, 16,	168) 0	concatenate_14[0][0] conv2d_17[0][0]
batch_normalization_17 (BatchNo	(None, 16, 16,	168) 672	concatenate_15[0][0]
activation_17 (Activation)	(None, 16, 16,	168) 0	batch_normalization_17[0][0]
conv2d_18 (Conv2D)	(None, 16, 16,	24) 36288	activation_17[0][0]
concatenate_16 (Concatenate)	(None, 16, 16,	192) 0	concatenate_15[0][0] conv2d_18[0][0]
batch_normalization_18 (BatchNo	(None, 16, 16,	192) 768	concatenate_16[0][0]
activation_18 (Activation)	(None, 16, 16,	192) 0	batch_normalization_18[0][0]
conv2d_19 (Conv2D)	(None, 16, 16,	24) 41472	activation_18[0][0]
concatenate_17 (Concatenate)	(None, 16, 16,	216) 0	concatenate_16[0][0] conv2d_19[0][0]
batch_normalization_19 (BatchNo	(None, 16, 16,	216) 864	concatenate_17[0][0]
activation_19 (Activation)	(None, 16, 16,	216) 0	batch_normalization_19[0][0]
conv2d_20 (Conv2D)	(None, 16, 16,	24) 46656	activation_19[0][0]
concatenate_18 (Concatenate)	(None, 16, 16,	240) 0	concatenate_17[0][0] conv2d_20[0][0]
batch_normalization_20 (BatchNo	(None, 16, 16,	240) 960	concatenate_18[0][0]
activation_20 (Activation)	(None, 16, 16,	240) 0	batch_normalization_20[0][0]
conv2d_21 (Conv2D)	(None, 16, 16,	24) 5760	activation_20[0][0]
average_pooling2d_2 (AveragePoo	(None, 8, 8, 24	1) 0	conv2d_21[0][0]
batch_normalization_21 (BatchNo	(None, 8, 8, 24	96	average_pooling2d_2[0][0]
activation_21 (Activation)	(None, 8, 8, 24) 0	batch_normalization_21[0][0]
conv2d_22 (Conv2D)	(None, 8, 8, 24	5184	activation_21[0][0]
concatenate_19 (Concatenate)	(None, 8, 8, 48	3) 0	average_pooling2d_2[0][0] conv2d_22[0][0]
batch_normalization_22 (BatchNo	(None, 8, 8, 48	3) 192	concatenate_19[0][0]
activation_22 (Activation)	(None, 8, 8, 48	3) 0	batch_normalization_22[0][0]
conv2d_23 (Conv2D)	(None, 8, 8, 24	10368	activation_22[0][0]
concatenate_20 (Concatenate)	(None, 8, 8, 72	2) 0	concatenate_19[0][0] conv2d_23[0][0]
batch_normalization_23 (BatchNo	(None, 8, 8, 72	2) 288	concatenate_20[0][0]
activation_23 (Activation)	(None, 8, 8, 72	2) 0	batch_normalization_23[0][0]
conv2d_24 (Conv2D)	(None, 8, 8, 24	15552	activation_23[0][0]
concatenate_21 (Concatenate)	(None, 8, 8, 96	5) 0	concatenate_20[0][0] conv2d_24[0][0]
	(None, 8, 8, 96	384	concatenate_21[0][0]
activation_24 (Activation)	(None, 8, 8, 96	5) 0	batch_normalization_24[0][0]
conv2d_25 (Conv2D)	(None, 8, 8, 24	20736	activation_24[0][0]

concatenate_22 (Concatenate)	(None,	8,	8,	120)	0	concatenate_21[0][0]
						conv2d_25[0][0]
batch_normalization_25 (BatchNo	(None,	8,	8,	120)	480	concatenate_22[0][0]
activation_25 (Activation)	(None,	8,	8,	120)	0	batch_normalization_25[0][0]
conv2d_26 (Conv2D)	(None,	8,	8,	24)	25920	activation_25[0][0]
concatenate_23 (Concatenate)	(None,	8,	8,	144)	0	concatenate_22[0][0] conv2d_26[0][0]
batch_normalization_26 (BatchNo	(None,	8,	8,	144)	576	concatenate_23[0][0]
activation_26 (Activation)	(None,	8,	8,	144)	0	batch_normalization_26[0][0]
conv2d_27 (Conv2D)	(None,	8,	8,	24)	31104	activation_26[0][0]
concatenate_24 (Concatenate)	(None,	8,	8,	168)	0	concatenate_23[0][0] conv2d_27[0][0]
batch_normalization_27 (BatchNo	(None,	8,	8,	168)	672	concatenate_24[0][0]
activation_27 (Activation)	(None,	8,	8,	168)	0	batch_normalization_27[0][0]
conv2d_28 (Conv2D)	(None,	8,	8,	24)	36288	activation_27[0][0]
concatenate_25 (Concatenate)	(None,	8,	8,	192)	0	concatenate_24[0][0] conv2d_28[0][0]
batch_normalization_28 (BatchNo	(None,	8,	8,	192)	768	concatenate_25[0][0]
activation_28 (Activation)	(None,	8,	8,	192)	0	batch_normalization_28[0][0]
conv2d_29 (Conv2D)	(None,	8,	8,	24)	41472	activation_28[0][0]
concatenate_26 (Concatenate)	(None,	8,	8,	216)	0	concatenate_25[0][0] conv2d_29[0][0]
batch_normalization_29 (BatchNo	(None,	8,	8,	216)	864	concatenate_26[0][0]
activation_29 (Activation)	(None,	8,	8,	216)	0	batch_normalization_29[0][0]
conv2d_30 (Conv2D)	(None,	8,	8,	24)	46656	activation_29[0][0]
concatenate_27 (Concatenate)	(None,	8,	8,	240)	0	concatenate_26[0][0] conv2d_30[0][0]
batch_normalization_30 (BatchNo	(None,	8,	8,	240)	960	concatenate_27[0][0]
activation_30 (Activation)	(None,	8,	8,	240)	0	batch_normalization_30[0][0]
conv2d_31 (Conv2D)	(None,	8,	8,	24)	5760	activation_30[0][0]
average_pooling2d_3 (AveragePoo	(None,	4,	4,	24)	0	conv2d_31[0][0]
batch_normalization_31 (BatchNo	(None,	4,	4,	24)	96	average_pooling2d_3[0][0]
activation_31 (Activation)	(None,	4,	4,	24)	0	batch_normalization_31[0][0]
conv2d_32 (Conv2D)	(None,	4,	4,	24)	5184	activation_31[0][0]
concatenate_28 (Concatenate)	(None,	4,	4,	48)	0	average_pooling2d_3[0][0] conv2d_32[0][0]
batch_normalization_32 (BatchNo	(None,	4,	4,	48)	192	concatenate_28[0][0]
activation_32 (Activation)	(None,	4,	4,	48)	0	batch_normalization_32[0][0]
conv2d_33 (Conv2D)	(None,	4,	4,	24)	10368	activation_32[0][0]
concatenate_29 (Concatenate)	(None,	4,	4,	72)	0	concatenate_28[0][0] conv2d_33[0][0]
batch_normalization_33 (BatchNo	(None,	4,	4,	72)	288	concatenate_29[0][0]

activation_33 (Activation)	(None,	4,	4,	72)	0	batch_normalization_33[0][0]
conv2d_34 (Conv2D)	(None,	4,	4,	24)	15552	activation_33[0][0]
concatenate_30 (Concatenate)	(None,	4,	4,	96)	0	concatenate_29[0][0] conv2d_34[0][0]
batch_normalization_34 (BatchNo	(None,	4,	4,	96)	384	concatenate_30[0][0]
activation_34 (Activation)	(None,	4,	4,	96)	0	batch_normalization_34[0][0]
conv2d_35 (Conv2D)	(None,	4,	4,	24)	20736	activation_34[0][0]
concatenate_31 (Concatenate)	(None,	4,	4,	120)	0	concatenate_30[0][0] conv2d_35[0][0]
batch_normalization_35 (BatchNo	(None,	4,	4,	120)	480	concatenate_31[0][0]
activation_35 (Activation)	(None,	4,	4,	120)	0	batch_normalization_35[0][0]
conv2d_36 (Conv2D)	(None,	4,	4,	24)	25920	activation_35[0][0]
concatenate_32 (Concatenate)	(None,	4,	4,	144)	0	concatenate_31[0][0] conv2d_36[0][0]
batch_normalization_36 (BatchNo	(None,	4,	4,	144)	576	concatenate_32[0][0]
activation_36 (Activation)	(None,	4,	4,	144)	0	batch_normalization_36[0][0]
conv2d_37 (Conv2D)	(None,	4,	4,	24)	31104	activation_36[0][0]
concatenate_33 (Concatenate)	(None,	4,	4,	168)	0	concatenate_32[0][0] conv2d_37[0][0]
batch_normalization_37 (BatchNo	(None,	4,	4,	168)	672	concatenate_33[0][0]
activation_37 (Activation)	(None,	4,	4,	168)	0	batch_normalization_37[0][0]
conv2d_38 (Conv2D)	(None,	4,	4,	24)	36288	activation_37[0][0]
concatenate_34 (Concatenate)	(None,	4,	4,	192)	0	concatenate_33[0][0] conv2d_38[0][0]
batch_normalization_38 (BatchNo	(None,	4,	4,	192)	768	concatenate_34[0][0]
activation_38 (Activation)	(None,	4,	4,	192)	0	batch_normalization_38[0][0]
conv2d_39 (Conv2D)	(None,	4,	4,	24)	41472	activation_38[0][0]
concatenate_35 (Concatenate)	(None,	4,	4,	216)	0	concatenate_34[0][0] conv2d_39[0][0]
batch_normalization_39 (BatchNo	(None,	4,	4,	216)	864	concatenate_35[0][0]
activation_39 (Activation)	(None,	4,	4,	216)	0	batch_normalization_39[0][0]
conv2d_40 (Conv2D)	(None,	4,	4,	24)	46656	activation_39[0][0]
concatenate_36 (Concatenate)	(None,	4,	4,	240)	0	concatenate_35[0][0] conv2d_40[0][0]
batch_normalization_40 (BatchNo	(None,	4,	4,	240)	960	concatenate_36[0][0]
activation_40 (Activation)	(None,	4,	4,	240)	0	batch_normalization_40[0][0]
average_pooling2d_4 (AveragePoo	(None,	2,	2,	240)	0	activation_40[0][0]
conv2d_41 (Conv2D)	(None,	2,	2,	10)	2400	average_pooling2d_4[0][0]
global_max_pooling2d_1 (GlobalM	(None,	10))		0	conv2d_41[0][0]
activation_41 (Activation)	(None,	10))		0	global_max_pooling2d_1[0][0]
Total params: 974.568	======			======		

Total params: 974,568
Trainable params: 964,008
Non-trainable params: 10,560

```
In [0]:
```

```
history = model.fit_generator(
    datagen.flow(X_train, y_train, batch size=batch size),
    steps per epoch=(len(X train)/batch size)*5,
    epochs=150.
    verbose = 2,
    validation_data=(X_test, y_test),
    callbacks = [model chkpt, early stop]
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-
packages/keras/backend/tensorflow backend.py:422: The name tf.global variables is deprecated. Plea
se use tf.compat.v1.global variables instead.
Epoch 1/150
 - 246s - loss: 1.0587 - accuracy: 0.6205 - val loss: 0.9501 - val accuracy: 0.6849
Epoch 00001: val accuracy improved from -inf to 0.68490, saving model to 001-0.685.hdf5
Epoch 2/150
- 232s - loss: 0.6046 - accuracy: 0.7899 - val loss: 0.9797 - val accuracy: 0.7065
Epoch 00002: val accuracy improved from 0.68490 to 0.70650, saving model to 002-0.706.hdf5
Epoch 3/150
- 232s - loss: 0.4676 - accuracy: 0.8376 - val loss: 0.5694 - val accuracy: 0.8145
Epoch 00003: val accuracy improved from 0.70650 to 0.81450, saving model to 003-0.814.hdf5
Epoch 4/150
- 232s - loss: 0.3899 - accuracy: 0.8647 - val loss: 0.4849 - val accuracy: 0.8446
Epoch 00004: val accuracy improved from 0.81450 to 0.84460, saving model to 004-0.845.hdf5
- 232s - loss: 0.3370 - accuracy: 0.8820 - val loss: 0.4464 - val accuracy: 0.8512
Epoch 00005: val accuracy improved from 0.84460 to 0.85120, saving model to 005-0.851.hdf5
Epoch 6/150
 - 231s - loss: 0.2964 - accuracy: 0.8967 - val loss: 0.5265 - val accuracy: 0.8343
Epoch 00006: val_accuracy did not improve from 0.85120
Epoch 7/150
 - 231s - loss: 0.2606 - accuracy: 0.9091 - val loss: 0.3698 - val accuracy: 0.8822
Epoch 00007: val accuracy improved from 0.85120 to 0.88220, saving model to 007-0.882.hdf5
Epoch 8/150
- 231s - loss: 0.2356 - accuracy: 0.9183 - val loss: 0.4749 - val accuracy: 0.8586
Epoch 00008: val_accuracy did not improve from 0.88220
Epoch 9/150
 - 233s - loss: 0.2140 - accuracy: 0.9246 - val loss: 0.5490 - val accuracy: 0.8462
Epoch 00009: val accuracy did not improve from 0.88220
Epoch 10/150
 - 234s - loss: 0.1963 - accuracy: 0.9310 - val loss: 0.3899 - val accuracy: 0.8797
Epoch 00010: val accuracy did not improve from 0.88220
Epoch 11/150
 - 233s - loss: 0.1792 - accuracy: 0.9365 - val loss: 0.4276 - val accuracy: 0.8777
Epoch 00011: val accuracy did not improve from 0.88220
Epoch 12/150
 - 232s - loss: 0.1671 - accuracy: 0.9412 - val_loss: 0.3884 - val_accuracy: 0.8864
Epoch 00012: val accuracy improved from 0.88220 to 0.88640, saving model to 012-0.886.hdf5
Epoch 13/150
 - 233s - loss: 0.1554 - accuracy: 0.9449 - val loss: 0.4333 - val accuracy: 0.8721
Epoch 00013: val accuracy did not improve from 0.88640
Epoch 14/150
 - 232s - loss: 0.1430 - accuracy: 0.9493 - val loss: 0.3686 - val accuracy: 0.8935
Epoch 00014: val_accuracy improved from 0.88640 to 0.89350, saving model to 014-0.893.hdf5
Epoch 15/150
- 232s - loss: 0.1340 - accuracy: 0.9522 - val loss: 0.3495 - val accuracy: 0.8967
```

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Epoch UUU15: Val accuracy improved from U.8935U to U.896/U, saving model to U15-U.89/.ndf5
Epoch 16/150
 - 232s - loss: 0.1242 - accuracy: 0.9559 - val loss: 0.3584 - val accuracy: 0.8995
Epoch 00016: val accuracy improved from 0.89670 to 0.89950, saving model to 016-0.900.hdf5
Epoch 17/150
- 233s - loss: 0.1176 - accuracy: 0.9584 - val loss: 0.3525 - val accuracy: 0.9027
Epoch 00017: val accuracy improved from 0.89950 to 0.90270, saving model to 017-0.903.hdf5
Epoch 18/150
 - 232s - loss: 0.1129 - accuracy: 0.9599 - val loss: 0.4571 - val accuracy: 0.8809
Epoch 00018: val accuracy did not improve from 0.90270
Epoch 19/150
- 232s - loss: 0.1051 - accuracy: 0.9625 - val loss: 0.4063 - val accuracy: 0.8971
Epoch 00019: val accuracy did not improve from 0.90270
Epoch 20/150
 - 232s - loss: 0.1005 - accuracy: 0.9645 - val loss: 0.3349 - val accuracy: 0.9074
Epoch 00020: val_accuracy improved from 0.90270 to 0.90740, saving model to 020-0.907.hdf5
- 232s - loss: 0.0963 - accuracy: 0.9661 - val_loss: 0.3973 - val_accuracy: 0.8994
Epoch 00021: val accuracy did not improve from 0.90740
Epoch 22/150
- 232s - loss: 0.0922 - accuracy: 0.9675 - val loss: 0.3651 - val accuracy: 0.9028
Epoch 00022: val accuracy did not improve from 0.90740
Epoch 23/150
 - 232s - loss: 0.0872 - accuracy: 0.9689 - val loss: 0.3826 - val accuracy: 0.9004
Epoch 00023: val accuracy did not improve from 0.90740
Epoch 24/150
- 232s - loss: 0.0840 - accuracy: 0.9704 - val loss: 0.4157 - val accuracy: 0.8979
Epoch 00024: val\_accuracy did not improve from 0.90740
Epoch 25/150
 - 234s - loss: 0.0801 - accuracy: 0.9714 - val loss: 0.4416 - val accuracy: 0.8958
Epoch 00025: val accuracy did not improve from 0.90740
Epoch 26/150
- 234s - loss: 0.0781 - accuracy: 0.9726 - val loss: 0.3744 - val accuracy: 0.9085
Epoch 00026: val accuracy improved from 0.90740 to 0.90850, saving model to 026-0.909.hdf5
Epoch 27/150
- 233s - loss: 0.0750 - accuracy: 0.9734 - val loss: 0.4448 - val accuracy: 0.8954
Epoch 00027: val accuracy did not improve from 0.90850
Epoch 28/150
 - 232s - loss: 0.0714 - accuracy: 0.9746 - val_loss: 0.4535 - val_accuracy: 0.8972
Epoch 00028: val accuracy did not improve from 0.90850
Epoch 29/150
 - 232s - loss: 0.0696 - accuracy: 0.9756 - val loss: 0.5164 - val accuracy: 0.8848
Epoch 00029: val accuracy did not improve from 0.90850
Epoch 30/150
- 232s - loss: 0.0664 - accuracy: 0.9766 - val loss: 0.5058 - val accuracy: 0.8879
Epoch 00030: val accuracy did not improve from 0.90850
In [0]:
score = model.evaluate(X test, y test, verbose=1)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
10000/10000 [=========== ] - 4s 417us/step
Test loss: 0.5058346040025353
Test accuracy: 0.8878999948501587
In [0]:
```

reduce lr = ReduceLROnPlateau(monitor = 'val loss', factor = 0.1, patience = 5, min lr = 0.00000

```
1)
early stop = EarlyStopping(monitor = "val loss", patience = 10)
def decay fn(epoch, lr):
   if epoch < 50:
        return 0.001
    elif epoch >= 50 and epoch < 75:</pre>
       return 0.0001
    else:
        return 0.00001
lr scheduler = LearningRateScheduler(decay fn)
csv_logger = CSVLogger('training.log')
filepath = "{epoch:03d}-{val_accuracy:.3f}.hdf5"
model chkpt = ModelCheckpoint(filepath, monitor = 'val accuracy', save best only=True, verbose = 1)
#checkpoint = ModelCheckpoint('gdrive/My Drive/cnnoncifar/models/model-{epoch:03d}-{acc:03f}-{val
acc:03f}.h5',
                                       # verbose=1, monitor='val acc', save best only=True, mode='aut
model.load_weights('026-0.909.hdf5')
model.compile(loss='categorical crossentropy',
              optimizer=Adam(),
              metrics=['accuracy'])
# model.fit(xtrain, y_train,
                      batch size=batch size,
                      epochs=epochs,
                      verbose=1,
                      validation data=(xtest, y test))
print(model.summary())
model.fit generator(
   datagen.flow(X_train, y_train, batch_size=batch_size),
   steps_per_epoch=(len(X_train)/batch_size)*5,
    epochs=150, verbose = 1,initial_epoch = 30,
    validation_data=(X_test, y_test),
    callbacks=[model_chkpt,early_stop])
```

Model: "model 1"

Layer (type)	Output	Shap	e		Param #	Connected to
input_1 (InputLayer)	(None,	32 ,	32,	3)	0	
conv2d_1 (Conv2D)	(None,	32,	32,	24)	648	input_1[0][0]
batch_normalization_1 (BatchNor	(None,	32,	32,	24)	96	conv2d_1[0][0]
activation_1 (Activation)	(None,	32,	32,	24)	0	batch_normalization_1[0][0]
conv2d_2 (Conv2D)	(None,	32,	32,	24)	5184	activation_1[0][0]
concatenate_1 (Concatenate)	(None,	32,	32,	48)	0	conv2d_1[0][0] conv2d_2[0][0]
batch_normalization_2 (BatchNor	(None,	32,	32,	48)	192	concatenate_1[0][0]
activation_2 (Activation)	(None,	32,	32,	48)	0	batch_normalization_2[0][0]
conv2d_3 (Conv2D)	(None,	32,	32,	24)	10368	activation_2[0][0]
concatenate_2 (Concatenate)	(None,	32,	32,	72)	0	concatenate_1[0][0] conv2d_3[0][0]
batch_normalization_3 (BatchNor	(None,	32,	32,	72)	288	concatenate_2[0][0]

activation_3 (Activation)	(None,	32,	32,	72)	0	batch_normalization_3[0][0]
conv2d_4 (Conv2D)	(None,	32,	32,	24)	15552	activation_3[0][0]
concatenate_3 (Concatenate)	(None,	32,	32,	96)	0	concatenate_2[0][0] conv2d_4[0][0]
batch_normalization_4 (BatchNor	(None,	32,	32,	96)	384	concatenate_3[0][0]
activation_4 (Activation)	(None,	32,	32,	96)	0	batch_normalization_4[0][0]
conv2d_5 (Conv2D)	(None,	32,	32,	24)	20736	activation_4[0][0]
concatenate_4 (Concatenate)	(None,	32,	32,	120)	0	concatenate_3[0][0] conv2d_5[0][0]
batch_normalization_5 (BatchNor	(None,	32,	32,	120)	480	concatenate_4[0][0]
activation_5 (Activation)	(None,	32,	32,	120)	0	batch_normalization_5[0][0]
conv2d_6 (Conv2D)	(None,	32,	32,	24)	25920	activation_5[0][0]
concatenate_5 (Concatenate)	(None,	32,	32,	144)	0	concatenate_4[0][0] conv2d_6[0][0]
<pre>batch_normalization_6 (BatchNor</pre>	(None,	32,	32,	144)	576	concatenate_5[0][0]
activation_6 (Activation)	(None,	32,	32,	144)	0	batch_normalization_6[0][0]
conv2d_7 (Conv2D)	(None,	32,	32,	24)	31104	activation_6[0][0]
concatenate_6 (Concatenate)	(None,	32,	32,	168)	0	concatenate_5[0][0] conv2d_7[0][0]
batch_normalization_7 (BatchNor	(None,	32,	32,	168)	672	concatenate_6[0][0]
activation_7 (Activation)	(None,	32,	32,	168)	0	batch_normalization_7[0][0]
conv2d_8 (Conv2D)	(None,	32,	32,	24)	36288	activation_7[0][0]
concatenate_7 (Concatenate)	(None,	32,	32,	192)	0	concatenate_6[0][0] conv2d_8[0][0]
batch_normalization_8 (BatchNor	(None,	32,	32,	192)	768	concatenate_7[0][0]
activation_8 (Activation)	(None,	32,	32,	192)	0	batch_normalization_8[0][0]
conv2d_9 (Conv2D)	(None,	32,	32,	24)	41472	activation_8[0][0]
concatenate_8 (Concatenate)	(None,	32,	32,	216)	0	concatenate_7[0][0] conv2d_9[0][0]
batch_normalization_9 (BatchNor	(None,	32,	32,	216)	864	concatenate_8[0][0]
activation_9 (Activation)	(None,	32,	32,	216)	0	batch_normalization_9[0][0]
conv2d_10 (Conv2D)	(None,	32,	32,	24)	46656	activation_9[0][0]
concatenate_9 (Concatenate)	(None,	32,	32,	240)	0	concatenate_8[0][0] conv2d_10[0][0]
batch_normalization_10 (BatchNo	(None,	32,	32,	240)	960	concatenate_9[0][0]
activation_10 (Activation)	(None,	32,	32,	240)	0	batch_normalization_10[0][0]
conv2d_11 (Conv2D)	(None,	32,	32,	24)	5760	activation_10[0][0]
average_pooling2d_1 (AveragePoo	(None,	16,	16,	24)	0	conv2d_11[0][0]
batch_normalization_11 (BatchNo	(None,	16,	16,	24)	96	average_pooling2d_1[0][0]
activation_11 (Activation)	(None,	16,	16,	24)	0	batch_normalization_11[0][0]
conv2d_12 (Conv2D)	(None,	16,	16,	24)	5184	activation_11[0][0]
concatenate_10 (Concatenate)	(None,	16,	16,	48)	0	average_pooling2d_1[0][0] conv2d_12[0][0]

batch_normalization_12 (BatchNo	(None,	16,	16,	48)	192	concatenate_10[0][0]
activation_12 (Activation)	(None,	16,	16,	48)	0	batch_normalization_12[0][0]
conv2d_13 (Conv2D)	(None,	16,	16,	24)	10368	activation_12[0][0]
concatenate_11 (Concatenate)	(None,	16,	16,	72)	0	concatenate_10[0][0] conv2d_13[0][0]
batch_normalization_13 (BatchNo	(None,	16,	16,	72)	288	concatenate_11[0][0]
activation_13 (Activation)	(None,	16,	16,	72)	0	batch_normalization_13[0][0]
conv2d_14 (Conv2D)	(None,	16,	16,	24)	15552	activation_13[0][0]
concatenate_12 (Concatenate)	(None,	16,	16,	96)	0	concatenate_11[0][0] conv2d_14[0][0]
batch_normalization_14 (BatchNo	(None,	16,	16,	96)	384	concatenate_12[0][0]
activation_14 (Activation)	(None,	16,	16,	96)	0	batch_normalization_14[0][0]
conv2d_15 (Conv2D)	(None,	16,	16,	24)	20736	activation_14[0][0]
concatenate_13 (Concatenate)	(None,	16,	16,	120)	0	concatenate_12[0][0] conv2d_15[0][0]
batch_normalization_15 (BatchNo	(None,	16,	16,	120)	480	concatenate_13[0][0]
activation_15 (Activation)	(None,	16,	16,	120)	0	batch_normalization_15[0][0]
conv2d_16 (Conv2D)	(None,	16,	16,	24)	25920	activation_15[0][0]
concatenate_14 (Concatenate)	(None,	16,	16,	144)	0	concatenate_13[0][0] conv2d_16[0][0]
batch_normalization_16 (BatchNo	(None,	16,	16,	144)	576	concatenate_14[0][0]
activation_16 (Activation)	(None,	16,	16,	144)	0	batch_normalization_16[0][0]
conv2d_17 (Conv2D)	(None,	16,	16,	24)	31104	activation_16[0][0]
concatenate_15 (Concatenate)	(None,	16,	16,	168)	0	concatenate_14[0][0] conv2d_17[0][0]
batch_normalization_17 (BatchNo	(None,	16,	16,	168)	672	concatenate_15[0][0]
activation_17 (Activation)	(None,	16,	16,	168)	0	batch_normalization_17[0][0]
conv2d_18 (Conv2D)	(None,	16,	16,	24)	36288	activation_17[0][0]
concatenate_16 (Concatenate)	(None,	16,	16,	192)	0	concatenate_15[0][0] conv2d_18[0][0]
batch_normalization_18 (BatchNo	(None,	16,	16,	192)	768	concatenate_16[0][0]
activation_18 (Activation)	(None,	16,	16,	192)	0	batch_normalization_18[0][0]
conv2d_19 (Conv2D)	(None,	16,	16,	24)	41472	activation_18[0][0]
concatenate_17 (Concatenate)	(None,	16,	16,	216)	0	concatenate_16[0][0] conv2d_19[0][0]
batch_normalization_19 (BatchNo	(None,	16,	16,	216)	864	concatenate_17[0][0]
activation_19 (Activation)	(None,	16,	16,	216)	0	batch_normalization_19[0][0]
conv2d_20 (Conv2D)	(None,	16,	16,	24)	46656	activation_19[0][0]
concatenate_18 (Concatenate)	(None,	16,	16,	240)	0	concatenate_17[0][0] conv2d_20[0][0]
batch_normalization_20 (BatchNo	(None,	16,	16,	240)	960	concatenate_18[0][0]
activation_20 (Activation)	(None,	16,	16,	240)	0	batch_normalization_20[0][0]

conv2d_21 (Conv2D)	(None,	16,	. 10	6, 24)	5760	activation_20[0][0]
average_pooling2d_2 (AveragePoo	(None,	8,	8,	24)	0	conv2d_21[0][0]
batch_normalization_21 (BatchNo	(None,	8,	8,	24)	96	average_pooling2d_2[0][0]
activation_21 (Activation)	(None,	8,	8,	24)	0	batch_normalization_21[0][0]
conv2d_22 (Conv2D)	(None,	8,	8,	24)	5184	activation_21[0][0]
concatenate_19 (Concatenate)	(None,	8,	8,	48)	0	average_pooling2d_2[0][0] conv2d_22[0][0]
batch_normalization_22 (BatchNo	(None,	8,	8,	48)	192	concatenate_19[0][0]
activation_22 (Activation)	(None,	8,	8,	48)	0	batch_normalization_22[0][0]
conv2d_23 (Conv2D)	(None,	8,	8,	24)	10368	activation_22[0][0]
concatenate_20 (Concatenate)	(None,	8,	8,	72)	0	concatenate_19[0][0] conv2d_23[0][0]
batch_normalization_23 (BatchNo	(None,	8,	8,	72)	288	concatenate_20[0][0]
activation_23 (Activation)	(None,	8,	8,	72)	0	batch_normalization_23[0][0]
conv2d_24 (Conv2D)	(None,	8,	8,	24)	15552	activation_23[0][0]
concatenate_21 (Concatenate)	(None,	8,	8,	96)	0	concatenate_20[0][0] conv2d_24[0][0]
batch_normalization_24 (BatchNo	(None,	8,	8,	96)	384	concatenate_21[0][0]
activation_24 (Activation)	(None,	8,	8,	96)	0	batch_normalization_24[0][0]
conv2d_25 (Conv2D)	(None,	8,	8,	24)	20736	activation_24[0][0]
concatenate_22 (Concatenate)	(None,	8,	8,	120)	0	concatenate_21[0][0] conv2d_25[0][0]
batch_normalization_25 (BatchNo	(None,	8,	8,	120)	480	concatenate_22[0][0]
activation_25 (Activation)	(None,	8,	8,	120)	0	batch_normalization_25[0][0]
conv2d_26 (Conv2D)	(None,	8,	8,	24)	25920	activation_25[0][0]
concatenate_23 (Concatenate)	(None,	8,	8,	144)	0	concatenate_22[0][0] conv2d_26[0][0]
batch_normalization_26 (BatchNo	(None,	8,	8,	144)	576	concatenate_23[0][0]
activation_26 (Activation)	(None,	8,	8,	144)	0	batch_normalization_26[0][0]
conv2d_27 (Conv2D)	(None,	8,	8,	24)	31104	activation_26[0][0]
concatenate_24 (Concatenate)	(None,	8,	8,	168)	0	concatenate_23[0][0] conv2d_27[0][0]
batch_normalization_27 (BatchNo	(None,	8,	8,	168)	672	concatenate_24[0][0]
activation_27 (Activation)	(None,	8,	8,	168)	0	batch_normalization_27[0][0]
conv2d_28 (Conv2D)	(None,	8,	8,	24)	36288	activation_27[0][0]
concatenate_25 (Concatenate)	(None,	8,	8,	192)	0	concatenate_24[0][0] conv2d_28[0][0]
batch_normalization_28 (BatchNo	(None,	8,	8,	192)	768	concatenate_25[0][0]
activation_28 (Activation)	(None,	8,	8,	192)	0	batch_normalization_28[0][0]
conv2d_29 (Conv2D)	(None,	8,	8,	24)	41472	activation_28[0][0]
concatenate_26 (Concatenate)	(None,	8,	8,	216)	0	concatenate_25[0][0] conv2d_29[0][0]
batch_normalization_29 (BatchNo	(None,	8,	8,	216)	864	concatenate_26[0][0]

activation_29 (Activation)	(None,	8,	8,	216)	0	batch_normalization_29[0][0]
conv2d_30 (Conv2D)	(None,	8,	8,	24)	46656	activation_29[0][0]
concatenate_27 (Concatenate)	(None,	8,	8,	240)	0	concatenate_26[0][0] conv2d_30[0][0]
batch_normalization_30 (BatchNo	(None,	8,	8,	240)	960	concatenate_27[0][0]
activation_30 (Activation)	(None,	8,	8,	240)	0	batch_normalization_30[0][0]
conv2d_31 (Conv2D)	(None,	8,	8,	24)	5760	activation_30[0][0]
average_pooling2d_3 (AveragePoo	(None,	4,	4,	24)	0	conv2d_31[0][0]
batch_normalization_31 (BatchNo	(None,	4,	4,	24)	96	average_pooling2d_3[0][0]
activation_31 (Activation)	(None,	4,	4,	24)	0	batch_normalization_31[0][0]
conv2d_32 (Conv2D)	(None,	4,	4,	24)	5184	activation_31[0][0]
concatenate_28 (Concatenate)	(None,	4,	4,	48)	0	average_pooling2d_3[0][0] conv2d_32[0][0]
batch_normalization_32 (BatchNo	(None,	4,	4,	48)	192	concatenate_28[0][0]
activation_32 (Activation)	(None,	4,	4,	48)	0	batch_normalization_32[0][0]
conv2d_33 (Conv2D)	(None,	4,	4,	24)	10368	activation_32[0][0]
concatenate_29 (Concatenate)	(None,	4,	4,	72)	0	concatenate_28[0][0] conv2d_33[0][0]
batch_normalization_33 (BatchNo	(None,	4,	4,	72)	288	concatenate_29[0][0]
activation_33 (Activation)	(None,	4,	4,	72)	0	batch_normalization_33[0][0]
conv2d_34 (Conv2D)	(None,	4,	4,	24)	15552	activation_33[0][0]
concatenate_30 (Concatenate)	(None,	4,	4,	96)	0	concatenate_29[0][0] conv2d_34[0][0]
batch_normalization_34 (BatchNo	(None,	4,	4,	96)	384	concatenate_30[0][0]
activation_34 (Activation)	(None,	4,	4,	96)	0	batch_normalization_34[0][0]
conv2d_35 (Conv2D)	(None,	4,	4,	24)	20736	activation_34[0][0]
concatenate_31 (Concatenate)	(None,	4,	4,	120)	0	concatenate_30[0][0] conv2d_35[0][0]
batch_normalization_35 (BatchNo	(None,	4,	4,	120)	480	concatenate_31[0][0]
activation_35 (Activation)	(None,	4,	4,	120)	0	batch_normalization_35[0][0]
conv2d_36 (Conv2D)	(None,	4,	4,	24)	25920	activation_35[0][0]
concatenate_32 (Concatenate)	(None,	4,	4,	144)	0	concatenate_31[0][0] conv2d_36[0][0]
batch_normalization_36 (BatchNo	(None,	4,	4,	144)	576	concatenate_32[0][0]
activation_36 (Activation)	(None,	4,	4,	144)	0	batch_normalization_36[0][0]
conv2d_37 (Conv2D)	(None,	4,	4,	24)	31104	activation_36[0][0]
concatenate_33 (Concatenate)	(None,	4,	4,	168)	0	concatenate_32[0][0] conv2d_37[0][0]
batch_normalization_37 (BatchNo	(None,	4,	4,	168)	672	concatenate_33[0][0]
activation_37 (Activation)	(None,	4,	4,	168)	0	batch_normalization_37[0][0]
conv2d_38 (Conv2D)	(None,	4,	4,	24)	36288	activation_37[0][0]
concatenate 34 (Concatenate)	(None,	4,	4,	192)	0	concatenate 33[0][0]

```
batch normalization 38 (BatchNo (None, 4, 4, 192)
                                           768
                                                     concatenate 34[0][0]
activation 38 (Activation)
                                                     batch normalization 38[0][0]
                          (None, 4, 4, 192)
                                           0
conv2d 39 (Conv2D)
                          (None, 4, 4, 24)
                                           41472
                                                     activation 38[0][0]
concatenate 35 (Concatenate)
                          (None, 4, 4, 216)
                                                     concatenate 34[0][0]
                                                     conv2d_39[0][0]
batch normalization 39 (BatchNo (None, 4, 4, 216)
                                           864
                                                     concatenate 35[0][0]
activation 39 (Activation)
                          (None, 4, 4, 216)
                                                     batch normalization 39[0][0]
                                           0
conv2d 40 (Conv2D)
                                           46656
                                                     activation 39[0][0]
                          (None, 4, 4, 24)
concatenate 36 (Concatenate)
                          (None, 4, 4, 240)
                                                     concatenate 35[0][0]
                                                     conv2d 40[0][0]
batch normalization 40 (BatchNo (None, 4, 4, 240)
                                           960
                                                     concatenate 36[0][0]
activation 40 (Activation)
                          (None, 4, 4, 240)
                                                     batch normalization 40[0][0]
average_pooling2d_4 (AveragePoo (None, 2, 2, 240)
                                                     activation_40[0][0]
                                           0
conv2d_41 (Conv2D)
                          (None, 2, 2, 10)
                                           2400
                                                     average_pooling2d_4[0][0]
global max pooling2d 1 (GlobalM (None, 10)
                                           0
                                                     conv2d 41[0][0]
activation 41 (Activation)
                         (None, 10)
                                                     global max pooling2d 1[0][0]
Total params: 974,568
Trainable params: 964,008
Non-trainable params: 10,560
Epoch 31/150
1954/1953 [============== ] - 241s 123ms/step - loss: 0.0748 - accuracy: 0.9738 - v
al_loss: 0.3629 - val_accuracy: 0.9081
Epoch 00031: val_accuracy improved from -inf to 0.90810, saving model to 031-0.908.hdf5
Epoch 32/150
al loss: 0.4749 - val accuracy: 0.8947
Epoch 00032: val_accuracy did not improve from 0.90810
Epoch 33/150
1954/1953 [============= ] - 237s 121ms/step - loss: 0.0700 - accuracy: 0.9752 - v
al loss: 0.4094 - val accuracy: 0.8994
Epoch 00033: val accuracy did not improve from 0.90810
Epoch 34/150
1954/1953 [============= ] - 240s 123ms/step - loss: 0.0667 - accuracy: 0.9766 - v
al_loss: 0.3743 - val_accuracy: 0.9124
Epoch 00034: val accuracy improved from 0.90810 to 0.91240, saving model to 034-0.912.hdf5
Epoch 35/150
al loss: 0.4359 - val accuracy: 0.8992
Epoch 00035: val accuracy did not improve from 0.91240
Epoch 36/150
al loss: 0.3888 - val accuracy: 0.9084
Epoch 00036: val accuracy did not improve from 0.91240
Epoch 37/150
al loss: 0.3959 - val accuracy: 0.9086
Epoch 00037: val accuracy did not improve from 0.91240
Epoch 38/150
al loss: 0.3856 - val accuracy: 0.9131
```

Epoch 00038: val accuracy improved from 0.91240 to 0.91310, saving model to 038-0.913.hdf5

```
Epoch 39/150
1954/1953 [============== ] - 236s 121ms/step - loss: 0.0581 - accuracy: 0.9796 - v
al loss: 0.5289 - val accuracy: 0.8884
Epoch 00039: val accuracy did not improve from 0.91310
Epoch 40/150
al loss: 0.4577 - val accuracy: 0.9012
Epoch 00040: val accuracy did not improve from 0.91310
Epoch 41/150
al loss: 0.3901 - val accuracy: 0.9104
Epoch 00041: val accuracy did not improve from 0.91310
Out[0]:
<keras.callbacks.callbacks.History at 0x7f1901df1fd0>
In [0]:
score = model.evaluate(X test, y_test, verbose=1)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
10000/10000 [============ ] - 4s 401us/step
Test loss: 0.39009565761350096
Test accuracy: 0.9103999733924866
In [0]:
model.load weights('038-0.913.hdf5')
model.compile(loss='categorical crossentropy',
                   optimizer=Adam(),
                   metrics=['accuracy'])
# model.fit(X_train,y_train)
train_acc = model.evaluate(X_train,y_train)
val acc = model.evaluate(X_test,y_test)
50000/50000 [======= ] - 21s 415us/step
10000/10000 [=========== ] - 4s 402us/step
In [0]:
print(train acc[1], val acc[1])
0.9830999970436096 0.913100004196167
```