

## CNN on CIFR Assignment:

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 [0]: # import keras
# from keras.datasets import cifar10
# from keras.models import Model, Sequential
# from keras.layers import Dense, Dropout, Flatten, Input, AveragePooling2D, merge, Activation
# from keras.layers import Conv2D, MaxPooling2D, BatchNormalization
# from keras.layers import Concatenate
# from keras.optimizers import Adam
from tensorflow.keras import models, layers
from tensorflow.keras.models import Model
from tensorflow.keras.layers import BatchNormalization, Activation, Flatten
from tensorflow.keras.optimizers import Adam
import matplotlib.pyplot as plt
import numpy as np
```

```
In [0]: # this part will prevent tensorflow to allocate all the available GPU Memory
# backend
import tensorflow as tf
# from tensorflow import keras

# from keras import backend as k

# Don't pre-allocate memory; allocate as-needed
# import tensorflow as tf
#tf.config.gpu.set_per_process_memory_fraction(0.75)
#tf.config.gpu.set_per_process_memory_growth(True)
# config = tf.ConfigProto()
# config.gpu_options.allow_growth = True

# Create a session with the above options specified.
# k.tensorflow_backend.set_session(tf.Session(config=config))
```

```
In [0]: # Hyperparameters
batch_size = 64
num_classes = 10
epochs = 85
l = 6
num_filter = 35
compression = 1
dropout_rate = 0.2
```

```
In [4]: # 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 encoding
y_train = tf.keras.utils.to_categorical(y_train, num_classes)
y_test = tf.keras.utils.to_categorical(y_test, num_classes)

Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
170500096/170498071 [=====] - 6s 0us/step
```

```
In [0]: X_train.shape
```

```
Out[0]: (50000, 32, 32, 3)
```

```
In [0]: X_test.shape
```

```
Out[0]: (10000, 32, 32, 3)
```

```

In [0]: # Dense Block
def densenblock(input, num_filter = 12, dropout_rate = 0.2):
    global compression
    temp = input
    for _ in range(1):
        BatchNorm = layers.BatchNormalization()(temp)
        relu = layers.Activation('relu')(BatchNorm)
        Conv2D_3_3 = layers.Conv2D(int(num_filter*compression), (3,3), use_bias=False, padding='same')(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 Block
def transition(input, num_filter = 12, dropout_rate = 0.2):
    global compression
    BatchNorm = layers.BatchNormalization()(input)
    relu = layers.Activation('relu')(BatchNorm)
    Conv2D_BottleNeck = layers.Conv2D(int(num_filter*compression), (1,1), use_bias=False, padding='same')(relu)
    if dropout_rate>0:
        Conv2D_BottleNeck = layers.Dropout(dropout_rate)(Conv2D_BottleNeck)
    avg = layers.AveragePooling2D(pool_size=(2,2))(Conv2D_BottleNeck)
    return avg

#output Layer
def output_layer(input):
    global compression
    BatchNorm = layers.BatchNormalization()(input)
    relu = layers.Activation('relu')(BatchNorm)
    AvgPooling = layers.AveragePooling2D(pool_size=(3,3))(relu)
    #flat = layers.Flatten()(AvgPooling)
    #a=tf.keras.layers.Reshape((None,312,1), input_shape=(2,2,78))
    #model=tf.reshape(flat,[None,312,1])
    output = layers.Conv2D(10,(1,1),strides=(1,1),activation='softmax',padding='valid')(AvgPooling)
    flat = layers.Flatten()(output)

    return flat

```

```
In [0]: #https://www.kaggle.com/genesis16/densenet-93-accuracy
input = layers.Input(shape=(img_height, img_width, channel,))
First_Conv2D = layers.Conv2D(num_filter, (3,3), use_bias=False ,padding='same'
)(input)

First_Block = denseblock(First_Conv2D, num_filter, dropout_rate)
First_Transition = transition(First_Block, num_filter, dropout_rate)

Second_Block = denseblock(First_Transition, num_filter, dropout_rate)
Second_Transition = transition(Second_Block, num_filter, dropout_rate)

Third_Block = denseblock(Second_Transition, num_filter, dropout_rate)
Third_Transition = transition(Third_Block, num_filter, dropout_rate)

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

```
In [11]: model = Model(inputs=[input], outputs=[output])  
         model.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_2 (InputLayer)	[(None, 32, 32, 3)]	0	
=====			
conv2d_53 (Conv2D)	(None, 32, 32, 35)	945	input_2[0]
=====			
batch_normalization_52 (Batch Normalization)	(None, 32, 32, 35)	140	conv2d_53[0]
=====			
activation_52 (Activation)	(None, 32, 32, 35)	0	batch_normalization_52[0][0]
=====			
conv2d_54 (Conv2D)	(None, 32, 32, 35)	11025	activation_52[0][0]
=====			
dropout_51 (Dropout)	(None, 32, 32, 35)	0	conv2d_54[0]
=====			
concatenate_48 (Concatenate)	(None, 32, 32, 70)	0	conv2d_53[0] dropout_51[0][0]
=====			
batch_normalization_53 (Batch Normalization)	(None, 32, 32, 70)	280	concatenate_48[0][0]
=====			
activation_53 (Activation)	(None, 32, 32, 70)	0	batch_normalization_53[0][0]
=====			
conv2d_55 (Conv2D)	(None, 32, 32, 35)	22050	activation_53[0][0]
=====			
dropout_52 (Dropout)	(None, 32, 32, 35)	0	conv2d_55[0]
=====			
concatenate_49 (Concatenate)	(None, 32, 32, 105)	0	concatenate_48[0][0] dropout_52[0][0]
=====			

batch_normalization_54 (BatchNo	(None, 32, 32, 105)	420	concatenate_49[0][0]
activation_54 (Activation)	(None, 32, 32, 105)	0	batch_normalization_54[0][0]
conv2d_56 (Conv2D)	(None, 32, 32, 35)	33075	activation_54[0][0]
dropout_53 (Dropout)	(None, 32, 32, 35)	0	conv2d_56[0][0]
concatenate_50 (Concatenate)	(None, 32, 32, 140)	0	concatenate_49[0][0]
			dropout_53[0][0]
batch_normalization_55 (BatchNo	(None, 32, 32, 140)	560	concatenate_50[0][0]
activation_55 (Activation)	(None, 32, 32, 140)	0	batch_normalization_55[0][0]
conv2d_57 (Conv2D)	(None, 32, 32, 35)	44100	activation_55[0][0]
dropout_54 (Dropout)	(None, 32, 32, 35)	0	conv2d_57[0][0]
concatenate_51 (Concatenate)	(None, 32, 32, 175)	0	concatenate_50[0][0]
			dropout_54[0][0]
batch_normalization_56 (BatchNo	(None, 32, 32, 175)	700	concatenate_51[0][0]
activation_56 (Activation)	(None, 32, 32, 175)	0	batch_normalization_56[0][0]
conv2d_58 (Conv2D)	(None, 32, 32, 35)	55125	activation_56[0][0]
dropout_55 (Dropout)	(None, 32, 32, 35)	0	conv2d_58[0][0]

[0]

concatenate_52 (Concatenate) 51[0][0]	(None, 32, 32, 210)	0	concatenate_ dropout_55
[0][0]			
batch_normalization_57 (BatchNo 52[0][0]	(None, 32, 32, 210)	840	concatenate_
activation_57 (Activation) ization_57[0][0]	(None, 32, 32, 210)	0	batch_normal
conv2d_59 (Conv2D) 7[0][0]	(None, 32, 32, 35)	66150	activation_5
dropout_56 (Dropout) [0]	(None, 32, 32, 35)	0	conv2d_59[0]
concatenate_53 (Concatenate) 52[0][0]	(None, 32, 32, 245)	0	concatenate_ dropout_56
[0][0]			
batch_normalization_58 (BatchNo 53[0][0]	(None, 32, 32, 245)	980	concatenate_
activation_58 (Activation) ization_58[0][0]	(None, 32, 32, 245)	0	batch_normal
conv2d_60 (Conv2D) 8[0][0]	(None, 32, 32, 35)	8575	activation_5
dropout_57 (Dropout) [0]	(None, 32, 32, 35)	0	conv2d_60[0]
average_pooling2d_4 (AveragePoo [0][0]	(None, 16, 16, 35)	0	dropout_57
batch_normalization_59 (BatchNo ing2d_4[0][0]	(None, 16, 16, 35)	140	average_pool
activation_59 (Activation) ization_59[0][0]	(None, 16, 16, 35)	0	batch_normal



conv2d_61 (Conv2D) 9[0][0]	(None, 16, 16, 35)	11025	activation_5
dropout_58 (Dropout) [0]	(None, 16, 16, 35)	0	conv2d_61[0]
concatenate_54 (Concatenate) ing2d_4[0][0] [0][0]	(None, 16, 16, 70)	0	average_pool dropout_58
batch_normalization_60 (BatchNo 54[0][0]	(None, 16, 16, 70)	280	concatenate_
activation_60 (Activation) ization_60[0][0]	(None, 16, 16, 70)	0	batch_normal
conv2d_62 (Conv2D) 0[0][0]	(None, 16, 16, 35)	22050	activation_6
dropout_59 (Dropout) [0]	(None, 16, 16, 35)	0	conv2d_62[0]
concatenate_55 (Concatenate) 54[0][0] [0][0]	(None, 16, 16, 105)	0	concatenate_ dropout_59
batch_normalization_61 (BatchNo 55[0][0]	(None, 16, 16, 105)	420	concatenate_
activation_61 (Activation) ization_61[0][0]	(None, 16, 16, 105)	0	batch_normal
conv2d_63 (Conv2D) 1[0][0]	(None, 16, 16, 35)	33075	activation_6
dropout_60 (Dropout) [0]	(None, 16, 16, 35)	0	conv2d_63[0]
concatenate_56 (Concatenate) 55[0][0]	(None, 16, 16, 140)	0	concatenate_ dropout_60

[0][0]

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batch_normalization_62 (BatchNo	(None, 16, 16, 140)	560	concatenate_56[0][0]
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activation_62 (Activation)	(None, 16, 16, 140)	0	batch_normalization_62[0][0]
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conv2d_64 (Conv2D)	(None, 16, 16, 35)	44100	activation_62[0][0]
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dropout_61 (Dropout)	(None, 16, 16, 35)	0	conv2d_64[0][0]
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concatenate_57 (Concatenate)	(None, 16, 16, 175)	0	concatenate_56[0][0]
			dropout_61[0][0]

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batch_normalization_63 (BatchNo	(None, 16, 16, 175)	700	concatenate_57[0][0]
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activation_63 (Activation)	(None, 16, 16, 175)	0	batch_normalization_63[0][0]
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conv2d_65 (Conv2D)	(None, 16, 16, 35)	55125	activation_63[0][0]
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dropout_62 (Dropout)	(None, 16, 16, 35)	0	conv2d_65[0][0]
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concatenate_58 (Concatenate)	(None, 16, 16, 210)	0	concatenate_57[0][0]
			dropout_62[0][0]

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batch_normalization_64 (BatchNo	(None, 16, 16, 210)	840	concatenate_58[0][0]
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activation_64 (Activation)	(None, 16, 16, 210)	0	batch_normalization_64[0][0]
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conv2d_66 (Conv2D)	(None, 16, 16, 35)	66150	activation_64[0][0]
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dropout_63 (Dropout) [0]	(None, 16, 16, 35)	0	conv2d_66[0]
concatenate_59 (Concatenate) 58[0][0]  [0][0]	(None, 16, 16, 245)	0	concatenate_ dropout_63
batch_normalization_65 (BatchNo 59[0][0]	(None, 16, 16, 245)	980	concatenate_
activation_65 (Activation) ization_65[0][0]	(None, 16, 16, 245)	0	batch_normal
conv2d_67 (Conv2D) 5[0][0]	(None, 16, 16, 35)	8575	activation_6
dropout_64 (Dropout) [0]	(None, 16, 16, 35)	0	conv2d_67[0]
average_pooling2d_5 (AveragePoo [0][0]	(None, 8, 8, 35)	0	dropout_64
batch_normalization_66 (BatchNo ing2d_5[0][0]	(None, 8, 8, 35)	140	average_pool
activation_66 (Activation) ization_66[0][0]	(None, 8, 8, 35)	0	batch_normal
conv2d_68 (Conv2D) 6[0][0]	(None, 8, 8, 35)	11025	activation_6
dropout_65 (Dropout) [0]	(None, 8, 8, 35)	0	conv2d_68[0]
concatenate_60 (Concatenate) ing2d_5[0][0]  [0][0]	(None, 8, 8, 70)	0	average_pool dropout_65
batch_normalization_67 (BatchNo 60[0][0]	(None, 8, 8, 70)	280	concatenate_

activation_67 (Activation) ization_67[0][0]	(None, 8, 8, 70)	0	batch_normal
conv2d_69 (Conv2D) 7[0][0]	(None, 8, 8, 35)	22050	activation_6
dropout_66 (Dropout) [0]	(None, 8, 8, 35)	0	conv2d_69[0]
concatenate_61 (Concatenate) 60[0][0]  [0][0]	(None, 8, 8, 105)	0	concatenate_ dropout_66
batch_normalization_68 (BatchNo 61[0][0])	(None, 8, 8, 105)	420	concatenate_
activation_68 (Activation) ization_68[0][0]	(None, 8, 8, 105)	0	batch_normal
conv2d_70 (Conv2D) 8[0][0]	(None, 8, 8, 35)	33075	activation_6
dropout_67 (Dropout) [0]	(None, 8, 8, 35)	0	conv2d_70[0]
concatenate_62 (Concatenate) 61[0][0]  [0][0]	(None, 8, 8, 140)	0	concatenate_ dropout_67
batch_normalization_69 (BatchNo 62[0][0])	(None, 8, 8, 140)	560	concatenate_
activation_69 (Activation) ization_69[0][0]	(None, 8, 8, 140)	0	batch_normal
conv2d_71 (Conv2D) 9[0][0]	(None, 8, 8, 35)	44100	activation_6
dropout_68 (Dropout) [0]	(None, 8, 8, 35)	0	conv2d_71[0]

concatenate_63 (Concatenate) 62[0][0] [0][0]	(None, 8, 8, 175)	0	concatenate_ dropout_68
batch_normalization_70 (BatchNo 63[0][0])	(None, 8, 8, 175)	700	concatenate_
activation_70 (Activation) ization_70[0][0]	(None, 8, 8, 175)	0	batch_normal
conv2d_72 (Conv2D) 0[0][0]	(None, 8, 8, 35)	55125	activation_7
dropout_69 (Dropout) [0]	(None, 8, 8, 35)	0	conv2d_72[0]
concatenate_64 (Concatenate) 63[0][0] [0][0]	(None, 8, 8, 210)	0	concatenate_ dropout_69
batch_normalization_71 (BatchNo 64[0][0])	(None, 8, 8, 210)	840	concatenate_
activation_71 (Activation) ization_71[0][0]	(None, 8, 8, 210)	0	batch_normal
conv2d_73 (Conv2D) 1[0][0]	(None, 8, 8, 35)	66150	activation_7
dropout_70 (Dropout) [0]	(None, 8, 8, 35)	0	conv2d_73[0]
concatenate_65 (Concatenate) 64[0][0] [0][0]	(None, 8, 8, 245)	0	concatenate_ dropout_70
batch_normalization_72 (BatchNo 65[0][0])	(None, 8, 8, 245)	980	concatenate_
activation_72 (Activation) ization_72[0][0]	(None, 8, 8, 245)	0	batch_normal

conv2d_74 (Conv2D) 2[0][0]	(None, 8, 8, 35)	8575	activation_7
dropout_71 (Dropout) [0]	(None, 8, 8, 35)	0	conv2d_74[0]
average_pooling2d_6 (AveragePool) [0][0]	(None, 4, 4, 35)	0	dropout_71
batch_normalization_73 (Batch Normalization) [0][0]	(None, 4, 4, 35)	140	average_pooling2d_6[0][0]
activation_73 (Activation) [0][0]	(None, 4, 4, 35)	0	batch_normalization_73[0][0]
conv2d_75 (Conv2D) 3[0][0]	(None, 4, 4, 35)	11025	activation_7
dropout_72 (Dropout) [0]	(None, 4, 4, 35)	0	conv2d_75[0]
concatenate_66 (Concatenate) [0][0]	(None, 4, 4, 70)	0	average_pooling2d_6[0][0] dropout_72
batch_normalization_74 (Batch Normalization) [0][0]	(None, 4, 4, 70)	280	concatenate_66[0][0]
activation_74 (Activation) [0][0]	(None, 4, 4, 70)	0	batch_normalization_74[0][0]
conv2d_76 (Conv2D) 4[0][0]	(None, 4, 4, 35)	22050	activation_7
dropout_73 (Dropout) [0]	(None, 4, 4, 35)	0	conv2d_76[0]
concatenate_67 (Concatenate) [0][0]	(None, 4, 4, 105)	0	concatenate_66[0][0] dropout_73

batch_normalization_75 (BatchNo	(None, 4, 4, 105)	420	concatenate_67[0][0]
activation_75 (Activation)	(None, 4, 4, 105)	0	batch_normalization_75[0][0]
conv2d_77 (Conv2D)	(None, 4, 4, 35)	33075	activation_75[0][0]
dropout_74 (Dropout)	(None, 4, 4, 35)	0	conv2d_77[0][0]
concatenate_68 (Concatenate)	(None, 4, 4, 140)	0	concatenate_67[0][0]
			dropout_74[0][0]
batch_normalization_76 (BatchNo	(None, 4, 4, 140)	560	concatenate_68[0][0]
activation_76 (Activation)	(None, 4, 4, 140)	0	batch_normalization_76[0][0]
conv2d_78 (Conv2D)	(None, 4, 4, 35)	44100	activation_76[0][0]
dropout_75 (Dropout)	(None, 4, 4, 35)	0	conv2d_78[0][0]
concatenate_69 (Concatenate)	(None, 4, 4, 175)	0	concatenate_68[0][0]
			dropout_75[0][0]
batch_normalization_77 (BatchNo	(None, 4, 4, 175)	700	concatenate_69[0][0]
activation_77 (Activation)	(None, 4, 4, 175)	0	batch_normalization_77[0][0]
conv2d_79 (Conv2D)	(None, 4, 4, 35)	55125	activation_77[0][0]
dropout_76 (Dropout)	(None, 4, 4, 35)	0	conv2d_79[0][0]

[0]

concatenate_70 (Concatenate)	(None, 4, 4, 210)	0	concatenate_70[0][0]
			dropout_76
[0][0]			

batch_normalization_78 (Batch Normalization)	(None, 4, 4, 210)	840	batch_normalization_78[0][0]
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activation_78 (Activation)	(None, 4, 4, 210)	0	activation_78[0][0]
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conv2d_80 (Conv2D)	(None, 4, 4, 35)	66150	conv2d_80[0][0]
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dropout_77 (Dropout)	(None, 4, 4, 35)	0	dropout_77[0]
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concatenate_71 (Concatenate)	(None, 4, 4, 245)	0	concatenate_71[0][0]
			dropout_77
[0][0]			

batch_normalization_79 (Batch Normalization)	(None, 4, 4, 245)	980	batch_normalization_79[0][0]
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activation_79 (Activation)	(None, 4, 4, 245)	0	activation_79[0][0]
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average_pooling2d_7 (Average Pooling)	(None, 1, 1, 245)	0	average_pooling2d_7[0][0]
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conv2d_81 (Conv2D)	(None, 1, 1, 10)	2460	conv2d_81[0][0]
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flatten_1 (Flatten)	(None, 10)	0	flatten_1[0]
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=====  
 Total params: 970,910  
 Trainable params: 963,070  
 Non-trainable params: 7,840



```
In [0]: # determine Loss function and Optimizer
model.compile(loss='categorical_crossentropy',
              optimizer=Adam(),
              metrics=['accuracy'])
```

```
In [0]: from keras.preprocessing.image import ImageDataGenerator
generator = ImageDataGenerator(rotation_range=20,
                              width_shift_range=0.125,
                              height_shift_range=0.125,
                              horizontal_flip=True,
                              fill_mode='nearest',
                              zoom_range=0.10)

generator.fit(X_train)
```

```
In [22]: #https://blog.keras.io/building-powerful-image-classification-models-using-very-little-data.html
model.fit_generator(generator.flow(X_train, y_train, batch_size=batch_size),
                    steps_per_epoch=(len(X_train)//batch_size)*5, epochs=20,
                    validation_data=(X_test, y_test))
```

```
Epoch 1/20
3905/3905 [=====] - 410s 105ms/step - loss: 0.4277 -
acc: 0.8520 - val_loss: 0.8166 - val_acc: 0.7626
Epoch 2/20
3905/3905 [=====] - 406s 104ms/step - loss: 0.3864 -
acc: 0.8654 - val_loss: 0.5648 - val_acc: 0.8379
Epoch 3/20
3905/3905 [=====] - 409s 105ms/step - loss: 0.3513 -
acc: 0.8778 - val_loss: 0.4767 - val_acc: 0.8535
Epoch 4/20
3905/3905 [=====] - 409s 105ms/step - loss: 0.3245 -
acc: 0.8873 - val_loss: 0.4529 - val_acc: 0.8631
Epoch 5/20
3905/3905 [=====] - 407s 104ms/step - loss: 0.3051 -
acc: 0.8936 - val_loss: 0.3657 - val_acc: 0.8901
Epoch 6/20
3905/3905 [=====] - 406s 104ms/step - loss: 0.2859 -
acc: 0.8998 - val_loss: 0.4397 - val_acc: 0.8722
Epoch 7/20
3905/3905 [=====] - 407s 104ms/step - loss: 0.2689 -
acc: 0.9054 - val_loss: 0.4023 - val_acc: 0.8840
Epoch 8/20
3905/3905 [=====] - 402s 103ms/step - loss: 0.2564 -
acc: 0.9105 - val_loss: 0.3955 - val_acc: 0.8888
Epoch 9/20
3905/3905 [=====] - 400s 103ms/step - loss: 0.2447 -
acc: 0.9146 - val_loss: 0.3739 - val_acc: 0.8931
Epoch 10/20
3905/3905 [=====] - 400s 102ms/step - loss: 0.2340 -
acc: 0.9180 - val_loss: 0.3771 - val_acc: 0.8959
Epoch 11/20
3905/3905 [=====] - 400s 103ms/step - loss: 0.2259 -
acc: 0.9203 - val_loss: 0.4087 - val_acc: 0.8845
Epoch 12/20
3905/3905 [=====] - 401s 103ms/step - loss: 0.2155 -
acc: 0.9244 - val_loss: 0.4331 - val_acc: 0.8867
Epoch 13/20
3905/3905 [=====] - 400s 102ms/step - loss: 0.2092 -
acc: 0.9262 - val_loss: 0.3012 - val_acc: 0.9111
Epoch 14/20
3905/3905 [=====] - 400s 102ms/step - loss: 0.2005 -
acc: 0.9293 - val_loss: 0.3843 - val_acc: 0.8952
Epoch 15/20
3905/3905 [=====] - 399s 102ms/step - loss: 0.1950 -
acc: 0.9312 - val_loss: 0.3442 - val_acc: 0.9079
Epoch 16/20
3905/3905 [=====] - 400s 102ms/step - loss: 0.1888 -
acc: 0.9331 - val_loss: 0.3730 - val_acc: 0.9056
Epoch 17/20
3905/3905 [=====] - 400s 102ms/step - loss: 0.1847 -
acc: 0.9351 - val_loss: 0.3603 - val_acc: 0.8998
Epoch 18/20
3905/3905 [=====] - 400s 102ms/step - loss: 0.1777 -
acc: 0.9373 - val_loss: 0.3620 - val_acc: 0.9038
Epoch 19/20
3905/3905 [=====] - 400s 103ms/step - loss: 0.1722 -
acc: 0.9394 - val_loss: 0.3658 - val_acc: 0.9072
```

```
Epoch 20/20  
3905/3905 [=====] - 400s 102ms/step - loss: 0.1679 -  
acc: 0.9407 - val_loss: 0.3436 - val_acc: 0.9071
```

```
Out[22]: <tensorflow.python.keras.callbacks.History at 0x7fb431f2aac8>
```

```
In [23]: # Test the model  
score = model.evaluate(X_test, y_test, verbose=1)  
print('Test loss:', score[0])  
print('Test accuracy:', score[1])
```

```
10000/10000 [=====] - 4s 444us/sample - loss: 0.3447  
- acc: 0.9071  
Test loss: 0.3446852895885706  
Test accuracy: 0.9071
```

```
In [0]: # Save the trained weights in to .h5 format  
model.save_weights("DNST_model.h5")  
print("Saved model to disk")
```

```
Saved model to disk
```

## Conclusions:

- 1- Used the given densenet to build the architecture of the model.
- 2- Used Image Augmentation techniques to make the model robust.
- 3- Used adam optimizer to optimize the loss.
- 4- Run the code for 20 epochs.
- 5- Test Loss=0.344
- 6- Test accuracy=90.71%