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July 1, 2019

```
In [1]: # import keras
        # from keras.datasets import cifar10
        # from keras.models import Model, Sequential
        # from keras.layers import Dense, Dropout, Flatten, Input, AveragePooling2D, merge, Ac
        # from keras.layers import Conv2D, MaxPooling2D, BatchNormalization
        # from keras.layers import Concatenate
        # from keras.optimizers import Adam

        !pip install -q keras

        import keras
        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, RMSprop
        import tensorflow as tf
```

Using TensorFlow backend.

```
In [0]: #!pip install tensorflow-gpu==1.5.0

In [0]: #import tensorflow as tf
        #tf.__version__

In [0]: # this part will prevent tensorflow to allocate all the available GPU Memory
        # backend

        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 = 128
batch_size = 64
num_classes = 10

```

```
epochs = 35
```

```
l = 6
```

```
num_filter = 35
```

```
compression = 1
dropout_rate = 0.3
```

In [5]: *# 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 [=====] - 11s 0us/step

In [0]: `import numpy as np`
`from sklearn.model_selection import train_test_split`

```

def preprocess_data(data_set):
    mean = np.array([125.3, 123.0, 113.9])
    std = np.array([63.0, 62.1, 66.7])

```

```

    data_set -= mean
    data_set /= std
    return data_set

```

```

X_train = X_train.astype('float32')
X_test = X_test.astype('float32')
X_train = preprocess_data(X_train)

```

```
X_test = preprocess_data(X_test)
```

```
X_train, X_val, y_train, y_val = train_test_split(X_train, y_train, test_size = 0.10)
```

```
In [0]: from keras.preprocessing.image import ImageDataGenerator
```

1 Using Image Augmentation

```
In [0]: #datagen = ImageDataGenerator(  
#     featurewise_center=True,  
#     featurewise_std_normalization=True,  
#     rotation_range=20,  
#     width_shift_range=0.1,  
#     height_shift_range=0.1,  
#     horizontal_flip=True)
```

```
#datagen.fit(X_train)
```

```
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  
)
```

```
datagen.fit(X_train)
```

```
In [9]: X_train.shape
```

```
Out[9]: (45000, 32, 32, 3)
```

```
In [10]: X_test.shape
```

```
Out[10]: (10000, 32, 32, 3)
```

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

#output layer
def output_layer(input):
    global compression
    BatchNorm = layers.BatchNormalization()(input)
    relu = layers.Activation('relu')(BatchNorm)
    AvgPooling = layers.AveragePooling2D(pool_size=(2,2))(relu)
    flat = layers.Flatten()(AvgPooling)
    output = layers.Dense(num_classes, activation='softmax')(flat)
    return output

```

2 Chaging the architecture

```

In [12]: #num_filter = 12
         #dropout_rate = 0.2

         #l = 12

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)

```

```
#Fourth_Block = denseblock(Third_Transition, num_filter, dropout_rate)
#Fourth_Transition = transition(Fourth_Block, num_filter, dropout_rate)
```

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

```
output = output_layer>Last_Block)
```

WARNING: Logging before flag parsing goes to stderr.

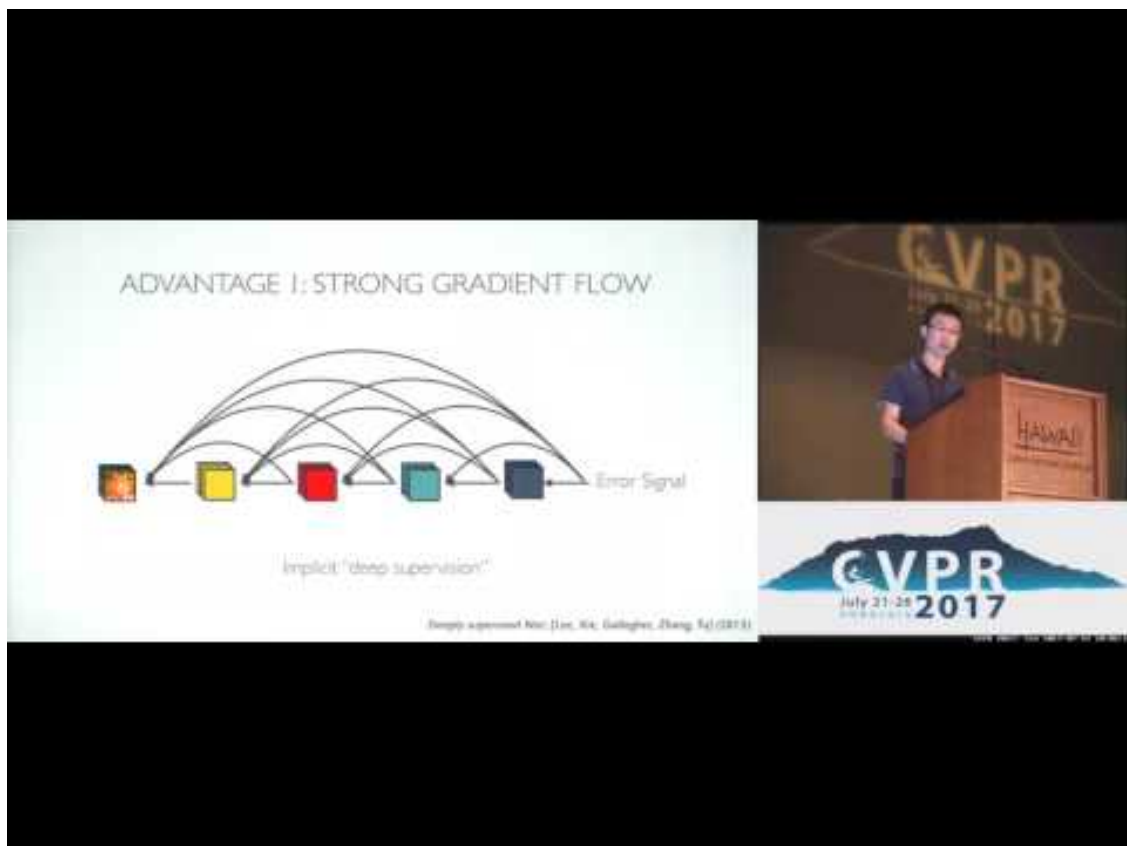
W0628 04:38:55.109632 140063208904576 deprecation.py:506] From /usr/local/lib/python3.6/dist-p

Instructions for updating:

Call initializer instance with the dtype argument instead of passing it to the constructor

```
In [0]: #https://arxiv.org/pdf/1608.06993.pdf
from IPython.display import IFrame, YouTubeVideo
YouTubeVideo(id='-W6y8xnd--U', width=600)
```

Out [0]:



```
In [13]: model = Model(inputs=[input], outputs=[output])
        model.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 32, 32, 3)]	0	
conv2d (Conv2D)	(None, 32, 32, 35)	945	input_1[0][0]
batch_normalization (BatchNorma	(None, 32, 32, 35)	140	conv2d[0][0]
activation (Activation)	(None, 32, 32, 35)	0	batch_normalization[0][0]
conv2d_1 (Conv2D)	(None, 32, 32, 35)	11025	activation[0][0]
dropout (Dropout)	(None, 32, 32, 35)	0	conv2d_1[0][0]
concatenate (Concatenate)	(None, 32, 32, 70)	0	conv2d[0][0] dropout[0][0]
batch_normalization_1 (BatchNor	(None, 32, 32, 70)	280	concatenate[0][0]
activation_1 (Activation)	(None, 32, 32, 70)	0	batch_normalization_1[0][0]
conv2d_2 (Conv2D)	(None, 32, 32, 35)	22050	activation_1[0][0]
dropout_1 (Dropout)	(None, 32, 32, 35)	0	conv2d_2[0][0]
concatenate_1 (Concatenate)	(None, 32, 32, 105)	0	concatenate[0][0] dropout_1[0][0]
batch_normalization_2 (BatchNor	(None, 32, 32, 105)	420	concatenate_1[0][0]
activation_2 (Activation)	(None, 32, 32, 105)	0	batch_normalization_2[0][0]
conv2d_3 (Conv2D)	(None, 32, 32, 35)	33075	activation_2[0][0]
dropout_2 (Dropout)	(None, 32, 32, 35)	0	conv2d_3[0][0]
concatenate_2 (Concatenate)	(None, 32, 32, 140)	0	concatenate_1[0][0] dropout_2[0][0]
batch_normalization_3 (BatchNor	(None, 32, 32, 140)	560	concatenate_2[0][0]
activation_3 (Activation)	(None, 32, 32, 140)	0	batch_normalization_3[0][0]

conv2d_4 (Conv2D)	(None, 32, 32, 35)	44100	activation_3[0][0]
dropout_3 (Dropout)	(None, 32, 32, 35)	0	conv2d_4[0][0]
concatenate_3 (Concatenate)	(None, 32, 32, 175)	0	concatenate_2[0][0] dropout_3[0][0]
batch_normalization_4 (BatchNor	(None, 32, 32, 175)	700	concatenate_3[0][0]
activation_4 (Activation)	(None, 32, 32, 175)	0	batch_normalization_4[0][0]
conv2d_5 (Conv2D)	(None, 32, 32, 35)	55125	activation_4[0][0]
dropout_4 (Dropout)	(None, 32, 32, 35)	0	conv2d_5[0][0]
concatenate_4 (Concatenate)	(None, 32, 32, 210)	0	concatenate_3[0][0] dropout_4[0][0]
batch_normalization_5 (BatchNor	(None, 32, 32, 210)	840	concatenate_4[0][0]
activation_5 (Activation)	(None, 32, 32, 210)	0	batch_normalization_5[0][0]
conv2d_6 (Conv2D)	(None, 32, 32, 35)	66150	activation_5[0][0]
dropout_5 (Dropout)	(None, 32, 32, 35)	0	conv2d_6[0][0]
concatenate_5 (Concatenate)	(None, 32, 32, 245)	0	concatenate_4[0][0] dropout_5[0][0]
batch_normalization_6 (BatchNor	(None, 32, 32, 245)	980	concatenate_5[0][0]
activation_6 (Activation)	(None, 32, 32, 245)	0	batch_normalization_6[0][0]
conv2d_7 (Conv2D)	(None, 32, 32, 35)	8575	activation_6[0][0]
dropout_6 (Dropout)	(None, 32, 32, 35)	0	conv2d_7[0][0]
average_pooling2d (AveragePooli	(None, 16, 16, 35)	0	dropout_6[0][0]
batch_normalization_7 (BatchNor	(None, 16, 16, 35)	140	average_pooling2d[0][0]
activation_7 (Activation)	(None, 16, 16, 35)	0	batch_normalization_7[0][0]
conv2d_8 (Conv2D)	(None, 16, 16, 35)	11025	activation_7[0][0]
dropout_7 (Dropout)	(None, 16, 16, 35)	0	conv2d_8[0][0]
concatenate_6 (Concatenate)	(None, 16, 16, 70)	0	average_pooling2d[0][0]

			dropout_7[0][0]
batch_normalization_8 (BatchNor	(None, 16, 16, 70)	280	concatenate_6[0][0]
activation_8 (Activation)	(None, 16, 16, 70)	0	batch_normalization_8[0][0]
conv2d_9 (Conv2D)	(None, 16, 16, 35)	22050	activation_8[0][0]
dropout_8 (Dropout)	(None, 16, 16, 35)	0	conv2d_9[0][0]
concatenate_7 (Concatenate)	(None, 16, 16, 105)	0	concatenate_6[0][0] dropout_8[0][0]
batch_normalization_9 (BatchNor	(None, 16, 16, 105)	420	concatenate_7[0][0]
activation_9 (Activation)	(None, 16, 16, 105)	0	batch_normalization_9[0][0]
conv2d_10 (Conv2D)	(None, 16, 16, 35)	33075	activation_9[0][0]
dropout_9 (Dropout)	(None, 16, 16, 35)	0	conv2d_10[0][0]
concatenate_8 (Concatenate)	(None, 16, 16, 140)	0	concatenate_7[0][0] dropout_9[0][0]
batch_normalization_10 (BatchNo	(None, 16, 16, 140)	560	concatenate_8[0][0]
activation_10 (Activation)	(None, 16, 16, 140)	0	batch_normalization_10[0][0]
conv2d_11 (Conv2D)	(None, 16, 16, 35)	44100	activation_10[0][0]
dropout_10 (Dropout)	(None, 16, 16, 35)	0	conv2d_11[0][0]
concatenate_9 (Concatenate)	(None, 16, 16, 175)	0	concatenate_8[0][0] dropout_10[0][0]
batch_normalization_11 (BatchNo	(None, 16, 16, 175)	700	concatenate_9[0][0]
activation_11 (Activation)	(None, 16, 16, 175)	0	batch_normalization_11[0][0]
conv2d_12 (Conv2D)	(None, 16, 16, 35)	55125	activation_11[0][0]
dropout_11 (Dropout)	(None, 16, 16, 35)	0	conv2d_12[0][0]
concatenate_10 (Concatenate)	(None, 16, 16, 210)	0	concatenate_9[0][0] dropout_11[0][0]
batch_normalization_12 (BatchNo	(None, 16, 16, 210)	840	concatenate_10[0][0]

activation_12 (Activation)	(None, 16, 16, 210)	0	batch_normalization_12[0][0]
conv2d_13 (Conv2D)	(None, 16, 16, 35)	66150	activation_12[0][0]
dropout_12 (Dropout)	(None, 16, 16, 35)	0	conv2d_13[0][0]
concatenate_11 (Concatenate)	(None, 16, 16, 245)	0	concatenate_10[0][0] dropout_12[0][0]
batch_normalization_13 (BatchNo	(None, 16, 16, 245)	980	concatenate_11[0][0]
activation_13 (Activation)	(None, 16, 16, 245)	0	batch_normalization_13[0][0]
conv2d_14 (Conv2D)	(None, 16, 16, 35)	8575	activation_13[0][0]
dropout_13 (Dropout)	(None, 16, 16, 35)	0	conv2d_14[0][0]
average_pooling2d_1 (AveragePoo	(None, 8, 8, 35)	0	dropout_13[0][0]
batch_normalization_14 (BatchNo	(None, 8, 8, 35)	140	average_pooling2d_1[0][0]
activation_14 (Activation)	(None, 8, 8, 35)	0	batch_normalization_14[0][0]
conv2d_15 (Conv2D)	(None, 8, 8, 35)	11025	activation_14[0][0]
dropout_14 (Dropout)	(None, 8, 8, 35)	0	conv2d_15[0][0]
concatenate_12 (Concatenate)	(None, 8, 8, 70)	0	average_pooling2d_1[0][0] dropout_14[0][0]
batch_normalization_15 (BatchNo	(None, 8, 8, 70)	280	concatenate_12[0][0]
activation_15 (Activation)	(None, 8, 8, 70)	0	batch_normalization_15[0][0]
conv2d_16 (Conv2D)	(None, 8, 8, 35)	22050	activation_15[0][0]
dropout_15 (Dropout)	(None, 8, 8, 35)	0	conv2d_16[0][0]
concatenate_13 (Concatenate)	(None, 8, 8, 105)	0	concatenate_12[0][0] dropout_15[0][0]
batch_normalization_16 (BatchNo	(None, 8, 8, 105)	420	concatenate_13[0][0]
activation_16 (Activation)	(None, 8, 8, 105)	0	batch_normalization_16[0][0]
conv2d_17 (Conv2D)	(None, 8, 8, 35)	33075	activation_16[0][0]
dropout_16 (Dropout)	(None, 8, 8, 35)	0	conv2d_17[0][0]

concatenate_14 (Concatenate)	(None, 8, 8, 140)	0	concatenate_13[0][0] dropout_16[0][0]
batch_normalization_17 (BatchNo	(None, 8, 8, 140)	560	concatenate_14[0][0]
activation_17 (Activation)	(None, 8, 8, 140)	0	batch_normalization_17[0][0]
conv2d_18 (Conv2D)	(None, 8, 8, 35)	44100	activation_17[0][0]
dropout_17 (Dropout)	(None, 8, 8, 35)	0	conv2d_18[0][0]
concatenate_15 (Concatenate)	(None, 8, 8, 175)	0	concatenate_14[0][0] dropout_17[0][0]
batch_normalization_18 (BatchNo	(None, 8, 8, 175)	700	concatenate_15[0][0]
activation_18 (Activation)	(None, 8, 8, 175)	0	batch_normalization_18[0][0]
conv2d_19 (Conv2D)	(None, 8, 8, 35)	55125	activation_18[0][0]
dropout_18 (Dropout)	(None, 8, 8, 35)	0	conv2d_19[0][0]
concatenate_16 (Concatenate)	(None, 8, 8, 210)	0	concatenate_15[0][0] dropout_18[0][0]
batch_normalization_19 (BatchNo	(None, 8, 8, 210)	840	concatenate_16[0][0]
activation_19 (Activation)	(None, 8, 8, 210)	0	batch_normalization_19[0][0]
conv2d_20 (Conv2D)	(None, 8, 8, 35)	66150	activation_19[0][0]
dropout_19 (Dropout)	(None, 8, 8, 35)	0	conv2d_20[0][0]
concatenate_17 (Concatenate)	(None, 8, 8, 245)	0	concatenate_16[0][0] dropout_19[0][0]
batch_normalization_20 (BatchNo	(None, 8, 8, 245)	980	concatenate_17[0][0]
activation_20 (Activation)	(None, 8, 8, 245)	0	batch_normalization_20[0][0]
conv2d_21 (Conv2D)	(None, 8, 8, 35)	8575	activation_20[0][0]
dropout_20 (Dropout)	(None, 8, 8, 35)	0	conv2d_21[0][0]
average_pooling2d_2 (AveragePoo	(None, 4, 4, 35)	0	dropout_20[0][0]
batch_normalization_21 (BatchNo	(None, 4, 4, 35)	140	average_pooling2d_2[0][0]

activation_21 (Activation)	(None, 4, 4, 35)	0	batch_normalization_21[0][0]
conv2d_22 (Conv2D)	(None, 4, 4, 35)	11025	activation_21[0][0]
dropout_21 (Dropout)	(None, 4, 4, 35)	0	conv2d_22[0][0]
concatenate_18 (Concatenate)	(None, 4, 4, 70)	0	average_pooling2d_2[0][0] dropout_21[0][0]
batch_normalization_22 (BatchNo	(None, 4, 4, 70)	280	concatenate_18[0][0]
activation_22 (Activation)	(None, 4, 4, 70)	0	batch_normalization_22[0][0]
conv2d_23 (Conv2D)	(None, 4, 4, 35)	22050	activation_22[0][0]
dropout_22 (Dropout)	(None, 4, 4, 35)	0	conv2d_23[0][0]
concatenate_19 (Concatenate)	(None, 4, 4, 105)	0	concatenate_18[0][0] dropout_22[0][0]
batch_normalization_23 (BatchNo	(None, 4, 4, 105)	420	concatenate_19[0][0]
activation_23 (Activation)	(None, 4, 4, 105)	0	batch_normalization_23[0][0]
conv2d_24 (Conv2D)	(None, 4, 4, 35)	33075	activation_23[0][0]
dropout_23 (Dropout)	(None, 4, 4, 35)	0	conv2d_24[0][0]
concatenate_20 (Concatenate)	(None, 4, 4, 140)	0	concatenate_19[0][0] dropout_23[0][0]
batch_normalization_24 (BatchNo	(None, 4, 4, 140)	560	concatenate_20[0][0]
activation_24 (Activation)	(None, 4, 4, 140)	0	batch_normalization_24[0][0]
conv2d_25 (Conv2D)	(None, 4, 4, 35)	44100	activation_24[0][0]
dropout_24 (Dropout)	(None, 4, 4, 35)	0	conv2d_25[0][0]
concatenate_21 (Concatenate)	(None, 4, 4, 175)	0	concatenate_20[0][0] dropout_24[0][0]
batch_normalization_25 (BatchNo	(None, 4, 4, 175)	700	concatenate_21[0][0]
activation_25 (Activation)	(None, 4, 4, 175)	0	batch_normalization_25[0][0]
conv2d_26 (Conv2D)	(None, 4, 4, 35)	55125	activation_25[0][0]

dropout_25 (Dropout)	(None, 4, 4, 35)	0	conv2d_26[0][0]
concatenate_22 (Concatenate)	(None, 4, 4, 210)	0	concatenate_21[0][0] dropout_25[0][0]
batch_normalization_26 (Batch Normalization)	(None, 4, 4, 210)	840	concatenate_22[0][0]
activation_26 (Activation)	(None, 4, 4, 210)	0	batch_normalization_26[0][0]
conv2d_27 (Conv2D)	(None, 4, 4, 35)	66150	activation_26[0][0]
dropout_26 (Dropout)	(None, 4, 4, 35)	0	conv2d_27[0][0]
concatenate_23 (Concatenate)	(None, 4, 4, 245)	0	concatenate_22[0][0] dropout_26[0][0]
batch_normalization_27 (Batch Normalization)	(None, 4, 4, 245)	980	concatenate_23[0][0]
activation_27 (Activation)	(None, 4, 4, 245)	0	batch_normalization_27[0][0]
average_pooling2d_3 (Average Pooling)	(None, 2, 2, 245)	0	activation_27[0][0]
flatten (Flatten)	(None, 980)	0	average_pooling2d_3[0][0]
dense (Dense)	(None, 10)	9810	flatten[0][0]

=====
 Total params: 978,260
 Trainable params: 970,420
 Non-trainable params: 7,840
 =====

In [0]: # determine Loss function and Optimizer

```
#rms = RMSprop(lr=0.001,decay=1e-6)
```

```
#adam_opt = Adam(lr=0.1,decay=1e-6)
```

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

In [0]: ##### <https://www.kaggle.com/genesis16/densenet-93-accuracy>

```

from tensorflow.keras.callbacks import ReduceLROnPlateau, ModelCheckpoint, EarlyStopping

reduce_lr = ReduceLROnPlateau(monitor = 'val_loss', factor = 0.1, patience = 5, min_lr

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:
        return 0.0001
    else:
        return 0.00001

lr_scheduler = LearningRateScheduler(decay_fn)

csv_logger = CSVLogger('training.log')

In [16]: #model.fit(X_train, y_train,
#               batch_size=batch_size,
#               epochs=epochs,
#               verbose=1,
#               validation_data=(X_test, y_test))

history = model.fit_generator(
    datagen.flow(X_train, y_train, batch_size=32),
    steps_per_epoch=(len(X_train)/batch_size)*5,
    epochs=epochs,
    verbose = 1,
    validation_data=(X_val, y_val),
    callbacks = [lr_scheduler, csv_logger]
)

Epoch 1/35
3516/3515 [=====] - 542s 154ms/step - loss: 1.3503 - acc: 0.5097 - va
Epoch 2/35
3516/3515 [=====] - 528s 150ms/step - loss: 0.9411 - acc: 0.6689 - va
Epoch 3/35
3516/3515 [=====] - 526s 149ms/step - loss: 0.7894 - acc: 0.7240 - va
Epoch 4/35
3516/3515 [=====] - 526s 149ms/step - loss: 0.6992 - acc: 0.7554 - va
Epoch 5/35
3516/3515 [=====] - 525s 149ms/step - loss: 0.6359 - acc: 0.7796 - va
Epoch 6/35
3516/3515 [=====] - 525s 149ms/step - loss: 0.5880 - acc: 0.7963 - va
Epoch 7/35
3516/3515 [=====] - 525s 149ms/step - loss: 0.5505 - acc: 0.8084 - va
Epoch 8/35

```

3516/3515 [=====] - 525s 149ms/step - loss: 0.5220 - acc: 0.8190 - va.
 Epoch 9/35
 3516/3515 [=====] - 525s 149ms/step - loss: 0.4961 - acc: 0.8269 - va.
 Epoch 10/35
 3516/3515 [=====] - 525s 149ms/step - loss: 0.4769 - acc: 0.8335 - va.
 Epoch 11/35
 3516/3515 [=====] - 525s 149ms/step - loss: 0.4579 - acc: 0.8396 - va.
 Epoch 12/35
 3516/3515 [=====] - 525s 149ms/step - loss: 0.4383 - acc: 0.8481 - va.
 Epoch 13/35
 3516/3515 [=====] - 525s 149ms/step - loss: 0.4251 - acc: 0.8531 - va.
 Epoch 14/35
 3516/3515 [=====] - 526s 150ms/step - loss: 0.4115 - acc: 0.8572 - va.
 Epoch 15/35
 3516/3515 [=====] - 524s 149ms/step - loss: 0.3971 - acc: 0.8616 - va.
 Epoch 16/35
 3516/3515 [=====] - 524s 149ms/step - loss: 0.3931 - acc: 0.8623 - va.
 Epoch 17/35
 3516/3515 [=====] - 527s 150ms/step - loss: 0.3807 - acc: 0.8681 - va.
 Epoch 18/35
 3516/3515 [=====] - 526s 150ms/step - loss: 0.3702 - acc: 0.8719 - va.
 Epoch 19/35
 3516/3515 [=====] - 526s 150ms/step - loss: 0.3645 - acc: 0.8736 - va.
 Epoch 20/35
 3516/3515 [=====] - 525s 149ms/step - loss: 0.3507 - acc: 0.8778 - va.
 Epoch 21/35
 3516/3515 [=====] - 526s 150ms/step - loss: 0.3450 - acc: 0.8795 - va.
 Epoch 22/35
 3516/3515 [=====] - 527s 150ms/step - loss: 0.3385 - acc: 0.8817 - va.
 Epoch 23/35
 3516/3515 [=====] - 526s 150ms/step - loss: 0.3349 - acc: 0.8829 - va.
 Epoch 24/35
 3516/3515 [=====] - 525s 149ms/step - loss: 0.3294 - acc: 0.8852 - va.
 Epoch 25/35
 3516/3515 [=====] - 526s 150ms/step - loss: 0.3226 - acc: 0.8884 - va.
 Epoch 26/35
 3516/3515 [=====] - 525s 149ms/step - loss: 0.3173 - acc: 0.8895 - va.
 Epoch 27/35
 3516/3515 [=====] - 524s 149ms/step - loss: 0.3120 - acc: 0.8909 - va.
 Epoch 28/35
 3516/3515 [=====] - 525s 149ms/step - loss: 0.3062 - acc: 0.8933 - va.
 Epoch 29/35
 3516/3515 [=====] - 526s 150ms/step - loss: 0.3020 - acc: 0.8948 - va.
 Epoch 30/35
 3516/3515 [=====] - 529s 150ms/step - loss: 0.2990 - acc: 0.8953 - va.
 Epoch 31/35
 3516/3515 [=====] - 531s 151ms/step - loss: 0.2925 - acc: 0.8981 - va.
 Epoch 32/35

```

3516/3515 [=====] - 534s 152ms/step - loss: 0.2885 - acc: 0.8986 - va
Epoch 33/35
3516/3515 [=====] - 535s 152ms/step - loss: 0.2847 - acc: 0.9006 - va
Epoch 34/35
3516/3515 [=====] - 533s 152ms/step - loss: 0.2831 - acc: 0.9017 - va
Epoch 35/35
3516/3515 [=====] - 530s 151ms/step - loss: 0.2793 - acc: 0.9030 - va

```

```
In [17]: # Test the model
```

```

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

```

```

10000/10000 [=====] - 10s 1ms/sample - loss: 0.3223 - acc: 0.9041
Test loss: 0.32226717756986617
Test accuracy: 0.9041

```

```
In [18]: # Save the trained weights in to .h5 format
```

```

    model.save_weights("DNST_model.h5")
    print("Saved model to disk")

```

```
Saved model to disk
```

3 Conclusion

1. The most important parameters are : the value "l" which is number of layers in dense block and "k" which is denoted number of filters.
2. I tried many values of l and k . Ultimately l=6 and k=35 proved to be effective. Also, the total number of parameters increase rapidly with these two parameters.
3. Also, the process of early stopping as well as using decay function for learning rate also helped.

```
In [0]:
```