DenseNet_cifar10

April 22, 2021

```
[]: # 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
     from tensorflow.keras.preprocessing.image import ImageDataGenerator
     from tensorflow.keras import regularizers
     import tensorboard
     import numpy as np
     import tensorflow as tf
     %load_ext tensorboard
```

The tensorboard extension is already loaded. To reload it, use: %reload_ext tensorboard

```
[]: # this part will prevent tensorflow to allocate all the avaliable GPU Memory # backend import tensorflow as tf
```

```
batch_size = 64
num_classes = 10
epochs = 85
l = 6 # no of layers in dense block
num_filter = 35 # growth rate k
compression = 1.0
dropout_rate = 0.20
weight_decay = 0.001
```

```
[]: # 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)
[]: # 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 ,padding='same', kernel_initializer='he_normal',
                                        kernel_regularizer=tf.keras.regularizers.
      →12(weight_decay))(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, 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', kernel_initializer='he_normal',
                                        kernel_regularizer=tf.keras.regularizers.
      →12(weight_decay))(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(momentum=0.9, epsilon=0.00001)(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',___

kernel_initializer='he_normal',
kernel_regularizer=regularizers.12(weight_decay))(flat)
return output

[]: input = layers.Input(shape=(img_height, img_width, channel,))
First_Conv2D = layers.Conv2D(num_filter, (3,3), use_bias=False__
, padding='same')(input)
```

[]:

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

[]:



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

Model: "functional_3"			
Layer (type)	Output Shape	Param #	Connected to
input_3 (InputLayer)	[(None, 32, 32, 3)]	0	
conv2d_29 (Conv2D)	(None, 32, 32, 35)	945	input_3[0][0]
batch_normalization_29 (BatchNo	(None, 32, 32, 35)	140	conv2d_29[0][0]
activation_29 (Activation) batch_normalization_29[0][0]	(None, 32, 32, 35)	0	

conv2d_30 (Conv2D) activation_29[0][0]	(None,					
dropout_27 (Dropout)	(None,				0	conv2d_30[0][0]
concatenate_24 (Concatenate) dropout_27[0][0]	(None,	32,	32,	70)	0	conv2d_29[0][0]
batch_normalization_30 (BatchNo concatenate_24[0][0]	(None,	32,	32,	70)	280	
activation_30 (Activation) batch_normalization_30[0][0]	(None,				0	
conv2d_31 (Conv2D) activation_30[0][0]	(None,				22050	
dropout_28 (Dropout)	(None,					conv2d_31[0][0]
concatenate_25 (Concatenate) concatenate_24[0][0] dropout_28[0][0]	(None,	32,	32,	105)	0	
batch_normalization_31 (BatchNo concatenate_25[0][0]	(None,	32,	32,	105)	420	
activation_31 (Activation) batch_normalization_31[0][0]	(None,					
conv2d_32 (Conv2D) activation_31[0][0]					33075	
dropout_29 (Dropout)	(None,					conv2d_32[0][0]
concatenate_26 (Concatenate)	(None,	32,	32,	140)	0	

concatenate_25[0][0] dropout_29[0][0]						
batch_normalization_32 (BatchNo concatenate_26[0][0]	(None,	32,	32,	140)	560	
activation_32 (Activation) batch_normalization_32[0][0]	(None,					
conv2d_33 (Conv2D) activation_32[0][0]	(None,		32,	35)	44100	
dropout_30 (Dropout)			32,	35)		conv2d_33[0][0]
concatenate_27 (Concatenate) concatenate_26[0][0] dropout_30[0][0]	4					
batch_normalization_33 (BatchNo concatenate_27[0][0]					700	
activation_33 (Activation) batch_normalization_33[0][0]						
conv2d_34 (Conv2D) activation_33[0][0]	(None,				55125	
dropout_31 (Dropout)	(None,	32,	32,	35)	0	conv2d_34[0][0]
concatenate_28 (Concatenate) concatenate_27[0][0] dropout_31[0][0]	. ,					
batch_normalization_34 (BatchNo concatenate_28[0][0]	(None,	32,	32,	210)	840	

activation_34 (Activation) batch_normalization_34[0][0]	(None,	32,	32,	210)	0	
conv2d_35 (Conv2D) activation_34[0][0]	(None,	32,	32,	35)	66150	
dropout_32 (Dropout)	(None,	32,	32,	35)	0	conv2d_35[0][0]
concatenate_29 (Concatenate) concatenate_28[0][0] dropout_32[0][0]	,					
batch_normalization_35 (BatchNo concatenate_29[0][0]						
activation_35 (Activation) batch_normalization_35[0][0]	(None,	32,	32,	245)	0	
 conv2d_36 (Conv2D) activation_35[0][0]	(None,					
dropout_33 (Dropout)	(None,	32,	32,	35)	0	conv2d_36[0][0]
average_pooling2d_4 (AveragePoodropout_33[0][0]	(None,	16,	16,	35)	0	
batch_normalization_36 (BatchNo average_pooling2d_4[0][0]	(None,	16,	16,	35)	140	
activation_36 (Activation) batch_normalization_36[0][0]	(None,					
conv2d_37 (Conv2D) activation_36[0][0]	(None,	16,	16,	35)	11025	
dropout_34 (Dropout)	(None,	16,	16,	35)	0	conv2d_37[0][0]

concatenate_30 (Concatenate) average_pooling2d_4[0][0] dropout_34[0][0]	(None,	16,	16,	70)	0	
batch_normalization_37 (BatchNo concatenate_30[0][0]		16,	16,	70)	280	
activation_37 (Activation) batch_normalization_37[0][0]	(None,	16,	16,	70)	0	
 conv2d_38 (Conv2D) activation_37[0][0]	(None,	16,	16,	35)	22050	
dropout_35 (Dropout)	(None,	16,	16,	35)	0	conv2d_38[0][0]
concatenate_31 (Concatenate) concatenate_30[0][0] dropout_35[0][0]	(None,					
batch_normalization_38 (BatchNo concatenate_31[0][0]						
activation_38 (Activation) batch_normalization_38[0][0]	(None,	16,	16,	105)	0	
conv2d_39 (Conv2D) activation_38[0][0]	(None,					
dropout_36 (Dropout)	(None,	16,	16,	35)	0	conv2d_39[0][0]
concatenate_32 (Concatenate) concatenate_31[0][0] dropout_36[0][0]	(None,	16,	16,	140)	0	
batch_normalization_39 (BatchNo						

concatenate_32[0][0]						
activation_39 (Activation) batch_normalization_39[0][0]	(None,	16,	16,	140)	0	
conv2d_40 (Conv2D) activation_39[0][0]	(None,	16,	16,	35)	44100	
dropout_37 (Dropout)	(None,	16,	16,	35)	0	conv2d_40[0][0]
concatenate_33 (Concatenate) concatenate_32[0][0] dropout_37[0][0]	(None,		16,	175)	0	
batch_normalization_40 (BatchNo concatenate_33[0][0]			16,	175)	700	
activation_40 (Activation) batch_normalization_40[0][0]	(None,	16,	16,	175)	0	
conv2d_41 (Conv2D) activation_40[0][0]	(None,	16,	16,	35)	55125	
dropout_38 (Dropout)	(None,	16,	16,	35)	0	conv2d_41[0][0]
concatenate_34 (Concatenate) concatenate_33[0][0] dropout_38[0][0]	(None,	16,	16,	210)	0	
batch_normalization_41 (BatchNo concatenate_34[0][0]						
activation_41 (Activation) batch_normalization_41[0][0]	(None,	16,	16,	210)	0	
 conv2d_42 (Conv2D)	(None,					

activation_41[0][0]			
dropout_39 (Dropout)	(None, 16, 16, 35)	0	conv2d_42[0][0]
concatenate_35 (Concatenate) concatenate_34[0][0] dropout_39[0][0]	(None, 16, 16, 245)	0	
batch_normalization_42 (BatchNo concatenate_35[0][0]	(None, 16, 16, 245)		
activation_42 (Activation) batch_normalization_42[0][0]	(None, 16, 16, 245)		
conv2d_43 (Conv2D) activation_42[0][0]	(None, 16, 16, 35)	8575	
dropout_40 (Dropout)	(None, 16, 16, 35)	0	conv2d_43[0][0]
average_pooling2d_5 (AveragePoodropout_40[0][0]		0	
batch_normalization_43 (BatchNo average_pooling2d_5[0][0]	(None, 8, 8, 35)	140	
activation_43 (Activation) batch_normalization_43[0][0]	(None, 8, 8, 35)	0	
 conv2d_44 (Conv2D) activation_43[0][0]	(None, 8, 8, 35)		
dropout_41 (Dropout)	(None, 8, 8, 35)		conv2d_44[0][0]
concatenate_36 (Concatenate) average_pooling2d_5[0][0] dropout_41[0][0]	(None, 8, 8, 70)	0	

batch_normalization_44 (BatchNo concatenate_36[0][0]	(None,	8,	8,	70)	280	
activation_44 (Activation) batch_normalization_44[0][0]	(None,	8,		70)	0	
conv2d_45 (Conv2D) activation_44[0][0]	(None,	8,		35)	22050	
dropout_42 (Dropout)	(None,	8,	8,	35)	0	conv2d_45[0][0]
concatenate_37 (Concatenate) concatenate_36[0][0] dropout_42[0][0]	(None,	8,	8,	105)	0	
batch_normalization_45 (BatchNo concatenate_37[0][0]					420	
activation_45 (Activation) batch_normalization_45[0][0]	(None,				0	
conv2d_46 (Conv2D) activation_45[0][0]	(None,	8,	8,	35)	33075	
dropout_43 (Dropout)				35)		conv2d_46[0][0]
concatenate_38 (Concatenate) concatenate_37[0][0] dropout_43[0][0]	(None,	8,	8,	140)	0	
batch_normalization_46 (BatchNo concatenate_38[0][0]	(None,	8,	8,	140)	560	
activation_46 (Activation) batch_normalization_46[0][0]	(None,				0	

	·		35)		
dropout_44 (Dropout)	(None, 8	8, 8,	35)	0	conv2d_47[0][0]
concatenate_39 (Concatenate) concatenate_38[0][0] dropout_44[0][0]	(None, 8			0	
batch_normalization_47 (BatchNo concatenate_39[0][0]	(None, 8	8, 8,	175)	700	
activation_47 (Activation) batch_normalization_47[0][0]	(None, 8	3, 8,	175)	0	
conv2d_48 (Conv2D) activation_47[0][0]	(None, 8	3, 8,	35)	55125	
dropout_45 (Dropout)				0	conv2d_48[0][0]
concatenate_40 (Concatenate) concatenate_39[0][0] dropout_45[0][0]	(None, 8	3, 8,	210)	0	
batch_normalization_48 (BatchNo concatenate_40[0][0]				840	
activation_48 (Activation) batch_normalization_48[0][0]	(None, 8	3, 8,	210)	0	
conv2d_49 (Conv2D) activation_48[0][0]	(None, 8	3, 8,	35)	66150	
dropout_46 (Dropout)			35)	0	conv2d_49[0][0]

concatenate_41 (Concatenate) concatenate_40[0][0] dropout_46[0][0]	(None,	8, 8	, 245)	0	
batch_normalization_49 (BatchNo concatenate_41[0][0]	(None,	8, 8	, 245)	980	
activation_49 (Activation) batch_normalization_49[0][0]	(None,	8, 8	, 245)	0	
conv2d_50 (Conv2D) activation_49[0][0]	(None,	8, 8	, 35)	8575	
dropout_47 (Dropout)			, 35)		conv2d_50[0][0]
average_pooling2d_6 (AveragePoodropout_47[0][0]	(None,	4, 4	, 35)	0	
batch_normalization_50 (BatchNo average_pooling2d_6[0][0]			, 35)	140	
activation_50 (Activation) batch_normalization_50[0][0]			, 35)	0	
conv2d_51 (Conv2D) activation_50[0][0]			, 35)		
dropout_48 (Dropout)	•		, 35)		conv2d_51[0][0]
concatenate_42 (Concatenate) average_pooling2d_6[0][0] dropout_48[0][0]	(None,	4, 4	, 70)		
batch_normalization_51 (BatchNo concatenate_42[0][0]				280	

activation_51 (Activation) batch_normalization_51[0][0]	(None,	4,	4,	70)	0	
	(None,	4,	4,	35)	22050	
dropout_49 (Dropout)	(None,	4,	4,	35)	0	conv2d_52[0][0]
concatenate_43 (Concatenate) concatenate_42[0][0] dropout_49[0][0]	(None,	4,	4,	105)	0	
batch_normalization_52 (BatchNo concatenate_43[0][0]				105)	420	
activation_52 (Activation) batch_normalization_52[0][0]	(None,	4,	4,		0	
	(None,	4,	4,	35)	33075	
dropout_50 (Dropout)	(None,	4,	4,	35)	0	conv2d_53[0][0]
concatenate_44 (Concatenate) concatenate_43[0][0] dropout_50[0][0]	(None,				0	
batch_normalization_53 (BatchNo concatenate_44[0][0]	(None,	4,	4,	140)	560	
activation_53 (Activation) batch_normalization_53[0][0]	(None,			140)	0	
conv2d_54 (Conv2D) activation_53[0][0]				35)	44100	

dropout_51 (Dropout)	-	4,	4,	35)	0	conv2d_54[0][0]
concatenate_45 (Concatenate) concatenate_44[0][0] dropout_51[0][0]	(None,	4,	4,	175)	0	
batch_normalization_54 (BatchNo concatenate_45[0][0]	(None,	4,	4,	175)	700	
activation_54 (Activation) batch_normalization_54[0][0]	(None,	4,	4,	175)	0	
conv2d_55 (Conv2D) activation_54[0][0]				35)	55125	
dropout_52 (Dropout)			4,		0	conv2d_55[0][0]
concatenate_46 (Concatenate) concatenate_45[0][0] dropout_52[0][0]	(None,	4,	4,	210)	0	
batch_normalization_55 (BatchNo concatenate_46[0][0]	(None,	4,	4,	210)	840	
activation_55 (Activation) batch_normalization_55[0][0]	(None,				0	
conv2d_56 (Conv2D) activation_55[0][0]		4,	4,	35)	66150	
dropout_53 (Dropout)	(None,	4,	4,	35)	0	conv2d_56[0][0]
concatenate_47 (Concatenate) concatenate_46[0][0] dropout_53[0][0]				245)		

```
batch_normalization_56 (BatchNo (None, 4, 4, 245)
                                               980
   concatenate_47[0][0]
   activation_56 (Activation)
                            (None, 4, 4, 245)
   batch_normalization_56[0][0]
   ______
   average_pooling2d_7 (AveragePoo (None, 2, 2, 245) 0
   activation_56[0][0]
   flatten_1 (Flatten)
                             (None, 980)
   average_pooling2d_7[0][0]
   dense_1 (Dense)
                             (None, 10)
                                         9810 flatten_1[0][0]
   ______
   _____
   Total params: 978,260
   Trainable params: 970,420
   Non-trainable params: 7,840
[]: print(len(model.layers))
   142
[]: X_train = X_train.astype('float32')
    X_test = X_test.astype('float32')
    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 = preprocess_data(X_train)
    # X_test = preprocess_data(X_test)
[]:  # optimizer = tf.keras.optimizers.Adam()
    optimizer = tf.keras.optimizers.Adam()
```

WARNING:tensorflow:`period` argument is deprecated. Please use `save_freq` to specify the frequency in number of batches seen.

```
[]:
    # Data augementation

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)
```

```
accuracy: 0.4355 - val_loss: 59.8250 - val_accuracy: 0.1857
Epoch 2/85
782/782 [============= ] - 68s 87ms/step - loss: 1.6475 -
accuracy: 0.5752 - val_loss: 60.2469 - val_accuracy: 0.1294
Epoch 3/85
accuracy: 0.6285 - val loss: 27.9607 - val accuracy: 0.2448
Epoch 4/85
782/782 [============= ] - 68s 87ms/step - loss: 1.2716 -
accuracy: 0.6621 - val_loss: 32.9015 - val_accuracy: 0.1750
Epoch 5/85
accuracy: 0.6845 - val_loss: 31.4301 - val_accuracy: 0.2376
Epoch 6/85
accuracy: 0.6982 - val_loss: 69.3460 - val_accuracy: 0.2062
Epoch 7/85
782/782 [============ ] - 68s 86ms/step - loss: 1.1173 -
accuracy: 0.7081 - val_loss: 52.1518 - val_accuracy: 0.1108
Epoch 8/85
782/782 [============= ] - 68s 87ms/step - loss: 1.0835 -
accuracy: 0.7197 - val_loss: 35.3169 - val_accuracy: 0.1624
Epoch 9/85
782/782 [============= ] - 68s 87ms/step - loss: 1.0553 -
accuracy: 0.7272 - val_loss: 17.7513 - val_accuracy: 0.3071
Epoch 10/85
782/782 [============== ] - 68s 87ms/step - loss: 1.0381 -
accuracy: 0.7333 - val_loss: 69.4971 - val_accuracy: 0.1180
accuracy: 0.7402 - val_loss: 23.9153 - val_accuracy: 0.2038
Epoch 12/85
782/782 [============= ] - 68s 86ms/step - loss: 0.9960 -
accuracy: 0.7444 - val_loss: 31.7959 - val_accuracy: 0.2458
Epoch 13/85
accuracy: 0.7503 - val loss: 20.4676 - val accuracy: 0.2682
Epoch 14/85
accuracy: 0.7510 - val_loss: 65.6729 - val_accuracy: 0.1270
Epoch 15/85
accuracy: 0.7556 - val_loss: 38.4958 - val_accuracy: 0.1216
Epoch 16/85
accuracy: 0.7593 - val_loss: 21.9149 - val_accuracy: 0.1659
Epoch 17/85
782/782 [============= ] - 68s 86ms/step - loss: 0.9286 -
```

```
accuracy: 0.7617 - val_loss: 29.8570 - val_accuracy: 0.2042
   Epoch 18/85
   782/782 [============== ] - 68s 87ms/step - loss: 0.9197 -
   accuracy: 0.7670 - val_loss: 31.2535 - val_accuracy: 0.1697
   Epoch 19/85
   782/782 [============== ] - 68s 87ms/step - loss: 0.9126 -
   accuracy: 0.7675 - val_loss: 38.3681 - val_accuracy: 0.1559
   Epoch 00019: early stopping
[]: %tensorboard --logdir logs/fit
   <IPython.core.display.Javascript object>
[]: # Test the model
    score = model.evaluate(X_test, y_test, verbose=1)
    print('Test loss:', score[0])
    print('Test accuracy:', score[1])
   accuracy: 0.1559
   Test loss: 38.36806106567383
   Test accuracy: 0.1559000015258789
[]: # Save the trained weights in to .h5 format
    model.save_weights("DNST_model.h5")
    print("Saved model to disk")
   Saved model to disk
[]:
```