

In []:

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

In [1]:

```
import tensorflow as tf
from tensorflow import keras
from keras.datasets import cifar10
from tensorflow.keras import models, layers
from tensorflow.keras.models import Model
from tensorflow.keras.layers import BatchNormalization, Activation, Flatten, Dense, Dropout, Flatten, Conv2D, MaxPooling2D, Concatenate, SeparableConv2D
from tensorflow.keras.optimizers import Adam
from keras.preprocessing.image import ImageDataGenerator

import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```

In [31]:

```
# Hyperparameters
batch_size = 32
num_classes = 10
epochs = 10
l = 12
num_filter = 18
compression = 1
dropout_rate = 0
```

In [32]:

```
# 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)
```

In [33]:

```
X_train = X_train.astype('float32')
X_test = X_test.astype('float32')
X_train /= 255
X_test /= 255

print(X_train.shape)
print(X_test.shape)
```

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

Image Augmentation

In [6]:

```
aug = ImageDataGenerator(shear_range=0.2,
                        zoom_range=0.2,
                        height_shift_range=0.1,
                        horizontal_flip=True)
```

Model Building and Training

In [45]:

```

# Dense Block
def denseblock(input, num_filter = 48, dropout_rate = 0.0):
    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)
        concat = layers.Concatenate(axis=-1)([temp,Conv2D_3_3])
        temp = concat

    return temp

## transition Block
def transition(input, num_filter = 48, dropout_rate = 0.0):
    global compression
    BatchNorm = layers.BatchNormalization()(input)
    relu = layers.Activation('relu')(BatchNorm)
    Conv2D_BottleNeck = layers.Conv2D(int(num_filter*compression), (1,1), use_bias=False
e ,padding='same')(relu)
    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

```

In [46]:

```

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 [47]:

```

model = Model(inputs=[input], outputs=[output])

```

In [48]:

```
model.summary()
```

Model: "model_5"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_6 (InputLayer)	[(None, 32, 32, 3)]	0	
conv2d_312 (Conv2D)	(None, 32, 32, 18)	486	input_6
batch_normalization_312 (Batch Normalization)	(None, 32, 32, 18)	72	conv2d_312[0][0]
activation_312 (Activation)	(None, 32, 32, 18)	0	batch_normalization_312[0][0]
conv2d_313 (Conv2D)	(None, 32, 32, 18)	2916	activation_312[0][0]
concatenate_288 (Concatenate)	(None, 32, 32, 36)	0	conv2d_312[0][0] conv2d_313[0][0]
batch_normalization_313 (Batch Normalization)	(None, 32, 32, 36)	144	concatenate_288[0][0]
activation_313 (Activation)	(None, 32, 32, 36)	0	batch_normalization_313[0][0]
conv2d_314 (Conv2D)	(None, 32, 32, 18)	5832	activation_313[0][0]
concatenate_289 (Concatenate)	(None, 32, 32, 54)	0	concatenate_288[0][0] conv2d_314[0][0]
batch_normalization_314 (Batch Normalization)	(None, 32, 32, 54)	216	concatenate_289[0][0]
activation_314 (Activation)	(None, 32, 32, 54)	0	batch_normalization_314[0][0]
conv2d_315 (Conv2D)	(None, 32, 32, 18)	8748	activation_314[0][0]

concatenate_290 (Concatenate) te_289[0][0]	(None, 32, 32, 72)	0	concatena conv2d_31 5[0][0]
batch_normalization_315 (BatchN te_290[0][0]	(None, 32, 32, 72)	288	concatena
activation_315 (Activation) malization_315[0][0]	(None, 32, 32, 72)	0	batch_nor
conv2d_316 (Conv2D) n_315[0][0]	(None, 32, 32, 18)	11664	activatio
concatenate_291 (Concatenate) te_290[0][0]	(None, 32, 32, 90)	0	concatena conv2d_31 6[0][0]
batch_normalization_316 (BatchN te_291[0][0]	(None, 32, 32, 90)	360	concatena
activation_316 (Activation) malization_316[0][0]	(None, 32, 32, 90)	0	batch_nor
conv2d_317 (Conv2D) n_316[0][0]	(None, 32, 32, 18)	14580	activatio
concatenate_292 (Concatenate) te_291[0][0]	(None, 32, 32, 108)	0	concatena conv2d_31 7[0][0]
batch_normalization_317 (BatchN te_292[0][0]	(None, 32, 32, 108)	432	concatena
activation_317 (Activation) malization_317[0][0]	(None, 32, 32, 108)	0	batch_nor
conv2d_318 (Conv2D) n_317[0][0]	(None, 32, 32, 18)	17496	activatio
concatenate_293 (Concatenate) te_292[0][0]	(None, 32, 32, 126)	0	concatena conv2d_31 8[0][0]

batch_normalization_318 (Batch Normalization)	(None, 32, 32, 126)	504	concatenate_293[0][0]
activation_318 (Activation)	(None, 32, 32, 126)	0	batch_normalization_318[0][0]
conv2d_319 (Conv2D)	(None, 32, 32, 18)	20412	activation_318[0][0]
concatenate_294 (Concatenate)	(None, 32, 32, 144)	0	concatenate_293[0][0]
			conv2d_319[0][0]
batch_normalization_319 (Batch Normalization)	(None, 32, 32, 144)	576	concatenate_294[0][0]
activation_319 (Activation)	(None, 32, 32, 144)	0	batch_normalization_319[0][0]
conv2d_320 (Conv2D)	(None, 32, 32, 18)	23328	activation_319[0][0]
concatenate_295 (Concatenate)	(None, 32, 32, 162)	0	concatenate_294[0][0]
			conv2d_320[0][0]
batch_normalization_320 (Batch Normalization)	(None, 32, 32, 162)	648	concatenate_295[0][0]
activation_320 (Activation)	(None, 32, 32, 162)	0	batch_normalization_320[0][0]
conv2d_321 (Conv2D)	(None, 32, 32, 18)	26244	activation_320[0][0]
concatenate_296 (Concatenate)	(None, 32, 32, 180)	0	concatenate_295[0][0]
			conv2d_321[0][0]
batch_normalization_321 (Batch Normalization)	(None, 32, 32, 180)	720	concatenate_296[0][0]
activation_321 (Activation)	(None, 32, 32, 180)	0	batch_normalization_321[0][0]

conv2d_322 (Conv2D) n_321[0][0]	(None, 32, 32, 18)	29160	activation_322[0][0]
concatenate_297 (Concatenate) te_296[0][0]	(None, 32, 32, 198)	0	concatenate_296[0][0]
batch_normalization_322 (Batch Normalization) te_297[0][0]	(None, 32, 32, 198)	792	concatenate_297[0][0]
activation_322 (Activation) malization_322[0][0]	(None, 32, 32, 198)	0	batch_normalization_322[0][0]
conv2d_323 (Conv2D) n_322[0][0]	(None, 32, 32, 18)	32076	activation_323[0][0]
concatenate_298 (Concatenate) te_297[0][0]	(None, 32, 32, 216)	0	concatenate_297[0][0]
batch_normalization_323 (Batch Normalization) te_298[0][0]	(None, 32, 32, 216)	864	concatenate_298[0][0]
activation_323 (Activation) malization_323[0][0]	(None, 32, 32, 216)	0	batch_normalization_323[0][0]
conv2d_324 (Conv2D) n_323[0][0]	(None, 32, 32, 18)	34992	activation_324[0][0]
concatenate_299 (Concatenate) te_298[0][0]	(None, 32, 32, 234)	0	concatenate_298[0][0]
batch_normalization_324 (Batch Normalization) te_299[0][0]	(None, 32, 32, 234)	936	concatenate_299[0][0]
activation_324 (Activation) malization_324[0][0]	(None, 32, 32, 234)	0	batch_normalization_324[0][0]
conv2d_325 (Conv2D) n_324[0][0]	(None, 32, 32, 18)	4212	activation_325[0][0]
average_pooling2d_24 (Average Pooling) 5[0][0]	(None, 16, 16, 18)	0	conv2d_325[0][0]

batch_normalization_325 (Batch Normalization)	(None, 16, 16, 18)	72	average_pooling2d_24[0][0]
activation_325 (Activation)	(None, 16, 16, 18)	0	batch_normalization_325[0][0]
conv2d_326 (Conv2D)	(None, 16, 16, 18)	2916	activation_325[0][0]
concatenate_300 (Concatenate)	(None, 16, 16, 36)	0	average_pooling2d_24[0][0]
			conv2d_326[0][0]
batch_normalization_326 (Batch Normalization)	(None, 16, 16, 36)	144	concatenate_300[0][0]
activation_326 (Activation)	(None, 16, 16, 36)	0	batch_normalization_326[0][0]
conv2d_327 (Conv2D)	(None, 16, 16, 18)	5832	activation_326[0][0]
concatenate_301 (Concatenate)	(None, 16, 16, 54)	0	concatenate_300[0][0]
			conv2d_327[0][0]
batch_normalization_327 (Batch Normalization)	(None, 16, 16, 54)	216	concatenate_301[0][0]
activation_327 (Activation)	(None, 16, 16, 54)	0	batch_normalization_327[0][0]
conv2d_328 (Conv2D)	(None, 16, 16, 18)	8748	activation_327[0][0]
concatenate_302 (Concatenate)	(None, 16, 16, 72)	0	concatenate_301[0][0]
			conv2d_328[0][0]
batch_normalization_328 (Batch Normalization)	(None, 16, 16, 72)	288	concatenate_302[0][0]
activation_328 (Activation)	(None, 16, 16, 72)	0	batch_normalization_328[0][0]

malization_328[0][0]

conv2d_329 (Conv2D) n_328[0][0]	(None, 16, 16, 18)	11664	activation_328[0][0]
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concatenate_303 (Concatenate) te_302[0][0]	(None, 16, 16, 90)	0	concatenate_302[0][0]
9[0][0]			conv2d_32

batch_normalization_329 (Batch Normalization) te_303[0][0]	(None, 16, 16, 90)	360	concatenate_303[0][0]
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activation_329 (Activation) malization_329[0][0]	(None, 16, 16, 90)	0	batch_normalization_329[0][0]
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conv2d_330 (Conv2D) n_329[0][0]	(None, 16, 16, 18)	14580	activation_329[0][0]
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concatenate_304 (Concatenate) te_303[0][0]	(None, 16, 16, 108)	0	concatenate_303[0][0]
0[0][0]			conv2d_33

batch_normalization_330 (Batch Normalization) te_304[0][0]	(None, 16, 16, 108)	432	concatenate_304[0][0]
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activation_330 (Activation) malization_330[0][0]	(None, 16, 16, 108)	0	batch_normalization_330[0][0]
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conv2d_331 (Conv2D) n_330[0][0]	(None, 16, 16, 18)	17496	activation_330[0][0]
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concatenate_305 (Concatenate) te_304[0][0]	(None, 16, 16, 126)	0	concatenate_304[0][0]
1[0][0]			conv2d_33

batch_normalization_331 (Batch Normalization) te_305[0][0]	(None, 16, 16, 126)	504	concatenate_305[0][0]
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activation_331 (Activation) malization_331[0][0]	(None, 16, 16, 126)	0	batch_normalization_331[0][0]
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conv2d_332 (Conv2D) n_331[0][0]	(None, 16, 16, 18)	20412	activation_331[0][0]
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concatenate_306 (Concatenate)	(None, 16, 16, 144)	0	concatenate_305[0][0]
2[0][0]			conv2d_33
batch_normalization_332 (Batch Normalization)	(None, 16, 16, 144)	576	concatenate_306[0][0]
activation_332 (Activation)	(None, 16, 16, 144)	0	batch_normalization_332[0][0]
conv2d_333 (Conv2D)	(None, 16, 16, 18)	23328	activation_332[0][0]
concatenate_307 (Concatenate)	(None, 16, 16, 162)	0	concatenate_306[0][0]
3[0][0]			conv2d_33
batch_normalization_333 (Batch Normalization)	(None, 16, 16, 162)	648	concatenate_307[0][0]
activation_333 (Activation)	(None, 16, 16, 162)	0	batch_normalization_333[0][0]
conv2d_334 (Conv2D)	(None, 16, 16, 18)	26244	activation_333[0][0]
concatenate_308 (Concatenate)	(None, 16, 16, 180)	0	concatenate_307[0][0]
4[0][0]			conv2d_33
batch_normalization_334 (Batch Normalization)	(None, 16, 16, 180)	720	concatenate_308[0][0]
activation_334 (Activation)	(None, 16, 16, 180)	0	batch_normalization_334[0][0]
conv2d_335 (Conv2D)	(None, 16, 16, 18)	29160	activation_334[0][0]
concatenate_309 (Concatenate)	(None, 16, 16, 198)	0	concatenate_308[0][0]
5[0][0]			conv2d_33
batch_normalization_335 (Batch Normalization)	(None, 16, 16, 198)	792	concatenate_309[0][0]

te_309[0][0]

activation_335 (Activation) malization_335[0][0]	(None, 16, 16, 198)	0	batch_nor
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conv2d_336 (Conv2D) n_335[0][0]	(None, 16, 16, 18)	32076	activatio
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concatenate_310 (Concatenate) te_309[0][0]	(None, 16, 16, 216)	0	concatena
6[0][0]			conv2d_33

batch_normalization_336 (BatchN te_310[0][0]	(None, 16, 16, 216)	864	concatena
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activation_336 (Activation) malization_336[0][0]	(None, 16, 16, 216)	0	batch_nor
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conv2d_337 (Conv2D) n_336[0][0]	(None, 16, 16, 18)	34992	activatio
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concatenate_311 (Concatenate) te_310[0][0]	(None, 16, 16, 234)	0	concatena
7[0][0]			conv2d_33

batch_normalization_337 (BatchN te_311[0][0]	(None, 16, 16, 234)	936	concatena
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activation_337 (Activation) malization_337[0][0]	(None, 16, 16, 234)	0	batch_nor
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conv2d_338 (Conv2D) n_337[0][0]	(None, 16, 16, 18)	4212	activatio
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average_pooling2d_25 (AveragePo 8[0][0]	(None, 8, 8, 18)	0	conv2d_33
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batch_normalization_338 (BatchN ooling2d_25[0][0]	(None, 8, 8, 18)	72	average_p
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activation_338 (Activation) malization_338[0][0]	(None, 8, 8, 18)	0	batch_nor
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conv2d_339 (Conv2D) n_338[0][0]	(None, 8, 8, 18)	2916	activatio
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concatenate_312 (Concatenate)	(None, 8, 8, 36)	0	average_p ooling2d_25[0][0]
			conv2d_33 9[0][0]
batch_normalization_339 (BatchN	(None, 8, 8, 36)	144	concatena te_312[0][0]
activation_339 (Activation)	(None, 8, 8, 36)	0	batch_nor malization_339[0][0]
conv2d_340 (Conv2D)	(None, 8, 8, 18)	5832	activatio n_339[0][0]
concatenate_313 (Concatenate)	(None, 8, 8, 54)	0	concatena te_312[0][0]
			conv2d_34 0[0][0]
batch_normalization_340 (BatchN	(None, 8, 8, 54)	216	concatena te_313[0][0]
activation_340 (Activation)	(None, 8, 8, 54)	0	batch_nor malization_340[0][0]
conv2d_341 (Conv2D)	(None, 8, 8, 18)	8748	activatio n_340[0][0]
concatenate_314 (Concatenate)	(None, 8, 8, 72)	0	concatena te_313[0][0]
			conv2d_34 1[0][0]
batch_normalization_341 (BatchN	(None, 8, 8, 72)	288	concatena te_314[0][0]
activation_341 (Activation)	(None, 8, 8, 72)	0	batch_nor malization_341[0][0]
conv2d_342 (Conv2D)	(None, 8, 8, 18)	11664	activatio n_341[0][0]
concatenate_315 (Concatenate)	(None, 8, 8, 90)	0	concatena te_314[0][0]
			conv2d_34 2[0][0]

batch_normalization_342 (Batch Normalization)	(None, 8, 8, 90)	360	concatenate_315[0][0]
activation_342 (Activation)	(None, 8, 8, 90)	0	batch_normalization_342[0][0]
conv2d_343 (Conv2D)	(None, 8, 8, 18)	14580	activation_342[0][0]
concatenate_316 (Concatenate)	(None, 8, 8, 108)	0	concatenate_315[0][0]
			conv2d_343[0][0]
batch_normalization_343 (Batch Normalization)	(None, 8, 8, 108)	432	concatenate_316[0][0]
activation_343 (Activation)	(None, 8, 8, 108)	0	batch_normalization_343[0][0]
conv2d_344 (Conv2D)	(None, 8, 8, 18)	17496	activation_343[0][0]
concatenate_317 (Concatenate)	(None, 8, 8, 126)	0	concatenate_316[0][0]
			conv2d_344[0][0]
batch_normalization_344 (Batch Normalization)	(None, 8, 8, 126)	504	concatenate_317[0][0]
activation_344 (Activation)	(None, 8, 8, 126)	0	batch_normalization_344[0][0]
conv2d_345 (Conv2D)	(None, 8, 8, 18)	20412	activation_344[0][0]
concatenate_318 (Concatenate)	(None, 8, 8, 144)	0	concatenate_317[0][0]
			conv2d_345[0][0]
batch_normalization_345 (Batch Normalization)	(None, 8, 8, 144)	576	concatenate_318[0][0]
activation_345 (Activation)	(None, 8, 8, 144)	0	batch_normalization_345[0][0]

conv2d_346 (Conv2D) n_345[0][0]	(None, 8, 8, 18)	23328	activation_346[0][0]
concatenate_319 (Concatenate) te_318[0][0] 6[0][0]	(None, 8, 8, 162)	0	concatenate_319[0][0] conv2d_346[0][0]
batch_normalization_346 (Batch Normalization) te_319[0][0]	(None, 8, 8, 162)	648	concatenate_319[0][0]
activation_346 (Activation) malization_346[0][0]	(None, 8, 8, 162)	0	batch_normalization_346[0][0]
conv2d_347 (Conv2D) n_346[0][0]	(None, 8, 8, 18)	26244	activation_347[0][0]
concatenate_320 (Concatenate) te_319[0][0] 7[0][0]	(None, 8, 8, 180)	0	concatenate_319[0][0] conv2d_347[0][0]
batch_normalization_347 (Batch Normalization) te_320[0][0]	(None, 8, 8, 180)	720	concatenate_320[0][0]
activation_347 (Activation) malization_347[0][0]	(None, 8, 8, 180)	0	batch_normalization_347[0][0]
conv2d_348 (Conv2D) n_347[0][0]	(None, 8, 8, 18)	29160	activation_348[0][0]
concatenate_321 (Concatenate) te_320[0][0] 8[0][0]	(None, 8, 8, 198)	0	concatenate_320[0][0] conv2d_348[0][0]
batch_normalization_348 (Batch Normalization) te_321[0][0]	(None, 8, 8, 198)	792	concatenate_321[0][0]
activation_348 (Activation) malization_348[0][0]	(None, 8, 8, 198)	0	batch_normalization_348[0][0]
conv2d_349 (Conv2D) n_348[0][0]	(None, 8, 8, 18)	32076	activation_349[0][0]
concatenate_322 (Concatenate)	(None, 8, 8, 216)	0	concatenate_322[0][0]

te_321[0][0]			conv2d_34
9[0][0]			
batch_normalization_349 (BatchN	(None, 8, 8, 216)	864	concatena
te_322[0][0]			
activation_349 (Activation)	(None, 8, 8, 216)	0	batch_nor
malization_349[0][0]			
conv2d_350 (Conv2D)	(None, 8, 8, 18)	34992	activatio
n_349[0][0]			
concatenate_323 (Concatenate)	(None, 8, 8, 234)	0	concatena
te_322[0][0]			conv2d_35
0[0][0]			
batch_normalization_350 (BatchN	(None, 8, 8, 234)	936	concatena
te_323[0][0]			
activation_350 (Activation)	(None, 8, 8, 234)	0	batch_nor
malization_350[0][0]			
conv2d_351 (Conv2D)	(None, 8, 8, 18)	4212	activatio
n_350[0][0]			
average_pooling2d_26 (AveragePo	(None, 4, 4, 18)	0	conv2d_35
1[0][0]			
batch_normalization_351 (BatchN	(None, 4, 4, 18)	72	average_p
ooling2d_26[0][0]			
activation_351 (Activation)	(None, 4, 4, 18)	0	batch_nor
malization_351[0][0]			
conv2d_352 (Conv2D)	(None, 4, 4, 18)	2916	activatio
n_351[0][0]			
concatenate_324 (Concatenate)	(None, 4, 4, 36)	0	average_p
ooling2d_26[0][0]			conv2d_35
2[0][0]			
batch_normalization_352 (BatchN	(None, 4, 4, 36)	144	concatena
te_324[0][0]			

activation_352 (Activation) malization_352[0][0]	(None, 4, 4, 36)	0	batch_nor
conv2d_353 (Conv2D) n_352[0][0]	(None, 4, 4, 18)	5832	activatio
concatenate_325 (Concatenate) te_324[0][0]	(None, 4, 4, 54)	0	concatena
3[0][0]			conv2d_35
batch_normalization_353 (BatchN te_325[0][0]	(None, 4, 4, 54)	216	concatena
activation_353 (Activation) malization_353[0][0]	(None, 4, 4, 54)	0	batch_nor
conv2d_354 (Conv2D) n_353[0][0]	(None, 4, 4, 18)	8748	activatio
concatenate_326 (Concatenate) te_325[0][0]	(None, 4, 4, 72)	0	concatena
4[0][0]			conv2d_35
batch_normalization_354 (BatchN te_326[0][0]	(None, 4, 4, 72)	288	concatena
activation_354 (Activation) malization_354[0][0]	(None, 4, 4, 72)	0	batch_nor
conv2d_355 (Conv2D) n_354[0][0]	(None, 4, 4, 18)	11664	activatio
concatenate_327 (Concatenate) te_326[0][0]	(None, 4, 4, 90)	0	concatena
5[0][0]			conv2d_35
batch_normalization_355 (BatchN te_327[0][0]	(None, 4, 4, 90)	360	concatena
activation_355 (Activation) malization_355[0][0]	(None, 4, 4, 90)	0	batch_nor
conv2d_356 (Conv2D) n_355[0][0]	(None, 4, 4, 18)	14580	activatio

concatenate_328 (Concatenate) te_327[0][0]	(None, 4, 4, 108)	0	concatena conv2d_35 6[0][0]
batch_normalization_356 (BatchN te_328[0][0]	(None, 4, 4, 108)	432	concatena
activation_356 (Activation) malization_356[0][0]	(None, 4, 4, 108)	0	batch_nor
conv2d_357 (Conv2D) n_356[0][0]	(None, 4, 4, 18)	17496	activatio
concatenate_329 (Concatenate) te_328[0][0]	(None, 4, 4, 126)	0	concatena conv2d_35 7[0][0]
batch_normalization_357 (BatchN te_329[0][0]	(None, 4, 4, 126)	504	concatena
activation_357 (Activation) malization_357[0][0]	(None, 4, 4, 126)	0	batch_nor
conv2d_358 (Conv2D) n_357[0][0]	(None, 4, 4, 18)	20412	activatio
concatenate_330 (Concatenate) te_329[0][0]	(None, 4, 4, 144)	0	concatena conv2d_35 8[0][0]
batch_normalization_358 (BatchN te_330[0][0]	(None, 4, 4, 144)	576	concatena
activation_358 (Activation) malization_358[0][0]	(None, 4, 4, 144)	0	batch_nor
conv2d_359 (Conv2D) n_358[0][0]	(None, 4, 4, 18)	23328	activatio
concatenate_331 (Concatenate) te_330[0][0]	(None, 4, 4, 162)	0	concatena conv2d_35 9[0][0]

batch_normalization_359 (Batch Normalization)	(None, 4, 4, 162)	648	concatenate_331[0][0]
activation_359 (Activation)	(None, 4, 4, 162)	0	batch_normalization_359[0][0]
conv2d_360 (Conv2D)	(None, 4, 4, 18)	26244	activation_359[0][0]
concatenate_332 (Concatenate)	(None, 4, 4, 180)	0	concatenate_331[0][0]
			conv2d_360[0][0]
batch_normalization_360 (Batch Normalization)	(None, 4, 4, 180)	720	concatenate_332[0][0]
activation_360 (Activation)	(None, 4, 4, 180)	0	batch_normalization_360[0][0]
conv2d_361 (Conv2D)	(None, 4, 4, 18)	29160	activation_360[0][0]
concatenate_333 (Concatenate)	(None, 4, 4, 198)	0	concatenate_332[0][0]
			conv2d_361[0][0]
batch_normalization_361 (Batch Normalization)	(None, 4, 4, 198)	792	concatenate_333[0][0]
activation_361 (Activation)	(None, 4, 4, 198)	0	batch_normalization_361[0][0]
conv2d_362 (Conv2D)	(None, 4, 4, 18)	32076	activation_361[0][0]
concatenate_334 (Concatenate)	(None, 4, 4, 216)	0	concatenate_333[0][0]
			conv2d_362[0][0]
batch_normalization_362 (Batch Normalization)	(None, 4, 4, 216)	864	concatenate_334[0][0]
activation_362 (Activation)	(None, 4, 4, 216)	0	batch_normalization_362[0][0]

conv2d_363 (Conv2D) n_362[0][0]	(None, 4, 4, 18)	34992	activation_363[0][0]
concatenate_335 (Concatenate) te_334[0][0]	(None, 4, 4, 234)	0	concatenate_335[0][0]
batch_normalization_363 (Batch Normalization) te_335[0][0]	(None, 4, 4, 234)	936	concatenate_335[0][0]
activation_363 (Activation) malization_363[0][0]	(None, 4, 4, 234)	0	batch_normalization_363[0][0]
average_pooling2d_27 (Average Pooling) n_363[0][0]	(None, 2, 2, 234)	0	activation_363[0][0]
flatten_5 (Flatten) ooling2d_27[0][0]	(None, 936)	0	average_pooling2d_27[0][0]
dense_5 (Dense) [0][0]	(None, 10)	9370	flatten_5[0][0]
=====			
Total params: 958,492			
Trainable params: 945,388			
Non-trainable params: 13,104			

In [21]:

```
print(len(model.layers))
```

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In [49]:

In [51]:

```
aug = ImageDataGenerator(shear_range=0.2,
                          zoom_range=0.2,
                          height_shift_range=0.1,
                          horizontal_flip=True)

# determine Loss function and Optimizer
model.compile(loss='categorical_crossentropy',
              optimizer=Adam(learning_rate=0.001),
              metrics=['accuracy'])
```

In [52]:

```
cp_callback = tf.keras.callbacks.ModelCheckpoint(  
    filepath="models/CNN_CIFR_APR_21_{epoch:04d}.hdf5",  
    verbose=1,  
    save_weights_only=True,  
    period=10)  
  
res_model = model.fit(  
    x = aug.flow(X_train, y_train, batch_size=batch_size,),  
    epochs=150,  
    verbose=1,  
    validation_data=(X_test, y_test),  
    callbacks = [cp_callback])
```

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

Epoch 1/150

1563/1563 [=====] - 73s 44ms/step - loss: 2.0559
- accuracy: 0.2771 - val_loss: 1.4818 - val_accuracy: 0.4724

Epoch 2/150

1563/1563 [=====] - 67s 43ms/step - loss: 1.3637
- accuracy: 0.5087 - val_loss: 1.3150 - val_accuracy: 0.5554

Epoch 3/150

1563/1563 [=====] - 67s 43ms/step - loss: 1.0332
- accuracy: 0.6286 - val_loss: 1.2683 - val_accuracy: 0.5775

Epoch 4/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.8787
- accuracy: 0.6897 - val_loss: 1.5207 - val_accuracy: 0.5580

Epoch 5/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.7662
- accuracy: 0.7294 - val_loss: 1.2011 - val_accuracy: 0.6211

Epoch 6/150

1563/1563 [=====] - 66s 42ms/step - loss: 0.6816
- accuracy: 0.7630 - val_loss: 0.9039 - val_accuracy: 0.7009

Epoch 7/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.6421
- accuracy: 0.7774 - val_loss: 0.7522 - val_accuracy: 0.7519

Epoch 8/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.5894
- accuracy: 0.7970 - val_loss: 0.8303 - val_accuracy: 0.7195

Epoch 9/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.5542
- accuracy: 0.8095 - val_loss: 0.7475 - val_accuracy: 0.7489

Epoch 10/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.5222
- accuracy: 0.8192 - val_loss: 0.5490 - val_accuracy: 0.8099

Epoch 00010: saving model to models/CNN_CIFR_APR_21_0010.hdf5

Epoch 11/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.4958
- accuracy: 0.8288 - val_loss: 0.6702 - val_accuracy: 0.7814

Epoch 12/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.4779
- accuracy: 0.8363 - val_loss: 0.7194 - val_accuracy: 0.7740

Epoch 13/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.4561
- accuracy: 0.8430 - val_loss: 0.7384 - val_accuracy: 0.7725

Epoch 14/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.4260
- accuracy: 0.8521 - val_loss: 0.5619 - val_accuracy: 0.8110

Epoch 15/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.4232
- accuracy: 0.8553 - val_loss: 0.5817 - val_accuracy: 0.8151

Epoch 16/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.4044
- accuracy: 0.8586 - val_loss: 0.6623 - val_accuracy: 0.7926

Epoch 17/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.3934
- accuracy: 0.8630 - val_loss: 0.5560 - val_accuracy: 0.8214

Epoch 18/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.3824
- accuracy: 0.8664 - val_loss: 0.5246 - val_accuracy: 0.8287

Epoch 19/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.3637
- accuracy: 0.8727 - val_loss: 0.8526 - val_accuracy: 0.7577

Epoch 20/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.3529
- accuracy: 0.8789 - val_loss: 0.4970 - val_accuracy: 0.8354

Epoch 00020: saving model to models/CNN_CIFR_APR_21_0020.hdf5

Epoch 21/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.3473
- accuracy: 0.8799 - val_loss: 0.5231 - val_accuracy: 0.8289

Epoch 22/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.3314
- accuracy: 0.8862 - val_loss: 0.4715 - val_accuracy: 0.8472

Epoch 23/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.3326
- accuracy: 0.8844 - val_loss: 0.4497 - val_accuracy: 0.8522

Epoch 24/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.3136
- accuracy: 0.8887 - val_loss: 0.4620 - val_accuracy: 0.8558

Epoch 25/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.3057
- accuracy: 0.8926 - val_loss: 0.5079 - val_accuracy: 0.8368

Epoch 26/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2968
- accuracy: 0.8960 - val_loss: 0.5256 - val_accuracy: 0.8370

Epoch 27/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2956
- accuracy: 0.8980 - val_loss: 0.3803 - val_accuracy: 0.8746

Epoch 28/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2905
- accuracy: 0.9009 - val_loss: 0.4478 - val_accuracy: 0.8580

Epoch 29/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2867
- accuracy: 0.9000 - val_loss: 0.4346 - val_accuracy: 0.8596

Epoch 30/150

1563/1563 [=====] - 68s 43ms/step - loss: 0.2743
- accuracy: 0.9045 - val_loss: 0.4546 - val_accuracy: 0.8608

Epoch 00030: saving model to models/CNN_CIFR_APR_21_0030.hdf5

Epoch 31/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2683
- accuracy: 0.9080 - val_loss: 0.5228 - val_accuracy: 0.8428

Epoch 32/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2710
- accuracy: 0.9055 - val_loss: 0.4431 - val_accuracy: 0.8633

Epoch 33/150

1563/1563 [=====] - 66s 42ms/step - loss: 0.2574
- accuracy: 0.9113 - val_loss: 0.5754 - val_accuracy: 0.8269

Epoch 34/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.2561
- accuracy: 0.9081 - val_loss: 0.3920 - val_accuracy: 0.8749

Epoch 35/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.2506
- accuracy: 0.9132 - val_loss: 0.5423 - val_accuracy: 0.8452

Epoch 36/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2508
- accuracy: 0.9131 - val_loss: 0.3805 - val_accuracy: 0.8788

Epoch 37/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2439
- accuracy: 0.9164 - val_loss: 0.4357 - val_accuracy: 0.8644

Epoch 38/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2356
- accuracy: 0.9168 - val_loss: 0.4718 - val_accuracy: 0.8556

Epoch 39/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2350
- accuracy: 0.9178 - val_loss: 0.4460 - val_accuracy: 0.8666

Epoch 40/150

1563/1563 [=====] - 66s 42ms/step - loss: 0.2256
- accuracy: 0.9202 - val_loss: 0.4293 - val_accuracy: 0.8715

Epoch 00040: saving model to models/CNN_CIFR_APR_21_0040.hdf5

Epoch 41/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2238
- accuracy: 0.9221 - val_loss: 0.3896 - val_accuracy: 0.8797

Epoch 42/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2212
- accuracy: 0.9235 - val_loss: 0.4427 - val_accuracy: 0.8668

Epoch 43/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2175
- accuracy: 0.9235 - val_loss: 0.3910 - val_accuracy: 0.8836

Epoch 44/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2178
- accuracy: 0.9247 - val_loss: 0.5024 - val_accuracy: 0.8535

Epoch 45/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2113
- accuracy: 0.9260 - val_loss: 0.4228 - val_accuracy: 0.8630

Epoch 46/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2051
- accuracy: 0.9284 - val_loss: 0.3894 - val_accuracy: 0.8773

Epoch 47/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2042
- accuracy: 0.9309 - val_loss: 0.3584 - val_accuracy: 0.8906

Epoch 48/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1979
- accuracy: 0.9295 - val_loss: 0.4130 - val_accuracy: 0.8801

Epoch 49/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.2004
- accuracy: 0.9297 - val_loss: 0.3788 - val_accuracy: 0.8801

Epoch 50/150

1563/1563 [=====] - 66s 43ms/step - loss: 0.1926
- accuracy: 0.9327 - val_loss: 0.3910 - val_accuracy: 0.8810

Epoch 00050: saving model to models/CNN_CIFR_APR_21_0050.hdf5

Epoch 51/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1924
- accuracy: 0.9326 - val_loss: 0.3595 - val_accuracy: 0.8904

Epoch 52/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1893
- accuracy: 0.9343 - val_loss: 0.4073 - val_accuracy: 0.8819

Epoch 53/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1886
- accuracy: 0.9343 - val_loss: 0.3884 - val_accuracy: 0.8874

Epoch 54/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1835
- accuracy: 0.9357 - val_loss: 0.4128 - val_accuracy: 0.8698

Epoch 55/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1859
- accuracy: 0.9356 - val_loss: 0.4159 - val_accuracy: 0.8852

Epoch 56/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1787
- accuracy: 0.9351 - val_loss: 0.4618 - val_accuracy: 0.8741

Epoch 57/150

1563/1563 [=====] - 66s 42ms/step - loss: 0.1783
- accuracy: 0.9356 - val_loss: 0.3833 - val_accuracy: 0.8864

Epoch 58/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1774
- accuracy: 0.9395 - val_loss: 0.4322 - val_accuracy: 0.8756

Epoch 59/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1746
- accuracy: 0.9386 - val_loss: 0.5668 - val_accuracy: 0.8438

Epoch 60/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1712
- accuracy: 0.9403 - val_loss: 0.5007 - val_accuracy: 0.8572

Epoch 00060: saving model to models/CNN_CIFR_APR_21_0060.hdf5

Epoch 61/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1727
- accuracy: 0.9412 - val_loss: 0.4146 - val_accuracy: 0.8800

Epoch 62/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1606
- accuracy: 0.9448 - val_loss: 0.3966 - val_accuracy: 0.8819

Epoch 63/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1624
- accuracy: 0.9431 - val_loss: 0.4694 - val_accuracy: 0.8750

Epoch 64/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1657
- accuracy: 0.9416 - val_loss: 0.3988 - val_accuracy: 0.8865

Epoch 65/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1577
- accuracy: 0.9452 - val_loss: 0.4243 - val_accuracy: 0.8873

Epoch 66/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1551
- accuracy: 0.9465 - val_loss: 0.3759 - val_accuracy: 0.8915

Epoch 67/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1589
- accuracy: 0.9436 - val_loss: 0.3728 - val_accuracy: 0.8912

Epoch 68/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1544
- accuracy: 0.9466 - val_loss: 0.4373 - val_accuracy: 0.8815

Epoch 69/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1531
- accuracy: 0.9462 - val_loss: 0.3604 - val_accuracy: 0.8975

Epoch 70/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1528
- accuracy: 0.9475 - val_loss: 0.4536 - val_accuracy: 0.8818

Epoch 00070: saving model to models/CNN_CIFR_APR_21_0070.hdf5

Epoch 71/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1522
- accuracy: 0.9464 - val_loss: 0.4337 - val_accuracy: 0.8819

Epoch 72/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1503
- accuracy: 0.9469 - val_loss: 0.3924 - val_accuracy: 0.8940

Epoch 73/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1411
- accuracy: 0.9512 - val_loss: 0.4700 - val_accuracy: 0.8790

Epoch 74/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1460
- accuracy: 0.9487 - val_loss: 0.3794 - val_accuracy: 0.8947

Epoch 75/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1416
- accuracy: 0.9499 - val_loss: 0.4935 - val_accuracy: 0.8671

Epoch 76/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1380
- accuracy: 0.9512 - val_loss: 0.3925 - val_accuracy: 0.8887

Epoch 77/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1455
- accuracy: 0.9503 - val_loss: 0.4475 - val_accuracy: 0.8886

Epoch 78/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1444
- accuracy: 0.9488 - val_loss: 0.4721 - val_accuracy: 0.8780

Epoch 79/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1383
- accuracy: 0.9532 - val_loss: 0.3739 - val_accuracy: 0.8961

Epoch 80/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1349
- accuracy: 0.9513 - val_loss: 0.4277 - val_accuracy: 0.8823

Epoch 00080: saving model to models/CNN_CIFR_APR_21_0080.hdf5

Epoch 81/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1345
- accuracy: 0.9530 - val_loss: 0.4701 - val_accuracy: 0.8802

Epoch 82/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1338
- accuracy: 0.9542 - val_loss: 0.4273 - val_accuracy: 0.8848

Epoch 83/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1376
- accuracy: 0.9516 - val_loss: 0.4584 - val_accuracy: 0.8809

Epoch 84/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1360
- accuracy: 0.9516 - val_loss: 0.3975 - val_accuracy: 0.8937

Epoch 85/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1304
- accuracy: 0.9536 - val_loss: 0.3940 - val_accuracy: 0.8953

Epoch 86/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1279
- accuracy: 0.9555 - val_loss: 0.4143 - val_accuracy: 0.8899

Epoch 87/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1232
- accuracy: 0.9564 - val_loss: 0.5233 - val_accuracy: 0.8780

Epoch 88/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1300
- accuracy: 0.9547 - val_loss: 0.3722 - val_accuracy: 0.8982

Epoch 89/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1255
- accuracy: 0.9556 - val_loss: 0.3879 - val_accuracy: 0.8980

Epoch 90/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1246
- accuracy: 0.9557 - val_loss: 0.4095 - val_accuracy: 0.8980

Epoch 00090: saving model to models/CNN_CIFR_APR_21_0090.hdf5

Epoch 91/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1215
- accuracy: 0.9573 - val_loss: 0.4667 - val_accuracy: 0.8832

Epoch 92/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1265
- accuracy: 0.9553 - val_loss: 0.4134 - val_accuracy: 0.8881

Epoch 93/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1178
- accuracy: 0.9585 - val_loss: 0.4407 - val_accuracy: 0.8924

Epoch 94/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1252
- accuracy: 0.9555 - val_loss: 0.4126 - val_accuracy: 0.8945

Epoch 95/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1240
- accuracy: 0.9578 - val_loss: 0.5021 - val_accuracy: 0.8775

Epoch 96/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1123
- accuracy: 0.9608 - val_loss: 0.4373 - val_accuracy: 0.8861

Epoch 97/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1188
- accuracy: 0.9590 - val_loss: 0.3692 - val_accuracy: 0.9013

Epoch 98/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1124
- accuracy: 0.9609 - val_loss: 0.4346 - val_accuracy: 0.8933

Epoch 99/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1223
- accuracy: 0.9584 - val_loss: 0.3783 - val_accuracy: 0.8982

Epoch 100/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1120
- accuracy: 0.9611 - val_loss: 0.3999 - val_accuracy: 0.8988

Epoch 00100: saving model to models/CNN_CIFR_APR_21_0100.hdf5

Epoch 101/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1081
- accuracy: 0.9615 - val_loss: 0.4202 - val_accuracy: 0.8976

Epoch 102/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1173
- accuracy: 0.9598 - val_loss: 0.4964 - val_accuracy: 0.8828

Epoch 103/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1106
- accuracy: 0.9610 - val_loss: 0.3922 - val_accuracy: 0.9002

Epoch 104/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1069
- accuracy: 0.9627 - val_loss: 0.3959 - val_accuracy: 0.8936

Epoch 105/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1078
- accuracy: 0.9615 - val_loss: 0.5215 - val_accuracy: 0.8794

Epoch 106/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1123
- accuracy: 0.9612 - val_loss: 0.3960 - val_accuracy: 0.9014

Epoch 107/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1150
- accuracy: 0.9598 - val_loss: 0.5091 - val_accuracy: 0.8793

Epoch 108/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1043
- accuracy: 0.9639 - val_loss: 0.4140 - val_accuracy: 0.8966

Epoch 109/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1090
- accuracy: 0.9628 - val_loss: 0.3764 - val_accuracy: 0.8987

Epoch 110/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1075
- accuracy: 0.9625 - val_loss: 0.4183 - val_accuracy: 0.8998

Epoch 00110: saving model to models/CNN_CIFR_APR_21_0110.hdf5

Epoch 111/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1094
- accuracy: 0.9617 - val_loss: 0.4139 - val_accuracy: 0.8919

Epoch 112/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1040
- accuracy: 0.9642 - val_loss: 0.4082 - val_accuracy: 0.8986

Epoch 113/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.1005
- accuracy: 0.9640 - val_loss: 0.4627 - val_accuracy: 0.8895

Epoch 114/150

1563/1563 [=====] - 67s 43ms/step - loss: 0.0984
- accuracy: 0.9657 - val_loss: 0.4565 - val_accuracy: 0.8856

Epoch 115/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.1029
- accuracy: 0.9643 - val_loss: 0.4881 - val_accuracy: 0.8857
Epoch 116/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.1068
- accuracy: 0.9624 - val_loss: 0.4593 - val_accuracy: 0.8909
Epoch 117/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.1031
- accuracy: 0.9639 - val_loss: 0.3854 - val_accuracy: 0.9030
Epoch 118/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0920
- accuracy: 0.9669 - val_loss: 0.4358 - val_accuracy: 0.8948
Epoch 119/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0980
- accuracy: 0.9661 - val_loss: 0.4369 - val_accuracy: 0.8930
Epoch 120/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0976
- accuracy: 0.9659 - val_loss: 0.4985 - val_accuracy: 0.8838

Epoch 00120: saving model to models/CNN_CIFR_APR_21_0120.hdf5
Epoch 121/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0970
- accuracy: 0.9671 - val_loss: 0.4758 - val_accuracy: 0.8839
Epoch 122/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0973
- accuracy: 0.9666 - val_loss: 0.4733 - val_accuracy: 0.8850
Epoch 123/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0940
- accuracy: 0.9672 - val_loss: 0.4737 - val_accuracy: 0.8935
Epoch 124/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0987
- accuracy: 0.9650 - val_loss: 0.4241 - val_accuracy: 0.9002
Epoch 125/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0959
- accuracy: 0.9671 - val_loss: 0.3767 - val_accuracy: 0.9051
Epoch 126/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0937
- accuracy: 0.9664 - val_loss: 0.4093 - val_accuracy: 0.9036
Epoch 127/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0913
- accuracy: 0.9687 - val_loss: 0.4677 - val_accuracy: 0.8884
Epoch 128/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0913
- accuracy: 0.9687 - val_loss: 0.4663 - val_accuracy: 0.8896
Epoch 129/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0961
- accuracy: 0.9660 - val_loss: 0.5348 - val_accuracy: 0.8726
Epoch 130/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0934
- accuracy: 0.9675 - val_loss: 0.4149 - val_accuracy: 0.8986

Epoch 00130: saving model to models/CNN_CIFR_APR_21_0130.hdf5
Epoch 131/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0895
- accuracy: 0.9691 - val_loss: 0.4354 - val_accuracy: 0.8897
Epoch 132/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0887
- accuracy: 0.9684 - val_loss: 0.4052 - val_accuracy: 0.9012
Epoch 133/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0890
- accuracy: 0.9693 - val_loss: 0.4147 - val_accuracy: 0.8941

```
Epoch 134/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0966
- accuracy: 0.9674 - val_loss: 0.3841 - val_accuracy: 0.9030
Epoch 135/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0865
- accuracy: 0.9701 - val_loss: 0.4504 - val_accuracy: 0.8951
Epoch 136/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0887
- accuracy: 0.9691 - val_loss: 0.4288 - val_accuracy: 0.8986
Epoch 137/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0901
- accuracy: 0.9685 - val_loss: 0.4134 - val_accuracy: 0.9014
Epoch 138/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0831
- accuracy: 0.9711 - val_loss: 0.4349 - val_accuracy: 0.8948
Epoch 139/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0841
- accuracy: 0.9712 - val_loss: 0.3907 - val_accuracy: 0.9019
Epoch 140/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0858
- accuracy: 0.9708 - val_loss: 0.4171 - val_accuracy: 0.9001

Epoch 00140: saving model to models/CNN_CIFR_APR_21_0140.hdf5
Epoch 141/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0846
- accuracy: 0.9700 - val_loss: 0.4477 - val_accuracy: 0.8952
Epoch 142/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0903
- accuracy: 0.9689 - val_loss: 0.4104 - val_accuracy: 0.9055
Epoch 143/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0849
- accuracy: 0.9711 - val_loss: 0.4430 - val_accuracy: 0.8979
Epoch 144/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0805
- accuracy: 0.9714 - val_loss: 0.4034 - val_accuracy: 0.9084
Epoch 145/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0850
- accuracy: 0.9699 - val_loss: 0.4622 - val_accuracy: 0.8938
Epoch 146/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0870
- accuracy: 0.9697 - val_loss: 0.3974 - val_accuracy: 0.8989
Epoch 147/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0787
- accuracy: 0.9721 - val_loss: 0.4772 - val_accuracy: 0.8939
Epoch 148/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0839
- accuracy: 0.9712 - val_loss: 0.3918 - val_accuracy: 0.9041
Epoch 149/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0790
- accuracy: 0.9730 - val_loss: 0.4260 - val_accuracy: 0.9013
Epoch 150/150
1563/1563 [=====] - 67s 43ms/step - loss: 0.0799
- accuracy: 0.9726 - val_loss: 0.4882 - val_accuracy: 0.8944

Epoch 00150: saving model to models/CNN_CIFR_APR_21_0150.hdf5
```

In [57]:

```
# Continue Trainin from 150th epoch + train for 50 more epochs

model.load_weights("/content/CNN_CIFR_APR_21_0150.hdf5")
model.compile(loss='categorical_crossentropy',
              optimizer=Adam(learning_rate=0.001),
              metrics=['accuracy'])
cp_callback = tf.keras.callbacks.ModelCheckpoint(
    filepath="models/CNN_CIFR_APR_21_{epoch:04d}.hdf5",
    verbose=1,
    save_weights_only=True,
    period=10)

res_model = model.fit(
    x = aug.flow(X_train, y_train, batch_size=batch_size,),
    epochs=20,
    verbose=1,
    validation_data=(X_test, y_test),
    callbacks = [cp_callback])
```

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

Epoch 1/20

1563/1563 [=====] - 71s 43ms/step - loss: 0.0626
- accuracy: 0.9779 - val_loss: 0.3811 - val_accuracy: 0.9143

Epoch 2/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0453
- accuracy: 0.9852 - val_loss: 0.3894 - val_accuracy: 0.9139

Epoch 3/20

1563/1563 [=====] - 66s 43ms/step - loss: 0.0392
- accuracy: 0.9858 - val_loss: 0.3807 - val_accuracy: 0.9155

Epoch 4/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0362
- accuracy: 0.9869 - val_loss: 0.3820 - val_accuracy: 0.9192

Epoch 5/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0363
- accuracy: 0.9873 - val_loss: 0.3781 - val_accuracy: 0.9206

Epoch 6/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0321
- accuracy: 0.9885 - val_loss: 0.3949 - val_accuracy: 0.9164

Epoch 7/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0314
- accuracy: 0.9889 - val_loss: 0.3933 - val_accuracy: 0.9170

Epoch 8/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0310
- accuracy: 0.9898 - val_loss: 0.4028 - val_accuracy: 0.9185

Epoch 9/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0336
- accuracy: 0.9893 - val_loss: 0.4021 - val_accuracy: 0.9168

Epoch 10/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0290
- accuracy: 0.9898 - val_loss: 0.4029 - val_accuracy: 0.9173

Epoch 00010: saving model to models/CNN_CIFR_APR_21_0010.hdf5

Epoch 11/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0295
- accuracy: 0.9899 - val_loss: 0.3988 - val_accuracy: 0.9176

Epoch 12/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0279
- accuracy: 0.9905 - val_loss: 0.4015 - val_accuracy: 0.9204

Epoch 13/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0270
- accuracy: 0.9898 - val_loss: 0.4116 - val_accuracy: 0.9191

Epoch 14/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0273
- accuracy: 0.9903 - val_loss: 0.4094 - val_accuracy: 0.9182

Epoch 15/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0275
- accuracy: 0.9907 - val_loss: 0.4345 - val_accuracy: 0.9153

Epoch 16/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0266
- accuracy: 0.9915 - val_loss: 0.4220 - val_accuracy: 0.9180

Epoch 17/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0245
- accuracy: 0.9914 - val_loss: 0.4257 - val_accuracy: 0.9197

Epoch 18/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0248
- accuracy: 0.9915 - val_loss: 0.4091 - val_accuracy: 0.9231

Epoch 19/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0255
- accuracy: 0.9911 - val_loss: 0.4164 - val_accuracy: 0.9213

Epoch 20/20

1563/1563 [=====] - 67s 43ms/step - loss: 0.0247

- accuracy: 0.9910 - val_loss: 0.4300 - val_accuracy: 0.9200

Epoch 00020: saving model to models/CNN_CIFR_APR_21_0020.hdf5

In [60]:

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

313/313 [=====] - 3s 11ms/step - loss: 0.4300 - accuracy: 0.9200

Test loss: 0.4300319254398346

Test accuracy: 0.9200000166893005

In [59]:

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

Saved model to disk