

CNN(Dense Net) CIFAR

May 7, 2021

0.0.1 Dense Net on CIFAR”

```
[2]: 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
      import matplotlib.pyplot as plt
      import numpy as np
      from tensorflow.keras.callbacks import ModelCheckpoint, ReduceLROnPlateau
```

```
[3]: import tensorflow as tf
```

```
[4]: from google.colab import drive
      drive.mount('/content/drive')
```

Go to this URL in a browser: https://accounts.google.com/o/oauth2/auth?client_id=947318989803-6bn6qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect_uri=urn%3aietf%3awg%3aoauth%3a2.0%3aob&response_type=code&scope=email%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdocs.test%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly%20https%3a%2f%2fwww.googleapis.com%2fauth%2fpeopleapi.readonly

Enter your authorization code:

.....

Mounted at /content/drive

0.0.2 DATA

```
[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, 10)
      y_test = tf.keras.utils.to_categorical(y_test, 10)
```

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

```
[6]: batch_size = 50
      num_filter = 35
      dropout_rate = 0
      num_classes = 10
      l = 6
      compression = 1.0
```

```
[7]: # Dense Block
def denseblock(input, num_filter = 12 , 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)
        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 = 0.0):
    global compression
    BatchNorm = layers.BatchNormalization()(input)
    relu = layers.Activation('relu')(BatchNorm)
    Conv2D_BottleNeck = layers.Conv2D(int(num_filter*compression), (1,1),
    ↪use_bias=False ,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=(2,2))(relu)
```

```

        output = layers.
        ↪Conv2D(filters=num_classes,kernel_size=(2,2),activation='softmax')(AvgPooling)

        flat = layers.Flatten()(output)
        return flat

```

```

[8]: dropout_rate=0
input = layers.Input(shape=(img_height, img_width, channel,))
First_Conv2D = layers.Conv2D(35, (3,3), use_bias=False ,padding='same')(input)

First_Block = denseblock(First_Conv2D, 35, dropout_rate)
First_Transition = transition(First_Block, 35, dropout_rate)

Second_Block = denseblock(First_Transition, 35, dropout_rate)
Second_Transition = transition(Second_Block, 35, dropout_rate)

Third_Block = denseblock(Second_Transition, 35, dropout_rate)
Third_Transition = transition(Third_Block, 35, dropout_rate)

Last_Block = denseblock(Third_Transition, 35, dropout_rate)
output = output_layer(Last_Block)

```

```

[9]: 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]

```

-----
concatenate (Concatenate)      (None, 32, 32, 70)    0      conv2d[0][0]
                                   conv2d_1[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]
-----
concatenate_1 (Concatenate)    (None, 32, 32, 105)   0
concatenate[0][0]
                                   conv2d_2[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]
-----
concatenate_2 (Concatenate)    (None, 32, 32, 140)   0
concatenate_1[0][0]
                                   conv2d_3[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

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activation_3[0][0]
-----
concatenate_3 (Concatenate)      (None, 32, 32, 175)  0
concatenate_2[0][0]
                                                    conv2d_4[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]
-----
concatenate_4 (Concatenate)      (None, 32, 32, 210)  0
concatenate_3[0][0]
                                                    conv2d_5[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]
-----
concatenate_5 (Concatenate)      (None, 32, 32, 245)  0
concatenate_4[0][0]
                                                    conv2d_6[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]

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-----
conv2d_7 (Conv2D)                (None, 32, 32, 35)    8575
activation_6[0][0]

-----
average_pooling2d (AveragePooli (None, 16, 16, 35)    0          conv2d_7[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]
-----
concatenate_6 (Concatenate)      (None, 16, 16, 70)    0
average_pooling2d[0][0]
                                          conv2d_8[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]
-----
concatenate_7 (Concatenate)      (None, 16, 16, 105)   0
concatenate_6[0][0]
                                          conv2d_9[0][0]
-----
batch_normalization_9 (BatchNor (None, 16, 16, 105)   420
concatenate_7[0][0]
-----
activation_9 (Activation)        (None, 16, 16, 105)   0

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batch_normalization_9[0][0]
-----
conv2d_10 (Conv2D)          (None, 16, 16, 35)    33075
activation_9[0][0]
-----
concatenate_8 (Concatenate) (None, 16, 16, 140)    0
concatenate_7[0][0]
conv2d_10[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]
-----
concatenate_9 (Concatenate) (None, 16, 16, 175)    0
concatenate_8[0][0]
conv2d_11[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]
-----
concatenate_10 (Concatenate) (None, 16, 16, 210)    0
concatenate_9[0][0]
conv2d_12[0][0]
-----
batch_normalization_12 (BatchNo (None, 16, 16, 210) 840
concatenate_10[0][0]

```

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-----
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]

-----

concatenate_11 (Concatenate)    (None, 16, 16, 245)  0
concatenate_10[0][0]

conv2d_13[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]

-----

average_pooling2d_1 (AveragePoo (None, 8, 8, 35)    0
conv2d_14[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]

-----

concatenate_12 (Concatenate)    (None, 8, 8, 70)     0
average_pooling2d_1[0][0]

conv2d_15[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]
-----
-----
concatenate_13 (Concatenate)    (None, 8, 8, 105)     0
concatenate_12[0][0]
conv2d_16[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]
-----
-----
concatenate_14 (Concatenate)    (None, 8, 8, 140)     0
concatenate_13[0][0]
conv2d_17[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]
-----
-----
concatenate_15 (Concatenate)    (None, 8, 8, 175)     0
concatenate_14[0][0]
conv2d_18[0][0]

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-----
-----
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]
-----
-----
concatenate_16 (Concatenate)    (None, 8, 8, 210)    0
concatenate_15[0][0]
conv2d_19[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]
-----
-----
concatenate_17 (Concatenate)    (None, 8, 8, 245)    0
concatenate_16[0][0]
conv2d_20[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]
-----
-----

```

average_pooling2d_2 (AveragePoo	(None, 4, 4, 35)	0	conv2d_21[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]			

concatenate_18 (Concatenate)	(None, 4, 4, 70)	0	
average_pooling2d_2[0][0]			
			conv2d_22[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]			

concatenate_19 (Concatenate)	(None, 4, 4, 105)	0	
concatenate_18[0][0]			
			conv2d_23[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]			

```

-----
concatenate_20 (Concatenate)      (None, 4, 4, 140)      0
concatenate_19[0][0]

conv2d_24[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]

-----
concatenate_21 (Concatenate)      (None, 4, 4, 175)      0
concatenate_20[0][0]

conv2d_25[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]

-----
concatenate_22 (Concatenate)      (None, 4, 4, 210)      0
concatenate_21[0][0]

conv2d_26[0][0]

-----
batch_normalization_26 (BatchNo (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]

-----

concatenate_23 (Concatenate)      (None, 4, 4, 245)    0
concatenate_22[0][0]

conv2d_27[0][0]

-----

batch_normalization_27 (BatchNo (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 (AveragePoo (None, 2, 2, 245)    0
activation_27[0][0]

-----

conv2d_28 (Conv2D)                (None, 1, 1, 10)    9810
average_pooling2d_3[0][0]

-----

flatten (Flatten)                (None, 10)           0
conv2d_28[0][0]
=====
Total params: 978,260
Trainable params: 970,420
Non-trainable params: 7,840
-----

```

```

[10]: def conv_pixel(train, test):

        train_norm = train.astype('float32')
        test_norm = test.astype('float32')
        train_norm = train_norm / 255.0
        test_norm = test_norm / 255.0
        return train_norm, test_norm

X_train,X_test = conv_pixel(X_train,X_test)

```

0.0.3 Data Augmentation

```
[11]: data_gen = ImageDataGenerator(  
        rotation_range=22,  
        width_shift_range=0.125,  
        height_shift_range=0.125,  
        horizontal_flip=True,  
        fill_mode = 'nearest',  
        zoom_range=0.01)  
data_gen.fit(X_train)
```

```
[12]: X_train.shape
```

```
[12]: (50000, 32, 32, 3)
```

```
[13]: for X_batch, y_batch in data_gen.flow(X_train, y_train, batch_size=9):  
        for i in range(0, 9):  
            plt.subplot(330 + 1 + i)  
            plt.imshow(X_batch[i], cmap=plt.get_cmap())  
        plt.show()  
        break
```



```
[14]: ls
```

```
drive/  sample_data/
```

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

[16]: reduce_lr = ReduceLRonPlateau(monitor='val_loss',factor=0.1,patience= 5,
    min_lr=0.000001)
    filepath = "drive/My Drive/best_model.hdf5"
    checkpoint = ModelCheckpoint(filepath, monitor='val_loss', verbose=1,
    ↪save_best_only=True, mode='max')

    callbacks = [checkpoint, reduce_lr]

[17]: history=model.fit_generator(data_gen.flow(X_train, y_train, batch_size=50),
    steps_per_epoch = (len(X_train) /50), epochs=50,
    ↪validation_data=(X_test, y_test),callbacks=callbacks)
```

WARNING:tensorflow:From <ipython-input-17-516e8d78b8f2>:2: Model.fit_generator (from tensorflow.python.keras.engine.training) is deprecated and will be removed in a future version.

Instructions for updating:

Please use Model.fit, which supports generators.

Epoch 1/50

1000/1000 [=====] - ETA: 0s - loss: 1.4667 - accuracy: 0.4679

Epoch 00001: val_loss improved from -inf to 1.62575, saving model to drive/My Drive/best_model.hdf5

1000/1000 [=====] - 65s 65ms/step - loss: 1.4667 - accuracy: 0.4679 - val_loss: 1.6258 - val_accuracy: 0.4306 - lr: 0.0010

Epoch 2/50

1000/1000 [=====] - ETA: 0s - loss: 1.0271 - accuracy: 0.6341

Epoch 00002: val_loss did not improve from 1.62575

1000/1000 [=====] - 64s 64ms/step - loss: 1.0271 - accuracy: 0.6341 - val_loss: 1.1933 - val_accuracy: 0.6039 - lr: 0.0010

Epoch 3/50

1000/1000 [=====] - ETA: 0s - loss: 0.8457 - accuracy: 0.7010

Epoch 00003: val_loss did not improve from 1.62575

1000/1000 [=====] - 64s 64ms/step - loss: 0.8457 - accuracy: 0.7010 - val_loss: 0.8319 - val_accuracy: 0.7154 - lr: 0.0010

Epoch 4/50

1000/1000 [=====] - ETA: 0s - loss: 0.7349 - accuracy: 0.7442

Epoch 00004: val_loss did not improve from 1.62575

1000/1000 [=====] - 64s 64ms/step - loss: 0.7349 - accuracy: 0.7442 - val_loss: 0.6837 - val_accuracy: 0.7613 - lr: 0.0010

Epoch 5/50

1000/1000 [=====] - ETA: 0s - loss: 0.6655 - accuracy:

0.7676
Epoch 00005: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.6655 -
accuracy: 0.7676 - val_loss: 0.7177 - val_accuracy: 0.7538 - lr: 0.0010
Epoch 6/50
1000/1000 [=====] - ETA: 0s - loss: 0.6068 - accuracy:
0.7883
Epoch 00006: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.6068 -
accuracy: 0.7883 - val_loss: 0.9050 - val_accuracy: 0.7316 - lr: 0.0010
Epoch 7/50
1000/1000 [=====] - ETA: 0s - loss: 0.5591 - accuracy:
0.8050
Epoch 00007: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.5591 -
accuracy: 0.8050 - val_loss: 0.6049 - val_accuracy: 0.7985 - lr: 0.0010
Epoch 8/50
1000/1000 [=====] - ETA: 0s - loss: 0.5222 - accuracy:
0.8198
Epoch 00008: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.5222 -
accuracy: 0.8198 - val_loss: 0.6428 - val_accuracy: 0.7976 - lr: 0.0010
Epoch 9/50
1000/1000 [=====] - ETA: 0s - loss: 0.4918 - accuracy:
0.8292
Epoch 00009: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.4918 -
accuracy: 0.8292 - val_loss: 0.5549 - val_accuracy: 0.8087 - lr: 0.0010
Epoch 10/50
1000/1000 [=====] - ETA: 0s - loss: 0.4681 - accuracy:
0.8367
Epoch 00010: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.4681 -
accuracy: 0.8367 - val_loss: 0.5687 - val_accuracy: 0.8111 - lr: 0.0010
Epoch 11/50
1000/1000 [=====] - ETA: 0s - loss: 0.4399 - accuracy:
0.8478
Epoch 00011: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.4399 -
accuracy: 0.8478 - val_loss: 0.5621 - val_accuracy: 0.8129 - lr: 0.0010
Epoch 12/50
1000/1000 [=====] - ETA: 0s - loss: 0.4226 - accuracy:
0.8532
Epoch 00012: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.4226 -
accuracy: 0.8532 - val_loss: 0.5773 - val_accuracy: 0.8046 - lr: 0.0010
Epoch 13/50
1000/1000 [=====] - ETA: 0s - loss: 0.4011 - accuracy:

0.8601
Epoch 00013: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.4011 -
accuracy: 0.8601 - val_loss: 0.4529 - val_accuracy: 0.8510 - lr: 0.0010
Epoch 14/50
1000/1000 [=====] - ETA: 0s - loss: 0.3872 - accuracy:
0.8665
Epoch 00014: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.3872 -
accuracy: 0.8665 - val_loss: 0.4723 - val_accuracy: 0.8447 - lr: 0.0010
Epoch 15/50
1000/1000 [=====] - ETA: 0s - loss: 0.3690 - accuracy:
0.8707
Epoch 00015: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.3690 -
accuracy: 0.8707 - val_loss: 0.4254 - val_accuracy: 0.8568 - lr: 0.0010
Epoch 16/50
1000/1000 [=====] - ETA: 0s - loss: 0.3604 - accuracy:
0.8742
Epoch 00016: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.3604 -
accuracy: 0.8742 - val_loss: 0.4475 - val_accuracy: 0.8504 - lr: 0.0010
Epoch 17/50
1000/1000 [=====] - ETA: 0s - loss: 0.3441 - accuracy:
0.8818
Epoch 00017: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.3441 -
accuracy: 0.8818 - val_loss: 0.6499 - val_accuracy: 0.8091 - lr: 0.0010
Epoch 18/50
1000/1000 [=====] - ETA: 0s - loss: 0.3340 - accuracy:
0.8835
Epoch 00018: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.3340 -
accuracy: 0.8835 - val_loss: 0.4438 - val_accuracy: 0.8564 - lr: 0.0010
Epoch 19/50
1000/1000 [=====] - ETA: 0s - loss: 0.3197 - accuracy:
0.8888
Epoch 00019: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.3197 -
accuracy: 0.8888 - val_loss: 0.5516 - val_accuracy: 0.8288 - lr: 0.0010
Epoch 20/50
1000/1000 [=====] - ETA: 0s - loss: 0.3101 - accuracy:
0.8905
Epoch 00020: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.3101 -
accuracy: 0.8905 - val_loss: 0.5998 - val_accuracy: 0.8202 - lr: 0.0010
Epoch 21/50
1000/1000 [=====] - ETA: 0s - loss: 0.2401 - accuracy:

0.9154
Epoch 00021: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.2401 -
accuracy: 0.9154 - val_loss: 0.3210 - val_accuracy: 0.8929 - lr: 1.0000e-04
Epoch 22/50
1000/1000 [=====] - ETA: 0s - loss: 0.2170 - accuracy:
0.9245
Epoch 00022: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.2170 -
accuracy: 0.9245 - val_loss: 0.3114 - val_accuracy: 0.8985 - lr: 1.0000e-04
Epoch 23/50
1000/1000 [=====] - ETA: 0s - loss: 0.2091 - accuracy:
0.9266
Epoch 00023: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.2091 -
accuracy: 0.9266 - val_loss: 0.3216 - val_accuracy: 0.8963 - lr: 1.0000e-04
Epoch 24/50
1000/1000 [=====] - ETA: 0s - loss: 0.2013 - accuracy:
0.9298
Epoch 00024: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.2013 -
accuracy: 0.9298 - val_loss: 0.3001 - val_accuracy: 0.9019 - lr: 1.0000e-04
Epoch 25/50
1000/1000 [=====] - ETA: 0s - loss: 0.1980 - accuracy:
0.9306
Epoch 00025: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1980 -
accuracy: 0.9306 - val_loss: 0.3060 - val_accuracy: 0.9021 - lr: 1.0000e-04
Epoch 26/50
1000/1000 [=====] - ETA: 0s - loss: 0.1916 - accuracy:
0.9349
Epoch 00026: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1916 -
accuracy: 0.9349 - val_loss: 0.3103 - val_accuracy: 0.9017 - lr: 1.0000e-04
Epoch 27/50
1000/1000 [=====] - ETA: 0s - loss: 0.1868 - accuracy:
0.9353
Epoch 00027: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1868 -
accuracy: 0.9353 - val_loss: 0.3157 - val_accuracy: 0.8984 - lr: 1.0000e-04
Epoch 28/50
1000/1000 [=====] - ETA: 0s - loss: 0.1846 - accuracy:
0.9364
Epoch 00028: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1846 -
accuracy: 0.9364 - val_loss: 0.3158 - val_accuracy: 0.9008 - lr: 1.0000e-04
Epoch 29/50
1000/1000 [=====] - ETA: 0s - loss: 0.1826 - accuracy:

0.9354
Epoch 00029: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1826 -
accuracy: 0.9354 - val_loss: 0.3154 - val_accuracy: 0.9015 - lr: 1.0000e-04
Epoch 30/50
1000/1000 [=====] - ETA: 0s - loss: 0.1751 - accuracy:
0.9381
Epoch 00030: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1751 -
accuracy: 0.9381 - val_loss: 0.3073 - val_accuracy: 0.9015 - lr: 1.0000e-05
Epoch 31/50
1000/1000 [=====] - ETA: 0s - loss: 0.1717 - accuracy:
0.9395
Epoch 00031: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1717 -
accuracy: 0.9395 - val_loss: 0.3041 - val_accuracy: 0.9034 - lr: 1.0000e-05
Epoch 32/50
1000/1000 [=====] - ETA: 0s - loss: 0.1742 - accuracy:
0.9380
Epoch 00032: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1742 -
accuracy: 0.9380 - val_loss: 0.3077 - val_accuracy: 0.9036 - lr: 1.0000e-05
Epoch 33/50
1000/1000 [=====] - ETA: 0s - loss: 0.1717 - accuracy:
0.9390
Epoch 00033: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1717 -
accuracy: 0.9390 - val_loss: 0.3051 - val_accuracy: 0.9034 - lr: 1.0000e-05
Epoch 34/50
1000/1000 [=====] - ETA: 0s - loss: 0.1737 - accuracy:
0.9398
Epoch 00034: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1737 -
accuracy: 0.9398 - val_loss: 0.3061 - val_accuracy: 0.9031 - lr: 1.0000e-05
Epoch 35/50
1000/1000 [=====] - ETA: 0s - loss: 0.1722 - accuracy:
0.9401
Epoch 00035: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1722 -
accuracy: 0.9401 - val_loss: 0.3059 - val_accuracy: 0.9028 - lr: 1.0000e-06
Epoch 36/50
1000/1000 [=====] - ETA: 0s - loss: 0.1708 - accuracy:
0.9398
Epoch 00036: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1708 -
accuracy: 0.9398 - val_loss: 0.3060 - val_accuracy: 0.9034 - lr: 1.0000e-06
Epoch 37/50
1000/1000 [=====] - ETA: 0s - loss: 0.1703 - accuracy:

0.9406
Epoch 00037: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1703 -
accuracy: 0.9406 - val_loss: 0.3076 - val_accuracy: 0.9030 - lr: 1.0000e-06
Epoch 38/50
1000/1000 [=====] - ETA: 0s - loss: 0.1685 - accuracy:
0.9421
Epoch 00038: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1685 -
accuracy: 0.9421 - val_loss: 0.3063 - val_accuracy: 0.9038 - lr: 1.0000e-06
Epoch 39/50
1000/1000 [=====] - ETA: 0s - loss: 0.1712 - accuracy:
0.9401
Epoch 00039: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1712 -
accuracy: 0.9401 - val_loss: 0.3062 - val_accuracy: 0.9040 - lr: 1.0000e-06
Epoch 40/50
1000/1000 [=====] - ETA: 0s - loss: 0.1699 - accuracy:
0.9394
Epoch 00040: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1699 -
accuracy: 0.9394 - val_loss: 0.3037 - val_accuracy: 0.9044 - lr: 1.0000e-06
Epoch 41/50
1000/1000 [=====] - ETA: 0s - loss: 0.1711 - accuracy:
0.9397
Epoch 00041: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1711 -
accuracy: 0.9397 - val_loss: 0.3053 - val_accuracy: 0.9035 - lr: 1.0000e-06
Epoch 42/50
1000/1000 [=====] - ETA: 0s - loss: 0.1690 - accuracy:
0.9422
Epoch 00042: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1690 -
accuracy: 0.9422 - val_loss: 0.3044 - val_accuracy: 0.9046 - lr: 1.0000e-06
Epoch 43/50
1000/1000 [=====] - ETA: 0s - loss: 0.1702 - accuracy:
0.9402
Epoch 00043: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1702 -
accuracy: 0.9402 - val_loss: 0.3027 - val_accuracy: 0.9044 - lr: 1.0000e-06
Epoch 44/50
1000/1000 [=====] - ETA: 0s - loss: 0.1691 - accuracy:
0.9412
Epoch 00044: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1691 -
accuracy: 0.9412 - val_loss: 0.3049 - val_accuracy: 0.9043 - lr: 1.0000e-06
Epoch 45/50
1000/1000 [=====] - ETA: 0s - loss: 0.1702 - accuracy:

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0.9414
Epoch 00045: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1702 -
accuracy: 0.9414 - val_loss: 0.3049 - val_accuracy: 0.9047 - lr: 1.0000e-06
Epoch 46/50
1000/1000 [=====] - ETA: 0s - loss: 0.1654 - accuracy:
0.9422
Epoch 00046: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1654 -
accuracy: 0.9422 - val_loss: 0.3059 - val_accuracy: 0.9037 - lr: 1.0000e-06
Epoch 47/50
1000/1000 [=====] - ETA: 0s - loss: 0.1699 - accuracy:
0.9403
Epoch 00047: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1699 -
accuracy: 0.9403 - val_loss: 0.3062 - val_accuracy: 0.9037 - lr: 1.0000e-06
Epoch 48/50
1000/1000 [=====] - ETA: 0s - loss: 0.1699 - accuracy:
0.9410
Epoch 00048: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1699 -
accuracy: 0.9410 - val_loss: 0.3034 - val_accuracy: 0.9041 - lr: 1.0000e-06
Epoch 49/50
1000/1000 [=====] - ETA: 0s - loss: 0.1716 - accuracy:
0.9400
Epoch 00049: val_loss did not improve from 1.62575
1000/1000 [=====] - 63s 63ms/step - loss: 0.1716 -
accuracy: 0.9400 - val_loss: 0.3054 - val_accuracy: 0.9037 - lr: 1.0000e-06
Epoch 50/50
1000/1000 [=====] - ETA: 0s - loss: 0.1682 - accuracy:
0.9421
Epoch 00050: val_loss did not improve from 1.62575
1000/1000 [=====] - 64s 64ms/step - loss: 0.1682 -
accuracy: 0.9421 - val_loss: 0.3058 - val_accuracy: 0.9032 - lr: 1.0000e-06

```

0.1 Observation

- 1) Here i tried with variation of different values of number filters and conclusion was that the number of filters should be small.
- 2)I got accuracy as 94.21% and validation accuracy as 90.32 % after 50 epochs
- 3) Reduce lr on plateau helped to reach the convergence in less number of epochs
- 4) Data augumention helped to improve the results

[]: