

# Task 16

Registration ID : SIRSS1038

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## Importing Libraries

In [1]:

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import tensorflow as tf
from datetime import datetime
from keras.preprocessing import image
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.layers import Dense, Flatten, Conv2D, UpSampling2D, Dropout, BatchNormalization, GlobalAveragePooling2D
```

## Importing the Cifar 100 Dataset

In [2]:

```
cifar100 = tf.keras.datasets.cifar100
(X_train, Y_train), (X_test, Y_test) = cifar100.load_data()
```

Downloading data from <https://www.cs.toronto.edu/~kriz/cifar-100-python.tar.gz>  
169009152/169001437 [=====] - 6s 0us/step

In [3]:

```
def timer(start_time=None):
    #function to track time
    if not start_time:
        print(datetime.now())
        start_time = datetime.now()
        return start_time
    elif start_time:
        thour, temp_sec = divmod((datetime.now() - start_time).total_seconds(), 3600)
        tmin, tsec = divmod(temp_sec, 60)
        print('Time taken: %i hours %i minutes and %s seconds.' % (thour, tmin, round(tsec, 2)))

def plot_acc_loss(result):
    # function to plot the accuracy and loss graphs
    acc = result.history['accuracy']
    val_acc = result.history['val_accuracy']
    loss = result.history['loss']
    val_loss = result.history['val_loss']

    plt.figure(figsize=(20, 10))
    plt.subplot(1, 2, 1)
    plt.title("Training and Validation Accuracy")
    plt.plot(acc, color = 'green', label = 'Training Accuracy')
    plt.plot(val_acc, color = 'red', label = 'Validation Accuracy')
    plt.legend(loc='lower right')
    plt.ylabel('accuracy')
    plt.xlabel('epoch')
    plt.subplot(1, 2, 2)
    plt.title("Training and Validation Loss")
    plt.plot(loss, color = 'blue', label = 'Training Loss')
    plt.plot(val_loss, color = 'purple', label = 'Validation Loss')
```



In [6]:

```
from keras.utils.np_utils import to_categorical
y_train=to_categorical(y_train,num_classes=100)
y_val=to_categorical(y_val,num_classes=100)
y_test=to_categorical(Y_test,num_classes=100)
```

In [7]:

```
x_train = x_train*1.0/255
x_val = x_val*1.0/255
X_test = X_test*1.0/255
```

In [8]:

```
print(x_train.shape,x_val.shape,X_test.shape)
print(y_train.shape,y_val.shape,y_test.shape)

(40000, 32, 32, 3) (10000, 32, 32, 3) (10000, 32, 32, 3)
(40000, 100) (10000, 100) (10000, 100)
```

## Image Data Augmentation

In [9]:

```
train_datagen = ImageDataGenerator(
    rotation_range = 10,
    zoom_range = 0.1,
    width_shift_range = 0.1,
    height_shift_range = 0.1,
    shear_range = 0.1,
    horizontal_flip = True,
    vertical_flip = False
)
train_datagen.fit(x_train)
```

## Reduce Learning Rate if accuracy is not improving for 3 epochs

In [10]:

```
from keras.callbacks import ReduceLROnPlateau
learning_rate_reduction = ReduceLROnPlateau(
    monitor='val_accuracy',
    patience=3,
    verbose=1,
    factor=0.6,
    min_lr=1e-6)
```

## Importing the Resnet Model

In [11]:

```
from tensorflow.keras.applications.resnet50 import ResNet50
resnet_model = ResNet50(
    include_top=False,
    weights='imagenet',
    input_shape=(224,224,3)
)

for layer in resnet_model.layers:
    if isinstance(layer, BatchNormalization):
        layer.trainable = True
    else:
        layer.trainable = False
```

Downloading data from [https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet50\\_weights\\_tf\\_dim\\_ordering\\_tf\\_kernels\\_notop.h5](https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet50_weights_tf_dim_ordering_tf_kernels_notop.h5)

In [12]:

resnet\_model.summary()

Model: "resnet50"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	[ (None, 224, 224, 3) 0		
-----			
conv1_pad (ZeroPadding2D)	(None, 230, 230, 3) 0		input_1[0][0]
-----			
conv1_conv (Conv2D)	(None, 112, 112, 64) 9472		conv1_pad[0][0]
-----			
conv1_bn (BatchNormalization)	(None, 112, 112, 64) 256		conv1_conv[0][0]
-----			
conv1_relu (Activation)	(None, 112, 112, 64) 0		conv1_bn[0][0]
-----			
pool1_pad (ZeroPadding2D)	(None, 114, 114, 64) 0		conv1_relu[0][0]
-----			
pool1_pool (MaxPooling2D)	(None, 56, 56, 64) 0		pool1_pad[0][0]
-----			
conv2_block1_1_conv (Conv2D)	(None, 56, 56, 64) 4160		pool1_pool[0][0]
-----			
conv2_block1_1_bn (BatchNormali ]	(None, 56, 56, 64) 256		conv2_block1_1_conv[0][0]
-----			
conv2_block1_1_relu (Activation	(None, 56, 56, 64) 0		conv2_block1_1_bn[0][0]
-----			
conv2_block1_2_conv (Conv2D) ]	(None, 56, 56, 64) 36928		conv2_block1_1_relu[0][0]
-----			
conv2_block1_2_bn (BatchNormali ]	(None, 56, 56, 64) 256		conv2_block1_2_conv[0][0]
-----			
conv2_block1_2_relu (Activation	(None, 56, 56, 64) 0		conv2_block1_2_bn[0][0]
-----			
conv2_block1_0_conv (Conv2D)	(None, 56, 56, 256) 16640		pool1_pool[0][0]
-----			
conv2_block1_3_conv (Conv2D) ]	(None, 56, 56, 256) 16640		conv2_block1_2_relu[0][0]

conv2_block1_0_bn	(BatchNormali	(None, 56, 56, 256)	1024	conv2_block1_0_conv[0][0]
conv2_block1_3_bn	(BatchNormali	(None, 56, 56, 256)	1024	conv2_block1_3_conv[0][0]
conv2_block1_add	(Add)	(None, 56, 56, 256)	0	conv2_block1_0_bn[0][0] conv2_block1_3_bn[0][0]
conv2_block1_out	(Activation)	(None, 56, 56, 256)	0	conv2_block1_add[0][0]
conv2_block2_1_conv	(Conv2D)	(None, 56, 56, 64)	16448	conv2_block1_out[0][0]
conv2_block2_1_bn	(BatchNormali	(None, 56, 56, 64)	256	conv2_block2_1_conv[0][0]
conv2_block2_1_relu	(Activation	(None, 56, 56, 64)	0	conv2_block2_1_bn[0][0]
conv2_block2_2_conv	(Conv2D)	(None, 56, 56, 64)	36928	conv2_block2_1_relu[0][0]
conv2_block2_2_bn	(BatchNormali	(None, 56, 56, 64)	256	conv2_block2_2_conv[0][0]
conv2_block2_2_relu	(Activation	(None, 56, 56, 64)	0	conv2_block2_2_bn[0][0]
conv2_block2_3_conv	(Conv2D)	(None, 56, 56, 256)	16640	conv2_block2_2_relu[0][0]
conv2_block2_3_bn	(BatchNormali	(None, 56, 56, 256)	1024	conv2_block2_3_conv[0][0]
conv2_block2_add	(Add)	(None, 56, 56, 256)	0	conv2_block1_out[0][0] conv2_block2_3_bn[0][0]
conv2_block2_out	(Activation)	(None, 56, 56, 256)	0	conv2_block2_add[0][0]
conv2_block3_1_conv	(Conv2D)	(None, 56, 56, 64)	16448	conv2_block2_out[0][0]
conv2_block3_1_bn	(BatchNormali	(None, 56, 56, 64)	256	conv2_block3_1_conv[0][0]
conv2_block3_1_relu	(Activation	(None, 56, 56, 64)	0	conv2_block3_1_bn[0][0]

conv2_block3_2_conv (Conv2D)	(None, 56, 56, 64)	36928	conv2_block3_1_relu[0][0]
conv2_block3_2_bn (BatchNormali	(None, 56, 56, 64)	256	conv2_block3_2_conv[0][0]
conv2_block3_2_relu (Activation	(None, 56, 56, 64)	0	conv2_block3_2_bn[0][0]
conv2_block3_3_conv (Conv2D)	(None, 56, 56, 256)	16640	conv2_block3_2_relu[0][0]
conv2_block3_3_bn (BatchNormali	(None, 56, 56, 256)	1024	conv2_block3_3_conv[0][0]
conv2_block3_add (Add)	(None, 56, 56, 256)	0	conv2_block2_out[0][0] conv2_block3_3_bn[0][0]
conv2_block3_out (Activation)	(None, 56, 56, 256)	0	conv2_block3_add[0][0]
conv3_block1_1_conv (Conv2D)	(None, 28, 28, 128)	32896	conv2_block3_out[0][0]
conv3_block1_1_bn (BatchNormali	(None, 28, 28, 128)	512	conv3_block1_1_conv[0][0]
conv3_block1_1_relu (Activation	(None, 28, 28, 128)	0	conv3_block1_1_bn[0][0]
conv3_block1_2_conv (Conv2D)	(None, 28, 28, 128)	147584	conv3_block1_1_relu[0][0]
conv3_block1_2_bn (BatchNormali	(None, 28, 28, 128)	512	conv3_block1_2_conv[0][0]
conv3_block1_2_relu (Activation	(None, 28, 28, 128)	0	conv3_block1_2_bn[0][0]
conv3_block1_0_conv (Conv2D)	(None, 28, 28, 512)	131584	conv2_block3_out[0][0]
conv3_block1_3_conv (Conv2D)	(None, 28, 28, 512)	66048	conv3_block1_2_relu[0][0]
conv3_block1_0_bn (BatchNormali	(None, 28, 28, 512)	2048	conv3_block1_0_conv[0][0]
conv3_block1_3_bn (BatchNormali	(None, 28, 28, 512)	2048	conv3_block1_3_conv[0][0]
conv3_block1_add (Add)	(None, 28, 28, 512)	0	conv3_block1_0_bn[0][0]

				conv3_block1_3_bn[0][0]
]				
<hr/>				
conv3_block1_out	(Activation)	(None, 28, 28, 512)	0	conv3_block1_add[0][0]
<hr/>				
conv3_block2_1_conv	(Conv2D)	(None, 28, 28, 128)	65664	conv3_block1_out[0][0]
<hr/>				
conv3_block2_1_bn	(BatchNormali	(None, 28, 28, 128)	512	conv3_block2_1_conv[0][0]
]				
<hr/>				
conv3_block2_1_relu	(Activation	(None, 28, 28, 128)	0	conv3_block2_1_bn[0][0]
<hr/>				
conv3_block2_2_conv	(Conv2D)	(None, 28, 28, 128)	147584	conv3_block2_1_relu[0][0]
]				
<hr/>				
conv3_block2_2_bn	(BatchNormali	(None, 28, 28, 128)	512	conv3_block2_2_conv[0][0]
]				
<hr/>				
conv3_block2_2_relu	(Activation	(None, 28, 28, 128)	0	conv3_block2_2_bn[0][0]
<hr/>				
conv3_block2_3_conv	(Conv2D)	(None, 28, 28, 512)	66048	conv3_block2_2_relu[0][0]
]				
<hr/>				
conv3_block2_3_bn	(BatchNormali	(None, 28, 28, 512)	2048	conv3_block2_3_conv[0][0]
]				
<hr/>				
conv3_block2_add	(Add)	(None, 28, 28, 512)	0	conv3_block1_out[0][0]
				conv3_block2_3_bn[0][0]
]				
<hr/>				
conv3_block2_out	(Activation)	(None, 28, 28, 512)	0	conv3_block2_add[0][0]
<hr/>				
conv3_block3_1_conv	(Conv2D)	(None, 28, 28, 128)	65664	conv3_block2_out[0][0]
<hr/>				
conv3_block3_1_bn	(BatchNormali	(None, 28, 28, 128)	512	conv3_block3_1_conv[0][0]
]				
<hr/>				
conv3_block3_1_relu	(Activation	(None, 28, 28, 128)	0	conv3_block3_1_bn[0][0]
<hr/>				
conv3_block3_2_conv	(Conv2D)	(None, 28, 28, 128)	147584	conv3_block3_1_relu[0][0]
]				
<hr/>				
conv3_block3_2_bn	(BatchNormali	(None, 28, 28, 128)	512	conv3_block3_2_conv[0][0]
]				
<hr/>				
conv3_block3_2_relu	(Activation	(None, 28, 28, 128)	0	conv3_block3_2_bn[0][0]

conv3_block3_3_conv (Conv2D)	(None, 28, 28, 512)	66048	conv3_block3_2_relu[0][0]
conv3_block3_3_bn (BatchNormali	(None, 28, 28, 512)	2048	conv3_block3_3_conv[0][0]
conv3_block3_add (Add)	(None, 28, 28, 512)	0	conv3_block2_out[0][0] conv3_block3_3_bn[0][0]
conv3_block3_out (Activation)	(None, 28, 28, 512)	0	conv3_block3_add[0][0]
conv3_block4_1_conv (Conv2D)	(None, 28, 28, 128)	65664	conv3_block3_out[0][0]
conv3_block4_1_bn (BatchNormali	(None, 28, 28, 128)	512	conv3_block4_1_conv[0][0]
conv3_block4_1_relu (Activation	(None, 28, 28, 128)	0	conv3_block4_1_bn[0][0]
conv3_block4_2_conv (Conv2D)	(None, 28, 28, 128)	147584	conv3_block4_1_relu[0][0]
conv3_block4_2_bn (BatchNormali	(None, 28, 28, 128)	512	conv3_block4_2_conv[0][0]
conv3_block4_2_relu (Activation	(None, 28, 28, 128)	0	conv3_block4_2_bn[0][0]
conv3_block4_3_conv (Conv2D)	(None, 28, 28, 512)	66048	conv3_block4_2_relu[0][0]
conv3_block4_3_bn (BatchNormali	(None, 28, 28, 512)	2048	conv3_block4_3_conv[0][0]
conv3_block4_add (Add)	(None, 28, 28, 512)	0	conv3_block3_out[0][0] conv3_block4_3_bn[0][0]
conv3_block4_out (Activation)	(None, 28, 28, 512)	0	conv3_block4_add[0][0]
conv4_block1_1_conv (Conv2D)	(None, 14, 14, 256)	131328	conv3_block4_out[0][0]
conv4_block1_1_bn (BatchNormali	(None, 14, 14, 256)	1024	conv4_block1_1_conv[0][0]
conv4_block1_1_relu (Activation	(None, 14, 14, 256)	0	conv4_block1_1_bn[0][0]



conv4_block1_2_conv (Conv2D)	(None, 14, 14, 256)	590080	conv4_block1_1_relu[0][0]
conv4_block1_2_bn (BatchNormali	(None, 14, 14, 256)	1024	conv4_block1_2_conv[0][0]
conv4_block1_2_relu (Activation	(None, 14, 14, 256)	0	conv4_block1_2_bn[0][0]
conv4_block1_0_conv (Conv2D)	(None, 14, 14, 1024)	525312	conv3_block4_out[0][0]
conv4_block1_3_conv (Conv2D)	(None, 14, 14, 1024)	263168	conv4_block1_2_relu[0][0]
conv4_block1_0_bn (BatchNormali	(None, 14, 14, 1024)	4096	conv4_block1_0_conv[0][0]
conv4_block1_3_bn (BatchNormali	(None, 14, 14, 1024)	4096	conv4_block1_3_conv[0][0]
conv4_block1_add (Add)	(None, 14, 14, 1024)	0	conv4_block1_0_bn[0][0]
			conv4_block1_3_bn[0][0]
conv4_block1_out (Activation)	(None, 14, 14, 1024)	0	conv4_block1_add[0][0]
conv4_block2_1_conv (Conv2D)	(None, 14, 14, 256)	262400	conv4_block1_out[0][0]
conv4_block2_1_bn (BatchNormali	(None, 14, 14, 256)	1024	conv4_block2_1_conv[0][0]
conv4_block2_1_relu (Activation	(None, 14, 14, 256)	0	conv4_block2_1_bn[0][0]
conv4_block2_2_conv (Conv2D)	(None, 14, 14, 256)	590080	conv4_block2_1_relu[0][0]
conv4_block2_2_bn (BatchNormali	(None, 14, 14, 256)	1024	conv4_block2_2_conv[0][0]
conv4_block2_2_relu (Activation	(None, 14, 14, 256)	0	conv4_block2_2_bn[0][0]
conv4_block2_3_conv (Conv2D)	(None, 14, 14, 1024)	263168	conv4_block2_2_relu[0][0]
conv4_block2_3_bn (BatchNormali	(None, 14, 14, 1024)	4096	conv4_block2_3_conv[0][0]

conv4_block2_add (Add)	(None, 14, 14, 1024)	0	conv4_block1_out[0][0]
			conv4_block2_3_bn[0][0]
			]
conv4_block2_out (Activation)	(None, 14, 14, 1024)	0	conv4_block2_add[0][0]
conv4_block3_1_conv (Conv2D)	(None, 14, 14, 256)	262400	conv4_block2_out[0][0]
conv4_block3_1_bn (BatchNormali	(None, 14, 14, 256)	1024	conv4_block3_1_conv[0][0]
			]
conv4_block3_1_relu (Activation	(None, 14, 14, 256)	0	conv4_block3_1_bn[0][0]
conv4_block3_2_conv (Conv2D)	(None, 14, 14, 256)	590080	conv4_block3_1_relu[0][0]
			]
conv4_block3_2_bn (BatchNormali	(None, 14, 14, 256)	1024	conv4_block3_2_conv[0][0]
			]
conv4_block3_2_relu (Activation	(None, 14, 14, 256)	0	conv4_block3_2_bn[0][0]
conv4_block3_3_conv (Conv2D)	(None, 14, 14, 1024)	263168	conv4_block3_2_relu[0][0]
			]
conv4_block3_3_bn (BatchNormali	(None, 14, 14, 1024)	4096	conv4_block3_3_conv[0][0]
			]
conv4_block3_add (Add)	(None, 14, 14, 1024)	0	conv4_block2_out[0][0]
			conv4_block3_3_bn[0][0]
			]
conv4_block3_out (Activation)	(None, 14, 14, 1024)	0	conv4_block3_add[0][0]
conv4_block4_1_conv (Conv2D)	(None, 14, 14, 256)	262400	conv4_block3_out[0][0]
conv4_block4_1_bn (BatchNormali	(None, 14, 14, 256)	1024	conv4_block4_1_conv[0][0]
			]
conv4_block4_1_relu (Activation	(None, 14, 14, 256)	0	conv4_block4_1_bn[0][0]
conv4_block4_2_conv (Conv2D)	(None, 14, 14, 256)	590080	conv4_block4_1_relu[0][0]
			]
conv4_block4_2_bn (BatchNormali	(None, 14, 14, 256)	1024	conv4_block4_2_conv[0][0]
			]

conv4_block4_2_relu	(Activation (None, 14, 14, 256)	0	conv4_block4_2_bn[0][0]
<hr/>			
conv4_block4_3_conv	(Conv2D (None, 14, 14, 1024)	263168	conv4_block4_2_relu[0][0]
<hr/>			
conv4_block4_3_bn	(BatchNormali (None, 14, 14, 1024)	4096	conv4_block4_3_conv[0][0]
<hr/>			
conv4_block4_add	(Add) (None, 14, 14, 1024)	0	conv4_block3_out[0][0]
			conv4_block4_3_bn[0][0]
<hr/>			
conv4_block4_out	(Activation) (None, 14, 14, 1024)	0	conv4_block4_add[0][0]
<hr/>			
conv4_block5_1_conv	(Conv2D (None, 14, 14, 256)	262400	conv4_block4_out[0][0]
<hr/>			
conv4_block5_1_bn	(BatchNormali (None, 14, 14, 256)	1024	conv4_block5_1_conv[0][0]
<hr/>			
conv4_block5_1_relu	(Activation (None, 14, 14, 256)	0	conv4_block5_1_bn[0][0]
<hr/>			
conv4_block5_2_conv	(Conv2D (None, 14, 14, 256)	590080	conv4_block5_1_relu[0][0]
<hr/>			
conv4_block5_2_bn	(BatchNormali (None, 14, 14, 256)	1024	conv4_block5_2_conv[0][0]
<hr/>			
conv4_block5_2_relu	(Activation (None, 14, 14, 256)	0	conv4_block5_2_bn[0][0]
<hr/>			
conv4_block5_3_conv	(Conv2D (None, 14, 14, 1024)	263168	conv4_block5_2_relu[0][0]
<hr/>			
conv4_block5_3_bn	(BatchNormali (None, 14, 14, 1024)	4096	conv4_block5_3_conv[0][0]
<hr/>			
conv4_block5_add	(Add) (None, 14, 14, 1024)	0	conv4_block4_out[0][0]
			conv4_block5_3_bn[0][0]
<hr/>			
conv4_block5_out	(Activation) (None, 14, 14, 1024)	0	conv4_block5_add[0][0]
<hr/>			
conv4_block6_1_conv	(Conv2D (None, 14, 14, 256)	262400	conv4_block5_out[0][0]
<hr/>			
conv4_block6_1_bn	(BatchNormali (None, 14, 14, 256)	1024	conv4_block6_1_conv[0][0]

conv4_block6_1_relu	(Activation (None, 14, 14, 256))	0	conv4_block6_1_bn[0][0]
conv4_block6_2_conv	(Conv2D (None, 14, 14, 256))	590080	conv4_block6_1_relu[0][0]
conv4_block6_2_bn	(BatchNormali (None, 14, 14, 256))	1024	conv4_block6_2_conv[0][0]
conv4_block6_2_relu	(Activation (None, 14, 14, 256))	0	conv4_block6_2_bn[0][0]
conv4_block6_3_conv	(Conv2D (None, 14, 14, 1024))	263168	conv4_block6_2_relu[0][0]
conv4_block6_3_bn	(BatchNormali (None, 14, 14, 1024))	4096	conv4_block6_3_conv[0][0]
conv4_block6_add	(Add (None, 14, 14, 1024))	0	conv4_block5_out[0][0]
			conv4_block6_3_bn[0][0]
conv4_block6_out	(Activation (None, 14, 14, 1024))	0	conv4_block6_add[0][0]
conv5_block1_1_conv	(Conv2D (None, 7, 7, 512))	524800	conv4_block6_out[0][0]
conv5_block1_1_bn	(BatchNormali (None, 7, 7, 512))	2048	conv5_block1_1_conv[0][0]
conv5_block1_1_relu	(Activation (None, 7, 7, 512))	0	conv5_block1_1_bn[0][0]
conv5_block1_2_conv	(Conv2D (None, 7, 7, 512))	2359808	conv5_block1_1_relu[0][0]
conv5_block1_2_bn	(BatchNormali (None, 7, 7, 512))	2048	conv5_block1_2_conv[0][0]
conv5_block1_2_relu	(Activation (None, 7, 7, 512))	0	conv5_block1_2_bn[0][0]
conv5_block1_0_conv	(Conv2D (None, 7, 7, 2048))	2099200	conv4_block6_out[0][0]
conv5_block1_3_conv	(Conv2D (None, 7, 7, 2048))	1050624	conv5_block1_2_relu[0][0]
conv5_block1_0_bn	(BatchNormali (None, 7, 7, 2048))	8192	conv5_block1_0_conv[0][0]
conv5_block1_3_bn	(BatchNormali (None, 7, 7, 2048))	8192	conv5_block1_3_conv[0][0]

]			
<hr/>			
conv5_block1_add (Add)	(None, 7, 7, 2048)	0	conv5_block1_0_bn[0][0] conv5_block1_3_bn[0][0]
]			
<hr/>			
conv5_block1_out (Activation)	(None, 7, 7, 2048)	0	conv5_block1_add[0][0]
<hr/>			
conv5_block2_1_conv (Conv2D)	(None, 7, 7, 512)	1049088	conv5_block1_out[0][0]
<hr/>			
conv5_block2_1_bn (BatchNormali ]	(None, 7, 7, 512)	2048	conv5_block2_1_conv[0][0]
<hr/>			
conv5_block2_1_relu (Activation	(None, 7, 7, 512)	0	conv5_block2_1_bn[0][0]
<hr/>			
conv5_block2_2_conv (Conv2D) ]	(None, 7, 7, 512)	2359808	conv5_block2_1_relu[0][0]
<hr/>			
conv5_block2_2_bn (BatchNormali ]	(None, 7, 7, 512)	2048	conv5_block2_2_conv[0][0]
<hr/>			
conv5_block2_2_relu (Activation	(None, 7, 7, 512)	0	conv5_block2_2_bn[0][0]
<hr/>			
conv5_block2_3_conv (Conv2D) ]	(None, 7, 7, 2048)	1050624	conv5_block2_2_relu[0][0]
<hr/>			
conv5_block2_3_bn (BatchNormali ]	(None, 7, 7, 2048)	8192	conv5_block2_3_conv[0][0]
<hr/>			
conv5_block2_add (Add)	(None, 7, 7, 2048)	0	conv5_block1_out[0][0] conv5_block2_3_bn[0][0]
]			
<hr/>			
conv5_block2_out (Activation)	(None, 7, 7, 2048)	0	conv5_block2_add[0][0]
<hr/>			
conv5_block3_1_conv (Conv2D)	(None, 7, 7, 512)	1049088	conv5_block2_out[0][0]
<hr/>			
conv5_block3_1_bn (BatchNormali ]	(None, 7, 7, 512)	2048	conv5_block3_1_conv[0][0]
<hr/>			
conv5_block3_1_relu (Activation	(None, 7, 7, 512)	0	conv5_block3_1_bn[0][0]
<hr/>			
conv5_block3_2_conv (Conv2D) ]	(None, 7, 7, 512)	2359808	conv5_block3_1_relu[0][0]
<hr/>			
conv5_block3_2_bn (BatchNormali	(None, 7, 7, 512)	2048	conv5_block3_2_conv[0][0]

```

]

conv5_block3_2_relu (Activation (None, 7, 7, 512)) 0 conv5_block3_2_bn[0][0]

conv5_block3_3_conv (Conv2D) (None, 7, 7, 2048) 1050624 conv5_block3_2_relu[0][0]

conv5_block3_3_bn (BatchNormali (None, 7, 7, 2048) 8192 conv5_block3_3_conv[0][0]

conv5_block3_add (Add) (None, 7, 7, 2048) 0 conv5_block2_out[0][0]
conv5_block3_3_bn[0][0]

conv5_block3_out (Activation) (None, 7, 7, 2048) 0 conv5_block3_add[0][0]

=====
=====
Total params: 23,587,712
Trainable params: 53,120
Non-trainable params: 23,534,592

```

## Converting the output layer as per our dataset

In [13]:

```

model=tf.keras.models.Sequential()
model.add(UpSampling2D(size=(7, 7), interpolation='bilinear'))
model.add(resnet_model)
model.add(GlobalAveragePooling2D())
model.add(Dropout(.25))
model.add(Dense(256, activation='relu'))
model.add(BatchNormalization())
model.add(Dense(100, activation='softmax'))

```

In [14]:

```

optimizer = tf.keras.optimizers.SGD(learning_rate=1e-3, momentum=0.9)

```

In [15]:

```

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

```

In [16]:

```

start_time=timer(None)
result=model.fit(
    train_datagen.flow(x_train, y_train, batch_size = 128),
    validation_data = (x_val, y_val),
    epochs = 50,
    verbose = 1,
    callbacks = [learning_rate_reduction]
)
timer(start_time)

```

2021-08-13 22:27:11.100467  
Epoch 1/50

313/313 [=====] - 169s 525ms/step - loss: 4.7197 - accuracy: 0.0  
497 - val\_loss: 12.7910 - val\_accuracy: 0.0102  
Epoch 2/50  
313/313 [=====] - 162s 517ms/step - loss: 3.2714 - accuracy: 0.2  
299 - val\_loss: 6.4920 - val\_accuracy: 0.0116  
Epoch 3/50  
313/313 [=====] - 162s 517ms/step - loss: 2.7051 - accuracy: 0.3  
344 - val\_loss: 3.2080 - val\_accuracy: 0.2347  
Epoch 4/50  
313/313 [=====] - 162s 517ms/step - loss: 2.3692 - accuracy: 0.4  
049 - val\_loss: 2.0134 - val\_accuracy: 0.4748  
Epoch 5/50  
313/313 [=====] - 162s 517ms/step - loss: 2.1650 - accuracy: 0.4  
417 - val\_loss: 1.8416 - val\_accuracy: 0.5116  
Epoch 6/50  
313/313 [=====] - 162s 517ms/step - loss: 1.9954 - accuracy: 0.4  
801 - val\_loss: 1.7502 - val\_accuracy: 0.5280  
Epoch 7/50  
313/313 [=====] - 162s 517ms/step - loss: 1.8640 - accuracy: 0.5  
034 - val\_loss: 1.6559 - val\_accuracy: 0.5506  
Epoch 8/50  
313/313 [=====] - 162s 517ms/step - loss: 1.7697 - accuracy: 0.5  
286 - val\_loss: 1.5733 - val\_accuracy: 0.5723  
Epoch 9/50  
313/313 [=====] - 162s 517ms/step - loss: 1.6839 - accuracy: 0.5  
445 - val\_loss: 1.5109 - val\_accuracy: 0.5861  
Epoch 10/50  
313/313 [=====] - 162s 518ms/step - loss: 1.6186 - accuracy: 0.5  
599 - val\_loss: 1.4509 - val\_accuracy: 0.5997  
Epoch 11/50  
313/313 [=====] - 162s 517ms/step - loss: 1.5624 - accuracy: 0.5  
749 - val\_loss: 1.4044 - val\_accuracy: 0.6083  
Epoch 12/50  
313/313 [=====] - 162s 517ms/step - loss: 1.5044 - accuracy: 0.5  
848 - val\_loss: 1.3655 - val\_accuracy: 0.6195  
Epoch 13/50  
313/313 [=====] - 162s 517ms/step - loss: 1.4539 - accuracy: 0.5  
971 - val\_loss: 1.3278 - val\_accuracy: 0.6288  
Epoch 14/50  
313/313 [=====] - 162s 517ms/step - loss: 1.4150 - accuracy: 0.6  
058 - val\_loss: 1.2960 - val\_accuracy: 0.6368  
Epoch 15/50  
313/313 [=====] - 162s 517ms/step - loss: 1.3943 - accuracy: 0.6  
100 - val\_loss: 1.2632 - val\_accuracy: 0.6430  
Epoch 16/50  
313/313 [=====] - 161s 516ms/step - loss: 1.3572 - accuracy: 0.6  
201 - val\_loss: 1.2345 - val\_accuracy: 0.6523  
Epoch 17/50  
313/313 [=====] - 161s 516ms/step - loss: 1.3109 - accuracy: 0.6  
288 - val\_loss: 1.2104 - val\_accuracy: 0.6568  
Epoch 18/50  
313/313 [=====] - 162s 516ms/step - loss: 1.2867 - accuracy: 0.6  
378 - val\_loss: 1.1907 - val\_accuracy: 0.6613  
Epoch 19/50  
313/313 [=====] - 161s 516ms/step - loss: 1.2602 - accuracy: 0.6  
420 - val\_loss: 1.1700 - val\_accuracy: 0.6660  
Epoch 20/50  
313/313 [=====] - 162s 516ms/step - loss: 1.2423 - accuracy: 0.6  
461 - val\_loss: 1.1539 - val\_accuracy: 0.6710  
Epoch 21/50  
313/313 [=====] - 162s 516ms/step - loss: 1.2215 - accuracy: 0.6  
506 - val\_loss: 1.1359 - val\_accuracy: 0.6734  
Epoch 22/50  
313/313 [=====] - 162s 517ms/step - loss: 1.1983 - accuracy: 0.6  
591 - val\_loss: 1.1219 - val\_accuracy: 0.6768  
Epoch 23/50  
313/313 [=====] - 162s 516ms/step - loss: 1.1791 - accuracy: 0.6  
630 - val\_loss: 1.1059 - val\_accuracy: 0.6793  
Epoch 24/50  
313/313 [=====] - 162s 516ms/step - loss: 1.1657 - accuracy: 0.6  
650 - val\_loss: 1.0918 - val\_accuracy: 0.6825  
Epoch 25/50

313/313 [=====] - 162s 516ms/step - loss: 1.1404 - accuracy: 0.6  
715 - val\_loss: 1.0809 - val\_accuracy: 0.6857  
Epoch 26/50  
313/313 [=====] - 162s 516ms/step - loss: 1.1255 - accuracy: 0.6  
753 - val\_loss: 1.0686 - val\_accuracy: 0.6888  
Epoch 27/50  
313/313 [=====] - 162s 516ms/step - loss: 1.1262 - accuracy: 0.6  
732 - val\_loss: 1.0565 - val\_accuracy: 0.6928  
Epoch 28/50  
313/313 [=====] - 162s 517ms/step - loss: 1.0848 - accuracy: 0.6  
855 - val\_loss: 1.0448 - val\_accuracy: 0.6941  
Epoch 29/50  
313/313 [=====] - 162s 516ms/step - loss: 1.0941 - accuracy: 0.6  
871 - val\_loss: 1.0365 - val\_accuracy: 0.6980  
Epoch 30/50  
313/313 [=====] - 162s 517ms/step - loss: 1.0771 - accuracy: 0.6  
869 - val\_loss: 1.0258 - val\_accuracy: 0.6990  
Epoch 31/50  
313/313 [=====] - 162s 516ms/step - loss: 1.0673 - accuracy: 0.6  
907 - val\_loss: 1.0173 - val\_accuracy: 0.7024  
Epoch 32/50  
313/313 [=====] - 162s 517ms/step - loss: 1.0362 - accuracy: 0.7  
001 - val\_loss: 1.0111 - val\_accuracy: 0.7040  
Epoch 33/50  
313/313 [=====] - 162s 516ms/step - loss: 1.0268 - accuracy: 0.7  
024 - val\_loss: 1.0026 - val\_accuracy: 0.7066  
Epoch 34/50  
313/313 [=====] - 162s 517ms/step - loss: 1.0151 - accuracy: 0.7  
076 - val\_loss: 0.9936 - val\_accuracy: 0.7085  
Epoch 35/50  
313/313 [=====] - 162s 517ms/step - loss: 1.0074 - accuracy: 0.7  
026 - val\_loss: 0.9865 - val\_accuracy: 0.7107  
Epoch 36/50  
313/313 [=====] - 162s 517ms/step - loss: 1.0065 - accuracy: 0.7  
046 - val\_loss: 0.9836 - val\_accuracy: 0.7125  
Epoch 37/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9891 - accuracy: 0.7  
117 - val\_loss: 0.9720 - val\_accuracy: 0.7171  
Epoch 38/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9835 - accuracy: 0.7  
086 - val\_loss: 0.9715 - val\_accuracy: 0.7166  
Epoch 39/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9734 - accuracy: 0.7  
155 - val\_loss: 0.9626 - val\_accuracy: 0.7182  
Epoch 40/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9610 - accuracy: 0.7  
212 - val\_loss: 0.9585 - val\_accuracy: 0.7185  
Epoch 41/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9576 - accuracy: 0.7  
187 - val\_loss: 0.9522 - val\_accuracy: 0.7205  
Epoch 42/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9417 - accuracy: 0.7  
225 - val\_loss: 0.9427 - val\_accuracy: 0.7229  
Epoch 43/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9391 - accuracy: 0.7  
209 - val\_loss: 0.9394 - val\_accuracy: 0.7238  
Epoch 44/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9363 - accuracy: 0.7  
217 - val\_loss: 0.9325 - val\_accuracy: 0.7256  
Epoch 45/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9105 - accuracy: 0.7  
316 - val\_loss: 0.9308 - val\_accuracy: 0.7278  
Epoch 46/50  
313/313 [=====] - 162s 517ms/step - loss: 0.8966 - accuracy: 0.7  
355 - val\_loss: 0.9274 - val\_accuracy: 0.7265  
Epoch 47/50  
313/313 [=====] - 162s 517ms/step - loss: 0.9023 - accuracy: 0.7  
337 - val\_loss: 0.9256 - val\_accuracy: 0.7271  
Epoch 48/50  
313/313 [=====] - 161s 516ms/step - loss: 0.8994 - accuracy: 0.7  
326 - val\_loss: 0.9194 - val\_accuracy: 0.7282  
Epoch 49/50



313/313 [=====] - 162s 517ms/step - loss: 0.8868 - accuracy: 0.7373 - val\_loss: 0.9162 - val\_accuracy: 0.7279  
Epoch 50/50  
313/313 [=====] - 162s 516ms/step - loss: 0.8902 - accuracy: 0.7358 - val\_loss: 0.9101 - val\_accuracy: 0.7309  
Time taken: 2 hours 15 minutes and 4.86 seconds.

In [17]:

```
model.summary()
```

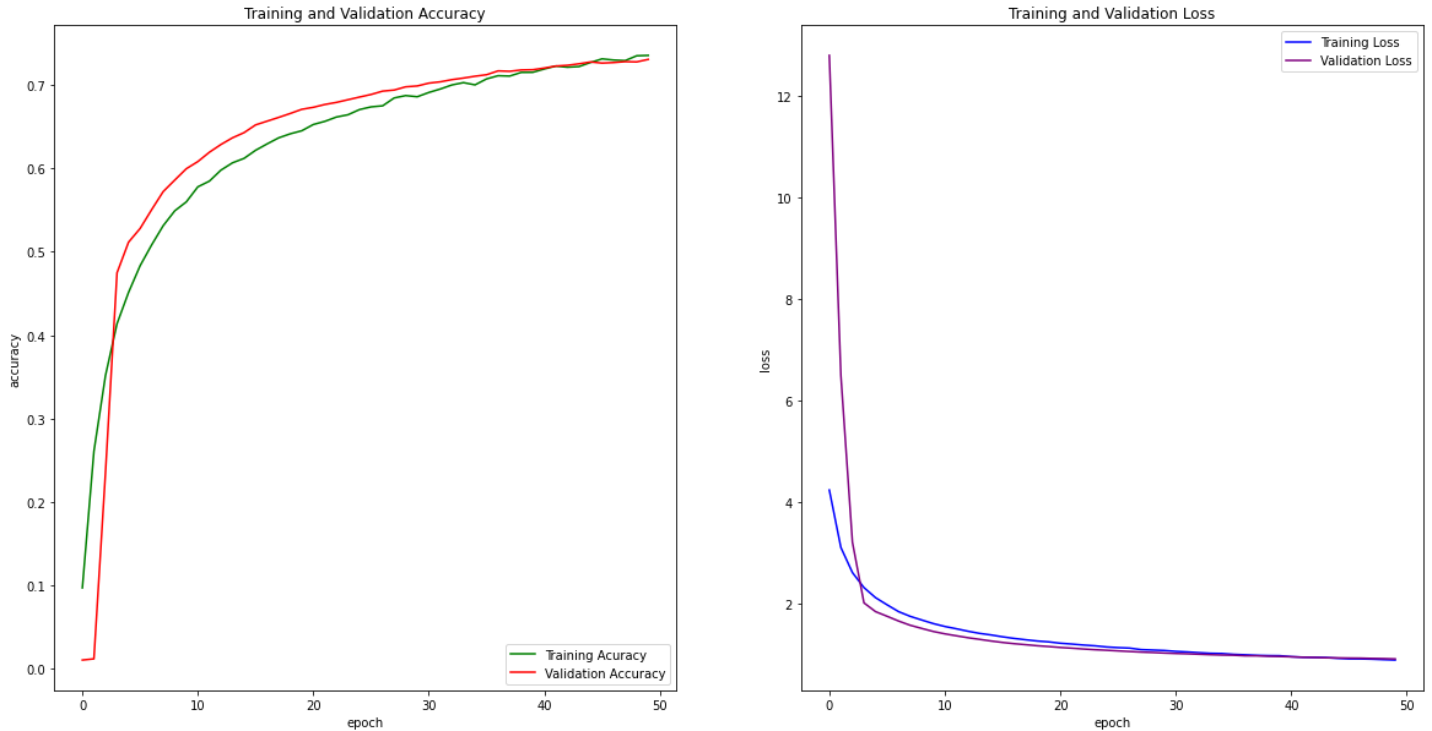
Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
up_sampling2d (UpSampling2D)	(None, None, None, None)	0
resnet50 (Functional)	(None, 7, 7, 2048)	23587712
global_average_pooling2d (Gl	(None, 2048)	0
dropout (Dropout)	(None, 2048)	0
dense (Dense)	(None, 256)	524544
batch_normalization (BatchNo	(None, 256)	1024
dense_1 (Dense)	(None, 100)	25700
=====		
Total params: 24,138,980		
Trainable params: 603,876		
Non-trainable params: 23,535,104		

## Plot accuracy and Loss

In [18]:

```
plot_acc_loss(result)
```



## Predictions,Accuracy and Confusion Matrix

In [19]:

```
y_pred = model.predict_classes(X_test)
y_true = Y_test.ravel()
print(y_pred.shape)
print(y_true.shape)
```

```
/opt/conda/lib/python3.7/site-packages/tensorflow/python/keras/engine/sequential.py:450:
UserWarning: `model.predict_classes()` is deprecated and will be removed after 2021-01-01
. Please use instead: * `np.argmax(model.predict(x), axis=-1)`, if your model does multi
-class classification (e.g. if it uses a `softmax` last-layer activation). * `(model.pre
dict(x) > 0.5).astype("int32")`, if your model does binary classification (e.g. if it
uses a `sigmoid` last-layer activation).
  warnings.warn("`model.predict_classes()` is deprecated and '
```

```
(10000,)
(10000,)
```

In [20]:

```
from sklearn.metrics import accuracy_score, confusion_matrix
print("Testing Accuracy: ", accuracy_score(y_true, y_pred))
```

Testing Accuracy: 0.7193

In [21]:

```
cm = confusion_matrix(y_true, y_pred)
cm
```

Out[21]:

```
array([[88, 0, 0, ..., 0, 0, 0],
       [ 0, 75, 1, ..., 0, 0, 0],
       [ 0, 1, 67, ..., 0, 2, 0],
       ...,
       [ 0, 0, 0, ..., 68, 0, 0],
       [ 0, 0, 4, ..., 0, 48, 0],
       [ 0, 0, 1, ..., 0, 0, 79]])
```