```
In [1]: # Import libraries
        import os, cv2
        import time
        import numpy as np
        import matplotlib.pyplot as plt
        from sklearn.utils import shuffle
        from sklearn.model_selection import train_test_split
        from keras.preprocessing import image
        from keras.utils import np utils
        from keras.models import Sequential
        from keras.layers import Input
        from keras.layers.core import Dense, Dropout, Activation, Flatten
        from keras.layers.convolutional import Convolution2D, MaxPooling2D
        from keras import callbacks
        from keras import backend as K
        K.set_image_data_format('channels_last')
        from sklearn.metrics import classification_report,confusion_matrix
        import itertools
        from keras.models import Model
        from tensorflow.keras.applications.efficientnet import EfficientNetB0
        from tensorflow.keras.applications.inception_v3 import decode_predictions
```

#### Set path for application

```
In [2]: data_path = 'D:/Harold/MyDNN/DataSet/Chest_xray_seperate'
    data_dir_list = os.listdir(data_path)
    print(data_path)
```

D:/Harold/MyDNN/DataSet/Chest xray seperate

# Set Image Size and Epocs

```
In [3]: img_rows=128
   img_cols=128
   num_channel=3
   num_epoch=100
```

## Define the number of classes

```
In [4]: num_classes = 2
   img_data_list=[]
```

```
In [5]: def preprocess_input(x):
            x[:, :, :, 0] = 103.939
            x[:, :, :, 1] = 116.779
            x[:, :, :, 2] = 123.68
            # 'RGB'->'BGR'
            x = x[:, :, :, ::-1]
            return x
        def data preparation():
            for dataset in data dir list:
                img list=os.listdir(data path+'/'+ dataset)
                print ('Loading the images of dataset-'+'{}\n'.format(dataset))
                for img in img list:
                    img path = data path + '/'+ dataset + '/'+ img
                    img = image.load_img(img_path, target_size=(224, 224))
                    x = image.img_to_array(img)
                    x = np.expand dims(x, axis=0)
                    x = preprocess input(x)
                      print('Input image shape:', x.shape)
                    img data list.append(x)
                print('Loading Complete')
              for dataset in data dir list:
                  img list=os.listdir(data path+'/'+ dataset)
                  print ('Loading the images of dataset-'+'{}\n'.format(dataset))
                  for img in img_list:
                      img path = data path + '/'+ dataset + '/'+ img
                      img = image.load_img(img_path, target_size=(224, 224))
                      x = image.img_to_array(img)
                      x = np.expand_dims(x, axis=0)
                      x = preprocess_input(x)
        #
                        print('Input image shape:', x.shape)
        #
                      img data list.append(x)
                  print('Loading Complete')
        def display loss accuracy(hist):
            train loss=hist.history['loss']
            val loss=hist.history['val loss']
            train acc=hist.history['accuracy']
            val acc=hist.history['val accuracy']
            xc=range(num_epoch)
            plt.figure(1, figsize=(7,5))
            plt.plot(xc, train loss)
            plt.plot(xc, val loss)
            plt.xlabel('num of Epochs')
            plt.ylabel('loss')
            plt.title('train loss vs val loss')
            plt.grid(True)
            plt.legend(['train','val'])
            #print plt.style.available # use bmh, classic,ggplot for big pictures
            plt.style.use(['classic'])
            plt.figure(2, figsize=(7,5))
            plt.plot(xc, train_acc)
            plt.plot(xc, val acc)
            plt.xlabel('num of Epochs')
            plt.ylabel('accuracy')
            plt.title('train_acc vs val_acc')
            plt.grid(True)
            plt.legend(['train','val'],loc=4)
            #print plt.style.available # use bmh, classic,ggplot for big pictures
            plt.style.use(['classic'])
```

```
def get featuremaps(model, layer idx, X batch):
    get activations = K.function([model.layers[0].input, K.learning phase()],[mode
1.layers[layer idx].output,])
    activations = get activations([X batch,0])
    return activations
def plot_featuremap_activations(activations):
    print (np.shape(activations))
    feature maps = activations[0][0]
    print (np.shape(feature maps))
   print (feature_maps.shape)
    fig=plt.figure(figsize=(16,16))
    plt.imshow(feature maps[:,:,filter num],cmap='gray')
    plt.savefig("featuremaps-layer-{}".format(layer_num) + "-filternum-{}".format(f
ilter num)+'.jpg')
    num of featuremaps=feature maps.shape[2]
    fig=plt.figure(figsize=(16,16))
    plt.title("featuremaps-layer-{}".format(layer num))
    subplot num=int(np.ceil(np.sqrt(num of featuremaps)))
    for i in range(int(num of featuremaps)):
        ax = fig.add subplot(subplot num, subplot num, i+1)
        \#ax.imshow(output\ image[0,:,:,i],interpolation='nearest')\ \#to\ see\ the\ firs
t filter
        ax.imshow(feature maps[:,:,i],cmap='gray')
        plt.xticks([])
        plt.yticks([])
        plt.tight layout()
    fig.savefig("featuremaps-layer-{}".format(layer num) + '.jpg')
# Plotting the confusion matrix
def plot confusion matrix(cm, classes,
                          normalize=False,
                          title='Confusion matrix',
                          cmap=plt.cm.Blues):
    This function prints and plots the confusion matrix.
    Normalization can be applied by setting `normalize=True`.
    plt.figure()
    plt.imshow(cm, interpolation='nearest', cmap=cmap)
    plt.title(title)
   plt.colorbar()
   tick_marks = np.arange(len(classes))
    plt.xticks(tick marks, classes, rotation=45)
    plt.yticks(tick marks, classes)
    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
        print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')
    print(cm)
    thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j, i, cm[i, j],
                 horizontalalignment="center",
                 color="white" if cm[i, j] > thresh else "black")
```

```
plt.tight_layout()
plt.ylabel('True label')
plt.xlabel('Predicted label')
plt.show()
```

#### **Data Preperation**

```
In [6]: # Calling Data Preperation
        data preperation()
        Loading the images of dataset-NORMAL
        Loading Complete
        Loading the images of dataset-PNEUMONIA
        Loading Complete
In [7]: print (len(img_data_list))
        img_data = np.array(img_data_list)
        #img_data = img_data.astype('float32')
        print (img_data.shape)
        img data=np.rollaxis(img data,1,0)
        print (img data.shape)
        img_data=img_data[0]
        print (img data.shape)
        5856
        (5856, 1, 224, 224, 3)
        (1, 5856, 224, 224, 3)
        (5856, 224, 224, 3)
```

## Assiging Labels

```
In [8]: num_of_samples = img_data.shape[0]
    labels = np.ones((num_of_samples,),dtype='int64')

    labels[0:1582]=0
    labels[1583:5856]=1

    names = ['normal','pneumonia']
```

### Creating clasas labels to one-hot encoding

```
In [9]: # convert class labels to on-hot encoding
Y = np_utils.to_categorical(labels, num_classes)
```

## Split Data set into training and validation set

```
In [10]: #Shuffle the dataset
    x,y = shuffle(img_data,Y, random_state=2)
    # Split the dataset
    X_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=2)
```

# Model Definition

Training the classifier alone

```
In [11]: image_input = Input(shape=(224, 224, 3))
    model = EfficientNetB0(input_tensor=image_input, include_top=True, weights='imagenet
')
    model.summary()
    last_layer = model.get_layer('avg_pool').output
    x= Flatten(name='flatten')(last_layer)
    out = Dense(num_classes, activation='softmax', name='output_layer')(x)
    custom_resnet_model = Model(inputs=image_input,outputs= out)
    custom_resnet_model.summary()
```

Model: "efficientnetb0"

Layer (type)						Connected to
input_1 (InputLayer)	[(None	, 224,	224,	3)	0	
rescaling (Rescaling)	(None,	224,	224,	3)	0	input_1[0][0]
normalization (Normalization)	(None,	224,	224,	3)	7	rescaling[0][0]
stem_conv_pad (ZeroPadding2D) [0][0]	(None,	225,	225,	3)	0	normalization
stem_conv (Conv2D) [0][0]	(None,	112,	112,	32)	864	stem_conv_pad
stem_bn (BatchNormalization)	(None,	112,	112,	32)	128	stem_conv[0][0]
stem_activation (Activation)	(None,	112,	112,	32)	0	stem_bn[0][0]
block1a_dwconv (DepthwiseConv2D [0][0]	(None,	112,	112,	32)	288	stem_activation
block1a_bn (BatchNormalization) [0][0]	(None,	112,	112,	32)	128	block1a_dwconv
blockla_activation (Activation) [0][0]	(None,	112,	112,	32)	0	block1a_bn
block1a_se_squeeze (GlobalAveraion[0][0]	(None,	32)			0	blockla_activat
block1a_se_reshape (Reshape) eze[0][0]	(None,	1, 1,	32)		0	block1a_se_sque
block1a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1,	8)		264	blockla_se_resh
block1a_se_expand (Conv2D) ce[0][0]	(None,	1, 1,	32)		288	block1a_se_redu
block1a_se_excite (Multiply) ion[0][0] nd[0][0]	(None,	112,	112,	32)	0	block1a_activat block1a_se_expa

<pre>block1a_project_conv (Conv2D) te[0][0]</pre>	(None,	112, 112, 16)	512	blockla_se_exci
block1a_project_bn (BatchNormal _conv[0][0]	(None,	112, 112, 16)	64	block1a_project
block2a_expand_conv (Conv2D) _bn[0][0]	(None,	112, 112, 96)	1536	blockla_project
block2a_expand_bn (BatchNormali conv[0][0]	(None,	112, 112, 96)	384	block2a_expand_
block2a_expand_activation (Acti bn[0][0]	(None,	112, 112, 96)	0	block2a_expand_
block2a_dwconv_pad (ZeroPadding activation[0][0]	(None,	113, 113, 96)	0	block2a_expand_
block2a_dwconv (DepthwiseConv2D pad[0][0]	(None,	56, 56, 96)	864	block2a_dwconv_
block2a_bn (BatchNormalization) [0][0]	(None,	56, 56, 96)	384	block2a_dwconv
block2a_activation (Activation) [0][0]	(None,	56, 56, 96)	0	block2a_bn
block2a_se_squeeze (GlobalAvera ion[0][0]	(None,	96)	0	block2a_activat
block2a_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 96)	0	block2a_se_sque
block2a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 4)	388	block2a_se_resh
block2a_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 96)	480	block2a_se_redu
block2a_se_excite (Multiply) ion[0][0] nd[0][0]	(None,	56, 56, 96)	0	block2a_activat block2a_se_expa
block2a_project_conv (Conv2D) te[0][0]	(None,	56, 56, 24)	2304	block2a_se_exci
block2a_project_bn (BatchNormal _conv[0][0]	(None,	56, 56, 24)	96	block2a_project

block2b_expand_conv (Conv2D) _bn[0][0]	(None,	56, 56, 144)	3456	block2a_project
block2b_expand_bn (BatchNormali conv[0][0]	(None,	56, 56, 144)	576	block2b_expand_
block2b_expand_activation (Acti bn[0][0]	(None,	56, 56, 144)	0	block2b_expand_
block2b_dwconv (DepthwiseConv2D activation[0][0]	(None,	56, 56, 144)	1296	block2b_expand_
block2b_bn (BatchNormalization) [0][0]	(None,	56, 56, 144)	576	block2b_dwconv
block2b_activation (Activation) [0][0]	(None,	56, 56, 144)	0	block2b_bn
block2b_se_squeeze (GlobalAvera ion[0][0]	(None,	144)	0	block2b_activat
block2b_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 144)	0	block2b_se_sque
block2b_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 6)	870	block2b_se_resh
block2b_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 144)	1008	block2b_se_redu
block2b_se_excite (Multiply) ion[0][0] nd[0][0]	(None,	56, 56, 144)	0	block2b_activat block2b_se_expa
block2b_project_conv (Conv2D) te[0][0]	(None,	56, 56, 24)	3456	block2b_se_exci
block2b_project_bn (BatchNormal _conv[0][0]	(None,	56, 56, 24)	96	block2b_project
block2b_drop (Dropout) _bn[0][0]	(None,	56, 56, 24)	0	block2b_project
block2b_add (Add) [0][0] _bn[0][0]	(None,	56, 56, 24)	0	block2b_drop block2a_project

block3a_expand_conv (Conv2D) [0][0]	(None,	56, 56, 144)	3456	block2b_add
block3a_expand_bn (BatchNormali conv[0][0]	(None,	56, 56, 144)	576	block3a_expand_
block3a_expand_activation (Acti bn[0][0]	(None,	56, 56, 144)	0	block3a_expand_
block3a_dwconv_pad (ZeroPadding activation[0][0]	(None,	59, 59, 144)	0	block3a_expand_
block3a_dwconv (DepthwiseConv2D pad[0][0]	(None,	28, 28, 144)	3600	block3a_dwconv_
block3a_bn (BatchNormalization) [0][0]	(None,	28, 28, 144)	576	block3a_dwconv
block3a_activation (Activation) [0][0]	(None,	28, 28, 144)	0	block3a_bn
block3a_se_squeeze (GlobalAvera ion[0][0]	(None,	144)	0	block3a_activat
block3a_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 144)	0	block3a_se_sque
block3a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 6)	870	block3a_se_resh
block3a_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 144)	1008	block3a_se_redu
block3a_se_excite (Multiply) ion[0][0]	(None,	28, 28, 144)	0	block3a_activat
nd[0][0]				block3a_se_expa
block3a_project_conv (Conv2D) te[0][0]	(None,	28, 28, 40)	5760	block3a_se_exci
block3a_project_bn (BatchNormal _conv[0][0]	(None,	28, 28, 40)	160	block3a_project
block3b_expand_conv (Conv2D) bn[0][0]	(None,	28, 28, 240)	9600	block3a_project

<pre>block3b_expand_bn (BatchNormali conv[0][0]</pre>	(None,	28, 28, 240	) 960	block3b_expand_
block3b_expand_activation (Acti bn[0][0]	(None,	28, 28, 240	) 0	block3b_expand_
block3b_dwconv (DepthwiseConv2D activation[0][0]	(None,	28, 28, 240	) 6000	block3b_expand_
block3b_bn (BatchNormalization) [0][0]	(None,	28, 28, 240	) 960	block3b_dwconv
block3b_activation (Activation) [0][0]	(None,	28, 28, 240	) 0	block3b_bn
block3b_se_squeeze (GlobalAvera ion[0][0]	(None,	240)	0	block3b_activat
block3b_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 240)	0	block3b_se_sque
block3b_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 10)	2410	block3b_se_resh
block3b_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 240)	2640	block3b_se_redu
block3b_se_excite (Multiply) ion[0][0]  nd[0][0]	(None,	28, 28, 240	) 0	block3b_activat block3b_se_expa
block3b_project_conv (Conv2D) te[0][0]	(None,	28, 28, 40)	9600	block3b_se_exci
block3b_project_bn (BatchNormal _conv[0][0]	(None,	28, 28, 40)	160	block3b_project
block3b_drop (Dropout) _bn[0][0]	(None,	28, 28, 40)	0	block3b_project
block3b_add (Add) [0][0]	(None,	28, 28, 40)	0	block3b_drop block3a_project
_bn[0][0]				·
block4a_expand_conv (Conv2D) [0][0]	(None,	28, 28, 240	) 9600	block3b_add

<pre>block4a_expand_bn (BatchNormali conv[0][0]</pre>	(None,	28, 28, 240)	960	block4a_expand_
block4a_expand_activation (Acti bn[0][0]	(None,	28, 28, 240)	0	block4a_expand_
block4a_dwconv_pad (ZeroPadding activation[0][0]	(None,	29, 29, 240)	0	block4a_expand_
block4a_dwconv (DepthwiseConv2D pad[0][0]	(None,	14, 14, 240)	2160	block4a_dwconv_
block4a_bn (BatchNormalization) [0][0]	(None,	14, 14, 240)	960	block4a_dwconv
block4a_activation (Activation) [0][0]	(None,	14, 14, 240)	0	block4a_bn
block4a_se_squeeze (GlobalAvera ion[0][0]	(None,	240)	0	block4a_activat
block4a_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 240)	0	block4a_se_sque
block4a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 10)	2410	block4a_se_resh
block4a_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 240)	2640	block4a_se_redu
block4a_se_excite (Multiply) ion[0][0]	(None,	14, 14, 240)	0	block4a_activat
nd[0][0]				
block4a_project_conv (Conv2D) te[0][0]	(None,	14, 14, 80)	19200	block4a_se_exci
block4a_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 80)	320	block4a_project
block4b_expand_conv (Conv2D) _bn[0][0]	(None,	14, 14, 480)	38400	block4a_project
block4b_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 480)	1920	block4b_expand_
block4b_expand_activation (Acti bn[0][0]	(None,	14, 14, 480)	0	block4b_expand_

block4b_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 480)	4320	block4b_expand_
block4b_bn (BatchNormalization) [0][0]	(None,	14, 14, 480)	1920	block4b_dwconv
block4b_activation (Activation) [0][0]	(None,	14, 14, 480)	0	block4b_bn
block4b_se_squeeze (GlobalAvera ion[0][0]	(None,	480)	0	block4b_activat
block4b_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 480)	0	block4b_se_sque
block4b_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 20)	9620	block4b_se_resh
block4b_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 480)	10080	block4b_se_redu
block4b_se_excite (Multiply) ion[0][0]	(None,	14, 14, 480)	0	block4b_activat
nd[0][0]				DIOCK4D_Se_expa
block4b_project_conv (Conv2D) te[0][0]	(None,	14, 14, 80)	38400	block4b_se_exci
block4b_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 80)	320	block4b_project
block4b_drop (Dropout) _bn[0][0]	(None,	14, 14, 80)	0	block4b_project
block4b_add (Add) [0][0]	(None,	14, 14, 80)	0	block4b_drop
_bn[0][0]				block4a_project
block4c_expand_conv (Conv2D) [0][0]	(None,	14, 14, 480)	38400	block4b_add
block4c_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 480)	1920	block4c_expand_
block4c_expand_activation (Acti bn[0][0]	(None,	14, 14, 480)	0	block4c_expand_

block4c_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 480)	4320	block4c_expand_
block4c_bn (BatchNormalization) [0][0]	(None,	14, 14, 480)	1920	block4c_dwconv
block4c_activation (Activation) [0][0]	(None,	14, 14, 480)	0	block4c_bn
block4c_se_squeeze (GlobalAvera ion[0][0]	(None,	480)	0	block4c_activat
block4c_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 480)	0	block4c_se_sque
block4c_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 20)	9620	block4c_se_resh
block4c_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 480)	10080	block4c_se_redu
block4c_se_excite (Multiply) ion[0][0]	(None,	14, 14, 480)	0	block4c_activat block4c_se_expa
nd[0][0] block4c_project_conv (Conv2D) te[0][0]	(None,	14, 14, 80)	38400	block4c_se_exci
block4c_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 80)	320	block4c_project
block4c_drop (Dropout) _bn[0][0]	(None,	14, 14, 80)	0	block4c_project
block4c_add (Add) [0][0]	(None,	14, 14, 80)	0	block4c_drop
[0][0]				block4b_add
block5a_expand_conv (Conv2D) [0][0]	(None,	14, 14, 480)	38400	block4c_add
block5a_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 480)	1920	block5a_expand_
block5a_expand_activation (Acti bn[0][0]	(None,	14, 14, 480)	0	block5a_expand_

block5a_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 480)	12000	block5a_expand_
block5a_bn (BatchNormalization) [0][0]	(None,	14, 14, 480)	1920	block5a_dwconv
block5a_activation (Activation) [0][0]	(None,	14, 14, 480)	0	block5a_bn
block5a_se_squeeze (GlobalAvera ion[0][0]	(None,	480)	0	block5a_activat
block5a_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 480)	0	block5a_se_sque
block5a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 20)	9620	block5a_se_resh
block5a_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 480)	10080	block5a_se_redu
block5a_se_excite (Multiply) ion[0][0]	(None,	14, 14, 480)	0	block5a_activat
nd[0][0]				block5a_se_expa
block5a_project_conv (Conv2D) te[0][0]	(None,	14, 14, 112)	53760	block5a_se_exci
block5a_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 112)	448	block5a_project
block5b_expand_conv (Conv2D) _bn[0][0]	(None,	14, 14, 672)	75264	block5a_project
block5b_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 672)	2688	block5b_expand_
block5b_expand_activation (Acti bn[0][0]	(None,	14, 14, 672)	0	block5b_expand_
block5b_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 672)	16800	block5b_expand_
block5b_bn (BatchNormalization) [0][0]	(None,	14, 14, 672)	2688	block5b_dwconv

block5b_activation (Activation) [0][0]	(None,	14, 14, 672)	0	block5b_bn
block5b_se_squeeze (GlobalAvera ion[0][0]	(None,	672)	0	block5b_activat
block5b_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 672)	0	block5b_se_sque
block5b_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 28)	18844	block5b_se_resh
block5b_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 672)	19488	block5b_se_redu
block5b_se_excite (Multiply) ion[0][0]	(None,	14, 14, 672)	0	block5b_activat
nd[0][0]				block5b_se_expa
block5b_project_conv (Conv2D) te[0][0]	(None,	14, 14, 112)	75264	block5b_se_exci
block5b_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 112)	448	block5b_project
block5b_drop (Dropout) _bn[0][0]	(None,	14, 14, 112)	0	block5b_project
block5b_add (Add) [0][0]	(None,	14, 14, 112)	0	block5b_drop
_bn[0][0]				block5a_project
block5c_expand_conv (Conv2D) [0][0]	(None,	14, 14, 672)	75264	block5b_add
block5c_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 672)	2688	block5c_expand_
block5c_expand_activation (Acti bn[0][0]	(None,	14, 14, 672)	0	block5c_expand_
block5c_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 672)	16800	block5c_expand_
block5c_bn (BatchNormalization) [0][0]	(None,	14, 14, 672)	2688	block5c_dwconv

block5c_activation (Activation) [0][0]	(None,	14, 14, 672)	0	block5c_bn
block5c_se_squeeze (GlobalAvera ion[0][0]	(None,	672)	0	block5c_activat
block5c_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 672)	0	block5c_se_sque
block5c_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 28)	18844	block5c_se_resh
block5c_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 672)	19488	block5c_se_redu
block5c_se_excite (Multiply) ion[0][0]	(None,	14, 14, 672)	0	block5c_activat
nd[0][0]				prockac_se_expa
block5c_project_conv (Conv2D) te[0][0]	(None,	14, 14, 112)	75264	block5c_se_exci
block5c_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 112)	448	block5c_project
block5c_drop (Dropout) _bn[0][0]	(None,	14, 14, 112)	0	block5c_project
block5c_add (Add) [0][0]	(None,	14, 14, 112)	0	block5c_drop
[0][0]				block5b_add
block6a_expand_conv (Conv2D) [0][0]	(None,	14, 14, 672)	75264	block5c_add
block6a_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 672)	2688	block6a_expand_
block6a_expand_activation (Acti bn[0][0]	(None,	14, 14, 672)	0	block6a_expand_
block6a_dwconv_pad (ZeroPadding activation[0][0]	(None,	17, 17, 672)	0	block6a_expand_
block6a_dwconv (DepthwiseConv2D pad[0][0]	(None,	7, 7, 672)	16800	block6a_dwconv_

<pre>block6a_bn (BatchNormalization) [0][0]</pre>	(None,	7, 7,	672)	2688	block6a_dwconv
block6a_activation (Activation) [0][0]	(None,	7, 7,	672)	0	block6a_bn
block6a_se_squeeze (GlobalAvera ion[0][0]	(None,	672)		0	block6a_activat
block6a_se_reshape (Reshape) eze[0][0]	(None,	1, 1,	672)	0	block6a_se_sque
block6a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1,	28)	18844	block6a_se_resh
block6a_se_expand (Conv2D) ce[0][0]	(None,	1, 1,	672)	19488	block6a_se_redu
block6a_se_excite (Multiply) ion[0][0]	(None,	7, 7,	672)	0	block6a_activat
nd[0][0]					block6a_se_expa
block6a_project_conv (Conv2D) te[0][0]	(None,	7, 7,	192)	129024	block6a_se_exci
block6a_project_bn (BatchNormal _conv[0][0]	(None,	7, 7,	192)	768	block6a_project
block6b_expand_conv (Conv2D) _bn[0][0]	(None,	7, 7,	1152)	221184	block6a_project
block6b_expand_bn (BatchNormali conv[0][0]	(None,	7, 7,	1152)	4608	block6b_expand_
block6b_expand_activation (Acti bn[0][0]	(None,	7, 7,	1152)	0	block6b_expand_
block6b_dwconv (DepthwiseConv2D activation[0][0]	(None,	7, 7,	1152)	28800	block6b_expand_
block6b_bn (BatchNormalization) [0][0]	(None,	7, 7,	1152)	4608	block6b_dwconv
block6b_activation (Activation) [0][0]	(None,	7, 7,	1152)	0	block6b_bn
block6b_se_squeeze (GlobalAvera ion[0][0]	(None,	1152)		0	block6b_activat

block6b_se_reshape (Reshape) eze[0][0]	(None,	1,	1,	1152)	0	block6b_se_sque
block6b_se_reduce (Conv2D) ape[0][0]	(None,	1,	1,	48)	55344	block6b_se_resh
block6b_se_expand (Conv2D) ce[0][0]	(None,	1,	1,	1152)	56448	block6b_se_redu
block6b_se_excite (Multiply) ion[0][0]	(None,	7,	7,	1152)	0	block6b_activat
nd[0][0]						DIOCKOD_Se_expa
block6b_project_conv (Conv2D) te[0][0]	(None,	7,	7,	192)	221184	block6b_se_exci
block6b_project_bn (BatchNormal _conv[0][0]	(None,	7,	7,	192)	768	block6b_project
block6b_drop (Dropout) _bn[0][0]	(None,	7,	7,	192)	0	block6b_project
block6b_add (Add) [0][0]	(None,	7,	7,	192)	0	block6b_drop
_bn[0][0]						block6a_project
block6c_expand_conv (Conv2D) [0][0]	(None,	7,	7,	1152)	221184	block6b_add
block6c_expand_bn (BatchNormali conv[0][0]	(None,	7,	7,	1152)	4608	block6c_expand_
block6c_expand_activation (Actibn[0][0]	(None,	7,	7,	1152)	0	block6c_expand_
block6c_dwconv (DepthwiseConv2D activation[0][0]	(None,	7,	7,	1152)	28800	block6c_expand_
block6c_bn (BatchNormalization) [0][0]	(None,	7,	7,	1152)	4608	block6c_dwconv
block6c_activation (Activation) [0][0]	(None,	7,	7,	1152)	0	block6c_bn
block6c_se_squeeze (GlobalAvera ion[0][0]	(None,	11	52)		0	block6c_activat

block6c_se_reshape (Reshape) eze[0][0]	(None,	1,	1,	1152)	0	block6c_se_sque
block6c_se_reduce (Conv2D) ape[0][0]	(None,	1,	1,	48)	55344	block6c_se_resh
block6c_se_expand (Conv2D) ce[0][0]	(None,	1,	1,	1152)	56448	block6c_se_redu
block6c_se_excite (Multiply) ion[0][0]	(None,	7,	7,	1152)	0	block6c_activat
nd[0][0]						DIOCKUC_SE_EXPA
block6c_project_conv (Conv2D) te[0][0]	(None,	7,	7,	192)	221184	block6c_se_exci
block6c_project_bn (BatchNormal _conv[0][0]	(None,	7,	7,	192)	768	block6c_project
block6c_drop (Dropout) _bn[0][0]	(None,	7,	7,	192)	0	block6c_project
block6c_add (Add) [0][0]	(None,	7,	7,	192)	0	block6c_drop
[0][0]						block6b_add
block6d_expand_conv (Conv2D) [0][0]	(None,	7,	7,	1152)	221184	block6c_add
block6d_expand_bn (BatchNormali conv[0][0]	(None,	7,	7,	1152)	4608	block6d_expand_
block6d_expand_activation (Actibn[0][0]	(None,	7,	7,	1152)	0	block6d_expand_
block6d_dwconv (DepthwiseConv2D activation[0][0]	(None,	7,	7,	1152)	28800	block6d_expand_
block6d_bn (BatchNormalization) [0][0]	(None,	7,	7,	1152)	4608	block6d_dwconv
block6d_activation (Activation) [0][0]	(None,	7,	7,	1152)	0	block6d_bn
block6d_se_squeeze (GlobalAvera ion[0][0]	(None,	11	52)		0	block6d_activat

block6d_se_reshape (Reshape) eze[0][0]	(None,	1,	1,	1152)	0	block6d_se_sque
block6d_se_reduce (Conv2D) ape[0][0]	(None,	1,	1,	48)	55344	block6d_se_resh
block6d_se_expand (Conv2D) ce[0][0]	(None,	1,	1,	1152)	56448	block6d_se_redu
block6d_se_excite (Multiply) ion[0][0]	(None,	7,	7,	1152)	0	block6d_activat block6d_se_expa
nd[0][0]						
block6d_project_conv (Conv2D) te[0][0]	(None,	7,	7,	192)	221184	block6d_se_exci
block6d_project_bn (BatchNormal _conv[0][0]	(None,	7,	7,	192)	768	block6d_project
block6d_drop (Dropout) _bn[0][0]	(None,	7,	7,	192)	0	block6d_project
block6d_add (Add) [0][0]	(None,	7,	7,	192)	0	block6d_drop
[0][0]						block6c_add
block7a_expand_conv (Conv2D) [0][0]	(None,	7,	7,	1152)	221184	block6d_add
block7a_expand_bn (BatchNormali conv[0][0]	(None,	7,	7,	1152)	4608	block7a_expand_
block7a_expand_activation (Acti bn[0][0]	(None,	7,	7,	1152)	0	block7a_expand_
block7a_dwconv (DepthwiseConv2D activation[0][0]	(None,	7,	7,	1152)	10368	block7a_expand_
block7a_bn (BatchNormalization) [0][0]	(None,	7,	7,	1152)	4608	block7a_dwconv
block7a_activation (Activation) [0][0]	(None,	7,	7,	1152)	0	block7a_bn
block7a_se_squeeze (GlobalAvera ion[0][0]	(None,	11	52)		0	block7a_activat

block7a_se_reshape (Reshape) eze[0][0]	(None,	1,	1,	1152)	0	block7a_se_sque
block7a_se_reduce (Conv2D) ape[0][0]	(None,	1,	1,	48)	55344	block7a_se_resh
block7a_se_expand (Conv2D) ce[0][0]	(None,	1,	1,	1152)	56448	block7a_se_redu
block7a_se_excite (Multiply) ion[0][0]	(None,	7,	7,	1152)	0	block7a_activat
nd[0][0]						block7a_se_expa
block7a_project_conv (Conv2D) te[0][0]	(None,	7,	7,	320)	368640	block7a_se_exci
block7a_project_bn (BatchNormal _conv[0][0]	(None,	7,	7,	320)	1280	block7a_project
top_conv (Conv2D) _bn[0][0]	(None,	7,	7,	1280)	409600	block7a_project
top_bn (BatchNormalization)	(None,	7,	7,	1280)	5120	top_conv[0][0]
top_activation (Activation)	(None,	7,	7,	1280)	0	top_bn[0][0]
avg_pool (GlobalAveragePooling2 [0][0]	(None,	128	30)		0	top_activation
top_dropout (Dropout)	(None,	128	30)		0	avg_pool[0][0]
predictions (Dense) [0][0]	(None,	100	00)		1281000	top_dropout
Total params: 5,330,571 Trainable params: 5,288,548 Non-trainable params: 42,023						
Model: "functional_1"						
Layer (type)	Output	Sha	ape		Param #	Connected to
input_1 (InputLayer)	[(None	, 22	24,	224, 3)	0	<b>_</b> _
rescaling (Rescaling)	(None,	224	4,	224, 3)	0	input_1[0][0]

normalization (Normalization)	(None,	224,	224,	3)	7	rescaling[0][0]
stem_conv_pad (ZeroPadding2D) [0][0]	(None,	225,	225,	3)	0	normalization
stem_conv (Conv2D) [0][0]	(None,	112,	112,	32)	864	stem_conv_pad
stem_bn (BatchNormalization)	(None,	112,	112,	32)	128	stem_conv[0][0]
stem_activation (Activation)	(None,	112,	112,	32)	0	stem_bn[0][0]
block1a_dwconv (DepthwiseConv2D [0][0]	(None,	112,	112,	32)	288	stem_activation
block1a_bn (BatchNormalization) [0][0]	(None,	112,	112,	32)	128	block1a_dwconv
block1a_activation (Activation) [0][0]	(None,	112,	112,	32)	0	block1a_bn
block1a_se_squeeze (GlobalAvera ion[0][0]	(None,	32)			0	blockla_activat
block1a_se_reshape (Reshape) eze[0][0]	(None,	1, 1,	, 32)		0	block1a_se_sque
block1a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1,	, 8)		264	block1a_se_resh
block1a_se_expand (Conv2D) ce[0][0]	(None,	1, 1,	, 32)		288	block1a_se_redu
block1a_se_excite (Multiply) ion[0][0]	(None,	112,	112,	32)	0	block1a_activat
nd[0][0]						block1a_se_expa
block1a_project_conv (Conv2D) te[0][0]	(None,	112,	112,	16)	512	blockla_se_exci
block1a_project_bn (BatchNormal _conv[0][0]	(None,	112,	112,	16)	64	block1a_project
block2a_expand_conv (Conv2D) _bn[0][0]	(None,	112,	112,	96)	1536	block1a_project

block2a_expand_bn (BatchNormali conv[0][0]	(None,	112, 112, 96)	384	block2a_expand_
block2a_expand_activation (Acti bn[0][0]	(None,	112, 112, 96)	0	block2a_expand_
block2a_dwconv_pad (ZeroPadding activation[0][0]	(None,	113, 113, 96)	0	block2a_expand_
block2a_dwconv (DepthwiseConv2D pad[0][0]	(None,	56, 56, 96)	864	block2a_dwconv_
block2a_bn (BatchNormalization) [0][0]	(None,	56, 56, 96)	384	block2a_dwconv
block2a_activation (Activation) [0][0]	(None,	56, 56, 96)	0	block2a_bn
block2a_se_squeeze (GlobalAvera ion[0][0]	(None,	96)	0	block2a_activat
block2a_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 96)	0	block2a_se_sque
block2a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 4)	388	block2a_se_resh
block2a_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 96)	480	block2a_se_redu
block2a_se_excite (Multiply) ion[0][0]	(None,	56, 56, 96)	0	block2a_activat
nd[0][0]  block2a_project_conv (Conv2D) te[0][0]	(None,	56, 56, 24)	2304	block2a_se_exci
block2a_project_bn (BatchNormal _conv[0][0]	(None,	56, 56, 24)	96	block2a_project
block2b_expand_conv (Conv2D) _bn[0][0]	(None,	56, 56, 144)	3456	block2a_project
block2b_expand_bn (BatchNormali conv[0][0]	(None,	56, 56, 144)	576	block2b_expand_
block2b_expand_activation (Acti	(None,	56, 56, 144)	0	block2b_expand_

bn	0	]	[	0	]

block2b_dwconv (DepthwiseConv2D activation[0][0]	(None,	56, 56, 144)	1296	block2b_expand_
block2b_bn (BatchNormalization) [0][0]	(None,	56, 56, 144)	576	block2b_dwconv
block2b_activation (Activation) [0][0]	(None,	56, 56, 144)	0	block2b_bn
block2b_se_squeeze (GlobalAvera ion[0][0]	(None,	144)	0	block2b_activat
block2b_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 144)	0	block2b_se_sque
block2b_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 6)	870	block2b_se_resh
block2b_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 144)	1008	block2b_se_redu
block2b_se_excite (Multiply) ion[0][0] nd[0][0]	(None,	56, 56, 144)	0	block2b_activat block2b_se_expa
block2b_project_conv (Conv2D) te[0][0]	(None,	56, 56, 24)	3456	block2b_se_exci
block2b_project_bn (BatchNormal _conv[0][0]	(None,	56, 56, 24)	96	block2b_project
block2b_drop (Dropout) _bn[0][0]	(None,	56, 56, 24)	0	block2b_project
block2b_add (Add) [0][0] _bn[0][0]	(None,	56, 56, 24)	0	block2b_drop block2a_project
block3a_expand_conv (Conv2D) [0][0]	(None,	56, 56, 144)	3456	block2b_add
block3a_expand_bn (BatchNormali conv[0][0]	(None,	56, 56, 144)	576	block3a_expand_
block3a_expand_activation (Acti	(None,	56, 56, 144)	0	block3a_expand_

bn	0	]	[	0	]

block3a_dwconv_pad (ZeroPadding activation[0][0]	(None,	59, 59, 144)	0	block3a_expand_
block3a_dwconv (DepthwiseConv2D pad[0][0]	(None,	28, 28, 144)	3600	block3a_dwconv_
block3a_bn (BatchNormalization) [0][0]	(None,	28, 28, 144)	576	block3a_dwconv
block3a_activation (Activation) [0][0]	(None,	28, 28, 144)	0	block3a_bn
block3a_se_squeeze (GlobalAvera ion[0][0]	(None,	144)	0	block3a_activat
block3a_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 144)	0	block3a_se_sque
block3a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 6)	870	block3a_se_resh
block3a_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 144)	1008	block3a_se_redu
block3a_se_excite (Multiply) ion[0][0] nd[0][0]	(None,	28, 28, 144)	0	block3a_activat block3a_se_expa
block3a_project_conv (Conv2D) te[0][0]	(None,	28, 28, 40)	5760	block3a_se_exci
block3a_project_bn (BatchNormal _conv[0][0]	(None,	28, 28, 40)	160	block3a_project
block3b_expand_conv (Conv2D) _bn[0][0]	(None,	28, 28, 240)	9600	block3a_project
block3b_expand_bn (BatchNormali conv[0][0]	(None,	28, 28, 240)	960	block3b_expand_
block3b_expand_activation (Acti bn[0][0]	(None,	28, 28, 240)	0	block3b_expand_
block3b_dwconv (DepthwiseConv2D activation[0][0]	(None,	28, 28, 240)	6000	block3b_expand_

_bn
_activat
se_sque
_se_resh
_se_redu
_activat
_se_expa
_se_exci
_project
_project
_drop
_project
_add
_expand_
_expand_
expand_

block4a_dwconv (DepthwiseConv2D pad[0][0]	(None,	14, 14, 240)	2160	block4a_dwconv_
block4a_bn (BatchNormalization) [0][0]	(None,	14, 14, 240)	960	block4a_dwconv
block4a_activation (Activation) [0][0]	(None,	14, 14, 240)	0	block4a_bn
block4a_se_squeeze (GlobalAvera ion[0][0]	(None,	240)	0	block4a_activat
block4a_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 240)	0	block4a_se_sque
block4a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 10)	2410	block4a_se_resh
block4a_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 240)	2640	block4a_se_redu
block4a_se_excite (Multiply) ion[0][0] nd[0][0]	(None,	14, 14, 240)	0	block4a_activat block4a_se_expa
block4a_project_conv (Conv2D) te[0][0]	(None,	14, 14, 80)	19200	block4a_se_exci
block4a_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 80)	320	block4a_project
block4b_expand_conv (Conv2D) _bn[0][0]	(None,	14, 14, 480)	38400	block4a_project
block4b_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 480)	1920	block4b_expand_
block4b_expand_activation (Acti bn[0][0]	(None,	14, 14, 480)	0	block4b_expand_
block4b_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 480)	4320	block4b_expand_
block4b_bn (BatchNormalization) [0][0]	(None,	14, 14, 480)	1920	block4b_dwconv
block4b_activation (Activation)	(None,	14, 14, 480)	0	block4b_bn

[0][0]				
block4b_se_squeeze (GlobalAvera ion[0][0]	(None,	480)	0	block4b_activat
block4b_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 480)	0	block4b_se_sque
block4b_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 20)	9620	block4b_se_resh
block4b_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 480)	10080	block4b_se_redu
block4b_se_excite (Multiply) ion[0][0]  nd[0][0]	(None,	14, 14, 480)	0	block4b_activat
block4b_project_conv (Conv2D) te[0][0]	(None,	14, 14, 80)	38400	block4b_se_exci
block4b_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 80)	320	block4b_project
block4b_drop (Dropout) _bn[0][0]	(None,	14, 14, 80)	0	block4b_project
block4b_add (Add) [0][0] _bn[0][0]	(None,	14, 14, 80)	0	block4b_drop block4a_project
block4c_expand_conv (Conv2D) [0][0]	(None,	14, 14, 480)	38400	block4b_add
block4c_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 480)	1920	block4c_expand_
block4c_expand_activation (Acti bn[0][0]	(None,	14, 14, 480)	0	block4c_expand_
block4c_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 480)	4320	block4c_expand_
block4c_bn (BatchNormalization) [0][0]	(None,	14, 14, 480)	1920	block4c_dwconv
block4c_activation (Activation)	(None,	14, 14, 480)	0	block4c_bn

[0] [0]				
block4c_se_squeeze (GlobalAvera ion[0][0]	(None,	480)	0	block4c_activat
block4c_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 480)	0	block4c_se_sque
block4c_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 20)	9620	block4c_se_resh
block4c_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 480)	10080	block4c_se_redu
ion[0][0]	(None,	14, 14, 480)	0	block4c_activat
nd[0][0]				
block4c_project_conv (Conv2D) te[0][0]	(None,	14, 14, 80)	38400	block4c_se_exci
block4c_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 80)	320	block4c_project
block4c_drop (Dropout) _bn[0][0]	(None,	14, 14, 80)	0	block4c_project
block4c_add (Add) [0][0]	(None,	14, 14, 80)	0	block4c_drop block4b add
[0][0]				2100/112_444
block5a_expand_conv (Conv2D) [0][0]	(None,	14, 14, 480)	38400	block4c_add
block5a_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 480)	1920	block5a_expand_
block5a_expand_activation (Actibn[0][0]	(None,	14, 14, 480)	0	block5a_expand_
block5a_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 480)	12000	block5a_expand_
block5a_bn (BatchNormalization) [0][0]	(None,	14, 14, 480)	1920	block5a_dwconv
block5a_activation (Activation)	(None,	14, 14, 480)	0	block5a_bn

[0][0]				
block5a_se_squeeze (GlobalAvera ion[0][0]	(None,	480)	0	block5a_activat
block5a_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 480)	0	block5a_se_sque
block5a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 20)	9620	block5a_se_resh
block5a_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 480)	10080	block5a_se_redu
block5a_se_excite (Multiply) ion[0][0] nd[0][0]	(None,	14, 14, 480)	0	block5a_activat
block5a_project_conv (Conv2D) te[0][0]	(None,	14, 14, 112)	53760	block5a_se_exci
block5a_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 112)	448	block5a_project
block5b_expand_conv (Conv2D) _bn[0][0]	(None,	14, 14, 672)	75264	block5a_project
block5b_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 672)	2688	block5b_expand_
block5b_expand_activation (Acti bn[0][0]	(None,	14, 14, 672)	0	block5b_expand_
block5b_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 672)	16800	block5b_expand_
block5b_bn (BatchNormalization) [0][0]	(None,	14, 14, 672)	2688	block5b_dwconv
block5b_activation (Activation) [0][0]	(None,	14, 14, 672)	0	block5b_bn
block5b_se_squeeze (GlobalAvera ion[0][0]	(None,	672)	0	block5b_activat
block5b_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 672)	0	block5b_se_sque

block5b_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 28)	18844	block5b_se_resh
block5b_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 672)	19488	block5b_se_redu
block5b_se_excite (Multiply) ion[0][0]	(None,	14, 14, 672)	0	block5b_activat block5b_se_expa
nd[0][0]  block5b_project_conv (Conv2D)  te[0][0]	(None,	14, 14, 112)	75264	block5b_se_exci
block5b_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 112)	448	block5b_project
block5b_drop (Dropout) _bn[0][0]	(None,	14, 14, 112)	0	block5b_project
block5b_add (Add) [0][0]	(None,	14, 14, 112)	0	block5b_drop block5a_project
_bn[0][0]				
block5c_expand_conv (Conv2D) [0][0]	(None,	14, 14, 672)	75264	block5b_add
block5c_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 672)	2688	block5c_expand_
block5c_expand_activation (Acti bn[0][0]	(None,	14, 14, 672)	0	block5c_expand_
block5c_dwconv (DepthwiseConv2D activation[0][0]	(None,	14, 14, 672)	16800	block5c_expand_
block5c_bn (BatchNormalization) [0][0]	(None,	14, 14, 672)	2688	block5c_dwconv
block5c_activation (Activation) [0][0]	(None,	14, 14, 672)	0	block5c_bn
block5c_se_squeeze (GlobalAvera ion[0][0]	(None,	672)	0	block5c_activat
block5c_se_reshape (Reshape) eze[0][0]	(None,	1, 1, 672)	0	block5c_se_sque

block5c_se_reduce (Conv2D) ape[0][0]	(None,	1, 1, 28)	18844	block5c_se_resh
block5c_se_expand (Conv2D) ce[0][0]	(None,	1, 1, 672)	19488	block5c_se_redu
block5c_se_excite (Multiply) ion[0][0]	(None,	14, 14, 672)	0	block5c_activat
nd[0][0]				
block5c_project_conv (Conv2D) te[0][0]	(None,	14, 14, 112)	75264	block5c_se_exci
block5c_project_bn (BatchNormal _conv[0][0]	(None,	14, 14, 112)	448	block5c_project
block5c_drop (Dropout) _bn[0][0]	(None,	14, 14, 112)	0	block5c_project
block5c_add (Add) [0][0]	(None,	14, 14, 112)	0	block5c_drop
[0][0]				block5b_add
block6a_expand_conv (Conv2D) [0][0]	(None,	14, 14, 672)	75264	block5c_add
block6a_expand_bn (BatchNormali conv[0][0]	(None,	14, 14, 672)	2688	block6a_expand_
block6a_expand_activation (Acti bn[0][0]	(None,	14, 14, 672)	0	block6a_expand_
block6a_dwconv_pad (ZeroPadding activation[0][0]	(None,	17, 17, 672)	0	block6a_expand_
block6a_dwconv (DepthwiseConv2D pad[0][0]	(None,	7, 7, 672)	16800	block6a_dwconv_
block6a_bn (BatchNormalization) [0][0]	(None,	7, 7, 672)	2688	block6a_dwconv
block6a_activation (Activation) [0][0]	(None,	7, 7, 672)	0	block6a_bn
block6a_se_squeeze (GlobalAvera ion[0][0]	(None,	672)	0	block6a_activat

block6a_se_reshape (Reshape) eze[0][0]	(None,	1, 1	l <b>,</b> 672	2) 0	block6a_se_sque
block6a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1	1, 28)	18844	block6a_se_resh
block6a_se_expand (Conv2D) ce[0][0]	(None,	1, 1	1, 672	2) 19488	block6a_se_redu
block6a_se_excite (Multiply) ion[0][0]	(None,	7, 7	7, 672	2) 0	block6a_activat
nd[0][0]					
block6a_project_conv (Conv2D) te[0][0]	(None,	7, 7	7, 192	2) 129024	block6a_se_exci
block6a_project_bn (BatchNormal _conv[0][0]	(None,	7, 7	7, 192	2) 768	block6a_project
block6b_expand_conv (Conv2D) _bn[0][0]	(None,	7, 7	7, 115	52) 221184	block6a_project
block6b_expand_bn (BatchNormali conv[0][0]	(None,	7, 7	7, 115	52) 4608	block6b_expand_
block6b_expand_activation (Acti bn[0][0]	(None,	7, 7	7, 115	52) 0	block6b_expand_
block6b_dwconv (DepthwiseConv2D activation[0][0]	(None,	7, 7	7, 115	52) 28800	block6b_expand_
block6b_bn (BatchNormalization) [0][0]	(None,	7, 7	7, 115	52) 4608	block6b_dwconv
block6b_activation (Activation) [0][0]	(None,	7, 7	7, 115	52) 0	block6b_bn
block6b_se_squeeze (GlobalAvera ion[0][0]	(None,	1152	2)	0	block6b_activat
block6b_se_reshape (Reshape) eze[0][0]	(None,	1, 1	1, 115	52) 0	block6b_se_sque
block6b_se_reduce (Conv2D) ape[0][0]	(None,	1, 1	L, 48)	55344	block6b_se_resh
block6b_se_expand (Conv2D)	(None,	1, 1	1, 115	52) 56448	block6b_se_redu

ce[0][0]						
block6b_se_excite (Multiply) ion[0][0]	(None,	7,	7,	1152)	0	block6b_activat
nd[0][0]						block6b_se_expa
block6b_project_conv (Conv2D) te[0][0]	(None,	7,	7,	192)	221184	block6b_se_exci
block6b_project_bn (BatchNormal _conv[0][0]	(None,	7,	7,	192)	768	block6b_project
block6b_drop (Dropout) _bn[0][0]	(None,	7,	7,	192)	0	block6b_project
block6b_add (Add) [0][0]	(None,	7,	7,	192)	0	block6b_drop
_bn[0][0]						block6a_project
block6c_expand_conv (Conv2D) [0][0]	(None,	7,	7,	1152)	221184	block6b_add
block6c_expand_bn (BatchNormali conv[0][0]	(None,	7,	7,	1152)	4608	block6c_expand_
block6c_expand_activation (Acti bn[0][0]	(None,	7,	7,	1152)	0	block6c_expand_
block6c_dwconv (DepthwiseConv2D activation[0][0]	(None,	7,	7,	1152)	28800	block6c_expand_
block6c_bn (BatchNormalization) [0][0]	(None,	7,	7,	1152)	4608	block6c_dwconv
block6c_activation (Activation) [0][0]	(None,	7,	7,	1152)	0	block6c_bn
block6c_se_squeeze (GlobalAvera ion[0][0]	(None,	11	52)		0	block6c_activat
block6c_se_reshape (Reshape) eze[0][0]	(None,	1,	1,	1152)	0	block6c_se_sque
block6c_se_reduce (Conv2D) ape[0][0]	(None,	1,	1,	48)	55344	block6c_se_resh
block6c_se_expand (Conv2D)	(None,	1,	1,	1152)	56448	block6c_se_redu

ce[0][0]						
block6c_se_excite (Multiply) ion[0][0]	(None,	7,	7,	1152)	0	block6c_activat
nd[0][0]						block6c_se_expa
block6c_project_conv (Conv2D) te[0][0]	(None,	7,	7,	192)	221184	block6c_se_exci
block6c_project_bn (BatchNormal _conv[0][0]	(None,	7,	7,	192)	768	block6c_project
block6c_drop (Dropout) _bn[0][0]	(None,	7,	7,	192)	0	block6c_project
block6c_add (Add) [0][0]	(None,	7,	7,	192)	0	block6c_drop
[0][0]						block6b_add
block6d_expand_conv (Conv2D) [0][0]	(None,	7,	7,	1152)	221184	block6c_add
block6d_expand_bn (BatchNormali conv[0][0]	(None,	7,	7,	1152)	4608	block6d_expand_
block6d_expand_activation (Acti bn[0][0]	(None,	7,	7,	1152)	0	block6d_expand_
block6d_dwconv (DepthwiseConv2D activation[0][0]	(None,	7,	7,	1152)	28800	block6d_expand_
block6d_bn (BatchNormalization) [0][0]	(None,	7,	7,	1152)	4608	block6d_dwconv
block6d_activation (Activation) [0][0]	(None,	7,	7,	1152)	0	block6d_bn
block6d_se_squeeze (GlobalAvera ion[0][0]	(None,	11	52)		0	block6d_activat
block6d_se_reshape (Reshape) eze[0][0]	(None,	1,	1,	1152)	0	block6d_se_sque
block6d_se_reduce (Conv2D) ape[0][0]	(None,	1,	1,	48)	55344	block6d_se_resh
block6d_se_expand (Conv2D)	(None,	1,	1,	1152)	56448	block6d_se_redu

ce[0][0]					
block6d_se_excite (Multiply) ion[0][0]	(None,	7, 7	, 1152)	0	block6d_activat
nd[0][0]					block6d_se_expa
block6d_project_conv (Conv2D) te[0][0]	(None,	7, 7,	, 192)	221184	block6d_se_exci
block6d_project_bn (BatchNormal _conv[0][0]	(None,	7, 7,	, 192)	768	block6d_project
block6d_drop (Dropout) _bn[0][0]	(None,	7, 7,	, 192)	0	block6d_project
block6d_add (Add) [0][0]	(None,	7, 7,	, 192)	0	block6d_drop
[0][0]					block6c_add
block7a_expand_conv (Conv2D) [0][0]	(None,	7, 7,	, 1152)	221184	block6d_add
block7a_expand_bn (BatchNormali conv[0][0]	(None,	7, 7,	, 1152)	4608	block7a_expand_
block7a_expand_activation (Acti bn[0][0]	(None,	7, 7,	, 1152)	0	block7a_expand_
block7a_dwconv (DepthwiseConv2D activation[0][0]	(None,	7, 7,	, 1152)	10368	block7a_expand_
block7a_bn (BatchNormalization) [0][0]	(None,	7, 7,	, 1152)	4608	block7a_dwconv
block7a_activation (Activation) [0][0]	(None,	7, 7,	, 1152)	0	block7a_bn
block7a_se_squeeze (GlobalAvera ion[0][0]	(None,	1152)	)	0	block7a_activat
block7a_se_reshape (Reshape) eze[0][0]	(None,	1, 1,	, 1152)	0	block7a_se_sque
block7a_se_reduce (Conv2D) ape[0][0]	(None,	1, 1,	, 48)	55344	block7a_se_resh
block7a_se_expand (Conv2D)	(None,	1, 1,	, 1152)	56448	block7a_se_redu

ce[0][0]			
block7a_se_excite (Multiply) ion[0][0] nd[0][0]	(None, 7, 7, 1152)	0	block7a_activat block7a_se_expa
block7a_project_conv (Conv2D) te[0][0]	(None, 7, 7, 320)	368640	block7a_se_exci
block7a_project_bn (BatchNormal _conv[0][0]	(None, 7, 7, 320)	1280	block7a_project
top_conv (Conv2D) _bn[0][0]	(None, 7, 7, 1280)	409600	block7a_project
top_bn (BatchNormalization)	(None, 7, 7, 1280)	5120	top_conv[0][0]
top_activation (Activation)	(None, 7, 7, 1280)	0	top_bn[0][0]
avg_pool (GlobalAveragePooling2	(None, 1280)	0	top_activation
<pre>for layer in custom_resnet_mode.     layer.trainable = False</pre>	l.layers[:-1]:		
<pre>custom_resnet_model.layers[-1].</pre>	trainable		
True			

```
In [12]:
```

Out[12]:

```
In [13]: custom_resnet_model.compile(loss='categorical_crossentropy',optimizer='adam',metric
         s=['accuracy'])
```

```
In [14]: t=time.time()
    hist = custom_resnet_model.fit(X_train, y_train, batch_size=32, epochs=num_epoch, v
    erbose=1, validation_data=(X_test, y_test))
    print('Training time: %s' % (t - time.time()))
    (loss, accuracy) = custom_resnet_model.evaluate(X_test, y_test, batch_size=10, verb
    ose=1)
    print("[INFO] loss={:.4f}, accuracy: {:.4f}%".format(loss,accuracy * 100))
```

```
Epoch 1/100
cy: 0.8845 - val loss: 0.1604 - val accuracy: 0.9403
Epoch 2/100
147/147 [============= ] - 8s 56ms/step - loss: 0.1834 - accurac
y: 0.9313 - val loss: 0.1341 - val accuracy: 0.9497
Epoch 3/100
147/147 [============= ] - 8s 57ms/step - loss: 0.1541 - accurac
y: 0.9443 - val_loss: 0.1338 - val_accuracy: 0.9514
Epoch 4/100
y: 0.9498 - val_loss: 0.1146 - val_accuracy: 0.9548
Epoch 5/100
y: 0.9522 - val loss: 0.1142 - val accuracy: 0.9590
Epoch 6/100
y: 0.9537 - val_loss: 0.1158 - val_accuracy: 0.9608
Epoch 7/100
y: 0.9573 - val_loss: 0.1075 - val_accuracy: 0.9608
Epoch 8/100
y: 0.9586 - val loss: 0.1077 - val accuracy: 0.9616
Epoch 9/100
y: 0.9571 - val loss: 0.1019 - val_accuracy: 0.9616
Epoch 10/100
y: 0.9582 - val_loss: 0.1016 - val_accuracy: 0.9616
Epoch 11/100
147/147 [============== ] - 8s 57ms/step - loss: 0.1082 - accurac
y: 0.9596 - val loss: 0.1034 - val accuracy: 0.9625
Epoch 12/100
y: 0.9643 - val loss: 0.0983 - val accuracy: 0.9642
Epoch 13/100
y: 0.9607 - val_loss: 0.1088 - val_accuracy: 0.9573
Epoch 14/100
y: 0.9637 - val loss: 0.1113 - val accuracy: 0.9582
Epoch 15/100
y: 0.9667 - val loss: 0.0955 - val accuracy: 0.9650
Epoch 16/100
y: 0.9654 - val loss: 0.0941 - val accuracy: 0.96670.0966 - accura
Epoch 17/100
y: 0.9654 - val_loss: 0.0990 - val_accuracy: 0.9616
Epoch 18/100
y: 0.9661 - val_loss: 0.0937 - val_accuracy: 0.9659
Epoch 19/100
y: 0.9686 - val loss: 0.0941 - val accuracy: 0.9684
Epoch 20/100
y: 0.9637 - val loss: 0.0929 - val accuracy: 0.9667
Epoch 21/100
y: 0.9641 - val_loss: 0.0918 - val_accuracy: 0.9684
```

```
Epoch 22/100
y: 0.9690 - val loss: 0.0917 - val accuracy: 0.9684
Epoch 23/100
y: 0.9673 - val loss: 0.0922 - val accuracy: 0.9676
Epoch 24/100
147/147 [============] - 8s 57ms/step - loss: 0.0893 - accurac
y: 0.9673 - val loss: 0.0908 - val accuracy: 0.9659
Epoch 25/100
y: 0.9684 - val_loss: 0.0911 - val_accuracy: 0.9676
Epoch 26/100
y: 0.9675 - val_loss: 0.0901 - val_accuracy: 0.9659
Epoch 27/100
y: 0.9695 - val_loss: 0.0925 - val_accuracy: 0.9676
Epoch 28/100
y: 0.9695 - val loss: 0.0898 - val accuracy: 0.9701
Epoch 29/100
y: 0.9688 - val loss: 0.0890 - val accuracy: 0.9650
Epoch 30/100
y: 0.9725 - val loss: 0.0886 - val accuracy: 0.9684
Epoch 31/100
y: 0.9684 - val loss: 0.0878 - val accuracy: 0.9676
Epoch 32/100
147/147 [============= ] - 8s 57ms/step - loss: 0.0806 - accurac
y: 0.9729 - val_loss: 0.0886 - val_accuracy: 0.9667
Epoch 33/100
y: 0.9699 - val loss: 0.0874 - val accuracy: 0.9676
Epoch 34/100
147/147 [============== ] - 8s 57ms/step - loss: 0.0794 - accurac
y: 0.9686 - val loss: 0.0884 - val accuracy: 0.9667
y: 0.9701 - val loss: 0.0955 - val accuracy: 0.9667
Epoch 36/100
y: 0.9729 - val_loss: 0.0895 - val_accuracy: 0.9684
Epoch 37/100
y: 0.9688 - val loss: 0.0881 - val accuracy: 0.9693
Epoch 38/100
y: 0.9720 - val loss: 0.0945 - val accuracy: 0.9676
Epoch 39/100
y: 0.9725 - val loss: 0.0891 - val accuracy: 0.9693
Epoch 40/100
147/147 [============= ] - 8s 57ms/step - loss: 0.0775 - accurac
y: 0.9708 - val loss: 0.0882 - val accuracy: 0.9684
Epoch 41/100
y: 0.9710 - val loss: 0.0867 - val accuracy: 0.9667
y: 0.9752 - val loss: 0.0855 - val accuracy: 0.9676
Epoch 43/100
```

```
y: 0.9714 - val loss: 0.0879 - val accuracy: 0.9693
Epoch 44/100
y: 0.9710 - val loss: 0.0871 - val accuracy: 0.9667
147/147 [============== ] - 8s 57ms/step - loss: 0.0710 - accurac
y: 0.9761 - val loss: 0.0897 - val accuracy: 0.9710
Epoch 46/100
y: 0.9725 - val_loss: 0.0878 - val_accuracy: 0.9676
Epoch 47/100
147/147 [============== ] - 8s 57ms/step - loss: 0.0690 - accurac
y: 0.9720 - val loss: 0.0878 - val accuracy: 0.9676
Epoch 48/100
y: 0.9746 - val loss: 0.0866 - val accuracy: 0.9676
Epoch 49/100
y: 0.9746 - val loss: 0.0878 - val accuracy: 0.9684
Epoch 50/100
147/147 [============== ] - 8s 57ms/step - loss: 0.0692 - accurac
y: 0.9754 - val loss: 0.0878 - val accuracy: 0.9667
Epoch 51/100
y: 0.9765 - val loss: 0.0877 - val_accuracy: 0.9684
147/147 [============== ] - 8s 57ms/step - loss: 0.0695 - accurac
y: 0.9729 - val_loss: 0.0917 - val_accuracy: 0.9684
Epoch 53/100
147/147 [============== ] - 8s 57ms/step - loss: 0.0711 - accurac
y: 0.9740 - val loss: 0.0931 - val accuracy: 0.9676
Epoch 54/100
y: 0.9735 - val loss: 0.0887 - val_accuracy: 0.9693
Epoch 55/100
y: 0.9733 - val_loss: 0.0948 - val_accuracy: 0.9667
Epoch 56/100
y: 0.9750 - val_loss: 0.0872 - val_accuracy: 0.9676
Epoch 57/100
y: 0.9740 - val_loss: 0.0922 - val_accuracy: 0.9676
Epoch 58/100
y: 0.9735 - val_loss: 0.0884 - val_accuracy: 0.9676
Epoch 59/100
y: 0.9737 - val loss: 0.0883 - val accuracy: 0.9693
Epoch 60/100
y: 0.9761 - val loss: 0.0884 - val accuracy: 0.9693
Epoch 61/100
y: 0.9750 - val loss: 0.0867 - val accuracy: 0.9693
y: 0.9750 - val loss: 0.0859 - val accuracy: 0.9693
Epoch 63/100
y: 0.9744 - val loss: 0.0912 - val accuracy: 0.9693
Epoch 64/100
```

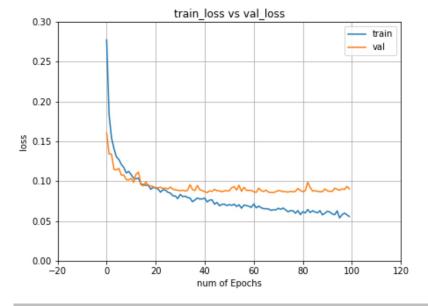
```
y: 0.9737 - val loss: 0.0882 - val accuracy: 0.9693
Epoch 65/100
y: 0.9767 - val loss: 0.0869 - val accuracy: 0.9693
Epoch 66/100
y: 0.9757 - val loss: 0.0887 - val accuracy: 0.9676
Epoch 67/100
147/147 [============] - 8s 57ms/step - loss: 0.0650 - accurac
y: 0.9761 - val loss: 0.0861 - val accuracy: 0.9693
Epoch 68/100
y: 0.9754 - val loss: 0.0858 - val accuracy: 0.9693
Epoch 69/100
y: 0.9763 - val loss: 0.0855 - val accuracy: 0.9693
Epoch 70/100
147/147 [============== ] - 8s 57ms/step - loss: 0.0638 - accurac
y: 0.9787 - val loss: 0.0867 - val accuracy: 0.9693
Epoch 71/100
y: 0.9757 - val loss: 0.0881 - val accuracy: 0.9701
147/147 [============] - 8s 57ms/step - loss: 0.0646 - accurac
y: 0.9765 - val loss: 0.0877 - val accuracy: 0.9676
Epoch 73/100
147/147 [============== ] - 8s 56ms/step - loss: 0.0662 - accurac
y: 0.9759 - val loss: 0.0867 - val accuracy: 0.9676
Epoch 74/100
y: 0.9767 - val loss: 0.0866 - val accuracy: 0.9684
Epoch 75/100
y: 0.9776 - val_loss: 0.0862 - val_accuracy: 0.9701
Epoch 76/100
y: 0.9761 - val_loss: 0.0871 - val_accuracy: 0.9693
Epoch 77/100
y: 0.9757 - val loss: 0.0865 - val accuracy: 0.9710
Epoch 78/100
y: 0.9784 - val_loss: 0.0871 - val_accuracy: 0.9693
Epoch 79/100
y: 0.9776 - val_loss: 0.0907 - val_accuracy: 0.9710
Epoch 80/100
y: 0.9784 - val loss: 0.0882 - val accuracy: 0.9684
Epoch 81/100
y: 0.9778 - val loss: 0.0867 - val_accuracy: 0.9693
Epoch 82/100
y: 0.9791 - val loss: 0.0879 - val accuracy: 0.9701
Epoch 83/100
y: 0.9746 - val loss: 0.0984 - val accuracy: 0.9693
Epoch 84/100
y: 0.9772 - val loss: 0.0921 - val accuracy: 0.9676
Epoch 85/100
y: 0.9744 - val_loss: 0.0873 - val_accuracy: 0.9701
```

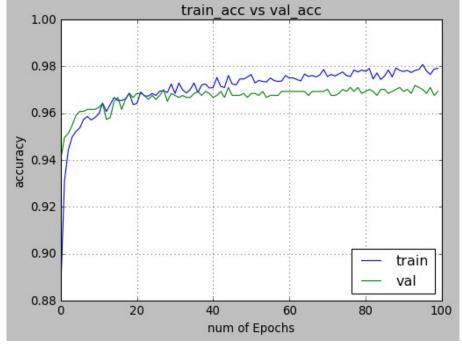
Epoch 86/100

```
y: 0.9759 - val loss: 0.0881 - val accuracy: 0.9701
     Epoch 87/100
     y: 0.9784 - val loss: 0.0871 - val accuracy: 0.9684
     Epoch 88/100
     147/147 [============] - 8s 57ms/step - loss: 0.0629 - accurac
     y: 0.9754 - val loss: 0.0866 - val accuracy: 0.9693
     y: 0.9793 - val_loss: 0.0867 - val_accuracy: 0.9701
     Epoch 90/100
     y: 0.9782 - val_loss: 0.0901 - val_accuracy: 0.9710
     Epoch 91/100
     y: 0.9778 - val_loss: 0.0878 - val_accuracy: 0.9693
     Epoch 92/100
     147/147 [=============] - 8s 57ms/step - loss: 0.0613 - accurac
     y: 0.9782 - val loss: 0.0872 - val accuracy: 0.9701
     Epoch 93/100
     147/147 [============== ] - 8s 57ms/step - loss: 0.0587 - accurac
     y: 0.9774 - val loss: 0.0872 - val accuracy: 0.9684
     Epoch 94/100
     y: 0.9782 - val loss: 0.0911 - val accuracy: 0.9718
     Epoch 95/100
     y: 0.9787 - val loss: 0.0897 - val accuracy: 0.9710
     Epoch 96/100
     147/147 [============= ] - 8s 57ms/step - loss: 0.0537 - accurac
     y: 0.9808 - val loss: 0.0885 - val accuracy: 0.9701
     Epoch 97/100
     y: 0.9780 - val_loss: 0.0902 - val_accuracy: 0.9684
     Epoch 98/100
     147/147 [============= ] - 8s 57ms/step - loss: 0.0599 - accurac
     y: 0.9765 - val loss: 0.0895 - val accuracy: 0.9710
     y: 0.9789 - val loss: 0.0933 - val accuracy: 0.9676
     Epoch 100/100
     y: 0.9791 - val_loss: 0.0901 - val_accuracy: 0.9693
     Training time: -848.8196833133698
     In [15]: (loss, accuracy) = custom resnet model.evaluate(X test, y test, batch size=10, verb
     ose=1)
     print("[INFO] loss={:.4f}, accuracy: {:.4f}%".format(loss,accuracy * 100))
     [INFO] loss=0.0901, accuracy: 96.9283%
```

visualizing losses and accuracy







## **Evaluating the model**

```
In [17]: score = custom_resnet_model.evaluate(X_test, y_test, verbose=0)
    print('Test Loss:', score[0])
    print('Test accuracy:', score[1])

test_image = X_test[0:1]
    print (test_image.shape)

print(model.predict(test_image))
    print(model.predict_classes(test_image))
    print(y_test[0:1])
```

```
Test Loss: 0.09007590264081955
Test accuracy: 0.9692832827568054
(1, 224, 224, 3)
[[1.14373943e-05 1.04924751e-04 2.51722900e-04 4.77186404e-04
  1.18551354e-04 2.09018617e-05 4.94372798e-05 3.72421637e-05
  2.81130688e-05 3.33237003e-05 5.39718130e-05 4.48332175e-05
  1.52935299e-05 2.69304637e-05 9.17184298e-06 1.39459144e-05
  3.57407189e-05 1.11406243e-05 9.04056869e-05 1.29782620e-05
  4.97367182e-05 1.52026812e-04 1.29650245e-04 9.28172449e-05
  5.12071965e-05 1.26058385e-05 2.34643376e-05 2.12379255e-05
  1.80296520e-05 4.98396868e-04 2.47400421e-05 6.25148459e-05
  1.78604005e-05 5.69082549e-05 7.87051613e-05 7.06591309e-05
  2.68311251e-05 3.09490679e-05 3.14168792e-05 2.49381701e-05
  3.04195401e-05 1.12252492e-05 7.05893126e-06 9.69405119e-06
  2.52910850e-05 3.18797393e-05 5.48759344e-05 8.84422025e-06
  5.99929308e-06 3.62554201e-05 4.74188018e-05 2.52091377e-05
  1.06253538e-04 5.40399014e-05 7.13719346e-05 5.98135775e-05
  2.53936560e-05 3.17523663e-05 3.84590785e-05 1.80896059e-05
  1.09258735e-05 6.93523816e-06 3.70208436e-05 3.30621988e-05
  3.41156119e-05 4.84199654e-06 4.28685780e-05 6.73844215e-06
  3.28146380e-05 1.46650229e-04 3.31878364e-05 2.56108760e-04
  2.61196430e-04 4.09581007e-05 7.56776790e-05 1.36237679e-04
  9.20068051e-05 2.07817684e-05 1.77334645e-04 1.37647934e-04
  4.13984781e-05 7.86930832e-05 4.24302016e-05 1.24560480e-04
  2.35329808e-05 5.86002352e-06 2.18944788e-05 8.65192269e-05
  8.45431496e-05 3.85376101e-04 9.83447353e-06 7.61144838e-05
  3.42622370e-05 7.31325781e-05 1.02619080e-04 2.22032431e-05
  5.87798240e-05 4.03036538e-05 6.36094046e-06 1.85568555e-04
  1.04721134e-04 1.46556995e-03 2.70735891e-05 2.10290254e-05
  9.79486140e-06 2.09725858e-05 4.52680506e-05 1.53035537e-04
  6.04435445e-05 2.56030016e-05 1.71550928e-05 3.08110699e-04
  1.67387456e-03 1.12096459e-04 2.55914565e-05 9.76902083e-06
  1.16005620e-04 4.60985902e-04 2.48370834e-05 3.65114938e-05
  3.35951845e-05 6.50190414e-05 1.97253947e-04 6.78287470e-05
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  1.00618592e-04 3.20101281e-05 7.87851386e-05 1.01110712e-03
  4.70257364e-05 7.69894468e-05 6.71668895e-05 1.08402739e-04
  1.48987558e-04 7.06426566e-04 4.69786319e-05 3.17751765e-05
  7.18288866e-05 7.85148150e-05 3.43337670e-05 1.47799292e-04
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  5.93844561e-05 4.66298407e-05 3.48248577e-05 1.75946974e-04
  4.10317414e-04 7.20948156e-05 1.07296561e-04 9.15680721e-05
  2.65757753e-05 6.20490537e-05 3.57686280e-04 4.92307408e-05
  7.88953694e-05 9.74631985e-05 9.77357340e-05 1.59143674e-04
  8.07626202e-05 4.58606664e-05 8.00844646e-05 2.82172841e-05
```

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3.26975896e-05 5.18276647e-05 1.59105966e-05 2.47692573e-04
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2.76584015e-05 4.85704193e-04 8.96239453e-06 8.32033838e-05
7.59714749e-05 5.65359805e-05 1.73191249e-04 6.54873802e-05
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2.23795490e-04 1.17462705e-05 8.37473490e-05 4.66217345e-04
9.20921302e-05 6.95714916e-05 5.27464772e-06 8.62578582e-03
1.39323893e-04 4.93507432e-05 1.76188332e-04 2.94449273e-02
1.55619404e-04 1.01388898e-03 1.95309069e-04 6.41724091e-06
1.14455390e-04 2.71182416e-05 6.47203298e-04 1.00095363e-04
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2.38124921e-04 6.03371416e-04 3.55180528e-05 2.16321168e-05
2.72363777e-05 8.89033035e-05 4.06742422e-03 1.29921682e-04
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```

```
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3.08887684e-05 1.84962082e-05 8.66356830e-04 1.23538586e-04
2.15153977e-05 2.23801049e-04 7.42496850e-05 6.04413508e-05
2.95608315e-05 1.18541066e-04 5.44077542e-04 4.79755603e-04
1.19070751e-06 2.86531867e-04 1.08347449e-04 1.23221835e-04
1.60122472e-05 3.52789066e-05 7.62058926e-06 1.49062726e-06
1.97045141e-04 3.07448827e-05 2.99867289e-03 1.13841954e-04
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7.74253858e-04 2.01023329e-04 1.54460280e-03 2.88035953e-04
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8.01541464e-05 2.58367072e-04 1.18391141e-02 2.76369654e-04
7.56522408e-04 1.20333594e-03 1.34738511e-04 3.40762839e-04
1.82323365e-05 7.10968015e-05 1.12378566e-05 5.10631944e-04
1.91196290e-04 1.09206350e-03 3.15800498e-05 2.18468718e-04
1.12467869e-05 4.41428600e-03 2.27257988e-04 6.61191953e-05
4.01627585e-05 1.24009266e-05 2.44076873e-05 4.03825616e-05
4.58932802e-04 1.87291380e-03 3.13152181e-04 3.02730768e-05
4.86531015e-03 1.40191099e-04 3.30081070e-03 1.58697047e-04
3.27433518e-04 4.06419967e-05 2.01998424e-04 1.32928126e-05
4.85229393e-05 1.75380628e-05 2.26789562e-04 4.04121878e-04
2.61310168e-04 4.74795932e-04 1.89746468e-04 3.73692531e-03
3.27527232e-04 9.13831318e-05 3.24898690e-04 4.24559839e-05
1.04334940e-04 2.81458633e-04 3.25866567e-04 3.70700407e-04
9.77324744e-05 4.96280300e-05 4.70058789e-04 7.70517101e-04
1.39880329e-04 3.93471419e-04 4.72605170e-05 1.75979440e-05
2.38323264e-05 3.17902995e-05 2.39583111e-04 4.03178186e-04
4.79454611e-05 7.74909800e-04 1.22122496e-04 7.42310367e-04
4.16312163e-04 4.45918304e-05 2.92662182e-04 1.42756937e-04
3.69849222e-05 1.75505877e-04 1.69684878e-04 2.01773928e-05
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9.67533968e-04 1.49089133e-03 1.81955820e-05 2.01489129e-05
6.49139620e-05 2.10260070e-04 1.31405532e-05 2.13075604e-04
4.66043202e-05 1.07829161e-04 1.37850569e-04 2.17440233e-04
9.95940281e-05 1.59908854e-03 4.40408097e-04 9.27102883e-05
1.46864229e-04 1.02436473e-03 1.35910641e-05 5.05308453e-05
9.87842668e-06 3.81041842e-04 1.07388140e-03 5.56326668e-05
8.97882041e-04 8.16292595e-05 3.35576275e-04 5.29770681e-04
2.01572650e-04 3.38251411e-05 1.80425523e-05 5.76800609e-04
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6.12997974e-05 3.24978668e-04 1.80621195e-04 3.68183901e-05
1.51478671e-04 4.82440781e-04 6.04991401e-05 7.77881432e-05
1.26551939e-04 1.84923367e-04 3.15761950e-04 5.93294317e-05
6.60373989e-05 6.65641564e-05 1.59843130e-05 1.58041785e-05
5.12444996e-04 4.13116722e-05 1.11431109e-05 1.06844418e-04
2.59616121e-04 2.77282033e-05 1.40357995e-04 1.17170248e-05
1.01555357e-04 7.50477266e-05 4.15015711e-05 6.11421565e-05
1.11586787e-03 1.75463210e-04 2.01662551e-04 2.15432628e-05
1.05642561e-04 8.10383644e-04 1.73605731e-05 7.83597934e-04
1.32822432e-03 1.24145346e-03 2.37979722e-04 1.58674651e-04
3.66873719e-05 2.97207967e-04 6.27743066e-05 5.37725755e-05
5.02291659e-05 4.35538124e-04 5.03882147e-05 5.22157527e-04
2.63669772e-05 3.11685086e-04 7.64837372e-04 1.75272813e-03
6.76614523e-04 5.19993082e-05 4.07306979e-05 1.40892621e-03
6.39285936e-05 3.50955548e-03 1.14453929e-02 2.54656974e-04
```

```
1.06379361e-04 2.97680050e-01 2.45035899e-05 3.06748225e-05
1.32625370e-04 4.29442771e-05 6.94501068e-05 1.06415697e-04
1.08547676e-04 6.69606961e-04 5.19007190e-05 1.91055569e-05
2.42599664e-04 5.64372109e-04 4.39090800e-05 1.56143127e-04
4.16793482e-05 8.86967537e-05 2.07324396e-04 3.20165709e-05
1.20084711e-04 3.51153547e-04 1.97403730e-04 2.12631872e-04
1.54889669e-04 6.05181776e-05 1.16133422e-03 5.36538319e-05
5.46015275e-04 9.81851845e-05 7.44502031e-05 4.70522762e-04
1.39306416e-04 4.22805641e-03 9.15047858e-05 2.49385266e-05
5.90733907e-05 4.27730993e-04 2.60059896e-05 5.41474583e-05
2.61146983e-04 8.86572598e-05 3.34721953e-01 3.97562602e-04
2.51476711e-04 1.31613610e-03 4.18234449e-05 2.01638992e-04
2.66266416e-05 8.53372330e-05 6.14622666e-04 2.26133983e-04
9.31883405e-05 8.49861390e-05 9.05829074e-04 3.95126153e-05
2.19148878e-05 8.25024472e-05 2.39794645e-05 1.10723381e-03
7.13291302e-05 2.36116393e-04 2.40695794e-04 2.60236062e-04
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3.61663464e-04 4.45229030e-04 2.72112986e-04 2.87087892e-06
3.61288403e-04 6.21178304e-04 2.04670141e-05 7.67436577e-05
8.71896627e-05 6.25145214e-04 3.63889936e-04 1.31533103e-04
3.38180194e-04 1.70185650e-03 2.65352952e-04 1.38429314e-04
9.57641714e-06 5.85757953e-04 2.93099583e-04 2.71516124e-04
1.10013694e-04 2.12021041e-05 5.32478502e-04 1.58851806e-04
7.28346786e-05 7.45117213e-05 5.06315439e-04 7.12172943e-04
2.33235605e-05 4.29070642e-04 2.58522086e-05 1.45645684e-03
1.60161668e-04 2.95481477e-05 1.37279043e-04 6.89423759e-05
2.97003971e-05 1.15219445e-04 3.95013085e-05 7.65176228e-05
8.95155099e-05 1.20341807e-04 1.30225468e-04 9.33943084e-05
6.54157624e-03 2.40754762e-05 4.81955431e-05 4.00307385e-04
6.05372363e-04 3.09340947e-04 2.67484138e-04 1.92242232e-03
1.71189918e-03 1.99379836e-04 3.41214036e-05 8.55870836e-04
1.34391375e-04 9.12115065e-05 2.63787092e-06 2.00020131e-05
3.23024142e-04 4.57847527e-05 6.59130455e-04 1.43710384e-03
1.65375485e-03 9.48748726e-04 7.93987885e-04 7.88530568e-04
1.26306771e-03 1.57873219e-05 6.23157975e-05 5.07640434e-05
1.00583668e-04 2.63427868e-02 2.79214786e-04 4.11254296e-04
2.00241338e-03 2.15464479e-05 2.20117727e-04 5.20894673e-06
1.06383231e-04 3.89995228e-04 1.18828044e-04 1.47819592e-05
1.56530696e-05 1.03744633e-05 8.23242299e-04 1.07548170e-04
8.51472083e-04 1.68794504e-05 6.11567113e-04 1.31508459e-05
1.45642407e-05 1.70212661e-05 4.47813227e-06 2.96817211e-06
7.97575922e-06 9.19564045e-04 5.69658412e-04 1.06845036e-04
1.38349315e-05 6.94129221e-06 3.45503300e-04 4.58483337e-05
5.70481236e-04 3.14873498e-04 2.07158388e-04 7.35614376e-05
3.47069254e-05 2.23567567e-05 3.26645531e-04 2.17293986e-04
6.19800921e-05 6.93862268e-04 2.12625891e-05 3.41955456e-05
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4.05132487e-05 8.79637955e-05 1.04369083e-03 1.91554471e-04
3.08821982e-05 3.44705804e-05 9.25602726e-05 2.59073931e-05
1.07711840e-05 6.71665708e-04 6.75932461e-05 6.22531879e-05
4.64484265e-06 7.00603268e-05 2.42308190e-04 1.96481833e-05
3.98481963e-04 2.74188315e-05 1.39456693e-04 5.24139730e-04
5.99602470e-04 9.94585389e-06 1.29866559e-04 8.90876327e-05
2.63735164e-05 2.77296203e-04 3.93479859e-05 9.76353040e-05
6.76064592e-05 3.16987825e-05 2.46331724e-03 8.19425768e-05
9.37245277e-05 1.14760885e-04 3.52426905e-05 9.93357971e-04
5.93172399e-06 1.44978758e-05 2.35269217e-05 2.61552759e-05
2.09589707e-05 6.33978652e-06 2.15005775e-06 5.56230952e-06
```

```
AttributeError Traceback (most recent call last)
<ipython-input-17-83421ec204f5> in <module>
7
8 print(model.predict(test_image))
----> 9 print(model.predict_classes(test_image))
```

Testing a new image

```
In [18]: test_image_path = 'D:/Harold/MyDNN/DataSet/Chest_xray_seperate/PNEUMONIA/person11_b
         acteria 45.jpeg'
         test_image = image.load_img(test_image_path, target_size=(224, 224))
         x = image.img_to_array(test_image)
         x = np.expand_dims(x, axis=0)
         x = preprocess input(x)
         print (x.shape)
         # if num channel==1:
              if (K.image data format() == 'channels first'):
                   test image= np.expand dims(test image, axis=0)
                  test image= np.expand dims(test image, axis=0)
                   print (test image.shape)
                  test_image= np.expand_dims(test_image, axis=3)
                  test image= np.expand dims(test image, axis=0)
                  print (test image.shape)
         # else:
               if (K.image_data_format() == 'channels_first'):
         #
                   test image=np.rollaxis(test image,2,0)
         #
                   test image= np.expand dims(test image, axis=0)
         #
                   print (test image.shape)
         #
              else:
         #
                  test_image= np.expand_dims(test_image, axis=0)
                   print (test_image.shape)
         # Predicting the test image
         yhat = custom resnet model.predict(x)
         print(yhat)
         # print(custom resnet model.predict classes(x))
         label = decode_predictions(yhat)
         # retrieve the most likely result, e.g. highest probability
         label = label[0][0]
```

```
(1, 224, 224, 3)
[[0.00159899 0.998401 ]]
-----
ValueError
                                      Traceback (most recent call last)
<ipython-input-18-505048f79341> in <module>
    30 print(yhat)
    31 # print(custom resnet model.predict classes(x))
---> 32 label = decode predictions (yhat)
    33 # retrieve the most likely result, e.g. highest probability
    34 label = label[0][0]
D:\Anaconda3\lib\site-packages\tensorflow\python\keras\applications\inception v
3.py in decode predictions(preds, top)
   412 @keras export ('keras.applications.inception v3.decode predictions')
   413 def decode predictions (preds, top=5):
--> 414 return imagenet_utils.decode_predictions(preds, top=top)
   415
   416
D:\Anaconda3\lib\site-packages\tensorflow\python\keras\applications\imagenet uti
ls.py in decode predictions(preds, top)
   149
                           'a batch of predictions '
   150
                           '(i.e. a 2D array of shape (samples, 1000)). '
--> 151
                           'Found array with shape: ' + str(preds.shape))
   if CLASS_INDEX is None:
   fpath = data_utils.get_file(
ValueError: `decode predictions` expects a batch of predictions (i.e. a 2D array
of shape (samples, 1000)). Found array with shape: (1, 2)
```

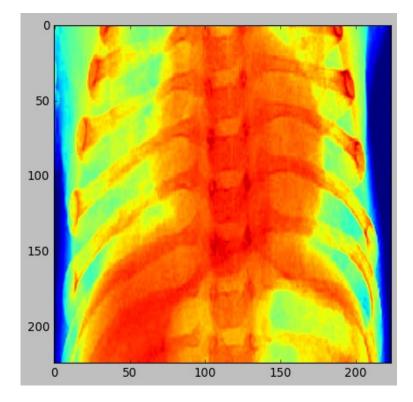
Visualizing the intermediate layer

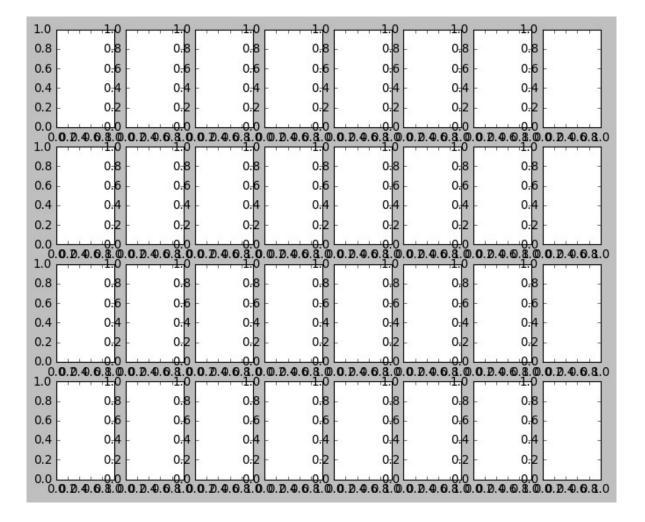
```
In [19]: | from keras.models import Model
         layer outputs = [layer.output for layer in model.layers]
         activation_model = Model(inputs=custom_resnet_model.input, outputs=layer_outputs)
         activations = custom_resnet_model.predict(X_train[10].reshape(1,224,224,3))
         print(activations.shape)
         def display activation (activations, col size, row size, act index):
             activation = activations[0, act index]
             activation index=1
             fig, ax = plt.subplots(row size, col size, figsize=(row size*2.5,col size*1))
             for row in range(0, row size):
                 for col in range(0,col size):
                     ax[row][col].imshow(activation[0, :, :, activation index], cmap='gray')
                     activation_index += 1
         plt.imshow(test_image)
         plt.imshow(X_train[10][:,:,0]);
         display_activation(activations, 8, 4, 1)
```

(1, 2)

```
IndexError
                                          Traceback (most recent call last)
<ipython-input-19-32e8200fb41b> in <module>
     14 plt.imshow(test_image)
     15 plt.imshow(X_train[10][:,:,0]);
---> 16 display activation (activations, 8, 4, 1)
<ipython-input-19-32e8200fb41b> in display activation(activations, col size, row
_size, act_index)
    10
          for row in range(0, row size):
     11
              for col in range(0,col_size):
---> 12
                    ax[row][col].imshow(activation[0, :, :, activation_index], c
map='gray')
     13
                    activation index += 1
     14 plt.imshow(test_image)
```

IndexError: invalid index to scalar variable.



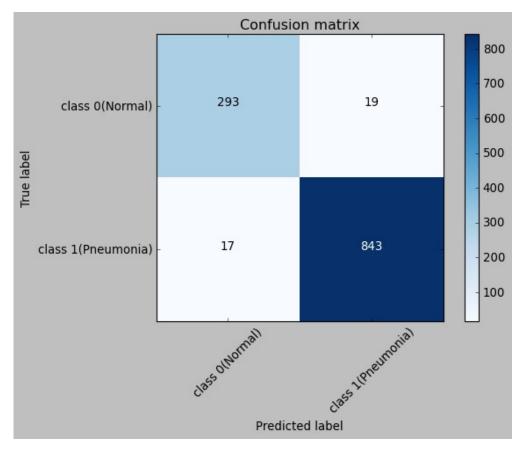


Confusion matrix

```
In [20]: Y_pred = custom_resnet_model.predict(X_test)
         print(Y pred)
         y_pred = np.argmax(Y_pred, axis=1)
         print(y_pred)
         #y_pred = model.predict_classes(X_test)
         #print(y pred)
         target names = ['class 0(Normal)', 'class 1(Pneumonia)']
         print(classification report(np.argmax(y test,axis=1), y pred,target names=target na
         print(confusion matrix(np.argmax(y test,axis=1), y pred))
         [[6.5542844e-07 9.9999940e-01]
         [9.9147195e-01 8.5280919e-03]
         [1.9132092e-05 9.9998093e-01]
         [2.7467880e-07 9.9999976e-01]
         [8.9294559e-08 9.9999988e-01]
         [9.9940789e-01 5.9217890e-04]]
         [1 0 1 ... 1 1 0]
                           precision recall f1-score support
           class 0(Normal) 0.95 0.94 0.94
                                                               312
                               0.98
                                        0.98
                                                   0.98
         class 1(Pneumonia)
                                                             860
                                                   0.97
                                                             1172
                  accuracy
                                                         1172
              macro avg 0.96 0.96 0.96 weighted avg 0.97 0.97 0.97
         [[293 19]
         [ 17 843]]
```

Compute confusion matrix

Confusion matrix, without normalization [[293 19] [ 17 843]]



In []: