

```
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.resnet import ResNet50
from tensorflow.keras.applications.inception_v3 import InceptionV3
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_preperation():
    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()], [model.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(filter_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 first filter
        ax.imshow(feature_maps[:, :, i], cmap='gray')
        plt.xticks([])
        plt.yticks([])
        plt.tight_layout()
    plt.show()
    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 Preparation

```
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_sta
te=2)
```

Model Definition

Training the classifier alone

```
In [11]: image_input = Input(shape=(224, 224, 3))
model = InceptionV3(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: "inception_v3"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 224, 224, 3) 0		
conv2d (Conv2D)	(None, 111, 111, 32) 864		input_1[0][0]
batch_normalization (BatchNorma	(None, 111, 111, 32) 96		conv2d[0][0]
activation (Activation)	(None, 111, 111, 32) 0		batch_normaliza tion[0][0]
conv2d_1 (Conv2D)	(None, 109, 109, 32) 9216		activation
batch_normalization_1 (BatchNor	(None, 109, 109, 32) 96		conv2d_1[0][0]
activation_1 (Activation)	(None, 109, 109, 32) 0		batch_normaliza tion_1[0][0]
conv2d_2 (Conv2D)	(None, 109, 109, 64) 18432		activation_1
batch_normalization_2 (BatchNor	(None, 109, 109, 64) 192		conv2d_2[0][0]
activation_2 (Activation)	(None, 109, 109, 64) 0		batch_normaliza tion_2[0][0]
max_pooling2d (MaxPooling2D)	(None, 54, 54, 64) 0		activation_2
conv2d_3 (Conv2D)	(None, 54, 54, 80) 5120		max_pooling2d
batch_normalization_3 (BatchNor	(None, 54, 54, 80) 240		conv2d_3[0][0]
activation_3 (Activation)	(None, 54, 54, 80) 0		batch_normaliza tion_3[0][0]
conv2d_4 (Conv2D)	(None, 52, 52, 192) 138240		activation_3
batch_normalization_4 (BatchNor	(None, 52, 52, 192) 576		conv2d_4[0][0]

activation_4 (Activation) activation_4[0][0]	(None, 52, 52, 192)	0	batch_normaliza tion_4[0][0]
max_pooling2d_1 (MaxPooling2D) [0][0]	(None, 25, 25, 192)	0	activation_4
conv2d_8 (Conv2D) [0][0]	(None, 25, 25, 64)	12288	max_pooling2d_1
batch_normalization_8 (BatchNor	(None, 25, 25, 64)	192	conv2d_8[0][0]
activation_8 (Activation) activation_8[0][0]	(None, 25, 25, 64)	0	batch_normaliza tion_8[0][0]
conv2d_6 (Conv2D) [0][0]	(None, 25, 25, 48)	9216	max_pooling2d_1
conv2d_9 (Conv2D) [0][0]	(None, 25, 25, 96)	55296	activation_8
batch_normalization_6 (BatchNor	(None, 25, 25, 48)	144	conv2d_6[0][0]
batch_normalization_9 (BatchNor	(None, 25, 25, 96)	288	conv2d_9[0][0]
activation_6 (Activation) activation_6[0][0]	(None, 25, 25, 48)	0	batch_normaliza tion_6[0][0]
activation_9 (Activation) activation_9[0][0]	(None, 25, 25, 96)	0	batch_normaliza tion_9[0][0]
average_pooling2d (AveragePooli [0][0]	(None, 25, 25, 192)	0	max_pooling2d_1
conv2d_5 (Conv2D) [0][0]	(None, 25, 25, 64)	12288	max_pooling2d_1
conv2d_7 (Conv2D) [0][0]	(None, 25, 25, 64)	76800	activation_6
conv2d_10 (Conv2D) [0][0]	(None, 25, 25, 96)	82944	activation_9
conv2d_11 (Conv2D) 2d[0][0]	(None, 25, 25, 32)	6144	average_pooling 2d[0][0]
batch_normalization_5 (BatchNor	(None, 25, 25, 64)	192	conv2d_5[0][0]

batch_normalization_7	(BatchNor	(None, 25, 25, 64)	192	conv2d_7[0][0]
batch_normalization_10	(BatchNo	(None, 25, 25, 96)	288	conv2d_10[0][0]
batch_normalization_11	(BatchNo	(None, 25, 25, 32)	96	conv2d_11[0][0]
activation_5	(Activation)	(None, 25, 25, 64)	0	batch_normaliza tion_5[0][0]
activation_7	(Activation)	(None, 25, 25, 64)	0	batch_normaliza tion_7[0][0]
activation_10	(Activation)	(None, 25, 25, 96)	0	batch_normaliza tion_10[0][0]
activation_11	(Activation)	(None, 25, 25, 32)	0	batch_normaliza tion_11[0][0]
mixed0	(Concatenate)	(None, 25, 25, 256)	0	activation_5 [0][0] activation_7 [0][0] activation_10 [0][0] activation_11 [0][0]
conv2d_15	(Conv2D)	(None, 25, 25, 64)	16384	mixed0[0][0]
batch_normalization_15	(BatchNo	(None, 25, 25, 64)	192	conv2d_15[0][0]
activation_15	(Activation)	(None, 25, 25, 64)	0	batch_normaliza tion_15[0][0]
conv2d_13	(Conv2D)	(None, 25, 25, 48)	12288	mixed0[0][0]
conv2d_16	(Conv2D)	(None, 25, 25, 96)	55296	activation_15 [0][0]
batch_normalization_13	(BatchNo	(None, 25, 25, 48)	144	conv2d_13[0][0]
batch_normalization_16	(BatchNo	(None, 25, 25, 96)	288	conv2d_16[0][0]
activation_13	(Activation)	(None, 25, 25, 48)	0	batch_normaliza tion_13[0][0]
activation_16	(Activation)	(None, 25, 25, 96)	0	batch_normaliza tion_16[0][0]

average_pooling2d_1 (AveragePoo	(None, 25, 25, 256)	0	mixed0[0][0]
conv2d_12 (Conv2D)	(None, 25, 25, 64)	16384	mixed0[0][0]
conv2d_14 (Conv2D) [0][0]	(None, 25, 25, 64)	76800	activation_13
conv2d_17 (Conv2D) [0][0]	(None, 25, 25, 96)	82944	activation_16
conv2d_18 (Conv2D) 2d_1[0][0]	(None, 25, 25, 64)	16384	average_pooling
batch_normalization_12 (BatchNo	(None, 25, 25, 64)	192	conv2d_12[0][0]
batch_normalization_14 (BatchNo	(None, 25, 25, 64)	192	conv2d_14[0][0]
batch_normalization_17 (BatchNo	(None, 25, 25, 96)	288	conv2d_17[0][0]
batch_normalization_18 (BatchNo	(None, 25, 25, 64)	192	conv2d_18[0][0]
activation_12 (Activation) tion_12[0][0]	(None, 25, 25, 64)	0	batch_normaliza
activation_14 (Activation) tion_14[0][0]	(None, 25, 25, 64)	0	batch_normaliza
activation_17 (Activation) tion_17[0][0]	(None, 25, 25, 96)	0	batch_normaliza
activation_18 (Activation) tion_18[0][0]	(None, 25, 25, 64)	0	batch_normaliza
mixed1 (Concatenate) [0][0] [0][0] [0][0] [0][0]	(None, 25, 25, 288)	0	activation_12 activation_14 activation_17 activation_18
conv2d_22 (Conv2D)	(None, 25, 25, 64)	18432	mixed1[0][0]
batch_normalization_22 (BatchNo	(None, 25, 25, 64)	192	conv2d_22[0][0]

activation_22 (Activation) tion_22[0][0]	(None, 25, 25, 64)	0	batch_normaliza
conv2d_20 (Conv2D)	(None, 25, 25, 48)	13824	mixed1[0][0]
conv2d_23 (Conv2D) [0][0]	(None, 25, 25, 96)	55296	activation_22
batch_normalization_20 (BatchNo	(None, 25, 25, 48)	144	conv2d_20[0][0]
batch_normalization_23 (BatchNo	(None, 25, 25, 96)	288	conv2d_23[0][0]
activation_20 (Activation) tion_20[0][0]	(None, 25, 25, 48)	0	batch_normaliza
activation_23 (Activation) tion_23[0][0]	(None, 25, 25, 96)	0	batch_normaliza
average_pooling2d_2 (AveragePoo	(None, 25, 25, 288)	0	mixed1[0][0]
conv2d_19 (Conv2D)	(None, 25, 25, 64)	18432	mixed1[0][0]
conv2d_21 (Conv2D) [0][0]	(None, 25, 25, 64)	76800	activation_20
conv2d_24 (Conv2D) [0][0]	(None, 25, 25, 96)	82944	activation_23
conv2d_25 (Conv2D) 2d_2[0][0]	(None, 25, 25, 64)	18432	average_pooling
batch_normalization_19 (BatchNo	(None, 25, 25, 64)	192	conv2d_19[0][0]
batch_normalization_21 (BatchNo	(None, 25, 25, 64)	192	conv2d_21[0][0]
batch_normalization_24 (BatchNo	(None, 25, 25, 96)	288	conv2d_24[0][0]
batch_normalization_25 (BatchNo	(None, 25, 25, 64)	192	conv2d_25[0][0]
activation_19 (Activation) tion_19[0][0]	(None, 25, 25, 64)	0	batch_normaliza
activation_21 (Activation) tion_21[0][0]	(None, 25, 25, 64)	0	batch_normaliza
activation_24 (Activation)	(None, 25, 25, 96)	0	batch_normaliza

tion_24[0][0]

activation_25 (Activation) tion_25[0][0]	(None, 25, 25, 64)	0	batch_normaliza
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mixed2 (Concatenate) [0][0]	(None, 25, 25, 288)	0	activation_19
[0][0]			activation_21
[0][0]			activation_24
[0][0]			activation_25

conv2d_27 (Conv2D)	(None, 25, 25, 64)	18432	mixed2[0][0]
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batch_normalization_27 (BatchNo	(None, 25, 25, 64)	192	conv2d_27[0][0]
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activation_27 (Activation) tion_27[0][0]	(None, 25, 25, 64)	0	batch_normaliza
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conv2d_28 (Conv2D) [0][0]	(None, 25, 25, 96)	55296	activation_27
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batch_normalization_28 (BatchNo	(None, 25, 25, 96)	288	conv2d_28[0][0]
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activation_28 (Activation) tion_28[0][0]	(None, 25, 25, 96)	0	batch_normaliza
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conv2d_26 (Conv2D)	(None, 12, 12, 384)	995328	mixed2[0][0]
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conv2d_29 (Conv2D) [0][0]	(None, 12, 12, 96)	82944	activation_28
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batch_normalization_26 (BatchNo	(None, 12, 12, 384)	1152	conv2d_26[0][0]
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batch_normalization_29 (BatchNo	(None, 12, 12, 96)	288	conv2d_29[0][0]
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activation_26 (Activation) tion_26[0][0]	(None, 12, 12, 384)	0	batch_normaliza
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activation_29 (Activation) tion_29[0][0]	(None, 12, 12, 96)	0	batch_normaliza
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max_pooling2d_2 (MaxPooling2D)	(None, 12, 12, 288)	0	mixed2[0][0]
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mixed3 (Concatenate) [0][0]	(None, 12, 12, 768)	0	activation_26
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			activation_29
[0][0]			max_pooling2d_2
[0][0]			
conv2d_34 (Conv2D)	(None, 12, 12, 128)	98304	mixed3[0][0]
batch_normalization_34 (BatchNo	(None, 12, 12, 128)	384	conv2d_34[0][0]
activation_34 (Activation)	(None, 12, 12, 128)	0	batch_normaliza
tion_34[0][0]			
conv2d_35 (Conv2D)	(None, 12, 12, 128)	114688	activation_34
[0][0]			
batch_normalization_35 (BatchNo	(None, 12, 12, 128)	384	conv2d_35[0][0]
activation_35 (Activation)	(None, 12, 12, 128)	0	batch_normaliza
tion_35[0][0]			
conv2d_31 (Conv2D)	(None, 12, 12, 128)	98304	mixed3[0][0]
conv2d_36 (Conv2D)	(None, 12, 12, 128)	114688	activation_35
[0][0]			
batch_normalization_31 (BatchNo	(None, 12, 12, 128)	384	conv2d_31[0][0]
batch_normalization_36 (BatchNo	(None, 12, 12, 128)	384	conv2d_36[0][0]
activation_31 (Activation)	(None, 12, 12, 128)	0	batch_normaliza
tion_31[0][0]			
activation_36 (Activation)	(None, 12, 12, 128)	0	batch_normaliza
tion_36[0][0]			
conv2d_32 (Conv2D)	(None, 12, 12, 128)	114688	activation_31
[0][0]			
conv2d_37 (Conv2D)	(None, 12, 12, 128)	114688	activation_36
[0][0]			
batch_normalization_32 (BatchNo	(None, 12, 12, 128)	384	conv2d_32[0][0]
batch_normalization_37 (BatchNo	(None, 12, 12, 128)	384	conv2d_37[0][0]
activation_32 (Activation)	(None, 12, 12, 128)	0	batch_normaliza
tion_32[0][0]			

activation_37 (Activation) tion_37[0][0]	(None, 12, 12, 128)	0	batch_normaliza
average_pooling2d_3 (AveragePoo	(None, 12, 12, 768)	0	mixed3[0][0]
conv2d_30 (Conv2D)	(None, 12, 12, 192)	147456	mixed3[0][0]
conv2d_33 (Conv2D) [0][0]	(None, 12, 12, 192)	172032	activation_32
conv2d_38 (Conv2D) [0][0]	(None, 12, 12, 192)	172032	activation_37
conv2d_39 (Conv2D) 2d_3[0][0]	(None, 12, 12, 192)	147456	average_pooling
batch_normalization_30 (BatchNo	(None, 12, 12, 192)	576	conv2d_30[0][0]
batch_normalization_33 (BatchNo	(None, 12, 12, 192)	576	conv2d_33[0][0]
batch_normalization_38 (BatchNo	(None, 12, 12, 192)	576	conv2d_38[0][0]
batch_normalization_39 (BatchNo	(None, 12, 12, 192)	576	conv2d_39[0][0]
activation_30 (Activation) tion_30[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_33 (Activation) tion_33[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_38 (Activation) tion_38[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_39 (Activation) tion_39[0][0]	(None, 12, 12, 192)	0	batch_normaliza
mixed4 (Concatenate) [0][0]	(None, 12, 12, 768)	0	activation_30
[0][0]			activation_33
[0][0]			activation_38
[0][0]			activation_39
conv2d_44 (Conv2D)	(None, 12, 12, 160)	122880	mixed4[0][0]

batch_normalization_44	(BatchNo	(None, 12, 12, 160)	480	conv2d_44[0][0]
activation_44	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_44[0][0]
conv2d_45	(Conv2D)	(None, 12, 12, 160)	179200	activation_44 [0][0]
batch_normalization_45	(BatchNo	(None, 12, 12, 160)	480	conv2d_45[0][0]
activation_45	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_45[0][0]
conv2d_41	(Conv2D)	(None, 12, 12, 160)	122880	mixed4[0][0]
conv2d_46	(Conv2D)	(None, 12, 12, 160)	179200	activation_45 [0][0]
batch_normalization_41	(BatchNo	(None, 12, 12, 160)	480	conv2d_41[0][0]
batch_normalization_46	(BatchNo	(None, 12, 12, 160)	480	conv2d_46[0][0]
activation_41	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_41[0][0]
activation_46	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_46[0][0]
conv2d_42	(Conv2D)	(None, 12, 12, 160)	179200	activation_41 [0][0]
conv2d_47	(Conv2D)	(None, 12, 12, 160)	179200	activation_46 [0][0]
batch_normalization_42	(BatchNo	(None, 12, 12, 160)	480	conv2d_42[0][0]
batch_normalization_47	(BatchNo	(None, 12, 12, 160)	480	conv2d_47[0][0]
activation_42	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_42[0][0]
activation_47	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_47[0][0]
average_pooling2d_4	(AveragePoo	(None, 12, 12, 768)	0	mixed4[0][0]

conv2d_40 (Conv2D)	(None, 12, 12, 192)	147456	mixed4[0][0]
conv2d_43 (Conv2D) [0][0]	(None, 12, 12, 192)	215040	activation_42
conv2d_48 (Conv2D) [0][0]	(None, 12, 12, 192)	215040	activation_47
conv2d_49 (Conv2D) 2d_4[0][0]	(None, 12, 12, 192)	147456	average_pooling
batch_normalization_40 (BatchNo	(None, 12, 12, 192)	576	conv2d_40[0][0]
batch_normalization_43 (BatchNo	(None, 12, 12, 192)	576	conv2d_43[0][0]
batch_normalization_48 (BatchNo	(None, 12, 12, 192)	576	conv2d_48[0][0]
batch_normalization_49 (BatchNo	(None, 12, 12, 192)	576	conv2d_49[0][0]
activation_40 (Activation) tion_40[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_43 (Activation) tion_43[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_48 (Activation) tion_48[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_49 (Activation) tion_49[0][0]	(None, 12, 12, 192)	0	batch_normaliza
mixed5 (Concatenate) [0][0] [0][0] [0][0] [0][0]	(None, 12, 12, 768)	0	activation_40 activation_43 activation_48 activation_49
conv2d_54 (Conv2D)	(None, 12, 12, 160)	122880	mixed5[0][0]
batch_normalization_54 (BatchNo	(None, 12, 12, 160)	480	conv2d_54[0][0]
activation_54 (Activation) tion_54[0][0]	(None, 12, 12, 160)	0	batch_normaliza

conv2d_55 (Conv2D) [0][0]	(None, 12, 12, 160)	179200	activation_54
batch_normalization_55 (BatchNo	(None, 12, 12, 160)	480	conv2d_55[0][0]
activation_55 (Activation) activation_55[0][0]	(None, 12, 12, 160)	0	batch_normaliza tion_55[0][0]
conv2d_51 (Conv2D)	(None, 12, 12, 160)	122880	mixed5[0][0]
conv2d_56 (Conv2D) [0][0]	(None, 12, 12, 160)	179200	activation_55
batch_normalization_51 (BatchNo	(None, 12, 12, 160)	480	conv2d_51[0][0]
batch_normalization_56 (BatchNo	(None, 12, 12, 160)	480	conv2d_56[0][0]
activation_51 (Activation) activation_51[0][0]	(None, 12, 12, 160)	0	batch_normaliza tion_51[0][0]
activation_56 (Activation) activation_56[0][0]	(None, 12, 12, 160)	0	batch_normaliza tion_56[0][0]
conv2d_52 (Conv2D) [0][0]	(None, 12, 12, 160)	179200	activation_51
conv2d_57 (Conv2D) [0][0]	(None, 12, 12, 160)	179200	activation_56
batch_normalization_52 (BatchNo	(None, 12, 12, 160)	480	conv2d_52[0][0]
batch_normalization_57 (BatchNo	(None, 12, 12, 160)	480	conv2d_57[0][0]
activation_52 (Activation) activation_52[0][0]	(None, 12, 12, 160)	0	batch_normaliza tion_52[0][0]
activation_57 (Activation) activation_57[0][0]	(None, 12, 12, 160)	0	batch_normaliza tion_57[0][0]
average_pooling2d_5 (AveragePoo	(None, 12, 12, 768)	0	mixed5[0][0]
conv2d_50 (Conv2D)	(None, 12, 12, 192)	147456	mixed5[0][0]
conv2d_53 (Conv2D) [0][0]	(None, 12, 12, 192)	215040	activation_52

conv2d_58 (Conv2D) [0][0]	(None, 12, 12, 192)	215040	activation_57
conv2d_59 (Conv2D) 2d_5[0][0]	(None, 12, 12, 192)	147456	average_pooling
batch_normalization_50 (BatchNo	(None, 12, 12, 192)	576	conv2d_50[0][0]
batch_normalization_53 (BatchNo	(None, 12, 12, 192)	576	conv2d_53[0][0]
batch_normalization_58 (BatchNo	(None, 12, 12, 192)	576	conv2d_58[0][0]
batch_normalization_59 (BatchNo	(None, 12, 12, 192)	576	conv2d_59[0][0]
activation_50 (Activation) tion_50[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_53 (Activation) tion_53[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_58 (Activation) tion_58[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_59 (Activation) tion_59[0][0]	(None, 12, 12, 192)	0	batch_normaliza
mixed6 (Concatenate) [0][0]	(None, 12, 12, 768)	0	activation_50
[0][0]			activation_53
[0][0]			activation_58
[0][0]			activation_59
conv2d_64 (Conv2D)	(None, 12, 12, 192)	147456	mixed6[0][0]
batch_normalization_64 (BatchNo	(None, 12, 12, 192)	576	conv2d_64[0][0]
activation_64 (Activation) tion_64[0][0]	(None, 12, 12, 192)	0	batch_normaliza
conv2d_65 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_64
batch_normalization_65 (BatchNo	(None, 12, 12, 192)	576	conv2d_65[0][0]
activation_65 (Activation)	(None, 12, 12, 192)	0	batch_normaliza

tion_65[0][0]

conv2d_61 (Conv2D)	(None, 12, 12, 192)	147456	mixed6[0][0]
conv2d_66 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_65
batch_normalization_61 (BatchNo	(None, 12, 12, 192)	576	conv2d_61[0][0]
batch_normalization_66 (BatchNo	(None, 12, 12, 192)	576	conv2d_66[0][0]
activation_61 (Activation) tion_61[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_66 (Activation) tion_66[0][0]	(None, 12, 12, 192)	0	batch_normaliza
conv2d_62 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_61
conv2d_67 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_66
batch_normalization_62 (BatchNo	(None, 12, 12, 192)	576	conv2d_62[0][0]
batch_normalization_67 (BatchNo	(None, 12, 12, 192)	576	conv2d_67[0][0]
activation_62 (Activation) tion_62[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_67 (Activation) tion_67[0][0]	(None, 12, 12, 192)	0	batch_normaliza
average_pooling2d_6 (AveragePoo	(None, 12, 12, 768)	0	mixed6[0][0]
conv2d_60 (Conv2D)	(None, 12, 12, 192)	147456	mixed6[0][0]
conv2d_63 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_62
conv2d_68 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_67
conv2d_69 (Conv2D) 2d_6[0][0]	(None, 12, 12, 192)	147456	average_pooling

batch_normalization_60	(BatchNo	(None, 12, 12, 192)	576	conv2d_60[0][0]
batch_normalization_63	(BatchNo	(None, 12, 12, 192)	576	conv2d_63[0][0]
batch_normalization_68	(BatchNo	(None, 12, 12, 192)	576	conv2d_68[0][0]
batch_normalization_69	(BatchNo	(None, 12, 12, 192)	576	conv2d_69[0][0]
activation_60 activation_60[0][0]	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_60[0][0]
activation_63 activation_63[0][0]	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_63[0][0]
activation_68 activation_68[0][0]	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_68[0][0]
activation_69 activation_69[0][0]	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_69[0][0]
mixed7 [0][0] [0][0] [0][0] [0][0]	(Concatenate)	(None, 12, 12, 768)	0	activation_60 activation_63 activation_68 activation_69
conv2d_72	(Conv2D)	(None, 12, 12, 192)	147456	mixed7[0][0]
batch_normalization_72	(BatchNo	(None, 12, 12, 192)	576	conv2d_72[0][0]
activation_72 activation_72[0][0]	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_72[0][0]
conv2d_73 [0][0]	(Conv2D)	(None, 12, 12, 192)	258048	activation_72
batch_normalization_73	(BatchNo	(None, 12, 12, 192)	576	conv2d_73[0][0]
activation_73 activation_73[0][0]	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_73[0][0]
conv2d_70	(Conv2D)	(None, 12, 12, 192)	147456	mixed7[0][0]
conv2d_74 [0][0]	(Conv2D)	(None, 12, 12, 192)	258048	activation_73

batch_normalization_70	(BatchNo	(None, 12, 12, 192)	576	conv2d_70[0][0]
batch_normalization_74	(BatchNo	(None, 12, 12, 192)	576	conv2d_74[0][0]
activation_70	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_70[0][0]
activation_74	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_74[0][0]
conv2d_71	(Conv2D)	(None, 5, 5, 320)	552960	activation_70 [0][0]
conv2d_75	(Conv2D)	(None, 5, 5, 192)	331776	activation_74 [0][0]
batch_normalization_71	(BatchNo	(None, 5, 5, 320)	960	conv2d_71[0][0]
batch_normalization_75	(BatchNo	(None, 5, 5, 192)	576	conv2d_75[0][0]
activation_71	(Activation)	(None, 5, 5, 320)	0	batch_normaliza tion_71[0][0]
activation_75	(Activation)	(None, 5, 5, 192)	0	batch_normaliza tion_75[0][0]
max_pooling2d_3	(MaxPooling2D)	(None, 5, 5, 768)	0	mixed7[0][0]
mixed8	(Concatenate)	(None, 5, 5, 1280)	0	activation_71 [0][0] activation_75 [0][0] max_pooling2d_3 [0][0]
conv2d_80	(Conv2D)	(None, 5, 5, 448)	573440	mixed8[0][0]
batch_normalization_80	(BatchNo	(None, 5, 5, 448)	1344	conv2d_80[0][0]
activation_80	(Activation)	(None, 5, 5, 448)	0	batch_normaliza tion_80[0][0]
conv2d_77	(Conv2D)	(None, 5, 5, 384)	491520	mixed8[0][0]
conv2d_81	(Conv2D)	(None, 5, 5, 384)	1548288	activation_80 [0][0]

batch_normalization_77	(BatchNo	(None, 5, 5, 384)	1152	conv2d_77[0][0]
batch_normalization_81	(BatchNo	(None, 5, 5, 384)	1152	conv2d_81[0][0]
activation_77	(Activation)	(None, 5, 5, 384)	0	batch_normaliza tion_77[0][0]
activation_81	(Activation)	(None, 5, 5, 384)	0	batch_normaliza tion_81[0][0]
conv2d_78	(Conv2D)	(None, 5, 5, 384)	442368	activation_77 [0][0]
conv2d_79	(Conv2D)	(None, 5, 5, 384)	442368	activation_77 [0][0]
conv2d_82	(Conv2D)	(None, 5, 5, 384)	442368	activation_81 [0][0]
conv2d_83	(Conv2D)	(None, 5, 5, 384)	442368	activation_81 [0][0]
average_pooling2d_7	(AveragePoo	(None, 5, 5, 1280)	0	mixed8[0][0]
conv2d_76	(Conv2D)	(None, 5, 5, 320)	409600	mixed8[0][0]
batch_normalization_78	(BatchNo	(None, 5, 5, 384)	1152	conv2d_78[0][0]
batch_normalization_79	(BatchNo	(None, 5, 5, 384)	1152	conv2d_79[0][0]
batch_normalization_82	(BatchNo	(None, 5, 5, 384)	1152	conv2d_82[0][0]
batch_normalization_83	(BatchNo	(None, 5, 5, 384)	1152	conv2d_83[0][0]
conv2d_84	(Conv2D)	(None, 5, 5, 192)	245760	average_pooling 2d_7[0][0]
batch_normalization_76	(BatchNo	(None, 5, 5, 320)	960	conv2d_76[0][0]
activation_78	(Activation)	(None, 5, 5, 384)	0	batch_normaliza tion_78[0][0]
activation_79	(Activation)	(None, 5, 5, 384)	0	batch_normaliza tion_79[0][0]

activation_82 (Activation) activation_82[0][0]	(None, 5, 5, 384)	0	batch_normaliza tion_82[0][0]
activation_83 (Activation) activation_83[0][0]	(None, 5, 5, 384)	0	batch_normaliza tion_83[0][0]
batch_normalization_84 (BatchNo	(None, 5, 5, 192)	576	conv2d_84[0][0]
activation_76 (Activation) activation_76[0][0]	(None, 5, 5, 320)	0	batch_normaliza tion_76[0][0]
mixed9_0 (Concatenate) [0][0]	(None, 5, 5, 768)	0	activation_78 activation_79 [0][0]
concatenate (Concatenate) [0][0]	(None, 5, 5, 768)	0	activation_82 activation_83 [0][0]
activation_84 (Activation) activation_84[0][0]	(None, 5, 5, 192)	0	batch_normaliza tion_84[0][0]
mixed9 (Concatenate) [0][0]	(None, 5, 5, 2048)	0	activation_76 mixed9_0[0][0] concatenate activation_84 [0][0]
conv2d_89 (Conv2D)	(None, 5, 5, 448)	917504	mixed9[0][0]
batch_normalization_89 (BatchNo	(None, 5, 5, 448)	1344	conv2d_89[0][0]
activation_89 (Activation) activation_89[0][0]	(None, 5, 5, 448)	0	batch_normaliza tion_89[0][0]
conv2d_86 (Conv2D)	(None, 5, 5, 384)	786432	mixed9[0][0]
conv2d_90 (Conv2D) [0][0]	(None, 5, 5, 384)	1548288	activation_89
batch_normalization_86 (BatchNo	(None, 5, 5, 384)	1152	conv2d_86[0][0]
batch_normalization_90 (BatchNo	(None, 5, 5, 384)	1152	conv2d_90[0][0]

activation_86 (Activation) tion_86[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_90 (Activation) tion_90[0][0]	(None, 5, 5, 384)	0	batch_normaliza
conv2d_87 (Conv2D) [0][0]	(None, 5, 5, 384)	442368	activation_86
conv2d_88 (Conv2D) [0][0]	(None, 5, 5, 384)	442368	activation_86
conv2d_91 (Conv2D) [0][0]	(None, 5, 5, 384)	442368	activation_90
conv2d_92 (Conv2D) [0][0]	(None, 5, 5, 384)	442368	activation_90
average_pooling2d_8 (AveragePoo	(None, 5, 5, 2048)	0	mixed9[0][0]
conv2d_85 (Conv2D)	(None, 5, 5, 320)	655360	mixed9[0][0]
batch_normalization_87 (BatchNo	(None, 5, 5, 384)	1152	conv2d_87[0][0]
batch_normalization_88 (BatchNo	(None, 5, 5, 384)	1152	conv2d_88[0][0]
batch_normalization_91 (BatchNo	(None, 5, 5, 384)	1152	conv2d_91[0][0]
batch_normalization_92 (BatchNo	(None, 5, 5, 384)	1152	conv2d_92[0][0]
conv2d_93 (Conv2D) 2d_8[0][0]	(None, 5, 5, 192)	393216	average_pooling
batch_normalization_85 (BatchNo	(None, 5, 5, 320)	960	conv2d_85[0][0]
activation_87 (Activation) tion_87[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_88 (Activation) tion_88[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_91 (Activation) tion_91[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_92 (Activation) tion_92[0][0]	(None, 5, 5, 384)	0	batch_normaliza

batch_normalization_93 (BatchNo	(None, 5, 5, 192)	576	conv2d_93[0][0]
activation_85 (Activation)	(None, 5, 5, 320)	0	batch_normaliza tion_85[0][0]
mixed9_1 (Concatenate)	(None, 5, 5, 768)	0	activation_87 [0][0] activation_88 [0][0]
concatenate_1 (Concatenate)	(None, 5, 5, 768)	0	activation_91 [0][0] activation_92 [0][0]
activation_93 (Activation)	(None, 5, 5, 192)	0	batch_normaliza tion_93[0][0]
mixed10 (Concatenate)	(None, 5, 5, 2048)	0	activation_85 [0][0] mixed9_1[0][0] concatenate_1 [0][0] activation_93 [0][0]
avg_pool (GlobalAveragePooling2	(None, 2048)	0	mixed10[0][0]
predictions (Dense)	(None, 1000)	2049000	avg_pool[0][0]
=====			
Total params: 23,851,784			
Trainable params: 23,817,352			
Non-trainable params: 34,432			

Model: "functional_1"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	[(None, 224, 224, 3)	0	
conv2d (Conv2D)	(None, 111, 111, 32)	864	input_1[0][0]
batch_normalization (BatchNorma	(None, 111, 111, 32)	96	conv2d[0][0]
activation (Activation)	(None, 111, 111, 32)	0	batch_normaliza tion[0][0]
conv2d_1 (Conv2D)	(None, 109, 109, 32)	9216	activation

[0][0]

batch_normalization_1 (BatchNor	(None, 109, 109, 32)	96	conv2d_1[0][0]
activation_1 (Activation)	(None, 109, 109, 32)	0	batch_normaliza
conv2d_2 (Conv2D)	(None, 109, 109, 64)	18432	activation_1
batch_normalization_2 (BatchNor	(None, 109, 109, 64)	192	conv2d_2[0][0]
activation_2 (Activation)	(None, 109, 109, 64)	0	batch_normaliza
max_pooling2d (MaxPooling2D)	(None, 54, 54, 64)	0	activation_2
conv2d_3 (Conv2D)	(None, 54, 54, 80)	5120	max_pooling2d
batch_normalization_3 (BatchNor	(None, 54, 54, 80)	240	conv2d_3[0][0]
activation_3 (Activation)	(None, 54, 54, 80)	0	batch_normaliza
conv2d_4 (Conv2D)	(None, 52, 52, 192)	138240	activation_3
batch_normalization_4 (BatchNor	(None, 52, 52, 192)	576	conv2d_4[0][0]
activation_4 (Activation)	(None, 52, 52, 192)	0	batch_normaliza
max_pooling2d_1 (MaxPooling2D)	(None, 25, 25, 192)	0	activation_4
conv2d_8 (Conv2D)	(None, 25, 25, 64)	12288	max_pooling2d_1
batch_normalization_8 (BatchNor	(None, 25, 25, 64)	192	conv2d_8[0][0]
activation_8 (Activation)	(None, 25, 25, 64)	0	batch_normaliza
conv2d_6 (Conv2D)	(None, 25, 25, 48)	9216	max_pooling2d_1

conv2d_9 (Conv2D) [0][0]	(None, 25, 25, 96)	55296	activation_8
batch_normalization_6 (BatchNor	(None, 25, 25, 48)	144	conv2d_6[0][0]
batch_normalization_9 (BatchNor	(None, 25, 25, 96)	288	conv2d_9[0][0]
activation_6 (Activation) tion_6[0][0]	(None, 25, 25, 48)	0	batch_normaliza
activation_9 (Activation) tion_9[0][0]	(None, 25, 25, 96)	0	batch_normaliza
average_pooling2d [0][0]	(AveragePooli (None, 25, 25, 192)	0	max_pooling2d_1
conv2d_5 (Conv2D) [0][0]	(None, 25, 25, 64)	12288	max_pooling2d_1
conv2d_7 (Conv2D) [0][0]	(None, 25, 25, 64)	76800	activation_6
conv2d_10 (Conv2D) [0][0]	(None, 25, 25, 96)	82944	activation_9
conv2d_11 (Conv2D) 2d[0][0]	(None, 25, 25, 32)	6144	average_pooling
batch_normalization_5 (BatchNor	(None, 25, 25, 64)	192	conv2d_5[0][0]
batch_normalization_7 (BatchNor	(None, 25, 25, 64)	192	conv2d_7[0][0]
batch_normalization_10 (BatchNo	(None, 25, 25, 96)	288	conv2d_10[0][0]
batch_normalization_11 (BatchNo	(None, 25, 25, 32)	96	conv2d_11[0][0]
activation_5 (Activation) tion_5[0][0]	(None, 25, 25, 64)	0	batch_normaliza
activation_7 (Activation) tion_7[0][0]	(None, 25, 25, 64)	0	batch_normaliza
activation_10 (Activation) tion_10[0][0]	(None, 25, 25, 96)	0	batch_normaliza

activation_11 (Activation) tion_11[0][0]	(None, 25, 25, 32)	0	batch_normaliza
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mixed0 (Concatenate) [0][0]	(None, 25, 25, 256)	0	activation_5
[0][0]			activation_7
[0][0]			activation_10
[0][0]			activation_11
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conv2d_15 (Conv2D)	(None, 25, 25, 64)	16384	mixed0[0][0]
<hr/>			
batch_normalization_15 (BatchNo	(None, 25, 25, 64)	192	conv2d_15[0][0]
<hr/>			
activation_15 (Activation) tion_15[0][0]	(None, 25, 25, 64)	0	batch_normaliza
<hr/>			
conv2d_13 (Conv2D)	(None, 25, 25, 48)	12288	mixed0[0][0]
<hr/>			
conv2d_16 (Conv2D) [0][0]	(None, 25, 25, 96)	55296	activation_15
<hr/>			
batch_normalization_13 (BatchNo	(None, 25, 25, 48)	144	conv2d_13[0][0]
<hr/>			
batch_normalization_16 (BatchNo	(None, 25, 25, 96)	288	conv2d_16[0][0]
<hr/>			
activation_13 (Activation) tion_13[0][0]	(None, 25, 25, 48)	0	batch_normaliza
<hr/>			
activation_16 (Activation) tion_16[0][0]	(None, 25, 25, 96)	0	batch_normaliza
<hr/>			
average_pooling2d_1 (AveragePoo	(None, 25, 25, 256)	0	mixed0[0][0]
<hr/>			
conv2d_12 (Conv2D)	(None, 25, 25, 64)	16384	mixed0[0][0]
<hr/>			
conv2d_14 (Conv2D) [0][0]	(None, 25, 25, 64)	76800	activation_13
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conv2d_17 (Conv2D) [0][0]	(None, 25, 25, 96)	82944	activation_16
<hr/>			
conv2d_18 (Conv2D) 2d_1[0][0]	(None, 25, 25, 64)	16384	average_pooling
<hr/>			
batch_normalization_12 (BatchNo	(None, 25, 25, 64)	192	conv2d_12[0][0]

batch_normalization_14	(BatchNo	(None, 25, 25, 64)	192	conv2d_14[0][0]
batch_normalization_17	(BatchNo	(None, 25, 25, 96)	288	conv2d_17[0][0]
batch_normalization_18	(BatchNo	(None, 25, 25, 64)	192	conv2d_18[0][0]
activation_12	(Activation)	(None, 25, 25, 64)	0	batch_normaliza tion_12[0][0]
activation_14	(Activation)	(None, 25, 25, 64)	0	batch_normaliza tion_14[0][0]
activation_17	(Activation)	(None, 25, 25, 96)	0	batch_normaliza tion_17[0][0]
activation_18	(Activation)	(None, 25, 25, 64)	0	batch_normaliza tion_18[0][0]
mixed1	(Concatenate)	(None, 25, 25, 288)	0	activation_12
				activation_14
				activation_17
				activation_18
conv2d_22	(Conv2D)	(None, 25, 25, 64)	18432	mixed1[0][0]
batch_normalization_22	(BatchNo	(None, 25, 25, 64)	192	conv2d_22[0][0]
activation_22	(Activation)	(None, 25, 25, 64)	0	batch_normaliza tion_22[0][0]
conv2d_20	(Conv2D)	(None, 25, 25, 48)	13824	mixed1[0][0]
conv2d_23	(Conv2D)	(None, 25, 25, 96)	55296	activation_22
batch_normalization_20	(BatchNo	(None, 25, 25, 48)	144	conv2d_20[0][0]
batch_normalization_23	(BatchNo	(None, 25, 25, 96)	288	conv2d_23[0][0]
activation_20	(Activation)	(None, 25, 25, 48)	0	batch_normaliza tion_20[0][0]

activation_23 (Activation) tion_23[0][0]	(None, 25, 25, 96)	0	batch_normaliza
average_pooling2d_2 (AveragePoo	(None, 25, 25, 288)	0	mixed1[0][0]
conv2d_19 (Conv2D)	(None, 25, 25, 64)	18432	mixed1[0][0]
conv2d_21 (Conv2D) [0][0]	(None, 25, 25, 64)	76800	activation_20
conv2d_24 (Conv2D) [0][0]	(None, 25, 25, 96)	82944	activation_23
conv2d_25 (Conv2D) 2d_2[0][0]	(None, 25, 25, 64)	18432	average_pooling
batch_normalization_19 (BatchNo	(None, 25, 25, 64)	192	conv2d_19[0][0]
batch_normalization_21 (BatchNo	(None, 25, 25, 64)	192	conv2d_21[0][0]
batch_normalization_24 (BatchNo	(None, 25, 25, 96)	288	conv2d_24[0][0]
batch_normalization_25 (BatchNo	(None, 25, 25, 64)	192	conv2d_25[0][0]
activation_19 (Activation) tion_19[0][0]	(None, 25, 25, 64)	0	batch_normaliza
activation_21 (Activation) tion_21[0][0]	(None, 25, 25, 64)	0	batch_normaliza
activation_24 (Activation) tion_24[0][0]	(None, 25, 25, 96)	0	batch_normaliza
activation_25 (Activation) tion_25[0][0]	(None, 25, 25, 64)	0	batch_normaliza
mixed2 (Concatenate) [0][0]	(None, 25, 25, 288)	0	activation_19
[0][0]			activation_21
[0][0]			activation_24
[0][0]			activation_25
conv2d_27 (Conv2D)	(None, 25, 25, 64)	18432	mixed2[0][0]
batch_normalization_27 (BatchNo	(None, 25, 25, 64)	192	conv2d_27[0][0]

<u>activation_27</u> (Activation)	(None, 25, 25, 64)	0	batch_normaliza
tion_27[0][0]			
<u>conv2d_28</u> (Conv2D)	(None, 25, 25, 96)	55296	activation_27
[0][0]			
<u>batch_normalization_28</u> (BatchNo	(None, 25, 25, 96)	288	conv2d_28[0][0]
<u>activation_28</u> (Activation)	(None, 25, 25, 96)	0	batch_normaliza
tion_28[0][0]			
<u>conv2d_26</u> (Conv2D)	(None, 12, 12, 384)	995328	mixed2[0][0]
<u>conv2d_29</u> (Conv2D)	(None, 12, 12, 96)	82944	activation_28
[0][0]			
<u>batch_normalization_26</u> (BatchNo	(None, 12, 12, 384)	1152	conv2d_26[0][0]
<u>batch_normalization_29</u> (BatchNo	(None, 12, 12, 96)	288	conv2d_29[0][0]
<u>activation_26</u> (Activation)	(None, 12, 12, 384)	0	batch_normaliza
tion_26[0][0]			
<u>activation_29</u> (Activation)	(None, 12, 12, 96)	0	batch_normaliza
tion_29[0][0]			
<u>max_pooling2d_2</u> (MaxPooling2D)	(None, 12, 12, 288)	0	mixed2[0][0]
<u>mixed3</u> (Concatenate)	(None, 12, 12, 768)	0	activation_26
[0][0]			
			activation_29
			max_pooling2d_2
<u>conv2d_34</u> (Conv2D)	(None, 12, 12, 128)	98304	mixed3[0][0]
<u>batch_normalization_34</u> (BatchNo	(None, 12, 12, 128)	384	conv2d_34[0][0]
<u>activation_34</u> (Activation)	(None, 12, 12, 128)	0	batch_normaliza
tion_34[0][0]			
<u>conv2d_35</u> (Conv2D)	(None, 12, 12, 128)	114688	activation_34
[0][0]			
<u>batch_normalization_35</u> (BatchNo	(None, 12, 12, 128)	384	conv2d_35[0][0]

activation_35 (Activation)	(None, 12, 12, 128)	0	batch_normaliza tion_35[0][0]
conv2d_31 (Conv2D)	(None, 12, 12, 128)	98304	mixed3[0][0]
conv2d_36 (Conv2D)	(None, 12, 12, 128)	114688	activation_35 [0][0]
batch_normalization_31 (BatchNo	(None, 12, 12, 128)	384	conv2d_31[0][0]
batch_normalization_36 (BatchNo	(None, 12, 12, 128)	384	conv2d_36[0][0]
activation_31 (Activation)	(None, 12, 12, 128)	0	batch_normaliza tion_31[0][0]
activation_36 (Activation)	(None, 12, 12, 128)	0	batch_normaliza tion_36[0][0]
conv2d_32 (Conv2D)	(None, 12, 12, 128)	114688	activation_31 [0][0]
conv2d_37 (Conv2D)	(None, 12, 12, 128)	114688	activation_36 [0][0]
batch_normalization_32 (BatchNo	(None, 12, 12, 128)	384	conv2d_32[0][0]
batch_normalization_37 (BatchNo	(None, 12, 12, 128)	384	conv2d_37[0][0]
activation_32 (Activation)	(None, 12, 12, 128)	0	batch_normaliza tion_32[0][0]
activation_37 (Activation)	(None, 12, 12, 128)	0	batch_normaliza tion_37[0][0]
average_pooling2d_3 (AveragePoo	(None, 12, 12, 768)	0	mixed3[0][0]
conv2d_30 (Conv2D)	(None, 12, 12, 192)	147456	mixed3[0][0]
conv2d_33 (Conv2D)	(None, 12, 12, 192)	172032	activation_32 [0][0]
conv2d_38 (Conv2D)	(None, 12, 12, 192)	172032	activation_37 [0][0]
conv2d_39 (Conv2D)	(None, 12, 12, 192)	147456	average_pooling

2d_3[0][0]

batch_normalization_30	(BatchNo	(None, 12, 12, 192)	576	conv2d_30[0][0]
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batch_normalization_33	(BatchNo	(None, 12, 12, 192)	576	conv2d_33[0][0]
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batch_normalization_38	(BatchNo	(None, 12, 12, 192)	576	conv2d_38[0][0]
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batch_normalization_39	(BatchNo	(None, 12, 12, 192)	576	conv2d_39[0][0]
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activation_30	(Activation)	(None, 12, 12, 192)	0	batch_normalization_30[0][0]
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activation_33	(Activation)	(None, 12, 12, 192)	0	batch_normalization_33[0][0]
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activation_38	(Activation)	(None, 12, 12, 192)	0	batch_normalization_38[0][0]
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activation_39	(Activation)	(None, 12, 12, 192)	0	batch_normalization_39[0][0]
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mixed4	(Concatenate)	(None, 12, 12, 768)	0	activation_30
[0][0]				activation_33
[0][0]				activation_38
[0][0]				activation_39

conv2d_44	(Conv2D)	(None, 12, 12, 160)	122880	mixed4[0][0]
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batch_normalization_44	(BatchNo	(None, 12, 12, 160)	480	conv2d_44[0][0]
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activation_44	(Activation)	(None, 12, 12, 160)	0	batch_normalization_44[0][0]
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conv2d_45	(Conv2D)	(None, 12, 12, 160)	179200	activation_44
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batch_normalization_45	(BatchNo	(None, 12, 12, 160)	480	conv2d_45[0][0]
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activation_45	(Activation)	(None, 12, 12, 160)	0	batch_normalization_45[0][0]
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conv2d_41	(Conv2D)	(None, 12, 12, 160)	122880	mixed4[0][0]
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conv2d_46 (Conv2D) [0][0]	(None, 12, 12, 160)	179200	activation_45
batch_normalization_41 (BatchNo	(None, 12, 12, 160)	480	conv2d_41[0][0]
batch_normalization_46 (BatchNo	(None, 12, 12, 160)	480	conv2d_46[0][0]
activation_41 (Activation) activation_41[0][0]	(None, 12, 12, 160)	0	batch_normaliza tion_41[0][0]
activation_46 (Activation) activation_46[0][0]	(None, 12, 12, 160)	0	batch_normaliza tion_46[0][0]
conv2d_42 (Conv2D) [0][0]	(None, 12, 12, 160)	179200	activation_41
conv2d_47 (Conv2D) [0][0]	(None, 12, 12, 160)	179200	activation_46
batch_normalization_42 (BatchNo	(None, 12, 12, 160)	480	conv2d_42[0][0]
batch_normalization_47 (BatchNo	(None, 12, 12, 160)	480	conv2d_47[0][0]
activation_42 (Activation) activation_42[0][0]	(None, 12, 12, 160)	0	batch_normaliza tion_42[0][0]
activation_47 (Activation) activation_47[0][0]	(None, 12, 12, 160)	0	batch_normaliza tion_47[0][0]
average_pooling2d_4 (AveragePoo	(None, 12, 12, 768)	0	mixed4[0][0]
conv2d_40 (Conv2D)	(None, 12, 12, 192)	147456	mixed4[0][0]
conv2d_43 (Conv2D) [0][0]	(None, 12, 12, 192)	215040	activation_42
conv2d_48 (Conv2D) [0][0]	(None, 12, 12, 192)	215040	activation_47
conv2d_49 (Conv2D) 2d_4[0][0]	(None, 12, 12, 192)	147456	average_pooling 2d_4[0][0]
batch_normalization_40 (BatchNo	(None, 12, 12, 192)	576	conv2d_40[0][0]
batch_normalization_43 (BatchNo	(None, 12, 12, 192)	576	conv2d_43[0][0]

batch_normalization_48	(BatchNo	(None, 12, 12, 192)	576	conv2d_48[0][0]
batch_normalization_49	(BatchNo	(None, 12, 12, 192)	576	conv2d_49[0][0]
activation_40	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_40[0][0]
activation_43	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_43[0][0]
activation_48	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_48[0][0]
activation_49	(Activation)	(None, 12, 12, 192)	0	batch_normaliza tion_49[0][0]
mixed5	(Concatenate)	(None, 12, 12, 768)	0	activation_40 [0][0] activation_43 [0][0] activation_48 [0][0] activation_49 [0][0]
conv2d_54	(Conv2D)	(None, 12, 12, 160)	122880	mixed5[0][0]
batch_normalization_54	(BatchNo	(None, 12, 12, 160)	480	conv2d_54[0][0]
activation_54	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_54[0][0]
conv2d_55	(Conv2D)	(None, 12, 12, 160)	179200	activation_54 [0][0]
batch_normalization_55	(BatchNo	(None, 12, 12, 160)	480	conv2d_55[0][0]
activation_55	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_55[0][0]
conv2d_51	(Conv2D)	(None, 12, 12, 160)	122880	mixed5[0][0]
conv2d_56	(Conv2D)	(None, 12, 12, 160)	179200	activation_55 [0][0]
batch_normalization_51	(BatchNo	(None, 12, 12, 160)	480	conv2d_51[0][0]

batch_normalization_56	(BatchNo	(None, 12, 12, 160)	480	conv2d_56[0][0]
activation_51	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_51[0][0]
activation_56	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_56[0][0]
conv2d_52	(Conv2D)	(None, 12, 12, 160)	179200	activation_51 [0][0]
conv2d_57	(Conv2D)	(None, 12, 12, 160)	179200	activation_56 [0][0]
batch_normalization_52	(BatchNo	(None, 12, 12, 160)	480	conv2d_52[0][0]
batch_normalization_57	(BatchNo	(None, 12, 12, 160)	480	conv2d_57[0][0]
activation_52	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_52[0][0]
activation_57	(Activation)	(None, 12, 12, 160)	0	batch_normaliza tion_57[0][0]
average_pooling2d_5	(AveragePoo	(None, 12, 12, 768)	0	mixed5[0][0]
conv2d_50	(Conv2D)	(None, 12, 12, 192)	147456	mixed5[0][0]
conv2d_53	(Conv2D)	(None, 12, 12, 192)	215040	activation_52 [0][0]
conv2d_58	(Conv2D)	(None, 12, 12, 192)	215040	activation_57 [0][0]
conv2d_59	(Conv2D)	(None, 12, 12, 192)	147456	average_pooling 2d_5[0][0]
batch_normalization_50	(BatchNo	(None, 12, 12, 192)	576	conv2d_50[0][0]
batch_normalization_53	(BatchNo	(None, 12, 12, 192)	576	conv2d_53[0][0]
batch_normalization_58	(BatchNo	(None, 12, 12, 192)	576	conv2d_58[0][0]
batch_normalization_59	(BatchNo	(None, 12, 12, 192)	576	conv2d_59[0][0]
activation_50	(Activation)	(None, 12, 12, 192)	0	batch_normaliza

tion_50[0][0]

activation_53 (Activation) tion_53[0][0]	(None, 12, 12, 192)	0	batch_normaliza
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activation_58 (Activation) tion_58[0][0]	(None, 12, 12, 192)	0	batch_normaliza
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activation_59 (Activation) tion_59[0][0]	(None, 12, 12, 192)	0	batch_normaliza
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mixed6 (Concatenate) [0][0]	(None, 12, 12, 768)	0	activation_50
[0][0]			activation_53
[0][0]			activation_58
[0][0]			activation_59

conv2d_64 (Conv2D)	(None, 12, 12, 192)	147456	mixed6[0][0]
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batch_normalization_64 (BatchNo	(None, 12, 12, 192)	576	conv2d_64[0][0]
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activation_64 (Activation) tion_64[0][0]	(None, 12, 12, 192)	0	batch_normaliza
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conv2d_65 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_64
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batch_normalization_65 (BatchNo	(None, 12, 12, 192)	576	conv2d_65[0][0]
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activation_65 (Activation) tion_65[0][0]	(None, 12, 12, 192)	0	batch_normaliza
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conv2d_61 (Conv2D)	(None, 12, 12, 192)	147456	mixed6[0][0]
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conv2d_66 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_65
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batch_normalization_61 (BatchNo	(None, 12, 12, 192)	576	conv2d_61[0][0]
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batch_normalization_66 (BatchNo	(None, 12, 12, 192)	576	conv2d_66[0][0]
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activation_61 (Activation) tion_61[0][0]	(None, 12, 12, 192)	0	batch_normaliza
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activation_66 (Activation)	(None, 12, 12, 192)	0	batch_normaliza
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tion_66[0][0]

conv2d_62 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_61
conv2d_67 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_66
batch_normalization_62 (BatchNo	(None, 12, 12, 192)	576	conv2d_62[0][0]
batch_normalization_67 (BatchNo	(None, 12, 12, 192)	576	conv2d_67[0][0]
activation_62 (Activation) tion_62[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_67 (Activation) tion_67[0][0]	(None, 12, 12, 192)	0	batch_normaliza
average_pooling2d_6 (AveragePoo	(None, 12, 12, 768)	0	mixed6[0][0]
conv2d_60 (Conv2D)	(None, 12, 12, 192)	147456	mixed6[0][0]
conv2d_63 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_62
conv2d_68 (Conv2D) [0][0]	(None, 12, 12, 192)	258048	activation_67
conv2d_69 (Conv2D) 2d_6[0][0]	(None, 12, 12, 192)	147456	average_pooling
batch_normalization_60 (BatchNo	(None, 12, 12, 192)	576	conv2d_60[0][0]
batch_normalization_63 (BatchNo	(None, 12, 12, 192)	576	conv2d_63[0][0]
batch_normalization_68 (BatchNo	(None, 12, 12, 192)	576	conv2d_68[0][0]
batch_normalization_69 (BatchNo	(None, 12, 12, 192)	576	conv2d_69[0][0]
activation_60 (Activation) tion_60[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_63 (Activation) tion_63[0][0]	(None, 12, 12, 192)	0	batch_normaliza
activation_68 (Activation)	(None, 12, 12, 192)	0	batch_normaliza

tion_68[0][0]

activation_69 (Activation)	(None, 12, 12, 192)	0	batch_normaliza
tion_69[0][0]			

mixed7 (Concatenate)	(None, 12, 12, 768)	0	activation_60
[0][0]			activation_63
[0][0]			activation_68
[0][0]			activation_69
[0][0]			

conv2d_72 (Conv2D)	(None, 12, 12, 192)	147456	mixed7[0][0]
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batch_normalization_72 (BatchNo	(None, 12, 12, 192)	576	conv2d_72[0][0]
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activation_72 (Activation)	(None, 12, 12, 192)	0	batch_normaliza
tion_72[0][0]			

conv2d_73 (Conv2D)	(None, 12, 12, 192)	258048	activation_72
[0][0]			

batch_normalization_73 (BatchNo	(None, 12, 12, 192)	576	conv2d_73[0][0]
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activation_73 (Activation)	(None, 12, 12, 192)	0	batch_normaliza
tion_73[0][0]			

conv2d_70 (Conv2D)	(None, 12, 12, 192)	147456	mixed7[0][0]
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conv2d_74 (Conv2D)	(None, 12, 12, 192)	258048	activation_73
[0][0]			

batch_normalization_70 (BatchNo	(None, 12, 12, 192)	576	conv2d_70[0][0]
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batch_normalization_74 (BatchNo	(None, 12, 12, 192)	576	conv2d_74[0][0]
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activation_70 (Activation)	(None, 12, 12, 192)	0	batch_normaliza
tion_70[0][0]			

activation_74 (Activation)	(None, 12, 12, 192)	0	batch_normaliza
tion_74[0][0]			

conv2d_71 (Conv2D)	(None, 5, 5, 320)	552960	activation_70
[0][0]			

conv2d_75 (Conv2D)	(None, 5, 5, 192)	331776	activation_74
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[0][0]

batch_normalization_71	(BatchNo (None, 5, 5, 320)	960	conv2d_71[0][0]
batch_normalization_75	(BatchNo (None, 5, 5, 192)	576	conv2d_75[0][0]
activation_71	(Activation) (None, 5, 5, 320)	0	batch_normaliza tion_71[0][0]
activation_75	(Activation) (None, 5, 5, 192)	0	batch_normaliza tion_75[0][0]
max_pooling2d_3	(MaxPooling2D) (None, 5, 5, 768)	0	mixed7[0][0]
mixed8	(Concatenate) (None, 5, 5, 1280)	0	activation_71 activation_75 max_pooling2d_3
conv2d_80	(Conv2D) (None, 5, 5, 448)	573440	mixed8[0][0]
batch_normalization_80	(BatchNo (None, 5, 5, 448)	1344	conv2d_80[0][0]
activation_80	(Activation) (None, 5, 5, 448)	0	batch_normaliza tion_80[0][0]
conv2d_77	(Conv2D) (None, 5, 5, 384)	491520	mixed8[0][0]
conv2d_81	(Conv2D) (None, 5, 5, 384)	1548288	activation_80
batch_normalization_77	(BatchNo (None, 5, 5, 384)	1152	conv2d_77[0][0]
batch_normalization_81	(BatchNo (None, 5, 5, 384)	1152	conv2d_81[0][0]
activation_77	(Activation) (None, 5, 5, 384)	0	batch_normaliza tion_77[0][0]
activation_81	(Activation) (None, 5, 5, 384)	0	batch_normaliza tion_81[0][0]
conv2d_78	(Conv2D) (None, 5, 5, 384)	442368	activation_77
conv2d_79	(Conv2D) (None, 5, 5, 384)	442368	activation_77

[0][0]

conv2d_82 (Conv2D) [0][0]	(None, 5, 5, 384)	442368	activation_81
conv2d_83 (Conv2D) [0][0]	(None, 5, 5, 384)	442368	activation_81
average_pooling2d_7 (AveragePool)	(None, 5, 5, 1280)	0	mixed8[0][0]
conv2d_76 (Conv2D)	(None, 5, 5, 320)	409600	mixed8[0][0]
batch_normalization_78 (Batch Normalization)	(None, 5, 5, 384)	1152	conv2d_78[0][0]
batch_normalization_79 (Batch Normalization)	(None, 5, 5, 384)	1152	conv2d_79[0][0]
batch_normalization_82 (Batch Normalization)	(None, 5, 5, 384)	1152	conv2d_82[0][0]
batch_normalization_83 (Batch Normalization)	(None, 5, 5, 384)	1152	conv2d_83[0][0]
conv2d_84 (Conv2D) 2d_7[0][0]	(None, 5, 5, 192)	245760	average_pooling
batch_normalization_76 (Batch Normalization)	(None, 5, 5, 320)	960	conv2d_76[0][0]
activation_78 (Activation) tion_78[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_79 (Activation) tion_79[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_82 (Activation) tion_82[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_83 (Activation) tion_83[0][0]	(None, 5, 5, 384)	0	batch_normaliza
batch_normalization_84 (Batch Normalization)	(None, 5, 5, 192)	576	conv2d_84[0][0]
activation_76 (Activation) tion_76[0][0]	(None, 5, 5, 320)	0	batch_normaliza
mixed9_0 (Concatenate) [0][0]	(None, 5, 5, 768)	0	activation_78
[0][0]			activation_79

concatenate (Concatenate) [0][0]	(None, 5, 5, 768)	0	activation_82
[0][0]			activation_83
activation_84 (Activation) activation_84[0][0]	(None, 5, 5, 192)	0	batch_normaliza tion_84[0][0]
mixed9 (Concatenate) [0][0]	(None, 5, 5, 2048)	0	activation_76
[0][0]			mixed9_0[0][0] concatenate
[0][0]			activation_84
conv2d_89 (Conv2D)	(None, 5, 5, 448)	917504	mixed9[0][0]
batch_normalization_89 (BatchNo	(None, 5, 5, 448)	1344	conv2d_89[0][0]
activation_89 (Activation) activation_89[0][0]	(None, 5, 5, 448)	0	batch_normaliza tion_89[0][0]
conv2d_86 (Conv2D)	(None, 5, 5, 384)	786432	mixed9[0][0]
conv2d_90 (Conv2D) [0][0]	(None, 5, 5, 384)	1548288	activation_89
batch_normalization_86 (BatchNo	(None, 5, 5, 384)	1152	conv2d_86[0][0]
batch_normalization_90 (BatchNo	(None, 5, 5, 384)	1152	conv2d_90[0][0]
activation_86 (Activation) activation_86[0][0]	(None, 5, 5, 384)	0	batch_normaliza tion_86[0][0]
activation_90 (Activation) activation_90[0][0]	(None, 5, 5, 384)	0	batch_normaliza tion_90[0][0]
conv2d_87 (Conv2D) [0][0]	(None, 5, 5, 384)	442368	activation_86
conv2d_88 (Conv2D) [0][0]	(None, 5, 5, 384)	442368	activation_86
conv2d_91 (Conv2D) [0][0]	(None, 5, 5, 384)	442368	activation_90
conv2d_92 (Conv2D)	(None, 5, 5, 384)	442368	activation_90

[0][0]

average_pooling2d_8 (AveragePoo	(None, 5, 5, 2048)	0	mixed9[0][0]
conv2d_85 (Conv2D)	(None, 5, 5, 320)	655360	mixed9[0][0]
batch_normalization_87 (BatchNo	(None, 5, 5, 384)	1152	conv2d_87[0][0]
batch_normalization_88 (BatchNo	(None, 5, 5, 384)	1152	conv2d_88[0][0]
batch_normalization_91 (BatchNo	(None, 5, 5, 384)	1152	conv2d_91[0][0]
batch_normalization_92 (BatchNo	(None, 5, 5, 384)	1152	conv2d_92[0][0]
conv2d_93 (Conv2D) 2d_8[0][0]	(None, 5, 5, 192)	393216	average_pooling
batch_normalization_85 (BatchNo	(None, 5, 5, 320)	960	conv2d_85[0][0]
activation_87 (Activation) tion_87[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_88 (Activation) tion_88[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_91 (Activation) tion_91[0][0]	(None, 5, 5, 384)	0	batch_normaliza
activation_92 (Activation) tion_92[0][0]	(None, 5, 5, 384)	0	batch_normaliza
batch_normalization_93 (BatchNo	(None, 5, 5, 192)	576	conv2d_93[0][0]
activation_85 (Activation) tion_85[0][0]	(None, 5, 5, 320)	0	batch_normaliza
mixed9_1 (Concatenate) [0][0]	(None, 5, 5, 768)	0	activation_87
			activation_88
			[0][0]
concatenate_1 (Concatenate) [0][0]	(None, 5, 5, 768)	0	activation_91
			activation_92
			[0][0]
activation_93 (Activation)	(None, 5, 5, 192)	0	batch_normaliza

```
tion_93[0][0]
```

```
mixed10 (Concatenate)
[0][0]
```

```
(None, 5, 5, 2048) 0
```

```
activation_85
```

```
mixed9_1[0][0]
concatenate_1
```

```
[0][0]
```

```
activation_93
```

```
[0][0]
```

```
In [12]: for layer in custom_resnet_model.layers[:-1]:
         layer.trainable = False
```

```
custom_resnet_model.layers[-1].trainable
```

```
Out[12]: True
```

```
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
147/147 [=====] - 11s 72ms/step - loss: 1.0208 - accuracy: 0.8461 - val_loss: 0.4900 - val_accuracy: 0.8993
Epoch 2/100
147/147 [=====] - 8s 58ms/step - loss: 0.6434 - accuracy: 0.8824 - val_loss: 0.7102 - val_accuracy: 0.8754
Epoch 3/100
147/147 [=====] - 9s 58ms/step - loss: 0.4469 - accuracy: 0.9084 - val_loss: 0.3235 - val_accuracy: 0.9121
Epoch 4/100
147/147 [=====] - 8s 58ms/step - loss: 0.3187 - accuracy: 0.9155 - val_loss: 0.3676 - val_accuracy: 0.9172
Epoch 5/100
147/147 [=====] - 8s 58ms/step - loss: 0.4683 - accuracy: 0.8999 - val_loss: 1.7047 - val_accuracy: 0.8362
Epoch 6/100
147/147 [=====] - 9s 58ms/step - loss: 0.4260 - accuracy: 0.9187 - val_loss: 0.3735 - val_accuracy: 0.9215
Epoch 7/100
147/147 [=====] - 9s 58ms/step - loss: 0.3582 - accuracy: 0.9214 - val_loss: 0.3865 - val_accuracy: 0.9172
Epoch 8/100
147/147 [=====] - 9s 58ms/step - loss: 0.3495 - accuracy: 0.9204 - val_loss: 0.7358 - val_accuracy: 0.8968
Epoch 9/100
147/147 [=====] - 9s 58ms/step - loss: 0.3342 - accuracy: 0.9263 - val_loss: 0.4181 - val_accuracy: 0.9189
Epoch 10/100
147/147 [=====] - 9s 58ms/step - loss: 0.2527 - accuracy: 0.9360 - val_loss: 0.3568 - val_accuracy: 0.9224
Epoch 11/100
147/147 [=====] - 9s 58ms/step - loss: 0.2203 - accuracy: 0.9402 - val_loss: 0.3575 - val_accuracy: 0.9164
Epoch 12/100
147/147 [=====] - 9s 58ms/step - loss: 0.2663 - accuracy: 0.9283 - val_loss: 0.4710 - val_accuracy: 0.8993
Epoch 13/100
147/147 [=====] - 9s 58ms/step - loss: 0.3595 - accuracy: 0.9281 - val_loss: 0.5862 - val_accuracy: 0.9036
Epoch 14/100
147/147 [=====] - 9s 58ms/step - loss: 0.3339 - accuracy: 0.9268 - val_loss: 0.3374 - val_accuracy: 0.9343
Epoch 15/100
147/147 [=====] - 9s 58ms/step - loss: 0.2060 - accuracy: 0.9441 - val_loss: 0.3339 - val_accuracy: 0.9275
Epoch 16/100
147/147 [=====] - 9s 58ms/step - loss: 0.1878 - accuracy: 0.9458 - val_loss: 0.3770 - val_accuracy: 0.9181
Epoch 17/100
147/147 [=====] - 9s 58ms/step - loss: 0.1934 - accuracy: 0.9466 - val_loss: 0.6175 - val_accuracy: 0.8814
Epoch 18/100
147/147 [=====] - 9s 58ms/step - loss: 0.3152 - accuracy: 0.9310 - val_loss: 0.5921 - val_accuracy: 0.8925
Epoch 19/100
147/147 [=====] - 9s 58ms/step - loss: 0.2916 - accuracy: 0.9353 - val_loss: 0.9115 - val_accuracy: 0.8626
Epoch 20/100
147/147 [=====] - 9s 58ms/step - loss: 0.3590 - accuracy: 0.9255 - val_loss: 0.5896 - val_accuracy: 0.9027
Epoch 21/100
147/147 [=====] - 9s 58ms/step - loss: 0.3249 - accuracy: 0.9327 - val_loss: 0.7492 - val_accuracy: 0.9027
```

```
Epoch 22/100
147/147 [=====] - 9s 58ms/step - loss: 0.2257 - accurac
y: 0.9466 - val_loss: 0.4125 - val_accuracy: 0.9189
Epoch 23/100
147/147 [=====] - 9s 58ms/step - loss: 0.1791 - accurac
y: 0.9494 - val_loss: 1.2673 - val_accuracy: 0.7986
Epoch 24/100
147/147 [=====] - 9s 58ms/step - loss: 0.1686 - accurac
y: 0.9547 - val_loss: 0.3550 - val_accuracy: 0.9198
Epoch 25/100
147/147 [=====] - 9s 58ms/step - loss: 0.2954 - accurac
y: 0.9323 - val_loss: 0.3836 - val_accuracy: 0.9386
Epoch 26/100
147/147 [=====] - 9s 58ms/step - loss: 0.2097 - accurac
y: 0.9445 - val_loss: 0.5146 - val_accuracy: 0.9198
Epoch 27/100
147/147 [=====] - 9s 58ms/step - loss: 0.2833 - accurac
y: 0.9381 - val_loss: 0.4192 - val_accuracy: 0.9249
Epoch 28/100
147/147 [=====] - 9s 58ms/step - loss: 0.2050 - accurac
y: 0.9507 - val_loss: 0.4302 - val_accuracy: 0.9275
Epoch 29/100
147/147 [=====] - 9s 58ms/step - loss: 0.1999 - accurac
y: 0.9515 - val_loss: 0.5634 - val_accuracy: 0.9096
Epoch 30/100
147/147 [=====] - 9s 58ms/step - loss: 0.2181 - accurac
y: 0.9524 - val_loss: 0.5809 - val_accuracy: 0.9027
Epoch 31/100
147/147 [=====] - 9s 58ms/step - loss: 0.2148 - accurac
y: 0.9434 - val_loss: 0.4878 - val_accuracy: 0.9224
Epoch 32/100
147/147 [=====] - 9s 58ms/step - loss: 0.1840 - accurac
y: 0.9503 - val_loss: 0.4427 - val_accuracy: 0.9275
Epoch 33/100
147/147 [=====] - 9s 58ms/step - loss: 0.2038 - accurac
y: 0.9451 - val_loss: 1.0047 - val_accuracy: 0.8899
Epoch 34/100
147/147 [=====] - 9s 58ms/step - loss: 0.2110 - accurac
y: 0.9503 - val_loss: 0.4074 - val_accuracy: 0.9283
Epoch 35/100
147/147 [=====] - 9s 58ms/step - loss: 0.1570 - accurac
y: 0.9539 - val_loss: 0.4282 - val_accuracy: 0.9275
Epoch 36/100
147/147 [=====] - 9s 58ms/step - loss: 0.1780 - accurac
y: 0.9528 - val_loss: 0.4919 - val_accuracy: 0.9258
Epoch 37/100
147/147 [=====] - 9s 59ms/step - loss: 0.1496 - accurac
y: 0.9594 - val_loss: 0.3967 - val_accuracy: 0.9334
Epoch 38/100
147/147 [=====] - 9s 58ms/step - loss: 0.1540 - accurac
y: 0.9573 - val_loss: 0.4740 - val_accuracy: 0.9198
Epoch 39/100
147/147 [=====] - 9s 58ms/step - loss: 0.1406 - accurac
y: 0.9556 - val_loss: 0.4325 - val_accuracy: 0.9300
Epoch 40/100
147/147 [=====] - 9s 58ms/step - loss: 0.1090 - accurac
y: 0.9667 - val_loss: 0.4443 - val_accuracy: 0.9206
Epoch 41/100
147/147 [=====] - 9s 58ms/step - loss: 0.1838 - accurac
y: 0.9513 - val_loss: 0.5371 - val_accuracy: 0.9189
Epoch 42/100
147/147 [=====] - 9s 58ms/step - loss: 0.1798 - accurac
y: 0.9543 - val_loss: 0.4792 - val_accuracy: 0.9275
Epoch 43/100
```

```
147/147 [=====] - 9s 58ms/step - loss: 0.1250 - accurac
y: 0.9635 - val_loss: 0.4698 - val_accuracy: 0.9198
Epoch 44/100
147/147 [=====] - 9s 58ms/step - loss: 0.1898 - accurac
y: 0.9518 - val_loss: 0.4537 - val_accuracy: 0.9266
Epoch 45/100
147/147 [=====] - 9s 58ms/step - loss: 0.1291 - accurac
y: 0.9639 - val_loss: 0.5048 - val_accuracy: 0.9181
Epoch 46/100
147/147 [=====] - 9s 58ms/step - loss: 0.1098 - accurac
y: 0.9667 - val_loss: 0.4585 - val_accuracy: 0.9258
Epoch 47/100
147/147 [=====] - 9s 58ms/step - loss: 0.4061 - accurac
y: 0.9300 - val_loss: 0.5552 - val_accuracy: 0.9309
Epoch 48/100
147/147 [=====] - 9s 58ms/step - loss: 0.1978 - accurac
y: 0.9513 - val_loss: 0.5558 - val_accuracy: 0.9309
Epoch 49/100
147/147 [=====] - 9s 58ms/step - loss: 0.1393 - accurac
y: 0.9611 - val_loss: 0.4993 - val_accuracy: 0.9326
Epoch 50/100
147/147 [=====] - 9s 58ms/step - loss: 0.2257 - accurac
y: 0.9503 - val_loss: 0.6781 - val_accuracy: 0.9138
Epoch 51/100
147/147 [=====] - 9s 58ms/step - loss: 0.1340 - accurac
y: 0.9622 - val_loss: 0.6502 - val_accuracy: 0.9224
Epoch 52/100
147/147 [=====] - 9s 58ms/step - loss: 0.1399 - accurac
y: 0.9614 - val_loss: 0.5504 - val_accuracy: 0.9189
Epoch 53/100
147/147 [=====] - 9s 58ms/step - loss: 0.1438 - accurac
y: 0.9614 - val_loss: 0.5303 - val_accuracy: 0.9224
Epoch 54/100
147/147 [=====] - 9s 58ms/step - loss: 0.1315 - accurac
y: 0.9637 - val_loss: 0.4481 - val_accuracy: 0.9258
Epoch 55/100
147/147 [=====] - 9s 58ms/step - loss: 0.1471 - accurac
y: 0.9603 - val_loss: 0.4904 - val_accuracy: 0.9369
Epoch 56/100
147/147 [=====] - 9s 58ms/step - loss: 0.1115 - accurac
y: 0.9663 - val_loss: 0.5338 - val_accuracy: 0.9206
Epoch 57/100
147/147 [=====] - 9s 58ms/step - loss: 0.1683 - accurac
y: 0.9592 - val_loss: 0.5840 - val_accuracy: 0.9155
Epoch 58/100
147/147 [=====] - 9s 58ms/step - loss: 0.1663 - accurac
y: 0.9567 - val_loss: 0.7930 - val_accuracy: 0.9019
Epoch 59/100
147/147 [=====] - 9s 58ms/step - loss: 0.1136 - accurac
y: 0.9669 - val_loss: 0.7413 - val_accuracy: 0.8891
Epoch 60/100
147/147 [=====] - 9s 58ms/step - loss: 0.1238 - accurac
y: 0.9616 - val_loss: 0.5597 - val_accuracy: 0.9283
Epoch 61/100
147/147 [=====] - 9s 58ms/step - loss: 0.1638 - accurac
y: 0.9562 - val_loss: 0.8839 - val_accuracy: 0.8848
Epoch 62/100
147/147 [=====] - 9s 58ms/step - loss: 0.0959 - accurac
y: 0.9720 - val_loss: 0.6772 - val_accuracy: 0.9053
Epoch 63/100
147/147 [=====] - 9s 58ms/step - loss: 0.1185 - accurac
y: 0.9669 - val_loss: 1.1789 - val_accuracy: 0.8387
Epoch 64/100
147/147 [=====] - 9s 58ms/step - loss: 0.1434 - accurac
```



```
y: 0.9631 - val_loss: 0.6470 - val_accuracy: 0.9121
Epoch 65/100
147/147 [=====] - 9s 58ms/step - loss: 0.0817 - accurac
y: 0.9752 - val_loss: 0.5366 - val_accuracy: 0.9198
Epoch 66/100
147/147 [=====] - 9s 58ms/step - loss: 0.1093 - accurac
y: 0.9686 - val_loss: 0.5944 - val_accuracy: 0.9292
Epoch 67/100
147/147 [=====] - 9s 58ms/step - loss: 0.1161 - accurac
y: 0.9684 - val_loss: 0.8564 - val_accuracy: 0.8720
Epoch 68/100
147/147 [=====] - 9s 58ms/step - loss: 0.1014 - accurac
y: 0.9708 - val_loss: 0.7628 - val_accuracy: 0.9130
Epoch 69/100
147/147 [=====] - 9s 58ms/step - loss: 0.2305 - accurac
y: 0.9518 - val_loss: 0.7955 - val_accuracy: 0.8985
Epoch 70/100
147/147 [=====] - 9s 58ms/step - loss: 0.1172 - accurac
y: 0.9680 - val_loss: 0.5821 - val_accuracy: 0.9215
Epoch 71/100
147/147 [=====] - 9s 58ms/step - loss: 0.1571 - accurac
y: 0.9609 - val_loss: 0.7850 - val_accuracy: 0.9036
Epoch 72/100
147/147 [=====] - 9s 58ms/step - loss: 0.1807 - accurac
y: 0.9571 - val_loss: 1.1599 - val_accuracy: 0.8567
Epoch 73/100
147/147 [=====] - 9s 58ms/step - loss: 0.1128 - accurac
y: 0.9661 - val_loss: 0.6321 - val_accuracy: 0.9258
Epoch 74/100
147/147 [=====] - 9s 58ms/step - loss: 0.0692 - accurac
y: 0.9767 - val_loss: 0.5624 - val_accuracy: 0.9317
Epoch 75/100
147/147 [=====] - 9s 58ms/step - loss: 0.1054 - accurac
y: 0.9686 - val_loss: 0.5877 - val_accuracy: 0.9249
Epoch 76/100
147/147 [=====] - 9s 58ms/step - loss: 0.1229 - accurac
y: 0.9652 - val_loss: 0.5991 - val_accuracy: 0.9241
Epoch 77/100
147/147 [=====] - 9s 58ms/step - loss: 0.1235 - accurac
y: 0.9678 - val_loss: 0.5696 - val_accuracy: 0.9224
Epoch 78/100
147/147 [=====] - 9s 58ms/step - loss: 0.1132 - accurac
y: 0.9703 - val_loss: 0.6181 - val_accuracy: 0.9258
Epoch 79/100
147/147 [=====] - 9s 58ms/step - loss: 0.2153 - accurac
y: 0.9550 - val_loss: 0.6405 - val_accuracy: 0.9266
Epoch 80/100
147/147 [=====] - 9s 58ms/step - loss: 0.1074 - accurac
y: 0.9658 - val_loss: 0.5567 - val_accuracy: 0.9249
Epoch 81/100
147/147 [=====] - 9s 58ms/step - loss: 0.1743 - accurac
y: 0.9599 - val_loss: 0.6895 - val_accuracy: 0.9258
Epoch 82/100
147/147 [=====] - 9s 58ms/step - loss: 0.0895 - accurac
y: 0.9740 - val_loss: 0.5883 - val_accuracy: 0.9224
Epoch 83/100
147/147 [=====] - 9s 58ms/step - loss: 0.0770 - accurac
y: 0.9750 - val_loss: 0.6008 - val_accuracy: 0.9215
Epoch 84/100
147/147 [=====] - 9s 58ms/step - loss: 0.0511 - accurac
y: 0.9836 - val_loss: 0.9341 - val_accuracy: 0.8908
Epoch 85/100
147/147 [=====] - 9s 58ms/step - loss: 0.1628 - accurac
y: 0.9639 - val_loss: 0.6709 - val_accuracy: 0.9155
```

```

Epoch 86/100
147/147 [=====] - 9s 58ms/step - loss: 0.0954 - accurac
y: 0.9727 - val_loss: 0.5795 - val_accuracy: 0.9249
Epoch 87/100
147/147 [=====] - 9s 58ms/step - loss: 0.1080 - accurac
y: 0.9671 - val_loss: 0.9625 - val_accuracy: 0.8942
Epoch 88/100
147/147 [=====] - 9s 58ms/step - loss: 0.1346 - accurac
y: 0.9633 - val_loss: 0.6700 - val_accuracy: 0.9164
Epoch 89/100
147/147 [=====] - 9s 58ms/step - loss: 0.1471 - accurac
y: 0.9665 - val_loss: 0.5886 - val_accuracy: 0.9275
Epoch 90/100
147/147 [=====] - 9s 58ms/step - loss: 0.1547 - accurac
y: 0.9616 - val_loss: 0.7518 - val_accuracy: 0.9130
Epoch 91/100
147/147 [=====] - 9s 58ms/step - loss: 0.0603 - accurac
y: 0.9791 - val_loss: 0.6492 - val_accuracy: 0.9232
Epoch 92/100
147/147 [=====] - 9s 58ms/step - loss: 0.0439 - accurac
y: 0.9829 - val_loss: 0.6758 - val_accuracy: 0.9172
Epoch 93/100
147/147 [=====] - 9s 58ms/step - loss: 0.0618 - accurac
y: 0.9778 - val_loss: 0.6672 - val_accuracy: 0.9147
Epoch 94/100
147/147 [=====] - 9s 58ms/step - loss: 0.2206 - accurac
y: 0.9554 - val_loss: 0.8791 - val_accuracy: 0.9053
Epoch 95/100
147/147 [=====] - 9s 58ms/step - loss: 0.0679 - accurac
y: 0.9780 - val_loss: 0.6570 - val_accuracy: 0.9172
Epoch 96/100
147/147 [=====] - 9s 58ms/step - loss: 0.0579 - accurac
y: 0.9812 - val_loss: 0.6256 - val_accuracy: 0.9275
Epoch 97/100
147/147 [=====] - 9s 58ms/step - loss: 0.0550 - accurac
y: 0.9793 - val_loss: 0.6989 - val_accuracy: 0.9241
Epoch 98/100
147/147 [=====] - 9s 58ms/step - loss: 0.1363 - accurac
y: 0.9654 - val_loss: 0.7348 - val_accuracy: 0.9113
Epoch 99/100
147/147 [=====] - 9s 58ms/step - loss: 0.0883 - accurac
y: 0.9752 - val_loss: 0.7566 - val_accuracy: 0.9147
Epoch 100/100
147/147 [=====] - 9s 58ms/step - loss: 0.1449 - accurac
y: 0.9665 - val_loss: 0.9426 - val_accuracy: 0.9044
Training time: -868.8096132278442
118/118 [=====] - 4s 30ms/step - loss: 0.9426 - accurac
y: 0.9044

```

```
In [15]: (loss, accuracy) = custom_resnet_model.evaluate(X_test, y_test, batch_size=10, verbose=1)
```

```
print("[INFO] loss={:.4f}, accuracy: {:.4f}%".format(loss, accuracy * 100))
```

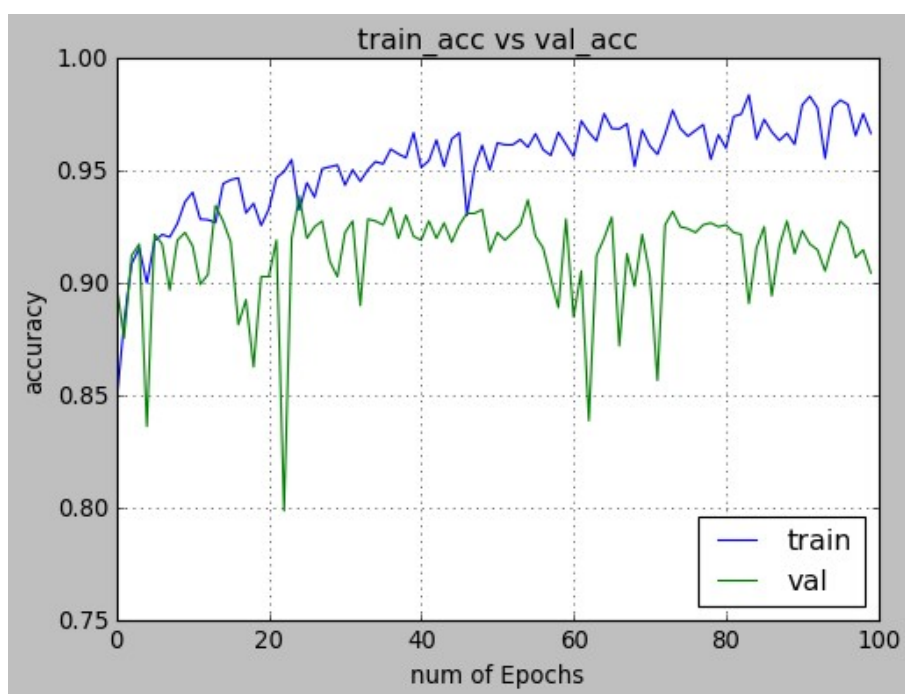
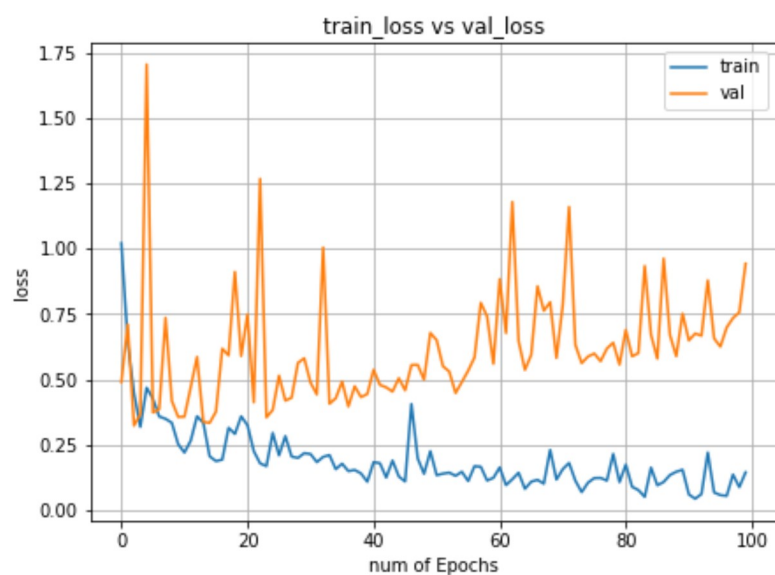
```

118/118 [=====] - 3s 25ms/step - loss: 0.9426 - accurac
y: 0.9044
[INFO] loss=0.9426, accuracy: 90.4437%

```

visualizing losses and accuracy

```
In [16]: display_loss_accuracy(hist)
```



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])
```

[illegible]

[illegible]

[illegible]

[illegible]


```
-----  
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_bacteria_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)
#     else:
#         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)
[[7.9543006e-20 1.0000000e+00]]
```

```
-----
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))
    152     if CLASS_INDEX is None:
    153         fpath = data_utils.get_file(

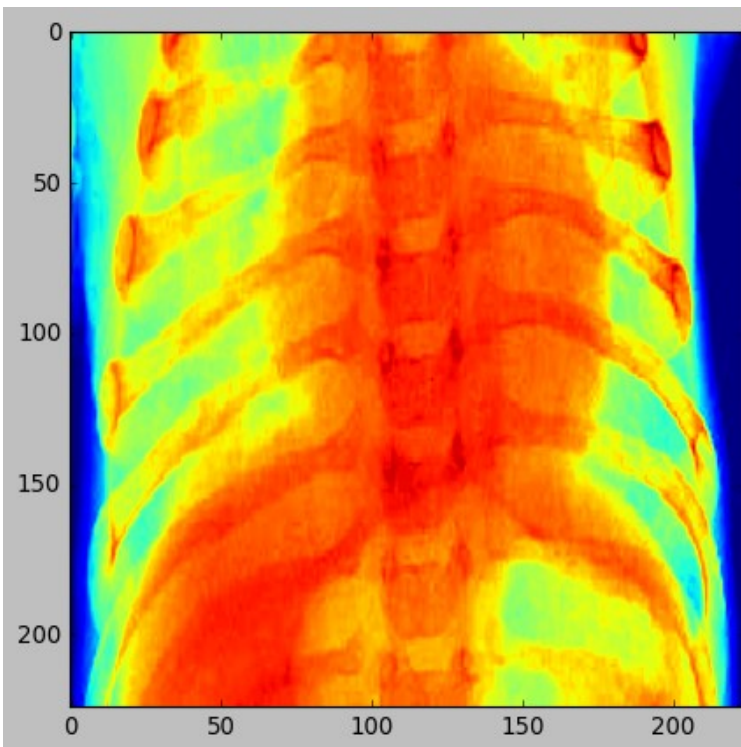
ValueError: `decode_predictions` expects a batch of predictions (i.e. a 2D array
of shape (samples, 1000)). Found array with shape: (1, 2)
```

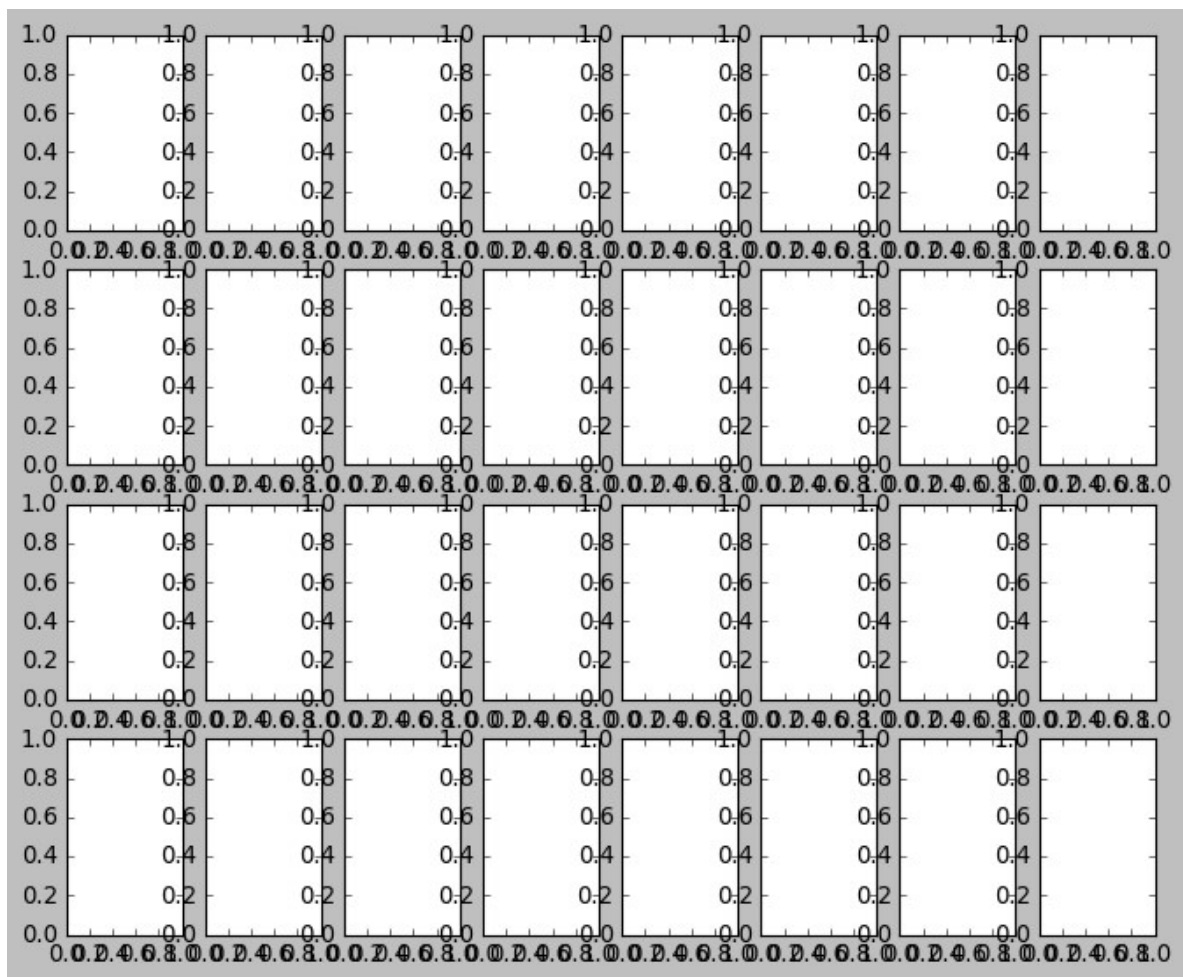
Visualizing the intermediate layer

```
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_names))
print(confusion_matrix(np.argmax(y_test,axis=1), y_pred))
```

```
[[2.1685484e-10 1.0000000e+00]
 [9.9941623e-01 5.8379851e-04]
 [2.1206019e-14 1.0000000e+00]
 ...
 [3.4830153e-22 1.0000000e+00]
 [1.7426691e-19 1.0000000e+00]
 [8.4951813e-07 9.9999917e-01]]
[1 0 1 ... 1 1 1]
```

	precision	recall	f1-score	support
class 0(Normal)	0.93	0.70	0.79	312
class 1(Pneumonia)	0.90	0.98	0.94	860
accuracy			0.90	1172
macro avg	0.91	0.84	0.87	1172
weighted avg	0.91	0.90	0.90	1172

```
[[217 95]
 [ 17 843]]
```

Compute confusion matrix

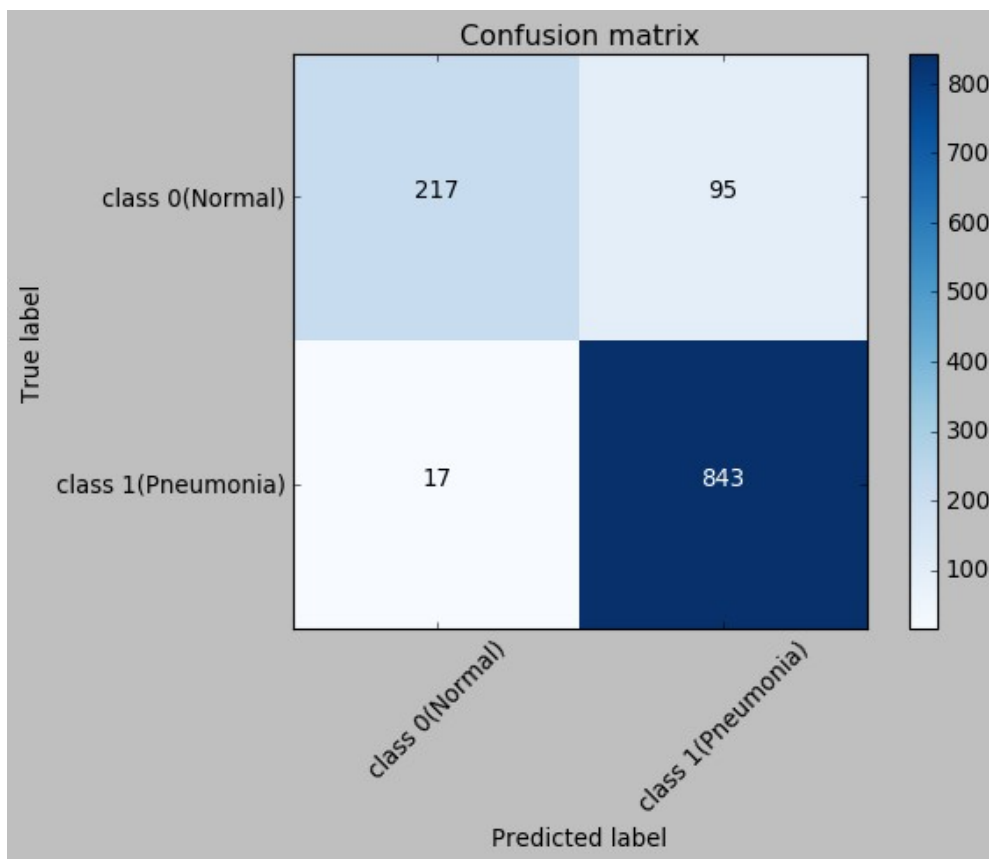
```
In [21]: # Compute confusion matrix
cnf_matrix = (confusion_matrix(np.argmax(y_test,axis=1), y_pred))

np.set_printoptions(precision=2)

# Plot non-normalized confusion matrix
plot_confusion_matrix(cnf_matrix, classes=target_names,
                      title='Confusion matrix')
```

Confusion matrix, without normalization

```
[[217  95]
 [ 17 843]]
```



In []:

In []: