

Import libraries

```
In [19]: import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
import os, cv2
from skimage.transform import resize
from sklearn.metrics import confusion_matrix, roc_curve, auc
import math
import smtplib
import pickle
import itertools

from keras.models import Sequential, Model, load_model
from keras.layers import Dense, Dropout, Flatten, GlobalAveragePooling2D
from keras.layers import Conv2D, MaxPooling2D, BatchNormalization
from keras.layers import Lambda, concatenate
from keras import backend as K
#from keras.utils import multi_gpu_model
from keras import applications
from keras import optimizers
from keras.callbacks import ModelCheckpoint, LearningRateScheduler, TensorBoard, EarlyStopping
from keras.preprocessing.image import ImageDataGenerator

print(K.tensorflow_backend._get_available_gpus())

['/job:localhost/replica:0/task:0/device:GPU:0', '/job:localhost/replica:0/task:0/device:GPU:1', '/job:localhost/replica:0/task:0/device:GPU:2', '/job:localhost/replica:0/task:0/device:GPU:3', '/job:localhost/replica:0/task:0/device:GPU:4', '/job:localhost/replica:0/task:0/device:GPU:5', '/job:localhost/replica:0/task:0/device:GPU:6', '/job:localhost/replica:0/task:0/device:GPU:7']
```

Patched multi_gpu_model function

This is needed to work around saving errors with parallel models. See

<https://stackoverflow.com/questions/47210811/can-not-save-model-using-model-save-following-multi-gpu-model-in-keras/48066771#48066771> (<https://stackoverflow.com/questions/47210811/can-not-save-model-using-model-save-following-multi-gpu-model-in-keras/48066771#48066771>). Note that loading the model is slightly different, as described in the link.

```

In [2]: def multi_gpu_model(model, gpus):
    if isinstance(gpus, (list, tuple)):
        num_gpus = len(gpus)
        target_gpu_ids = gpus
    else:
        num_gpus = gpus
        target_gpu_ids = range(num_gpus)

    def get_slice(data, i, parts):
        shape = tf.shape(data)
        batch_size = shape[0]
        input_shape = shape[1:]
        step = batch_size // parts
        if i == num_gpus - 1:
            size = batch_size - step * i
        else:
            size = step
        size = tf.concat([size, input_shape], axis=0)
        stride = tf.concat([step, input_shape * 0], axis=0)
        start = stride * i
        return tf.slice(data, start, size)

    all_outputs = []
    for i in range(len(model.outputs)):
        all_outputs.append([])

    # Place a copy of the model on each GPU,
    # each getting a slice of the inputs.
    for i, gpu_id in enumerate(target_gpu_ids):
        with tf.device('/gpu:%d' % gpu_id):
            with tf.name_scope('replica_%d' % gpu_id):
                inputs = []
                # Retrieve a slice of the input.
                for x in model.inputs:
                    input_shape = tuple(x.get_shape().as_list())[1:]
                    slice_i = Lambda(get_slice,
                                     output_shape=input_shape,
                                     arguments={'i': i,
                                                'parts': num_gpus})(x)
                    inputs.append(slice_i)

                # Apply model on slice
                # (creating a model replica on the target device).
                outputs = model(inputs)
                if not isinstance(outputs, list):
                    outputs = [outputs]

                # Save the outputs for merging back together later.
                for o in range(len(outputs)):
                    all_outputs[o].append(outputs[o])

    # Merge outputs on CPU.
    with tf.device('/cpu:0'):
        merged = []
        for name, outputs in zip(model.output_names, all_outputs):
            merged.append(concatenate(outputs,

```

```
axis=0, name=name))  
return Model(model.inputs, merged)
```

Train a VGGnet on image data

Use VGG19 pretrained on ImageNet to train on retinal OCT dataset using transfer learning, data augmentation

```
In [36]: # Constants  
img_width, img_height = 256, 256  
train_data_dir = "data/train"  
validation_data_dir = "data/val"  
test_data_dir = "data/test"  
nb_train_samples = 66813  
nb_validation_samples = 16703  
batch_size = 128  
epochs = 50  
learning_rate = 0.001  
mu = 0.9 # Momentum
```

```
In [17]: # Load pretrained model  
model = applications.VGG19(weights = "imagenet", include_top=False, input_shape = (img_width, img_height, 3))
```

We may need to modify the layers based on our dataset - it is small and different from ImageNet. See

<https://medium.com/@14prakash/transfer-learning-using-keras-d804b2e04ef8>

(<https://medium.com/@14prakash/transfer-learning-using-keras-d804b2e04ef8>).

```
In [18]: # Freeze the layers which you don't want to train. Here I am freezing the first 5 layers.
for layer in model.layers[:5]:
    layer.trainable = False

#Adding custom Layers
x = model.output
x = Flatten()(x)
x = Dense(1024, activation="relu")(x)
x = Dropout(0.5)(x)
x = Dense(1024, activation="relu")(x)
predictions = Dense(4, activation="softmax")(x)

# creating the final model
model_final = Model(input = model.input, output = predictions)

# Parallel computing
model_final = multi_gpu_model(model_final, gpus=8)

# compile the model
model_final.compile(loss = "categorical_crossentropy", optimizer = optimizers.SGD(lr=learning_rate, momentum=mu), metrics=["accuracy"])

/home/shared/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.py:14: UserWarning: Update your `Model` call to the Keras 2 API: `Model(inputs=Tensor("in...", outputs=Tensor("de...`
```

```

In [19]: # Initiate the train and test generators with data Augumentation
train_datagen = ImageDataGenerator(rescale = 1./255,
                                   horizontal_flip = True,
                                   fill_mode = "nearest",
                                   zoom_range = 0.3,
                                   width_shift_range = 0.3,
                                   height_shift_range=0.3,
                                   rotation_range=30)

test_datagen = ImageDataGenerator(rescale = 1./255,
                                   horizontal_flip = True,
                                   fill_mode = "nearest",
                                   zoom_range = 0.3,
                                   width_shift_range = 0.3,
                                   height_shift_range=0.3,
                                   rotation_range=30)

train_generator = train_datagen.flow_from_directory(train_data_dir,
                                                    target_size = (img_h
eight, img_width),
                                                    batch_size = batch_s
ize,
                                                    class_mode = "catego
rical")

validation_generator = test_datagen.flow_from_directory(validation_data_
dir,
                                                    target_size = (i
mg_height, img_width),
                                                    class_mode = "ca
tegorical")

```

Found 66813 images belonging to 4 classes.

Found 16703 images belonging to 4 classes.

```

In [20]: # Save the model according to the conditions
checkpoint = ModelCheckpoint("vgg19_1.h5", monitor='val_acc', verbose=1,
                             save_best_only=True, save_weights_only=False, mode='auto', period=1)
early = EarlyStopping(monitor='val_acc', min_delta=0, patience=10, verbo
se=1, mode='auto')

```

```
In [21]: # Train the model
vgg_history = model_final.fit_generator(train_generator,
                                       steps_per_epoch = math.floor(nb_
train_samples/batch_size),
                                       epochs = epochs,
                                       validation_data = validation_gen
erator,
                                       validation_steps = math.floor(nb
_validation_samples/batch_size),
                                       callbacks = [checkpoint, early])

# Save history
with open('vgg_train_history.p', 'wb') as f:
    pickle.dump(vgg_history.history,f)

# Email notification for when this is done
server = smtplib.SMTP('smtp.gmail.com', 587)
server.starttls()
server.login("raa421@gmail.com", "Reventon21!")

msg = "HELLO SIR YOUR NETWORK IS TRAINED"
server.sendmail("raa421@gmail.com", "rafiayub@stanford.edu", msg)
server.quit()
print('Done.')
```

```
Epoch 1/50
521/521 [=====] - 1389s 3s/step - loss: 0.4238
- acc: 0.8523 - val_loss: 0.4072 - val_acc: 0.8584

Epoch 00001: val_acc improved from -inf to 0.85841, saving model to vgg
16_1.h5
Epoch 2/50
521/521 [=====] - 1386s 3s/step - loss: 0.2097
- acc: 0.9295 - val_loss: 0.6079 - val_acc: 0.8195

Epoch 00002: val_acc did not improve
Epoch 3/50
521/521 [=====] - 1365s 3s/step - loss: 0.1776
- acc: 0.9404 - val_loss: 0.3422 - val_acc: 0.8880

Epoch 00003: val_acc improved from 0.85841 to 0.88798, saving model to
vgg16_1.h5
Epoch 4/50
521/521 [=====] - 1394s 3s/step - loss: 0.1595
- acc: 0.9471 - val_loss: 0.2628 - val_acc: 0.9224

Epoch 00004: val_acc improved from 0.88798 to 0.92236, saving model to
vgg16_1.h5
Epoch 5/50
521/521 [=====] - 1388s 3s/step - loss: 0.1517
- acc: 0.9492 - val_loss: 0.2740 - val_acc: 0.9142

Epoch 00005: val_acc did not improve
Epoch 6/50
521/521 [=====] - 1361s 3s/step - loss: 0.1430
- acc: 0.9519 - val_loss: 0.2186 - val_acc: 0.9262

Epoch 00006: val_acc improved from 0.92236 to 0.92620, saving model to
vgg16_1.h5
Epoch 7/50
521/521 [=====] - 1358s 3s/step - loss: 0.1344
- acc: 0.9554 - val_loss: 0.2923 - val_acc: 0.9026

Epoch 00007: val_acc did not improve
Epoch 8/50
521/521 [=====] - 1417s 3s/step - loss: 0.1269
- acc: 0.9572 - val_loss: 0.2447 - val_acc: 0.9238

Epoch 00008: val_acc did not improve
Epoch 9/50
521/521 [=====] - 1408s 3s/step - loss: 0.1263
- acc: 0.9580 - val_loss: 0.2713 - val_acc: 0.9125

Epoch 00009: val_acc did not improve
Epoch 10/50
521/521 [=====] - 1360s 3s/step - loss: 0.1245
- acc: 0.9585 - val_loss: 0.3252 - val_acc: 0.8959

Epoch 00010: val_acc did not improve
Epoch 11/50
521/521 [=====] - 1356s 3s/step - loss: 0.1182
- acc: 0.9605 - val_loss: 0.3221 - val_acc: 0.8942
```

Epoch 00011: val_acc did not improve
Epoch 12/50
521/521 [=====] - 1356s 3s/step - loss: 0.1149
- acc: 0.9615 - val_loss: 0.2750 - val_acc: 0.9175

Epoch 00012: val_acc did not improve
Epoch 13/50
521/521 [=====] - 1354s 3s/step - loss: 0.1166
- acc: 0.9608 - val_loss: 0.2377 - val_acc: 0.9286

Epoch 00013: val_acc improved from 0.92620 to 0.92861, saving model to vgg16_1.h5
Epoch 14/50
521/521 [=====] - 1383s 3s/step - loss: 0.1100
- acc: 0.9628 - val_loss: 0.2042 - val_acc: 0.9358

Epoch 00014: val_acc improved from 0.92861 to 0.93582, saving model to vgg16_1.h5
Epoch 15/50
521/521 [=====] - 1366s 3s/step - loss: 0.1085
- acc: 0.9632 - val_loss: 0.2833 - val_acc: 0.9094

Epoch 00015: val_acc did not improve
Epoch 16/50
521/521 [=====] - 1345s 3s/step - loss: 0.1071
- acc: 0.9639 - val_loss: 0.2447 - val_acc: 0.9228

Epoch 00016: val_acc did not improve
Epoch 17/50
521/521 [=====] - 1342s 3s/step - loss: 0.1043
- acc: 0.9649 - val_loss: 0.2981 - val_acc: 0.9077

Epoch 00017: val_acc did not improve
Epoch 18/50
521/521 [=====] - 1393s 3s/step - loss: 0.1033
- acc: 0.9652 - val_loss: 0.2500 - val_acc: 0.9219

Epoch 00018: val_acc did not improve
Epoch 19/50
521/521 [=====] - 1346s 3s/step - loss: 0.0997
- acc: 0.9661 - val_loss: 0.2329 - val_acc: 0.9231

Epoch 00019: val_acc did not improve
Epoch 20/50
521/521 [=====] - 1348s 3s/step - loss: 0.0973
- acc: 0.9665 - val_loss: 0.2506 - val_acc: 0.9180

Epoch 00020: val_acc did not improve
Epoch 21/50
521/521 [=====] - 1384s 3s/step - loss: 0.0980
- acc: 0.9676 - val_loss: 0.1837 - val_acc: 0.9404

Epoch 00021: val_acc improved from 0.93582 to 0.94038, saving model to vgg16_1.h5
Epoch 22/50
521/521 [=====] - 1349s 3s/step - loss: 0.0944


```

- acc: 0.9680 - val_loss: 0.2497 - val_acc: 0.9238

Epoch 00022: val_acc did not improve
Epoch 23/50
521/521 [=====] - 1359s 3s/step - loss: 0.0952
- acc: 0.9681 - val_loss: 0.2535 - val_acc: 0.9240

Epoch 00023: val_acc did not improve
Epoch 24/50
521/521 [=====] - 1364s 3s/step - loss: 0.0950
- acc: 0.9680 - val_loss: 0.2118 - val_acc: 0.9300

Epoch 00024: val_acc did not improve
Epoch 25/50
521/521 [=====] - 1347s 3s/step - loss: 0.0911
- acc: 0.9692 - val_loss: 0.2464 - val_acc: 0.9209

Epoch 00025: val_acc did not improve
Epoch 26/50
521/521 [=====] - 1376s 3s/step - loss: 0.0885
- acc: 0.9695 - val_loss: 0.2098 - val_acc: 0.9344

Epoch 00026: val_acc did not improve
Epoch 27/50
521/521 [=====] - 1369s 3s/step - loss: 0.0890
- acc: 0.9698 - val_loss: 0.2440 - val_acc: 0.9200

Epoch 00027: val_acc did not improve
Epoch 28/50
521/521 [=====] - 1363s 3s/step - loss: 0.0884
- acc: 0.9704 - val_loss: 0.2422 - val_acc: 0.9221

Epoch 00028: val_acc did not improve
Epoch 29/50
521/521 [=====] - 1398s 3s/step - loss: 0.0862
- acc: 0.9701 - val_loss: 0.3351 - val_acc: 0.9002

Epoch 00029: val_acc did not improve
Epoch 30/50
521/521 [=====] - 1429s 3s/step - loss: 0.0831
- acc: 0.9718 - val_loss: 0.2209 - val_acc: 0.9320

Epoch 00030: val_acc did not improve
Epoch 31/50
521/521 [=====] - 1386s 3s/step - loss: 0.0838
- acc: 0.9710 - val_loss: 0.2426 - val_acc: 0.9267

Epoch 00031: val_acc did not improve
Epoch 00031: early stopping
Done.

```

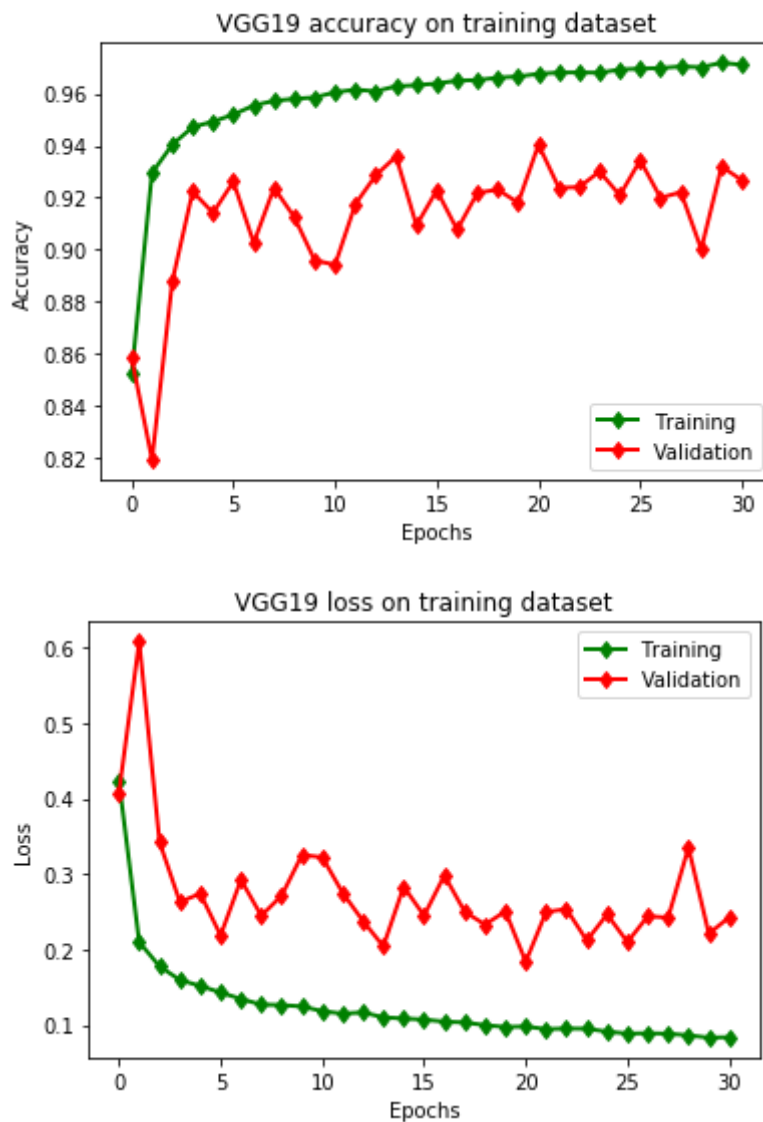
```

In [27]: # Load history
with open('vgg_train_history.p', 'rb') as f:
    train_history = pickle.load(f)

```

```
In [34]: # History is a dictionary with keys ['acc', 'loss', 'val_acc', 'val_loss']
         per epoch
plt.plot(train_history['acc'], label='Training', color='g', linewidth=2, marker='d')
plt.plot(train_history['val_acc'], label='Validation', color='r', linewidth=2, marker='d')
plt.legend()
plt.title('VGG19 accuracy on training dataset')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.savefig('vgg19_acc.jpg')
plt.show()

plt.plot(train_history['loss'], label='Training', color='g', linewidth=2, marker='d')
plt.plot(train_history['val_loss'], label='Validation', color='r', linewidth=2, marker='d')
plt.title('VGG19 loss on training dataset')
plt.legend()
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.savefig('vgg19_loss.jpg')
plt.show()
```



```
In [38]: # Test the network!! First initiate test generator
test_generator = test_datagen.flow_from_directory(test_data_dir,
                                                    target_size = (img_height, img_width),
                                                    class_mode = "categorical")
# Now test model
test_loss = model_final.evaluate_generator(test_generator)

Found 968 images belonging to 4 classes.
```

```
In [41]: model_final.metrics_names
```

```
Out[41]: ['loss', 'acc']
```

```
In [42]: print('Test loss: ', test_loss[0])
print('Test acc: ', test_loss[1])
with open('vgg_test_loss.p', 'wb') as f:
    pickle.dump(test_loss,f)
```

Test loss: 0.05007989506302993

Test acc: 0.9865702479338843

Train an InceptionV3 net

Configuration is based on previous work in Kermany et al.

```
In [17]: # Constants
img_width, img_height = 256, 256
train_data_dir = "data/train"
validation_data_dir = "data/val"
test_data_dir = "data/test"
nb_train_samples = 66813
nb_validation_samples = 16703
batch_size = 1000
epochs = 50
learning_rate = 0.001
#mu = 0.9 # Momentum
```

```
In [18]: # Load pretrained model
model = applications.InceptionV3(weights = "imagenet", include_top=False
, input_shape = (img_width, img_height, 3))
model.summary()
```

Layer (type) connected to	Output Shape	Param #	Connected to
input_4 (InputLayer)	(None, 256, 256, 3)	0	
conv2d_283 (Conv2D)	(None, 127, 127, 32)	864	input_4[0][0]
batch_normalization_283 (Batch Normalization)	(None, 127, 127, 32)	96	conv2d_283[0][0]
activation_283 (Activation)	(None, 127, 127, 32)	0	batch_normalization_283[0][0]
conv2d_284 (Conv2D)	(None, 125, 125, 32)	9216	activation_283[0][0]
batch_normalization_284 (Batch Normalization)	(None, 125, 125, 32)	96	conv2d_284[0][0]
activation_284 (Activation)	(None, 125, 125, 32)	0	batch_normalization_284[0][0]
conv2d_285 (Conv2D)	(None, 125, 125, 64)	18432	activation_284[0][0]
batch_normalization_285 (Batch Normalization)	(None, 125, 125, 64)	192	conv2d_285[0][0]
activation_285 (Activation)	(None, 125, 125, 64)	0	batch_normalization_285[0][0]
max_pooling2d_13 (Max Pooling)	(None, 62, 62, 64)	0	activation_285[0][0]
conv2d_286 (Conv2D)	(None, 62, 62, 80)	5120	max_pooling2d_13[0][0]
batch_normalization_286 (Batch Normalization)	(None, 62, 62, 80)	240	conv2d_286[0][0]

activation_286 (Activation) normalization_286[0][0]	(None, 62, 62, 80)	0	batch_
conv2d_287 (Conv2D) tion_286[0][0]	(None, 60, 60, 192)	138240	activa
batch_normalization_287 (BatchN _287[0][0]	(None, 60, 60, 192)	576	conv2d
activation_287 (Activation) normalization_287[0][0]	(None, 60, 60, 192)	0	batch_
max_pooling2d_14 (MaxPooling2D) tion_287[0][0]	(None, 29, 29, 192)	0	activa
conv2d_291 (Conv2D) oling2d_14[0][0]	(None, 29, 29, 64)	12288	max_po
batch_normalization_291 (BatchN _291[0][0]	(None, 29, 29, 64)	192	conv2d
activation_291 (Activation) normalization_291[0][0]	(None, 29, 29, 64)	0	batch_
conv2d_289 (Conv2D) oling2d_14[0][0]	(None, 29, 29, 48)	9216	max_po
conv2d_292 (Conv2D) tion_291[0][0]	(None, 29, 29, 96)	55296	activa
batch_normalization_289 (BatchN _289[0][0]	(None, 29, 29, 48)	144	conv2d
batch_normalization_292 (BatchN _292[0][0]	(None, 29, 29, 96)	288	conv2d
activation_289 (Activation) normalization_289[0][0]	(None, 29, 29, 48)	0	batch_
activation_292 (Activation) normalization_292[0][0]	(None, 29, 29, 96)	0	batch_

average_pooling2d_28 (AveragePo oling2d_14[0][0])	(None, 29, 29, 192)	0	max_po
conv2d_288 (Conv2D) oling2d_14[0][0])	(None, 29, 29, 64)	12288	max_po
conv2d_290 (Conv2D) tion_289[0][0])	(None, 29, 29, 64)	76800	activa
conv2d_293 (Conv2D) tion_292[0][0])	(None, 29, 29, 96)	82944	activa
conv2d_294 (Conv2D) e_pooling2d_28[0][0])	(None, 29, 29, 32)	6144	averag
batch_normalization_288 (BatchN _288[0][0])	(None, 29, 29, 64)	192	conv2d
batch_normalization_290 (BatchN _290[0][0])	(None, 29, 29, 64)	192	conv2d
batch_normalization_293 (BatchN _293[0][0])	(None, 29, 29, 96)	288	conv2d
batch_normalization_294 (BatchN _294[0][0])	(None, 29, 29, 32)	96	conv2d
activation_288 (Activation) normalization_288[0][0])	(None, 29, 29, 64)	0	batch_
activation_290 (Activation) normalization_290[0][0])	(None, 29, 29, 64)	0	batch_
activation_293 (Activation) normalization_293[0][0])	(None, 29, 29, 96)	0	batch_
activation_294 (Activation) normalization_294[0][0])	(None, 29, 29, 32)	0	batch_
mixed0 (Concatenate) tion_288[0][0])	(None, 29, 29, 256)	0	activa
			activa
tion_290[0][0])			activa

tion_293[0][0]			activa
tion_294[0][0]			
conv2d_298 (Conv2D) [0][0]	(None, 29, 29, 64)	16384	mixed0
batch_normalization_298 (BatchN _298[0][0])	(None, 29, 29, 64)	192	conv2d
activation_298 (Activation) normalization_298[0][0]	(None, 29, 29, 64)	0	batch_
conv2d_296 (Conv2D) [0][0]	(None, 29, 29, 48)	12288	mixed0
conv2d_299 (Conv2D) tion_298[0][0]	(None, 29, 29, 96)	55296	activa
batch_normalization_296 (BatchN _296[0][0])	(None, 29, 29, 48)	144	conv2d
batch_normalization_299 (BatchN _299[0][0])	(None, 29, 29, 96)	288	conv2d
activation_296 (Activation) normalization_296[0][0]	(None, 29, 29, 48)	0	batch_
activation_299 (Activation) normalization_299[0][0]	(None, 29, 29, 96)	0	batch_
average_pooling2d_29 (AveragePo [0][0])	(None, 29, 29, 256)	0	mixed0
conv2d_295 (Conv2D) [0][0]	(None, 29, 29, 64)	16384	mixed0
conv2d_297 (Conv2D) tion_296[0][0]	(None, 29, 29, 64)	76800	activa
conv2d_300 (Conv2D) tion_299[0][0]	(None, 29, 29, 96)	82944	activa

conv2d_301 (Conv2D) e_pooling2d_29[0][0]	(None, 29, 29, 64)	16384	average
batch_normalization_295 (Batch Normalization) _295[0][0]	(None, 29, 29, 64)	192	conv2d
batch_normalization_297 (Batch Normalization) _297[0][0]	(None, 29, 29, 64)	192	conv2d
batch_normalization_300 (Batch Normalization) _300[0][0]	(None, 29, 29, 96)	288	conv2d
batch_normalization_301 (Batch Normalization) _301[0][0]	(None, 29, 29, 64)	192	conv2d
activation_295 (Activation) normalization_295[0][0]	(None, 29, 29, 64)	0	batch_
activation_297 (Activation) normalization_297[0][0]	(None, 29, 29, 64)	0	batch_
activation_300 (Activation) normalization_300[0][0]	(None, 29, 29, 96)	0	batch_
activation_301 (Activation) normalization_301[0][0]	(None, 29, 29, 64)	0	batch_
mixed1 (Concatenate) tion_295[0][0]	(None, 29, 29, 288)	0	activa
tion_297[0][0]			activa
tion_300[0][0]			activa
tion_301[0][0]			activa
conv2d_305 (Conv2D) [0][0]	(None, 29, 29, 64)	18432	mixed1
batch_normalization_305 (Batch Normalization) _305[0][0]	(None, 29, 29, 64)	192	conv2d
activation_305 (Activation) normalization_305[0][0]	(None, 29, 29, 64)	0	batch_

conv2d_303 (Conv2D) [0][0]	(None, 29, 29, 48)	13824	mixed1
conv2d_306 (Conv2D) activation_305[0][0]	(None, 29, 29, 96)	55296	activation_305[0][0]
batch_normalization_303 (Batch Normalization) _303[0][0]	(None, 29, 29, 48)	144	conv2d_303[0][0]
batch_normalization_306 (Batch Normalization) _306[0][0]	(None, 29, 29, 96)	288	conv2d_306[0][0]
activation_303 (Activation) normalization_303[0][0]	(None, 29, 29, 48)	0	batch_normalization_303[0][0]
activation_306 (Activation) normalization_306[0][0]	(None, 29, 29, 96)	0	batch_normalization_306[0][0]
average_pooling2d_30 (Average Pooling) [0][0]	(None, 29, 29, 288)	0	mixed1[0][0]
conv2d_302 (Conv2D) [0][0]	(None, 29, 29, 64)	18432	mixed1[0][0]
conv2d_304 (Conv2D) activation_303[0][0]	(None, 29, 29, 64)	76800	activation_303[0][0]
conv2d_307 (Conv2D) activation_306[0][0]	(None, 29, 29, 96)	82944	activation_306[0][0]
conv2d_308 (Conv2D) average_pooling2d_30[0][0]	(None, 29, 29, 64)	18432	average_pooling2d_30[0][0]
batch_normalization_302 (Batch Normalization) _302[0][0]	(None, 29, 29, 64)	192	conv2d_302[0][0]
batch_normalization_304 (Batch Normalization) _304[0][0]	(None, 29, 29, 64)	192	conv2d_304[0][0]
batch_normalization_307 (Batch Normalization) _307[0][0]	(None, 29, 29, 96)	288	conv2d_307[0][0]

batch_normalization_308 (Batch Normalization)	(None, 29, 29, 64)	192	conv2d_308[0][0]
activation_302 (Activation)	(None, 29, 29, 64)	0	batch_normalization_302[0][0]
activation_304 (Activation)	(None, 29, 29, 64)	0	batch_normalization_304[0][0]
activation_307 (Activation)	(None, 29, 29, 96)	0	batch_normalization_307[0][0]
activation_308 (Activation)	(None, 29, 29, 64)	0	batch_normalization_308[0][0]
mixed2 (Concatenate)	(None, 29, 29, 288)	0	activation_302[0][0]
			activation_304[0][0]
			activation_307[0][0]
			activation_308[0][0]
conv2d_310 (Conv2D)	(None, 29, 29, 64)	18432	mixed2[0][0]
batch_normalization_310 (Batch Normalization)	(None, 29, 29, 64)	192	conv2d_310[0][0]
activation_310 (Activation)	(None, 29, 29, 64)	0	batch_normalization_310[0][0]
conv2d_311 (Conv2D)	(None, 29, 29, 96)	55296	activation_310[0][0]
batch_normalization_311 (Batch Normalization)	(None, 29, 29, 96)	288	conv2d_311[0][0]
activation_311 (Activation)	(None, 29, 29, 96)	0	batch_normalization_311[0][0]
conv2d_309 (Conv2D)	(None, 14, 14, 384)	995328	mixed2[0][0]

conv2d_312 (Conv2D)	(None, 14, 14, 96)	82944	activation_311[0][0]
batch_normalization_309 (Batch Normalization)	(None, 14, 14, 384)	1152	conv2d_309[0][0]
batch_normalization_312 (Batch Normalization)	(None, 14, 14, 96)	288	conv2d_312[0][0]
activation_309 (Activation)	(None, 14, 14, 384)	0	batch_normalization_309[0][0]
activation_312 (Activation)	(None, 14, 14, 96)	0	batch_normalization_312[0][0]
max_pooling2d_15 (MaxPooling2D)	(None, 14, 14, 288)	0	mixed2[0][0]
mixed3 (Concatenate)	(None, 14, 14, 768)	0	activation_309[0][0]
			activation_312[0][0]
			max_pooling2d_15[0][0]
conv2d_317 (Conv2D)	(None, 14, 14, 128)	98304	mixed3[0][0]
batch_normalization_317 (Batch Normalization)	(None, 14, 14, 128)	384	conv2d_317[0][0]
activation_317 (Activation)	(None, 14, 14, 128)	0	batch_normalization_317[0][0]
conv2d_318 (Conv2D)	(None, 14, 14, 128)	114688	activation_317[0][0]
batch_normalization_318 (Batch Normalization)	(None, 14, 14, 128)	384	conv2d_318[0][0]
activation_318 (Activation)	(None, 14, 14, 128)	0	batch_normalization_318[0][0]

conv2d_314 (Conv2D) [0][0]	(None, 14, 14, 128)	98304	mixed3
conv2d_319 (Conv2D) activation_318[0][0]	(None, 14, 14, 128)	114688	activation_318[0][0]
batch_normalization_314 (Batch Normalization) _314[0][0]	(None, 14, 14, 128)	384	conv2d_314[0][0]
batch_normalization_319 (Batch Normalization) _319[0][0]	(None, 14, 14, 128)	384	conv2d_319[0][0]
activation_314 (Activation) normalization_314[0][0]	(None, 14, 14, 128)	0	batch_normalization_314[0][0]
activation_319 (Activation) normalization_319[0][0]	(None, 14, 14, 128)	0	batch_normalization_319[0][0]
conv2d_315 (Conv2D) activation_314[0][0]	(None, 14, 14, 128)	114688	activation_314[0][0]
conv2d_320 (Conv2D) activation_319[0][0]	(None, 14, 14, 128)	114688	activation_319[0][0]
batch_normalization_315 (Batch Normalization) _315[0][0]	(None, 14, 14, 128)	384	conv2d_315[0][0]
batch_normalization_320 (Batch Normalization) _320[0][0]	(None, 14, 14, 128)	384	conv2d_320[0][0]
activation_315 (Activation) normalization_315[0][0]	(None, 14, 14, 128)	0	batch_normalization_315[0][0]
activation_320 (Activation) normalization_320[0][0]	(None, 14, 14, 128)	0	batch_normalization_320[0][0]
average_pooling2d_31 (Average Pooling) [0][0]	(None, 14, 14, 768)	0	mixed3[0][0]
conv2d_313 (Conv2D) [0][0]	(None, 14, 14, 192)	147456	mixed3[0][0]
conv2d_316 (Conv2D)	(None, 14, 14, 192)	172032	activation_316[0][0]

tion_315[0][0]

conv2d_321 (Conv2D)	(None, 14, 14, 192)	172032	activation_320[0][0]
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conv2d_322 (Conv2D)	(None, 14, 14, 192)	147456	average_pooling2d_31[0][0]
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batch_normalization_313 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_313[0][0]
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batch_normalization_316 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_316[0][0]
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batch_normalization_321 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_321[0][0]
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batch_normalization_322 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_322[0][0]
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activation_313 (Activation)	(None, 14, 14, 192)	0	batch_normalization_313[0][0]
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activation_316 (Activation)	(None, 14, 14, 192)	0	batch_normalization_316[0][0]
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activation_321 (Activation)	(None, 14, 14, 192)	0	batch_normalization_321[0][0]
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activation_322 (Activation)	(None, 14, 14, 192)	0	batch_normalization_322[0][0]
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mixed4 (Concatenate)	(None, 14, 14, 768)	0	activation_313[0][0]
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tion_316[0][0]

tion_321[0][0]

tion_322[0][0]

conv2d_327 (Conv2D)	(None, 14, 14, 160)	122880	mixed4[0][0]
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batch_normalization_327 (BatchN	(None, 14, 14, 160)	480	conv2d_327[0][0]
activation_327 (Activation)	(None, 14, 14, 160)	0	batch_normalization_327[0][0]
conv2d_328 (Conv2D)	(None, 14, 14, 160)	179200	activation_327[0][0]
batch_normalization_328 (BatchN	(None, 14, 14, 160)	480	conv2d_328[0][0]
activation_328 (Activation)	(None, 14, 14, 160)	0	batch_normalization_328[0][0]
conv2d_324 (Conv2D)	(None, 14, 14, 160)	122880	mixed4[0][0]
conv2d_329 (Conv2D)	(None, 14, 14, 160)	179200	activation_328[0][0]
batch_normalization_324 (BatchN	(None, 14, 14, 160)	480	conv2d_324[0][0]
batch_normalization_329 (BatchN	(None, 14, 14, 160)	480	conv2d_329[0][0]
activation_324 (Activation)	(None, 14, 14, 160)	0	batch_normalization_324[0][0]
activation_329 (Activation)	(None, 14, 14, 160)	0	batch_normalization_329[0][0]
conv2d_325 (Conv2D)	(None, 14, 14, 160)	179200	activation_324[0][0]
conv2d_330 (Conv2D)	(None, 14, 14, 160)	179200	activation_329[0][0]
batch_normalization_325 (BatchN	(None, 14, 14, 160)	480	conv2d_325[0][0]
batch_normalization_330 (BatchN	(None, 14, 14, 160)	480	conv2d_330[0][0]

_330[0][0]

activation_325 (Activation) normalization_325[0][0]	(None, 14, 14, 160)	0	batch_
activation_330 (Activation) normalization_330[0][0]	(None, 14, 14, 160)	0	batch_
average_pooling2d_32 (AveragePo [0][0])	(None, 14, 14, 768)	0	mixed4
conv2d_323 (Conv2D) [0][0]	(None, 14, 14, 192)	147456	mixed4
conv2d_326 (Conv2D) tion_325[0][0]	(None, 14, 14, 192)	215040	activa
conv2d_331 (Conv2D) tion_330[0][0]	(None, 14, 14, 192)	215040	activa
conv2d_332 (Conv2D) e_pooling2d_32[0][0]	(None, 14, 14, 192)	147456	averag
batch_normalization_323 (BatchN _323[0][0])	(None, 14, 14, 192)	576	conv2d
batch_normalization_326 (BatchN _326[0][0])	(None, 14, 14, 192)	576	conv2d
batch_normalization_331 (BatchN _331[0][0])	(None, 14, 14, 192)	576	conv2d
batch_normalization_332 (BatchN _332[0][0])	(None, 14, 14, 192)	576	conv2d
activation_323 (Activation) normalization_323[0][0]	(None, 14, 14, 192)	0	batch_
activation_326 (Activation) normalization_326[0][0]	(None, 14, 14, 192)	0	batch_
activation_331 (Activation) normalization_331[0][0]	(None, 14, 14, 192)	0	batch_

activation_332 (Activation) normalization_332[0][0]	(None, 14, 14, 192)	0	batch_
mixed5 (Concatenate) tion_323[0][0]	(None, 14, 14, 768)	0	activa
tion_326[0][0]			activa
tion_331[0][0]			activa
tion_332[0][0]			activa
conv2d_337 (Conv2D) [0][0]	(None, 14, 14, 160)	122880	mixed5
batch_normalization_337 (BatchN _337[0][0])	(None, 14, 14, 160)	480	conv2d
activation_337 (Activation) normalization_337[0][0]	(None, 14, 14, 160)	0	batch_
conv2d_338 (Conv2D) tion_337[0][0]	(None, 14, 14, 160)	179200	activa
batch_normalization_338 (BatchN _338[0][0])	(None, 14, 14, 160)	480	conv2d
activation_338 (Activation) normalization_338[0][0]	(None, 14, 14, 160)	0	batch_
conv2d_334 (Conv2D) [0][0]	(None, 14, 14, 160)	122880	mixed5
conv2d_339 (Conv2D) tion_338[0][0]	(None, 14, 14, 160)	179200	activa
batch_normalization_334 (BatchN _334[0][0])	(None, 14, 14, 160)	480	conv2d
batch_normalization_339 (BatchN _339[0][0])	(None, 14, 14, 160)	480	conv2d
activation_334 (Activation)	(None, 14, 14, 160)	0	batch_

normalization_334[0][0]

activation_339 (Activation) normalization_339[0][0]	(None, 14, 14, 160)	0	batch_
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conv2d_335 (Conv2D) tion_334[0][0]	(None, 14, 14, 160)	179200	activa
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conv2d_340 (Conv2D) tion_339[0][0]	(None, 14, 14, 160)	179200	activa
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batch_normalization_335 (BatchN _335[0][0]	(None, 14, 14, 160)	480	conv2d
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batch_normalization_340 (BatchN _340[0][0]	(None, 14, 14, 160)	480	conv2d
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activation_335 (Activation) normalization_335[0][0]	(None, 14, 14, 160)	0	batch_
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activation_340 (Activation) normalization_340[0][0]	(None, 14, 14, 160)	0	batch_
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average_pooling2d_33 (AveragePo [0][0]	(None, 14, 14, 768)	0	mixed5
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conv2d_333 (Conv2D) [0][0]	(None, 14, 14, 192)	147456	mixed5
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conv2d_336 (Conv2D) tion_335[0][0]	(None, 14, 14, 192)	215040	activa
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conv2d_341 (Conv2D) tion_340[0][0]	(None, 14, 14, 192)	215040	activa
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conv2d_342 (Conv2D) e_pooling2d_33[0][0]	(None, 14, 14, 192)	147456	averag
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batch_normalization_333 (BatchN _333[0][0]	(None, 14, 14, 192)	576	conv2d
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batch_normalization_336 (BatchN _336[0][0]	(None, 14, 14, 192)	576	conv2d
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batch_normalization_341 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_341[0][0]
batch_normalization_342 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_342[0][0]
activation_333 (Activation)	(None, 14, 14, 192)	0	batch_normalization_333[0][0]
activation_336 (Activation)	(None, 14, 14, 192)	0	batch_normalization_336[0][0]
activation_341 (Activation)	(None, 14, 14, 192)	0	batch_normalization_341[0][0]
activation_342 (Activation)	(None, 14, 14, 192)	0	batch_normalization_342[0][0]
mixed6 (Concatenate)	(None, 14, 14, 768)	0	activation_333[0][0]
			activation_336[0][0]
			activation_341[0][0]
			activation_342[0][0]
conv2d_347 (Conv2D)	(None, 14, 14, 192)	147456	mixed6[0][0]
batch_normalization_347 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_347[0][0]
activation_347 (Activation)	(None, 14, 14, 192)	0	batch_normalization_347[0][0]
conv2d_348 (Conv2D)	(None, 14, 14, 192)	258048	activation_347[0][0]
batch_normalization_348 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_348[0][0]
activation_348 (Activation)	(None, 14, 14, 192)	0	batch_normalization_348[0][0]

normalization_348[0][0]

conv2d_344 (Conv2D)	(None, 14, 14, 192)	147456	mixed6
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conv2d_349 (Conv2D)	(None, 14, 14, 192)	258048	activation_348[0][0]
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batch_normalization_344 (BatchN	(None, 14, 14, 192)	576	conv2d_344[0][0]
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batch_normalization_349 (BatchN	(None, 14, 14, 192)	576	conv2d_349[0][0]
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activation_344 (Activation)	(None, 14, 14, 192)	0	batch_normalization_344[0][0]
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activation_349 (Activation)	(None, 14, 14, 192)	0	batch_normalization_349[0][0]
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conv2d_345 (Conv2D)	(None, 14, 14, 192)	258048	activation_344[0][0]
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conv2d_350 (Conv2D)	(None, 14, 14, 192)	258048	activation_349[0][0]
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batch_normalization_345 (BatchN	(None, 14, 14, 192)	576	conv2d_345[0][0]
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batch_normalization_350 (BatchN	(None, 14, 14, 192)	576	conv2d_350[0][0]
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activation_345 (Activation)	(None, 14, 14, 192)	0	batch_normalization_345[0][0]
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activation_350 (Activation)	(None, 14, 14, 192)	0	batch_normalization_350[0][0]
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average_pooling2d_34 (AveragePo	(None, 14, 14, 768)	0	mixed6[0][0]
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conv2d_343 (Conv2D)	(None, 14, 14, 192)	147456	mixed6[0][0]
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conv2d_346 (Conv2D) activation_345[0][0]	(None, 14, 14, 192)	258048	activation_345[0][0]
conv2d_351 (Conv2D) activation_350[0][0]	(None, 14, 14, 192)	258048	activation_350[0][0]
conv2d_352 (Conv2D) average_pooling2d_34[0][0]	(None, 14, 14, 192)	147456	average_pooling2d_34[0][0]
batch_normalization_343 (Batch Normalization) _343[0][0]	(None, 14, 14, 192)	576	conv2d_343[0][0]
batch_normalization_346 (Batch Normalization) _346[0][0]	(None, 14, 14, 192)	576	conv2d_346[0][0]
batch_normalization_351 (Batch Normalization) _351[0][0]	(None, 14, 14, 192)	576	conv2d_351[0][0]
batch_normalization_352 (Batch Normalization) _352[0][0]	(None, 14, 14, 192)	576	conv2d_352[0][0]
activation_343 (Activation) normalization_343[0][0]	(None, 14, 14, 192)	0	batch_normalization_343[0][0]
activation_346 (Activation) normalization_346[0][0]	(None, 14, 14, 192)	0	batch_normalization_346[0][0]
activation_351 (Activation) normalization_351[0][0]	(None, 14, 14, 192)	0	batch_normalization_351[0][0]
activation_352 (Activation) normalization_352[0][0]	(None, 14, 14, 192)	0	batch_normalization_352[0][0]
mixed7 (Concatenate) activation_343[0][0] activation_346[0][0] activation_351[0][0] activation_352[0][0]	(None, 14, 14, 768)	0	activation_343[0][0] activation_346[0][0] activation_351[0][0] activation_352[0][0]
conv2d_355 (Conv2D)	(None, 14, 14, 192)	147456	mixed7

[0][0]

batch_normalization_355 (BatchN	(None, 14, 14, 192)	576	conv2d_355[0][0]
activation_355 (Activation)	(None, 14, 14, 192)	0	batch_normalization_355[0][0]
conv2d_356 (Conv2D)	(None, 14, 14, 192)	258048	activation_355[0][0]
batch_normalization_356 (BatchN	(None, 14, 14, 192)	576	conv2d_356[0][0]
activation_356 (Activation)	(None, 14, 14, 192)	0	batch_normalization_356[0][0]
conv2d_353 (Conv2D)	(None, 14, 14, 192)	147456	mixed7[0][0]
conv2d_357 (Conv2D)	(None, 14, 14, 192)	258048	activation_356[0][0]
batch_normalization_353 (BatchN	(None, 14, 14, 192)	576	conv2d_353[0][0]
batch_normalization_357 (BatchN	(None, 14, 14, 192)	576	conv2d_357[0][0]
activation_353 (Activation)	(None, 14, 14, 192)	0	batch_normalization_353[0][0]
activation_357 (Activation)	(None, 14, 14, 192)	0	batch_normalization_357[0][0]
conv2d_354 (Conv2D)	(None, 6, 6, 320)	552960	activation_353[0][0]
conv2d_358 (Conv2D)	(None, 6, 6, 192)	331776	activation_357[0][0]
batch_normalization_354 (BatchN	(None, 6, 6, 320)	960	conv2d_354[0][0]

batch_normalization_358 (Batch Normalization)	(None, 6, 6, 192)	576	conv2d_358[0][0]
activation_354 (Activation)	(None, 6, 6, 320)	0	batch_normalization_354[0][0]
activation_358 (Activation)	(None, 6, 6, 192)	0	batch_normalization_358[0][0]
max_pooling2d_16 (MaxPooling2D)	(None, 6, 6, 768)	0	mixed7[0][0]
mixed8 (Concatenate)	(None, 6, 6, 1280)	0	activation_354[0][0] activation_358[0][0] max_pooling2d_16[0][0]
conv2d_363 (Conv2D)	(None, 6, 6, 448)	573440	mixed8[0][0]
batch_normalization_363 (Batch Normalization)	(None, 6, 6, 448)	1344	conv2d_363[0][0]
activation_363 (Activation)	(None, 6, 6, 448)	0	batch_normalization_363[0][0]
conv2d_360 (Conv2D)	(None, 6, 6, 384)	491520	mixed8[0][0]
conv2d_364 (Conv2D)	(None, 6, 6, 384)	1548288	activation_363[0][0]
batch_normalization_360 (Batch Normalization)	(None, 6, 6, 384)	1152	conv2d_360[0][0]
batch_normalization_364 (Batch Normalization)	(None, 6, 6, 384)	1152	conv2d_364[0][0]
activation_360 (Activation)	(None, 6, 6, 384)	0	batch_normalization_360[0][0]

activation_364 (Activation) normalization_364[0][0]	(None, 6, 6, 384)	0	batch_
conv2d_361 (Conv2D) tion_360[0][0]	(None, 6, 6, 384)	442368	activa
conv2d_362 (Conv2D) tion_360[0][0]	(None, 6, 6, 384)	442368	activa
conv2d_365 (Conv2D) tion_364[0][0]	(None, 6, 6, 384)	442368	activa
conv2d_366 (Conv2D) tion_364[0][0]	(None, 6, 6, 384)	442368	activa
average_pooling2d_35 (AveragePo [0][0])	(None, 6, 6, 1280)	0	mixed8
conv2d_359 (Conv2D) [0][0]	(None, 6, 6, 320)	409600	mixed8
batch_normalization_361 (BatchN _361[0][0])	(None, 6, 6, 384)	1152	conv2d
batch_normalization_362 (BatchN _362[0][0])	(None, 6, 6, 384)	1152	conv2d
batch_normalization_365 (BatchN _365[0][0])	(None, 6, 6, 384)	1152	conv2d
batch_normalization_366 (BatchN _366[0][0])	(None, 6, 6, 384)	1152	conv2d
conv2d_367 (Conv2D) e_pooling2d_35[0][0]	(None, 6, 6, 192)	245760	averag
batch_normalization_359 (BatchN _359[0][0])	(None, 6, 6, 320)	960	conv2d
activation_361 (Activation) normalization_361[0][0]	(None, 6, 6, 384)	0	batch_

activation_362 (Activation) normalization_362[0][0]	(None, 6, 6, 384)	0	batch_
activation_365 (Activation) normalization_365[0][0]	(None, 6, 6, 384)	0	batch_
activation_366 (Activation) normalization_366[0][0]	(None, 6, 6, 384)	0	batch_
batch_normalization_367 (BatchN _367[0][0])	(None, 6, 6, 192)	576	conv2d
activation_359 (Activation) normalization_359[0][0]	(None, 6, 6, 320)	0	batch_
mixed9_0 (Concatenate) tion_361[0][0] tion_362[0][0]	(None, 6, 6, 768)	0	activa activa
concatenate_7 (Concatenate) tion_365[0][0] tion_366[0][0]	(None, 6, 6, 768)	0	activa activa
activation_367 (Activation) normalization_367[0][0]	(None, 6, 6, 192)	0	batch_
mixed9 (Concatenate) tion_359[0][0] _0[0][0] enate_7[0][0] tion_367[0][0]	(None, 6, 6, 2048)	0	activa mixed9 concat activa
conv2d_372 (Conv2D) [0][0]	(None, 6, 6, 448)	917504	mixed9
batch_normalization_372 (BatchN _372[0][0])	(None, 6, 6, 448)	1344	conv2d
activation_372 (Activation) normalization_372[0][0]	(None, 6, 6, 448)	0	batch_

conv2d_369 (Conv2D) [0][0]	(None, 6, 6, 384)	786432	mixed9
conv2d_373 (Conv2D) activation_372[0][0]	(None, 6, 6, 384)	1548288	activation_372[0][0]
batch_normalization_369 (Batch Normalization) _369[0][0]	(None, 6, 6, 384)	1152	conv2d_369[0][0]
batch_normalization_373 (Batch Normalization) _373[0][0]	(None, 6, 6, 384)	1152	conv2d_373[0][0]
activation_369 (Activation) normalization_369[0][0]	(None, 6, 6, 384)	0	batch_normalization_369[0][0]
activation_373 (Activation) normalization_373[0][0]	(None, 6, 6, 384)	0	batch_normalization_373[0][0]
conv2d_370 (Conv2D) activation_369[0][0]	(None, 6, 6, 384)	442368	activation_369[0][0]
conv2d_371 (Conv2D) activation_369[0][0]	(None, 6, 6, 384)	442368	activation_369[0][0]
conv2d_374 (Conv2D) activation_373[0][0]	(None, 6, 6, 384)	442368	activation_373[0][0]
conv2d_375 (Conv2D) activation_373[0][0]	(None, 6, 6, 384)	442368	activation_373[0][0]
average_pooling2d_36 (Average Pooling) [0][0]	(None, 6, 6, 2048)	0	mixed9
conv2d_368 (Conv2D) [0][0]	(None, 6, 6, 320)	655360	mixed9
batch_normalization_370 (Batch Normalization) _370[0][0]	(None, 6, 6, 384)	1152	conv2d_370[0][0]
batch_normalization_371 (Batch Normalization) _371[0][0]	(None, 6, 6, 384)	1152	conv2d_371[0][0]

batch_normalization_374 (BatchN _374[0][0])	(None, 6, 6, 384)	1152	conv2d
batch_normalization_375 (BatchN _375[0][0])	(None, 6, 6, 384)	1152	conv2d
conv2d_376 (Conv2D) e_pooling2d_36[0][0])	(None, 6, 6, 192)	393216	averag
batch_normalization_368 (BatchN _368[0][0])	(None, 6, 6, 320)	960	conv2d
activation_370 (Activation) normalization_370[0][0])	(None, 6, 6, 384)	0	batch_
activation_371 (Activation) normalization_371[0][0])	(None, 6, 6, 384)	0	batch_
activation_374 (Activation) normalization_374[0][0])	(None, 6, 6, 384)	0	batch_
activation_375 (Activation) normalization_375[0][0])	(None, 6, 6, 384)	0	batch_
batch_normalization_376 (BatchN _376[0][0])	(None, 6, 6, 192)	576	conv2d
activation_368 (Activation) normalization_368[0][0])	(None, 6, 6, 320)	0	batch_
mixed9_1 (Concatenate) tion_370[0][0])	(None, 6, 6, 768)	0	activa
			activa
concatenate_8 (Concatenate) tion_374[0][0])	(None, 6, 6, 768)	0	activa
			activa
activation_376 (Activation) normalization_376[0][0])	(None, 6, 6, 192)	0	batch_
mixed10 (Concatenate)	(None, 6, 6, 2048)	0	activa

tion_368[0][0]

_1[0][0]

enate_8[0][0]

tion_376[0][0]

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Total params: 21,802,784

Trainable params: 21,768,352

Non-trainable params: 34,432

```
In [19]: #Adding custom Layers
x = model.output
x = GlobalAveragePooling2D()(x)
predictions = Dense(4, activation="softmax")(x)

# creating the final model
model_final = Model(input = model.input, output = predictions)

model_final.summary()

# Freeze all layers except final Dense softmax layer
for layer in model_final.layers[:-1]:
    layer.trainable = False

# Parallel computing
model_final = multi_gpu_model(model_final, gpus=2)

# compile the model
model_final.compile(loss = "categorical_crossentropy", optimizer = optimizers.Adam(), metrics=["accuracy"])
```

Layer (type) connected to	Output Shape	Param #	Connected to
input_4 (InputLayer)	(None, 256, 256, 3)	0	
conv2d_283 (Conv2D)	(None, 127, 127, 32)	864	input_4[0][0]
batch_normalization_283 (Batch Normalization)	(None, 127, 127, 32)	96	conv2d_283[0][0]
activation_283 (Activation)	(None, 127, 127, 32)	0	batch_normalization_283[0][0]
conv2d_284 (Conv2D)	(None, 125, 125, 32)	9216	activation_283[0][0]
batch_normalization_284 (Batch Normalization)	(None, 125, 125, 32)	96	conv2d_284[0][0]
activation_284 (Activation)	(None, 125, 125, 32)	0	batch_normalization_284[0][0]
conv2d_285 (Conv2D)	(None, 125, 125, 64)	18432	activation_284[0][0]
batch_normalization_285 (Batch Normalization)	(None, 125, 125, 64)	192	conv2d_285[0][0]
activation_285 (Activation)	(None, 125, 125, 64)	0	batch_normalization_285[0][0]
max_pooling2d_13 (Max Pooling)	(None, 62, 62, 64)	0	activation_285[0][0]
conv2d_286 (Conv2D)	(None, 62, 62, 80)	5120	max_pooling2d_13[0][0]
batch_normalization_286 (Batch Normalization)	(None, 62, 62, 80)	240	conv2d_286[0][0]

activation_286 (Activation) normalization_286[0][0]	(None, 62, 62, 80)	0	batch_
conv2d_287 (Conv2D) tion_286[0][0]	(None, 60, 60, 192)	138240	activa
batch_normalization_287 (BatchN _287[0][0]	(None, 60, 60, 192)	576	conv2d
activation_287 (Activation) normalization_287[0][0]	(None, 60, 60, 192)	0	batch_
max_pooling2d_14 (MaxPooling2D) tion_287[0][0]	(None, 29, 29, 192)	0	activa
conv2d_291 (Conv2D) oling2d_14[0][0]	(None, 29, 29, 64)	12288	max_po
batch_normalization_291 (BatchN _291[0][0]	(None, 29, 29, 64)	192	conv2d
activation_291 (Activation) normalization_291[0][0]	(None, 29, 29, 64)	0	batch_
conv2d_289 (Conv2D) oling2d_14[0][0]	(None, 29, 29, 48)	9216	max_po
conv2d_292 (Conv2D) tion_291[0][0]	(None, 29, 29, 96)	55296	activa
batch_normalization_289 (BatchN _289[0][0]	(None, 29, 29, 48)	144	conv2d
batch_normalization_292 (BatchN _292[0][0]	(None, 29, 29, 96)	288	conv2d
activation_289 (Activation) normalization_289[0][0]	(None, 29, 29, 48)	0	batch_
activation_292 (Activation) normalization_292[0][0]	(None, 29, 29, 96)	0	batch_

average_pooling2d_28 (AveragePo oling2d_14[0][0])	(None, 29, 29, 192)	0	max_po
conv2d_288 (Conv2D) oling2d_14[0][0])	(None, 29, 29, 64)	12288	max_po
conv2d_290 (Conv2D) tion_289[0][0])	(None, 29, 29, 64)	76800	activa
conv2d_293 (Conv2D) tion_292[0][0])	(None, 29, 29, 96)	82944	activa
conv2d_294 (Conv2D) e_pooling2d_28[0][0])	(None, 29, 29, 32)	6144	averag
batch_normalization_288 (BatchN _288[0][0])	(None, 29, 29, 64)	192	conv2d
batch_normalization_290 (BatchN _290[0][0])	(None, 29, 29, 64)	192	conv2d
batch_normalization_293 (BatchN _293[0][0])	(None, 29, 29, 96)	288	conv2d
batch_normalization_294 (BatchN _294[0][0])	(None, 29, 29, 32)	96	conv2d
activation_288 (Activation) normalization_288[0][0])	(None, 29, 29, 64)	0	batch_
activation_290 (Activation) normalization_290[0][0])	(None, 29, 29, 64)	0	batch_
activation_293 (Activation) normalization_293[0][0])	(None, 29, 29, 96)	0	batch_
activation_294 (Activation) normalization_294[0][0])	(None, 29, 29, 32)	0	batch_
mixed0 (Concatenate) tion_288[0][0])	(None, 29, 29, 256)	0	activa
			activa
tion_290[0][0])			activa

tion_293[0][0]			activa
tion_294[0][0]			
conv2d_298 (Conv2D) [0][0]	(None, 29, 29, 64)	16384	mixed0
batch_normalization_298 (BatchN _298[0][0])	(None, 29, 29, 64)	192	conv2d
activation_298 (Activation) normalization_298[0][0]	(None, 29, 29, 64)	0	batch_
conv2d_296 (Conv2D) [0][0]	(None, 29, 29, 48)	12288	mixed0
conv2d_299 (Conv2D) tion_298[0][0]	(None, 29, 29, 96)	55296	activa
batch_normalization_296 (BatchN _296[0][0])	(None, 29, 29, 48)	144	conv2d
batch_normalization_299 (BatchN _299[0][0])	(None, 29, 29, 96)	288	conv2d
activation_296 (Activation) normalization_296[0][0]	(None, 29, 29, 48)	0	batch_
activation_299 (Activation) normalization_299[0][0]	(None, 29, 29, 96)	0	batch_
average_pooling2d_29 (AveragePo [0][0])	(None, 29, 29, 256)	0	mixed0
conv2d_295 (Conv2D) [0][0]	(None, 29, 29, 64)	16384	mixed0
conv2d_297 (Conv2D) tion_296[0][0]	(None, 29, 29, 64)	76800	activa
conv2d_300 (Conv2D) tion_299[0][0]	(None, 29, 29, 96)	82944	activa

conv2d_301 (Conv2D)	(None, 29, 29, 64)	16384	average
e_pooling2d_29[0][0]			
batch_normalization_295 (Batch Normalization)	(None, 29, 29, 64)	192	conv2d_295[0][0]
batch_normalization_297 (Batch Normalization)	(None, 29, 29, 64)	192	conv2d_297[0][0]
batch_normalization_300 (Batch Normalization)	(None, 29, 29, 96)	288	conv2d_300[0][0]
batch_normalization_301 (Batch Normalization)	(None, 29, 29, 64)	192	conv2d_301[0][0]
activation_295 (Activation)	(None, 29, 29, 64)	0	batch_normalization_295[0][0]
activation_297 (Activation)	(None, 29, 29, 64)	0	batch_normalization_297[0][0]
activation_300 (Activation)	(None, 29, 29, 96)	0	batch_normalization_300[0][0]
activation_301 (Activation)	(None, 29, 29, 64)	0	batch_normalization_301[0][0]
mixed1 (Concatenate)	(None, 29, 29, 288)	0	activation_295[0][0]
tion_295[0][0]			activation_297[0][0]
tion_297[0][0]			activation_300[0][0]
tion_300[0][0]			activation_301[0][0]
tion_301[0][0]			
conv2d_305 (Conv2D)	(None, 29, 29, 64)	18432	mixed1[0][0]
batch_normalization_305 (Batch Normalization)	(None, 29, 29, 64)	192	conv2d_305[0][0]
activation_305 (Activation)	(None, 29, 29, 64)	0	batch_normalization_305[0][0]

conv2d_303 (Conv2D) [0][0]	(None, 29, 29, 48)	13824	mixed1
conv2d_306 (Conv2D) activation_305[0][0]	(None, 29, 29, 96)	55296	activation_305[0][0]
batch_normalization_303 (Batch Normalization) _303[0][0]	(None, 29, 29, 48)	144	conv2d_303[0][0]
batch_normalization_306 (Batch Normalization) _306[0][0]	(None, 29, 29, 96)	288	conv2d_306[0][0]
activation_303 (Activation) normalization_303[0][0]	(None, 29, 29, 48)	0	batch_normalization_303[0][0]
activation_306 (Activation) normalization_306[0][0]	(None, 29, 29, 96)	0	batch_normalization_306[0][0]
average_pooling2d_30 (Average Pooling) [0][0]	(None, 29, 29, 288)	0	mixed1[0][0]
conv2d_302 (Conv2D) [0][0]	(None, 29, 29, 64)	18432	mixed1[0][0]
conv2d_304 (Conv2D) activation_303[0][0]	(None, 29, 29, 64)	76800	activation_303[0][0]
conv2d_307 (Conv2D) activation_306[0][0]	(None, 29, 29, 96)	82944	activation_306[0][0]
conv2d_308 (Conv2D) average_pooling2d_30[0][0]	(None, 29, 29, 64)	18432	average_pooling2d_30[0][0]
batch_normalization_302 (Batch Normalization) _302[0][0]	(None, 29, 29, 64)	192	conv2d_302[0][0]
batch_normalization_304 (Batch Normalization) _304[0][0]	(None, 29, 29, 64)	192	conv2d_304[0][0]
batch_normalization_307 (Batch Normalization) _307[0][0]	(None, 29, 29, 96)	288	conv2d_307[0][0]

batch_normalization_308 (Batch Normalization)	(None, 29, 29, 64)	192	conv2d_308[0][0]
activation_302 (Activation)	(None, 29, 29, 64)	0	batch_normalization_302[0][0]
activation_304 (Activation)	(None, 29, 29, 64)	0	batch_normalization_304[0][0]
activation_307 (Activation)	(None, 29, 29, 96)	0	batch_normalization_307[0][0]
activation_308 (Activation)	(None, 29, 29, 64)	0	batch_normalization_308[0][0]
mixed2 (Concatenate)	(None, 29, 29, 288)	0	activation_302[0][0]
			activation_304[0][0]
			activation_307[0][0]
			activation_308[0][0]
conv2d_310 (Conv2D)	(None, 29, 29, 64)	18432	mixed2[0][0]
batch_normalization_310 (Batch Normalization)	(None, 29, 29, 64)	192	conv2d_310[0][0]
activation_310 (Activation)	(None, 29, 29, 64)	0	batch_normalization_310[0][0]
conv2d_311 (Conv2D)	(None, 29, 29, 96)	55296	activation_310[0][0]
batch_normalization_311 (Batch Normalization)	(None, 29, 29, 96)	288	conv2d_311[0][0]
activation_311 (Activation)	(None, 29, 29, 96)	0	batch_normalization_311[0][0]
conv2d_309 (Conv2D)	(None, 14, 14, 384)	995328	mixed2[0][0]

conv2d_312 (Conv2D)	(None, 14, 14, 96)	82944	activation_311[0][0]
batch_normalization_309 (Batch Normalization)	(None, 14, 14, 384)	1152	conv2d_309[0][0]
batch_normalization_312 (Batch Normalization)	(None, 14, 14, 96)	288	conv2d_312[0][0]
activation_309 (Activation)	(None, 14, 14, 384)	0	batch_normalization_309[0][0]
activation_312 (Activation)	(None, 14, 14, 96)	0	batch_normalization_312[0][0]
max_pooling2d_15 (MaxPooling2D)	(None, 14, 14, 288)	0	mixed2[0][0]
mixed3 (Concatenate)	(None, 14, 14, 768)	0	activation_309[0][0]
			activation_312[0][0]
			max_pooling2d_15[0][0]
conv2d_317 (Conv2D)	(None, 14, 14, 128)	98304	mixed3[0][0]
batch_normalization_317 (Batch Normalization)	(None, 14, 14, 128)	384	conv2d_317[0][0]
activation_317 (Activation)	(None, 14, 14, 128)	0	batch_normalization_317[0][0]
conv2d_318 (Conv2D)	(None, 14, 14, 128)	114688	activation_317[0][0]
batch_normalization_318 (Batch Normalization)	(None, 14, 14, 128)	384	conv2d_318[0][0]
activation_318 (Activation)	(None, 14, 14, 128)	0	batch_normalization_318[0][0]

conv2d_314 (Conv2D) [0][0]	(None, 14, 14, 128)	98304	mixed3
conv2d_319 (Conv2D) activation_318[0][0]	(None, 14, 14, 128)	114688	activation_318[0][0]
batch_normalization_314 (Batch Normalization) _314[0][0]	(None, 14, 14, 128)	384	conv2d_314[0][0]
batch_normalization_319 (Batch Normalization) _319[0][0]	(None, 14, 14, 128)	384	conv2d_319[0][0]
activation_314 (Activation) normalization_314[0][0]	(None, 14, 14, 128)	0	batch_normalization_314[0][0]
activation_319 (Activation) normalization_319[0][0]	(None, 14, 14, 128)	0	batch_normalization_319[0][0]
conv2d_315 (Conv2D) activation_314[0][0]	(None, 14, 14, 128)	114688	activation_314[0][0]
conv2d_320 (Conv2D) activation_319[0][0]	(None, 14, 14, 128)	114688	activation_319[0][0]
batch_normalization_315 (Batch Normalization) _315[0][0]	(None, 14, 14, 128)	384	conv2d_315[0][0]
batch_normalization_320 (Batch Normalization) _320[0][0]	(None, 14, 14, 128)	384	conv2d_320[0][0]
activation_315 (Activation) normalization_315[0][0]	(None, 14, 14, 128)	0	batch_normalization_315[0][0]
activation_320 (Activation) normalization_320[0][0]	(None, 14, 14, 128)	0	batch_normalization_320[0][0]
average_pooling2d_31 (Average Pooling) [0][0]	(None, 14, 14, 768)	0	mixed3[0][0]
conv2d_313 (Conv2D) [0][0]	(None, 14, 14, 192)	147456	mixed3[0][0]
conv2d_316 (Conv2D)	(None, 14, 14, 192)	172032	activation_316[0][0]

tion_315[0][0]

conv2d_321 (Conv2D)	(None, 14, 14, 192)	172032	activation_320[0][0]
conv2d_322 (Conv2D)	(None, 14, 14, 192)	147456	average_pooling2d_31[0][0]
batch_normalization_313 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_313[0][0]
batch_normalization_316 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_316[0][0]
batch_normalization_321 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_321[0][0]
batch_normalization_322 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_322[0][0]
activation_313 (Activation)	(None, 14, 14, 192)	0	batch_normalization_313[0][0]
activation_316 (Activation)	(None, 14, 14, 192)	0	batch_normalization_316[0][0]
activation_321 (Activation)	(None, 14, 14, 192)	0	batch_normalization_321[0][0]
activation_322 (Activation)	(None, 14, 14, 192)	0	batch_normalization_322[0][0]
mixed4 (Concatenate)	(None, 14, 14, 768)	0	activation_313[0][0]
			activation_316[0][0]
			activation_321[0][0]
			activation_322[0][0]
conv2d_327 (Conv2D)	(None, 14, 14, 160)	122880	mixed4[0][0]

batch_normalization_327 (BatchN	(None, 14, 14, 160)	480	conv2d_327[0][0]
activation_327 (Activation)	(None, 14, 14, 160)	0	batch_normalization_327[0][0]
conv2d_328 (Conv2D)	(None, 14, 14, 160)	179200	activation_327[0][0]
batch_normalization_328 (BatchN	(None, 14, 14, 160)	480	conv2d_328[0][0]
activation_328 (Activation)	(None, 14, 14, 160)	0	batch_normalization_328[0][0]
conv2d_324 (Conv2D)	(None, 14, 14, 160)	122880	mixed4[0][0]
conv2d_329 (Conv2D)	(None, 14, 14, 160)	179200	activation_328[0][0]
batch_normalization_324 (BatchN	(None, 14, 14, 160)	480	conv2d_324[0][0]
batch_normalization_329 (BatchN	(None, 14, 14, 160)	480	conv2d_329[0][0]
activation_324 (Activation)	(None, 14, 14, 160)	0	batch_normalization_324[0][0]
activation_329 (Activation)	(None, 14, 14, 160)	0	batch_normalization_329[0][0]
conv2d_325 (Conv2D)	(None, 14, 14, 160)	179200	activation_324[0][0]
conv2d_330 (Conv2D)	(None, 14, 14, 160)	179200	activation_329[0][0]
batch_normalization_325 (BatchN	(None, 14, 14, 160)	480	conv2d_325[0][0]
batch_normalization_330 (BatchN	(None, 14, 14, 160)	480	conv2d_330[0][0]

_330[0][0]

activation_325 (Activation) normalization_325[0][0]	(None, 14, 14, 160)	0	batch_
activation_330 (Activation) normalization_330[0][0]	(None, 14, 14, 160)	0	batch_
average_pooling2d_32 (AveragePo [0][0])	(None, 14, 14, 768)	0	mixed4
conv2d_323 (Conv2D) [0][0]	(None, 14, 14, 192)	147456	mixed4
conv2d_326 (Conv2D) tion_325[0][0]	(None, 14, 14, 192)	215040	activa
conv2d_331 (Conv2D) tion_330[0][0]	(None, 14, 14, 192)	215040	activa
conv2d_332 (Conv2D) e_pooling2d_32[0][0]	(None, 14, 14, 192)	147456	averag
batch_normalization_323 (BatchN _323[0][0])	(None, 14, 14, 192)	576	conv2d
batch_normalization_326 (BatchN _326[0][0])	(None, 14, 14, 192)	576	conv2d
batch_normalization_331 (BatchN _331[0][0])	(None, 14, 14, 192)	576	conv2d
batch_normalization_332 (BatchN _332[0][0])	(None, 14, 14, 192)	576	conv2d
activation_323 (Activation) normalization_323[0][0]	(None, 14, 14, 192)	0	batch_
activation_326 (Activation) normalization_326[0][0]	(None, 14, 14, 192)	0	batch_
activation_331 (Activation) normalization_331[0][0]	(None, 14, 14, 192)	0	batch_

activation_332 (Activation) normalization_332[0][0]	(None, 14, 14, 192)	0	batch_
mixed5 (Concatenate) tion_323[0][0]	(None, 14, 14, 768)	0	activa
tion_326[0][0]			activa
tion_331[0][0]			activa
tion_332[0][0]			activa
conv2d_337 (Conv2D) [0][0]	(None, 14, 14, 160)	122880	mixed5
batch_normalization_337 (BatchN _337[0][0])	(None, 14, 14, 160)	480	conv2d
activation_337 (Activation) normalization_337[0][0]	(None, 14, 14, 160)	0	batch_
conv2d_338 (Conv2D) tion_337[0][0]	(None, 14, 14, 160)	179200	activa
batch_normalization_338 (BatchN _338[0][0])	(None, 14, 14, 160)	480	conv2d
activation_338 (Activation) normalization_338[0][0]	(None, 14, 14, 160)	0	batch_
conv2d_334 (Conv2D) [0][0]	(None, 14, 14, 160)	122880	mixed5
conv2d_339 (Conv2D) tion_338[0][0]	(None, 14, 14, 160)	179200	activa
batch_normalization_334 (BatchN _334[0][0])	(None, 14, 14, 160)	480	conv2d
batch_normalization_339 (BatchN _339[0][0])	(None, 14, 14, 160)	480	conv2d
activation_334 (Activation)	(None, 14, 14, 160)	0	batch_

normalization_334[0][0]

activation_339 (Activation) normalization_339[0][0]	(None, 14, 14, 160)	0	batch_
conv2d_335 (Conv2D) tion_334[0][0]	(None, 14, 14, 160)	179200	activa
conv2d_340 (Conv2D) tion_339[0][0]	(None, 14, 14, 160)	179200	activa
batch_normalization_335 (BatchN _335[0][0]	(None, 14, 14, 160)	480	conv2d
batch_normalization_340 (BatchN _340[0][0]	(None, 14, 14, 160)	480	conv2d
activation_335 (Activation) normalization_335[0][0]	(None, 14, 14, 160)	0	batch_
activation_340 (Activation) normalization_340[0][0]	(None, 14, 14, 160)	0	batch_
average_pooling2d_33 (AveragePo [0][0]	(None, 14, 14, 768)	0	mixed5
conv2d_333 (Conv2D) [0][0]	(None, 14, 14, 192)	147456	mixed5
conv2d_336 (Conv2D) tion_335[0][0]	(None, 14, 14, 192)	215040	activa
conv2d_341 (Conv2D) tion_340[0][0]	(None, 14, 14, 192)	215040	activa
conv2d_342 (Conv2D) e_pooling2d_33[0][0]	(None, 14, 14, 192)	147456	averag
batch_normalization_333 (BatchN _333[0][0]	(None, 14, 14, 192)	576	conv2d
batch_normalization_336 (BatchN _336[0][0]	(None, 14, 14, 192)	576	conv2d

batch_normalization_341 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_341[0][0]
batch_normalization_342 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_342[0][0]
activation_333 (Activation)	(None, 14, 14, 192)	0	batch_normalization_333[0][0]
activation_336 (Activation)	(None, 14, 14, 192)	0	batch_normalization_336[0][0]
activation_341 (Activation)	(None, 14, 14, 192)	0	batch_normalization_341[0][0]
activation_342 (Activation)	(None, 14, 14, 192)	0	batch_normalization_342[0][0]
mixed6 (Concatenate)	(None, 14, 14, 768)	0	activation_333[0][0]
			activation_336[0][0]
			activation_341[0][0]
			activation_342[0][0]
conv2d_347 (Conv2D)	(None, 14, 14, 192)	147456	mixed6[0][0]
batch_normalization_347 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_347[0][0]
activation_347 (Activation)	(None, 14, 14, 192)	0	batch_normalization_347[0][0]
conv2d_348 (Conv2D)	(None, 14, 14, 192)	258048	activation_347[0][0]
batch_normalization_348 (Batch Normalization)	(None, 14, 14, 192)	576	conv2d_348[0][0]
activation_348 (Activation)	(None, 14, 14, 192)	0	batch_normalization_348[0][0]

normalization_348[0][0]

conv2d_344 (Conv2D)	(None, 14, 14, 192)	147456	mixed6
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conv2d_349 (Conv2D)	(None, 14, 14, 192)	258048	activation_348[0][0]
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batch_normalization_344 (BatchN	(None, 14, 14, 192)	576	conv2d_344[0][0]
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batch_normalization_349 (BatchN	(None, 14, 14, 192)	576	conv2d_349[0][0]
---------------------------------	---------------------	-----	------------------

activation_344 (Activation)	(None, 14, 14, 192)	0	batch_normalization_344[0][0]
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activation_349 (Activation)	(None, 14, 14, 192)	0	batch_normalization_349[0][0]
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conv2d_345 (Conv2D)	(None, 14, 14, 192)	258048	activation_344[0][0]
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conv2d_350 (Conv2D)	(None, 14, 14, 192)	258048	activation_349[0][0]
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batch_normalization_345 (BatchN	(None, 14, 14, 192)	576	conv2d_345[0][0]
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batch_normalization_350 (BatchN	(None, 14, 14, 192)	576	conv2d_350[0][0]
---------------------------------	---------------------	-----	------------------

activation_345 (Activation)	(None, 14, 14, 192)	0	batch_normalization_345[0][0]
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activation_350 (Activation)	(None, 14, 14, 192)	0	batch_normalization_350[0][0]
-----------------------------	---------------------	---	-------------------------------

average_pooling2d_34 (AveragePo	(None, 14, 14, 768)	0	mixed6[0][0]
---------------------------------	---------------------	---	--------------

conv2d_343 (Conv2D)	(None, 14, 14, 192)	147456	mixed6[0][0]
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conv2d_346 (Conv2D) activation_345[0][0]	(None, 14, 14, 192)	258048	activation_345[0][0]
conv2d_351 (Conv2D) activation_350[0][0]	(None, 14, 14, 192)	258048	activation_350[0][0]
conv2d_352 (Conv2D) average_pooling2d_34[0][0]	(None, 14, 14, 192)	147456	average_pooling2d_34[0][0]
batch_normalization_343 (Batch Normalization) _343[0][0]	(None, 14, 14, 192)	576	conv2d_343[0][0]
batch_normalization_346 (Batch Normalization) _346[0][0]	(None, 14, 14, 192)	576	conv2d_346[0][0]
batch_normalization_351 (Batch Normalization) _351[0][0]	(None, 14, 14, 192)	576	conv2d_351[0][0]
batch_normalization_352 (Batch Normalization) _352[0][0]	(None, 14, 14, 192)	576	conv2d_352[0][0]
activation_343 (Activation) normalization_343[0][0]	(None, 14, 14, 192)	0	batch_normalization_343[0][0]
activation_346 (Activation) normalization_346[0][0]	(None, 14, 14, 192)	0	batch_normalization_346[0][0]
activation_351 (Activation) normalization_351[0][0]	(None, 14, 14, 192)	0	batch_normalization_351[0][0]
activation_352 (Activation) normalization_352[0][0]	(None, 14, 14, 192)	0	batch_normalization_352[0][0]
mixed7 (Concatenate) activation_343[0][0] activation_346[0][0] activation_351[0][0] activation_352[0][0]	(None, 14, 14, 768)	0	activation_343[0][0] activation_346[0][0] activation_351[0][0] activation_352[0][0]
conv2d_355 (Conv2D)	(None, 14, 14, 192)	147456	mixed7

[0][0]

batch_normalization_355 (BatchN	(None, 14, 14, 192)	576	conv2d_355[0][0]
activation_355 (Activation)	(None, 14, 14, 192)	0	batch_normalization_355[0][0]
conv2d_356 (Conv2D)	(None, 14, 14, 192)	258048	activation_355[0][0]
batch_normalization_356 (BatchN	(None, 14, 14, 192)	576	conv2d_356[0][0]
activation_356 (Activation)	(None, 14, 14, 192)	0	batch_normalization_356[0][0]
conv2d_353 (Conv2D)	(None, 14, 14, 192)	147456	mixed7[0][0]
conv2d_357 (Conv2D)	(None, 14, 14, 192)	258048	activation_356[0][0]
batch_normalization_353 (BatchN	(None, 14, 14, 192)	576	conv2d_353[0][0]
batch_normalization_357 (BatchN	(None, 14, 14, 192)	576	conv2d_357[0][0]
activation_353 (Activation)	(None, 14, 14, 192)	0	batch_normalization_353[0][0]
activation_357 (Activation)	(None, 14, 14, 192)	0	batch_normalization_357[0][0]
conv2d_354 (Conv2D)	(None, 6, 6, 320)	552960	activation_353[0][0]
conv2d_358 (Conv2D)	(None, 6, 6, 192)	331776	activation_357[0][0]
batch_normalization_354 (BatchN	(None, 6, 6, 320)	960	conv2d_354[0][0]

batch_normalization_358 (Batch Normalization)	(None, 6, 6, 192)	576	conv2d_358[0][0]
activation_354 (Activation)	(None, 6, 6, 320)	0	batch_normalization_354[0][0]
activation_358 (Activation)	(None, 6, 6, 192)	0	batch_normalization_358[0][0]
max_pooling2d_16 (MaxPooling2D)	(None, 6, 6, 768)	0	mixed7[0][0]
mixed8 (Concatenate)	(None, 6, 6, 1280)	0	activation_354[0][0] activation_358[0][0] max_pooling2d_16[0][0]
conv2d_363 (Conv2D)	(None, 6, 6, 448)	573440	mixed8[0][0]
batch_normalization_363 (Batch Normalization)	(None, 6, 6, 448)	1344	conv2d_363[0][0]
activation_363 (Activation)	(None, 6, 6, 448)	0	batch_normalization_363[0][0]
conv2d_360 (Conv2D)	(None, 6, 6, 384)	491520	mixed8[0][0]
conv2d_364 (Conv2D)	(None, 6, 6, 384)	1548288	activation_363[0][0]
batch_normalization_360 (Batch Normalization)	(None, 6, 6, 384)	1152	conv2d_360[0][0]
batch_normalization_364 (Batch Normalization)	(None, 6, 6, 384)	1152	conv2d_364[0][0]
activation_360 (Activation)	(None, 6, 6, 384)	0	batch_normalization_360[0][0]

activation_364 (Activation) normalization_364[0][0]	(None, 6, 6, 384)	0	batch_
conv2d_361 (Conv2D) tion_360[0][0]	(None, 6, 6, 384)	442368	activa
conv2d_362 (Conv2D) tion_360[0][0]	(None, 6, 6, 384)	442368	activa
conv2d_365 (Conv2D) tion_364[0][0]	(None, 6, 6, 384)	442368	activa
conv2d_366 (Conv2D) tion_364[0][0]	(None, 6, 6, 384)	442368	activa
average_pooling2d_35 (AveragePo [0][0])	(None, 6, 6, 1280)	0	mixed8
conv2d_359 (Conv2D) [0][0]	(None, 6, 6, 320)	409600	mixed8
batch_normalization_361 (BatchN _361[0][0])	(None, 6, 6, 384)	1152	conv2d
batch_normalization_362 (BatchN _362[0][0])	(None, 6, 6, 384)	1152	conv2d
batch_normalization_365 (BatchN _365[0][0])	(None, 6, 6, 384)	1152	conv2d
batch_normalization_366 (BatchN _366[0][0])	(None, 6, 6, 384)	1152	conv2d
conv2d_367 (Conv2D) e_pooling2d_35[0][0]	(None, 6, 6, 192)	245760	averag
batch_normalization_359 (BatchN _359[0][0])	(None, 6, 6, 320)	960	conv2d
activation_361 (Activation) normalization_361[0][0]	(None, 6, 6, 384)	0	batch_

activation_362 (Activation) normalization_362[0][0]	(None, 6, 6, 384)	0	batch_
activation_365 (Activation) normalization_365[0][0]	(None, 6, 6, 384)	0	batch_
activation_366 (Activation) normalization_366[0][0]	(None, 6, 6, 384)	0	batch_
batch_normalization_367 (BatchN _367[0][0])	(None, 6, 6, 192)	576	conv2d
activation_359 (Activation) normalization_359[0][0]	(None, 6, 6, 320)	0	batch_
mixed9_0 (Concatenate) tion_361[0][0] tion_362[0][0]	(None, 6, 6, 768)	0	activa activa
concatenate_7 (Concatenate) tion_365[0][0] tion_366[0][0]	(None, 6, 6, 768)	0	activa activa
activation_367 (Activation) normalization_367[0][0]	(None, 6, 6, 192)	0	batch_
mixed9 (Concatenate) tion_359[0][0] _0[0][0] enate_7[0][0] tion_367[0][0]	(None, 6, 6, 2048)	0	activa mixed9 concat activa
conv2d_372 (Conv2D) [0][0]	(None, 6, 6, 448)	917504	mixed9
batch_normalization_372 (BatchN _372[0][0])	(None, 6, 6, 448)	1344	conv2d
activation_372 (Activation) normalization_372[0][0]	(None, 6, 6, 448)	0	batch_

conv2d_369 (Conv2D) [0][0]	(None, 6, 6, 384)	786432	mixed9
conv2d_373 (Conv2D) activation_372[0][0]	(None, 6, 6, 384)	1548288	activation_372[0][0]
batch_normalization_369 (Batch Normalization) _369[0][0]	(None, 6, 6, 384)	1152	conv2d_369[0][0]
batch_normalization_373 (Batch Normalization) _373[0][0]	(None, 6, 6, 384)	1152	conv2d_373[0][0]
activation_369 (Activation) normalization_369[0][0]	(None, 6, 6, 384)	0	batch_normalization_369[0][0]
activation_373 (Activation) normalization_373[0][0]	(None, 6, 6, 384)	0	batch_normalization_373[0][0]
conv2d_370 (Conv2D) activation_369[0][0]	(None, 6, 6, 384)	442368	activation_369[0][0]
conv2d_371 (Conv2D) activation_369[0][0]	(None, 6, 6, 384)	442368	activation_369[0][0]
conv2d_374 (Conv2D) activation_373[0][0]	(None, 6, 6, 384)	442368	activation_373[0][0]
conv2d_375 (Conv2D) activation_373[0][0]	(None, 6, 6, 384)	442368	activation_373[0][0]
average_pooling2d_36 (Average Pooling) [0][0]	(None, 6, 6, 2048)	0	mixed9
conv2d_368 (Conv2D) [0][0]	(None, 6, 6, 320)	655360	mixed9
batch_normalization_370 (Batch Normalization) _370[0][0]	(None, 6, 6, 384)	1152	conv2d_370[0][0]
batch_normalization_371 (Batch Normalization) _371[0][0]	(None, 6, 6, 384)	1152	conv2d_371[0][0]

batch_normalization_374 (BatchN _374[0][0])	(None, 6, 6, 384)	1152	conv2d
batch_normalization_375 (BatchN _375[0][0])	(None, 6, 6, 384)	1152	conv2d
conv2d_376 (Conv2D) e_pooling2d_36[0][0])	(None, 6, 6, 192)	393216	averag
batch_normalization_368 (BatchN _368[0][0])	(None, 6, 6, 320)	960	conv2d
activation_370 (Activation) normalization_370[0][0])	(None, 6, 6, 384)	0	batch_
activation_371 (Activation) normalization_371[0][0])	(None, 6, 6, 384)	0	batch_
activation_374 (Activation) normalization_374[0][0])	(None, 6, 6, 384)	0	batch_
activation_375 (Activation) normalization_375[0][0])	(None, 6, 6, 384)	0	batch_
batch_normalization_376 (BatchN _376[0][0])	(None, 6, 6, 192)	576	conv2d
activation_368 (Activation) normalization_368[0][0])	(None, 6, 6, 320)	0	batch_
mixed9_1 (Concatenate) tion_370[0][0])	(None, 6, 6, 768)	0	activa
			activa
concatenate_8 (Concatenate) tion_374[0][0])	(None, 6, 6, 768)	0	activa
			activa
activation_376 (Activation) normalization_376[0][0])	(None, 6, 6, 192)	0	batch_
mixed10 (Concatenate)	(None, 6, 6, 2048)	0	activa

tion_368[0][0]			
_1[0][0]			mixed9
enate_8[0][0]			concat
tion_376[0][0]			activa

global_average_pooling2d_4 (Glo (None, 2048)	0		mixed1
0[0][0]			

dense_4 (Dense)	(None, 4)	8196	global
_average_pooling2d_4[0][0]			

=====

Total params: 21,810,980
 Trainable params: 21,776,548
 Non-trainable params: 34,432

```

/home/shared/anaconda3/lib/python3.6/site-packages/ipykernel_launcher.p
y:7: UserWarning: Update your `Model` call to the Keras 2 API: `Model(i
nputs=Tensor("in...", outputs=Tensor("de..."))`
import sys

```

```

In [20]: # Initiate the train and test generators with data Augmentation
train_datagen = ImageDataGenerator(rescale = 1./255,
                                   horizontal_flip = True,
                                   fill_mode = "nearest",
                                   zoom_range = 0.3,
                                   width_shift_range = 0.3,
                                   height_shift_range=0.3,
                                   rotation_range=30)

test_datagen = ImageDataGenerator(rescale = 1./255,
                                   horizontal_flip = True,
                                   fill_mode = "nearest",
                                   zoom_range = 0.3,
                                   width_shift_range = 0.3,
                                   height_shift_range=0.3,
                                   rotation_range=30)

train_generator = train_datagen.flow_from_directory(train_data_dir,
                                                    target_size = (img_h
eight, img_width),
                                                    batch_size = batch_s
ize,
                                                    class_mode = "catego
rical")

validation_generator = test_datagen.flow_from_directory(validation_data_
dir,
                                                    target_size = (i
mg_height, img_width),
                                                    class_mode = "ca
tegorical")

```

Found 66813 images belonging to 4 classes.

Found 16703 images belonging to 4 classes.

```

In [ ]: # Save the model according to the conditions
checkpoint = ModelCheckpoint("InceptionV3.h5", monitor='val_acc', verbose=1,
save_best_only=True, save_weights_only=False, mode='auto', period=1
)
early = EarlyStopping(monitor='val_acc', min_delta=0, patience=10, verbo
se=1, mode='auto')

```

```

In [ ]: # Train the model
inceptionv3_history = model_final.fit_generator(train_generator,
                                                steps_per_epoch = math.floor(nb_
train_samples/batch_size),
                                                epochs = epochs,
                                                validation_data = validation_gen
erator,
                                                validation_steps = math.floor(nb
_validation_samples/batch_size),
                                                callbacks = [checkpoint, early])

# Save history
with open('inceptionv3_train_history.p', 'wb') as f:
    pickle.dump(inceptionv3_history.history, f)

# Email notification for when this is done
server = smtplib.SMTP('smtp.gmail.com', 587)
server.starttls()
server.login("raa421@gmail.com", "Reventon21!")

msg = "HELLO SIR YOUR NETWORK IS TRAINED"
server.sendmail("raa421@gmail.com", "rafiayub@stanford.edu", msg)
server.quit()
print('Done.')

```

Epoch 1/50

20/66 [=====>.....] - ETA: 37:59 - loss: 1.1187 - ac
c: 0.5431

```

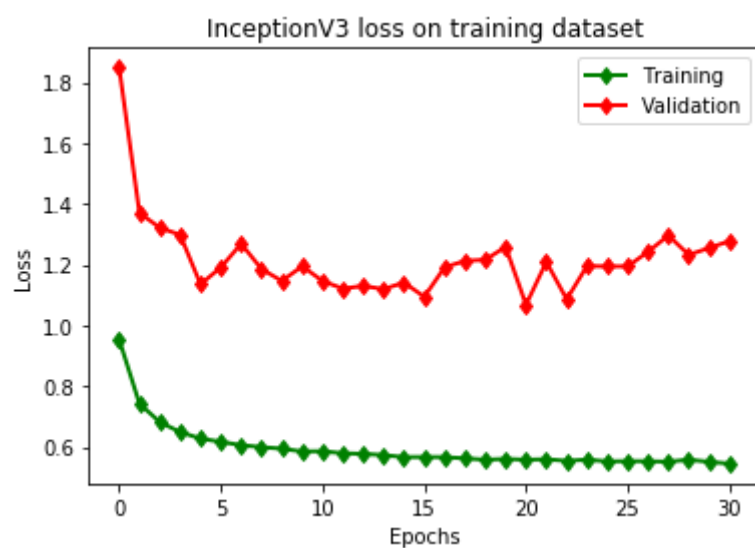
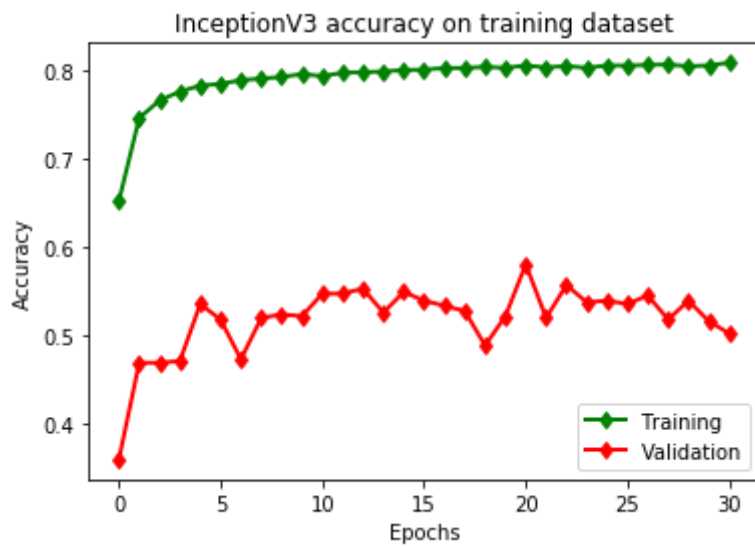
In [24]: # Load history
with open('inceptionv3_train_history.p', 'rb') as f:
    train_history = pickle.load(f)

```



```
In [25]: # History is a dictionary with keys ['acc', 'loss', 'val_acc', 'val_loss']
         per epoch
plt.plot(train_history['acc'], label='Training', color='g', linewidth=2, marker='d')
plt.plot(train_history['val_acc'], label='Validation', color='r', linewidth=2, marker='d')
plt.legend()
plt.title('InceptionV3 accuracy on training dataset')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.savefig('inceptionv3_acc.jpg')
plt.show()

plt.plot(train_history['loss'], label='Training', color='g', linewidth=2, marker='d')
plt.plot(train_history['val_loss'], label='Validation', color='r', linewidth=2, marker='d')
plt.title('InceptionV3 loss on training dataset')
plt.legend()
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.savefig('inceptionv3_loss.jpg')
plt.show()
```



```
In [26]: # Test the network!! First initiate test generator
test_generator = test_datagen.flow_from_directory(test_data_dir,
                                                    target_size = (img_height, img_width),
                                                    class_mode = "categorical")
# Now test model
test_loss = model_final.evaluate_generator(test_generator)

Found 968 images belonging to 4 classes.
```

```
In [27]: model_final.metrics_names
```

```
Out[27]: ['loss', 'acc']
```

```
In [28]: print('Test loss: ', test_loss[0])
print('Test acc: ', test_loss[1])
with open('inceptionv3_test_loss.p', 'wb') as f:
    pickle.dump(test_loss, f)
```

```
Test loss:  0.9668160736068221
Test acc:   0.6384297520661157
```

Compare performance of both nets

Load all variables

```
In [2]: with open('inceptionv3_test_loss.p', 'rb') as f:
        iv3testloss = pickle.load(f)
with open('vgg_test_loss.p', 'rb') as f:
        vggtestloss = pickle.load(f)
with open('inceptionv3_train_history.p', 'rb') as f:
        iv3hist = pickle.load(f)
with open('vgg_train_history.p', 'rb') as f:
        vgghist = pickle.load(f)

print('Done.')
```

```
Done.
```

Plot accuracy and loss

```

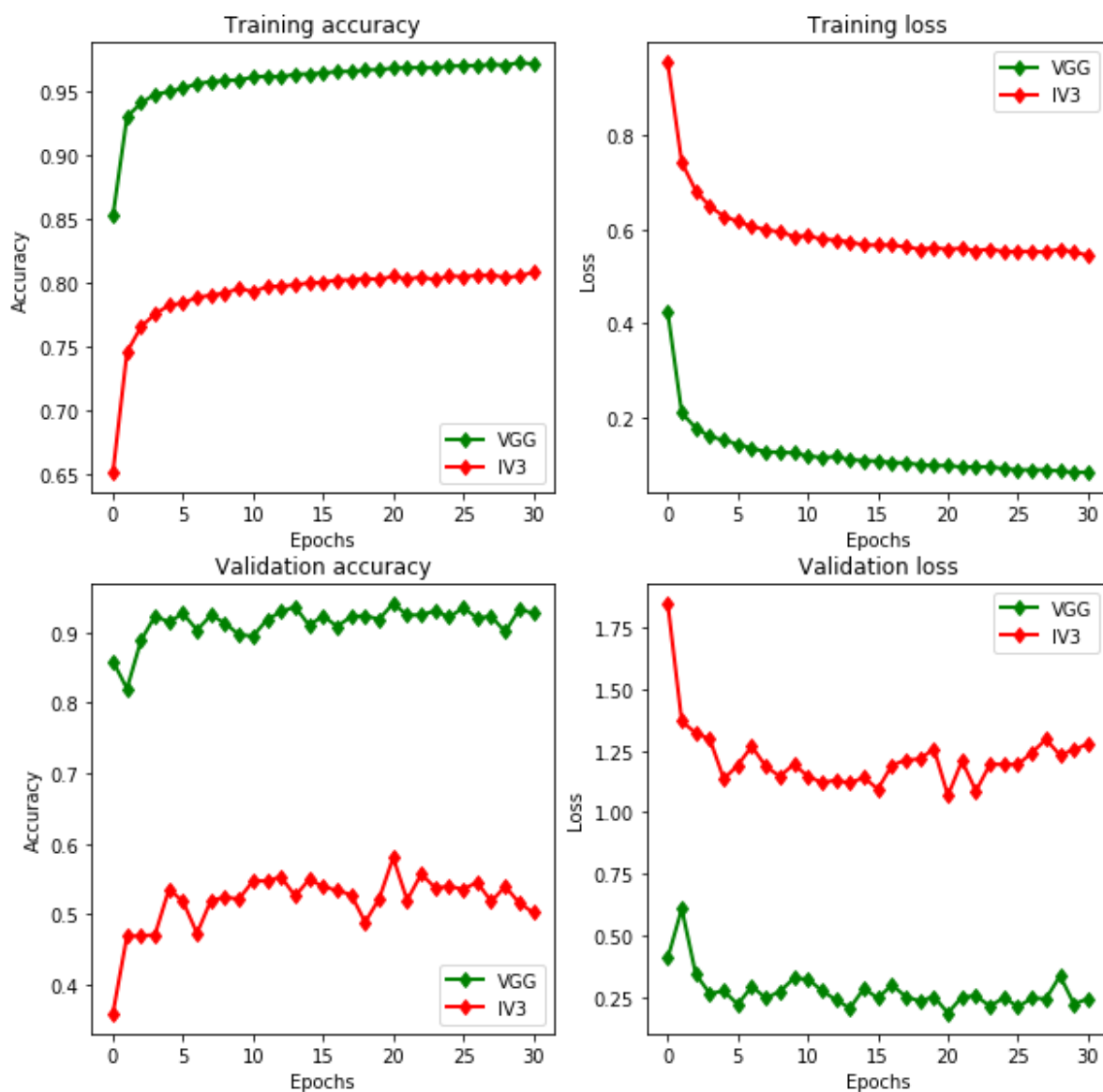
In [83]: plt.rcParams['figure.figsize'] = (10, 10)
plt.figure()
f, ax = plt.subplots(2,2)
# Training accuracy
ax[0,0].plot(vgghist['acc'], label='VGG', color='g',linewidth=2,marker=
'd')
ax[0,0].plot(iv3hist['acc'], label='IV3', color='r',linewidth=2,marker=
'd')
ax[0,0].set_title('Training accuracy')
ax[0,0].set_xlabel('Epochs')
ax[0,0].set_ylabel('Accuracy')
ax[0,0].legend()
# Training loss
ax[0,1].plot(vgghist['loss'], label='VGG', color='g',linewidth=2,marker=
'd')
ax[0,1].plot(iv3hist['loss'], label='IV3', color='r',linewidth=2,marker=
'd')
ax[0,1].set_title('Training loss')
ax[0,1].set_xlabel('Epochs')
ax[0,1].set_ylabel('Loss')
ax[0,1].legend()
# Validation accuracy
ax[1,0].plot(vgghist['val_acc'], label='VGG', color='g',linewidth=2,mark
er='d')
ax[1,0].plot(iv3hist['val_acc'], label='IV3', color='r',linewidth=2,mark
er='d')
ax[1,0].set_title('Validation accuracy')
ax[1,0].set_xlabel('Epochs')
ax[1,0].set_ylabel('Accuracy')
ax[1,0].legend()
# Validation loss
ax[1,1].plot(vgghist['val_loss'], label='VGG', color='g',linewidth=2,mar
ker='d')
ax[1,1].plot(iv3hist['val_loss'], label='IV3', color='r',linewidth=2,mar
ker='d')
ax[1,1].set_title('Validation loss')
ax[1,1].set_xlabel('Epochs')
ax[1,1].set_ylabel('Loss')
ax[1,1].legend()
plt.suptitle('Performance comparison of VGG19 and InceptionV3', fontsize
=16)

```

```
Out[83]: Text(0.5,0.98,'Performance comparison of VGG19 and InceptionV3')
```

```
<matplotlib.figure.Figure at 0x7f914fce3be0>
```

Performance comparison of VGG19 and InceptionV3



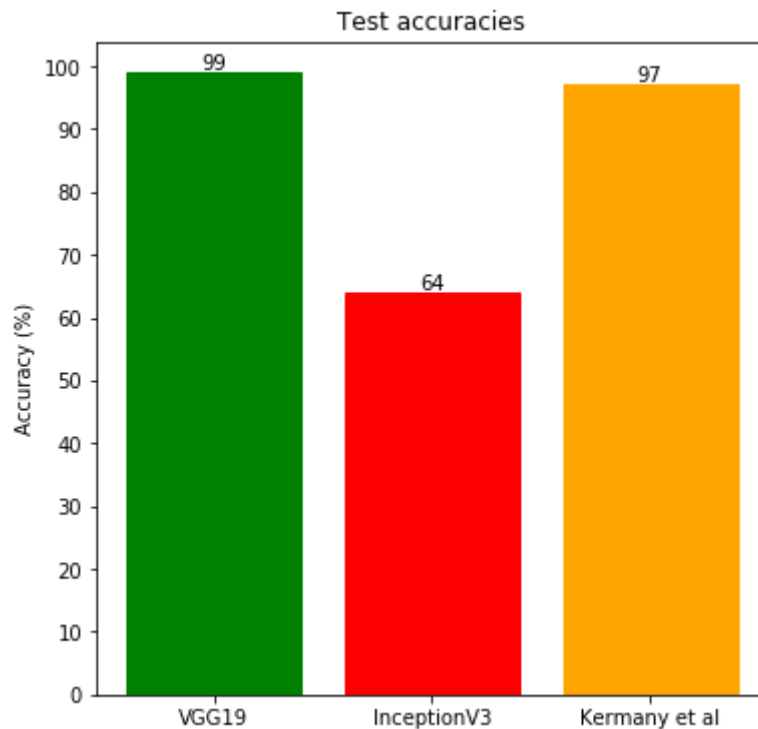
Compare accuracies with model in paper

```
In [4]: plt.rcParams['figure.figsize'] = (6, 6)

b = plt.bar([0,1,2],[round(100*vggtestloss[1]),round(100*iv3testloss[1]),97])

b[0].set_color('g')
b[1].set_color('r')
b[2].set_color('orange')
for rect in b:
    height = rect.get_height()
    plt.text(rect.get_x() + rect.get_width()/2.0, height, '%d' % int(height), ha='center', va='bottom')

plt.xticks([0,1,2],('VGG19','InceptionV3','Kermany et al'))
plt.yticks(np.linspace(0,100,11))
plt.ylabel('Accuracy (%)')
plt.title('Test accuracies')
plt.show()
```



Visualization

Load models and libraries needed - we use the help of a new module called keras-vis.

```
In [12]: from vis.utils import utils
from keras import activations
import matplotlib.cm as cm
from vis.visualization import visualize_cam, visualize_saliency, overlay

vgg_model = load_model('vgg16_1.h5', custom_objects={'tf':tf})
inceptionv3_model = load_model('InceptionV3.h5', custom_objects={'tf':tf})

# Extract the model from the parallel computing model
vgg_model = vgg_model.layers[-2]
inceptionv3_model = inceptionv3_model.layers[-2]

print('Models loaded.')
```

Models loaded.

Change the top layer of both models from softmax to linear for visualization.

```
In [14]: vgg_model.layers[-1].activation = activations.linear
vgg_model = utils.apply_modifications(vgg_model)

inceptionv3_model.layers[-1].activation = activations.linear
inceptionv3_model = utils.apply_modifications(inceptionv3_model)

print('Done.')
```

```
/home/shared/anaconda3/lib/python3.6/site-packages/keras/models.py:255:
UserWarning: No training configuration found in save file: the model was
not compiled. Compile it manually.
```

```
warnings.warn('No training configuration found in save file: '
```

Done.

Load images to view saliency maps for - two from each category

```
In [75]: %matplotlib inline
plt.rcParams['figure.figsize'] = (18, 20)

img1 = utils.load_img('data/test/CNV/CNV-6190971-1.jpeg', target_size=(256, 256, 3))
img2 = utils.load_img('data/test/CNV/CNV-5813701-3.jpeg', target_size=(256, 256, 3))
img3 = utils.load_img('data/test/DME/DME-7837305-5.jpeg', target_size=(256, 256, 3))
img4 = utils.load_img('data/test/DME/DME-9378346-1.jpeg', target_size=(256, 256, 3))
img5 = utils.load_img('data/test/DRUSEN/DRUSEN-8117834-1.jpeg', target_size=(256, 256, 3))
img6 = utils.load_img('data/test/DRUSEN/DRUSEN-8345703-2.jpeg', target_size=(256, 256, 3))
img7 = utils.load_img('data/test/NORMAL/NORMAL-1908313-1.jpeg', target_size=(256, 256, 3))
img8 = utils.load_img('data/test/NORMAL/NORMAL-2055634-1.jpeg', target_size=(256, 256, 3))

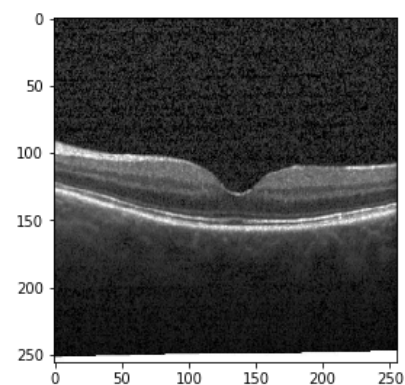
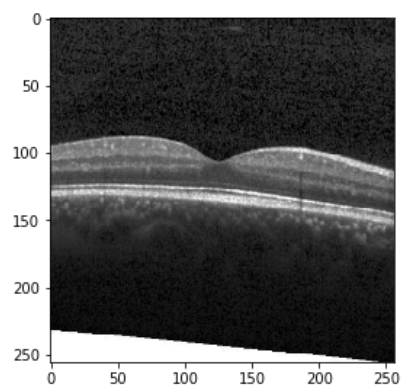
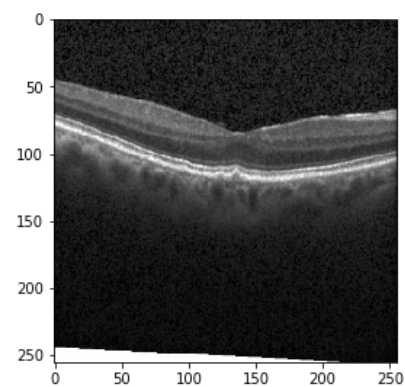
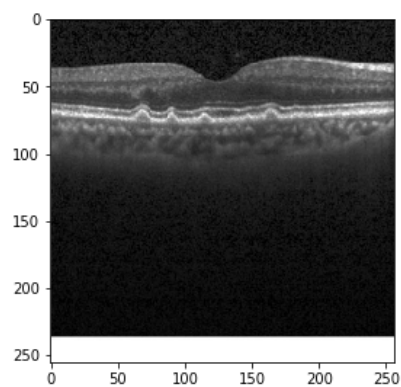
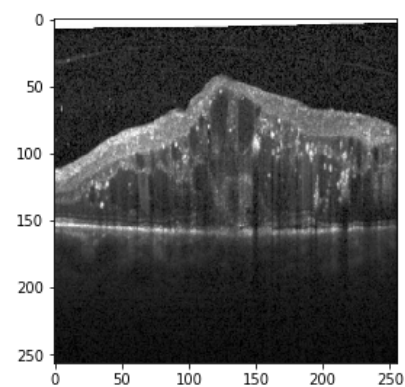
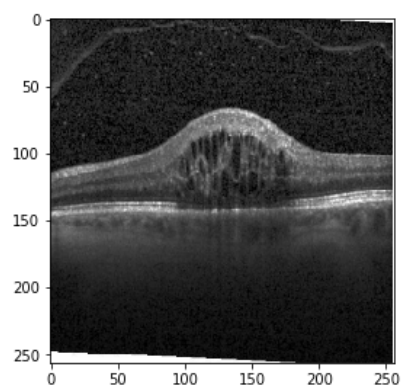
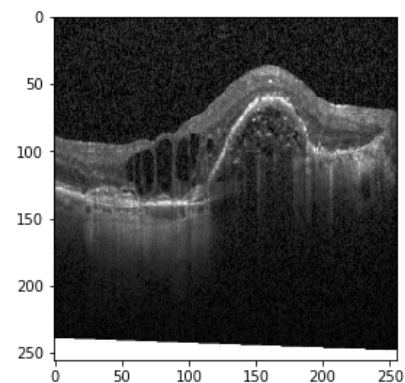
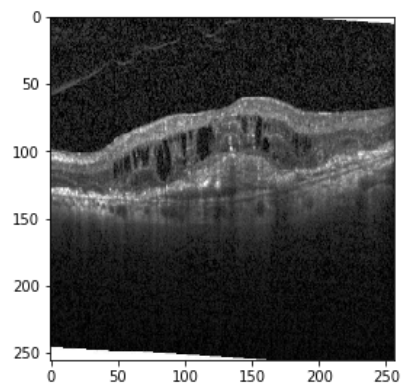
f, ax = plt.subplots(4, 2)
ax[0,0].imshow(img1[:,:,:1], cmap='gray')
ax[0,1].imshow(img2[:,:,:1], cmap='gray')
ax[1,0].imshow(img3[:,:,:1], cmap='gray')
ax[1,1].imshow(img4[:,:,:1], cmap='gray')
ax[2,0].imshow(img5[:,:,:1], cmap='gray')
ax[2,1].imshow(img6[:,:,:1], cmap='gray')
ax[3,0].imshow(img7[:,:,:1], cmap='gray')
ax[3,1].imshow(img8[:,:,:1], cmap='gray')
```



```
/home/shared/anaconda3/lib/python3.6/site-packages/skimage/transform/_warps.py:84: UserWarning: The default mode, 'constant', will be changed to 'reflect' in skimage 0.15.
```

```
warn("The default mode, 'constant', will be changed to 'reflect' in "
```

```
Out[75]: <matplotlib.image.AxesImage at 0x7f91510c9668>
```



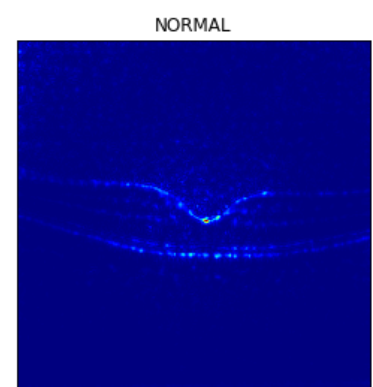
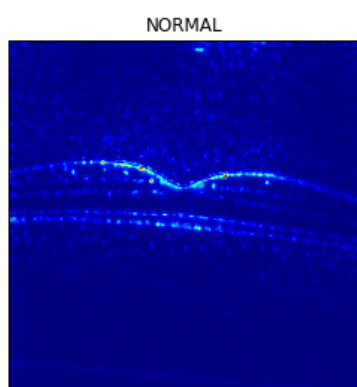
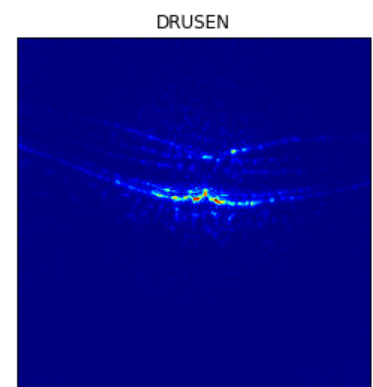
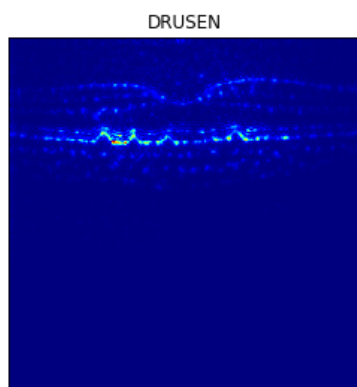
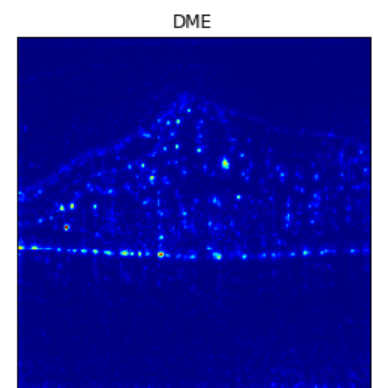
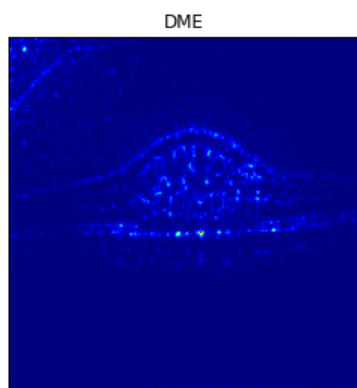
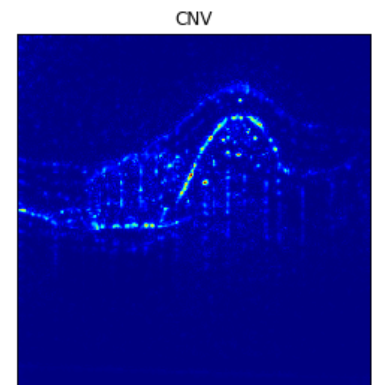
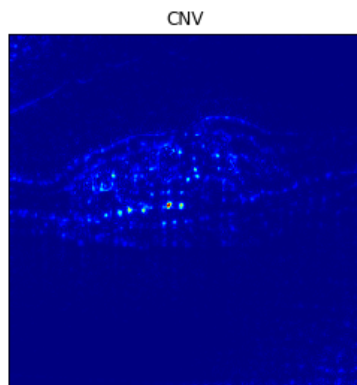
Visualize saliency for VGG19

```
In [33]: plt.figure()
f, ax = plt.subplots(4, 2)
plt.suptitle('VGG19 Saliency Maps', fontsize = 24)
for i, img in enumerate([img1, img2, img3, img4, img5, img6, img7, img8
]):
    print('Processing image %d...' % (i+1))
    grads = visualize_saliency(vgg_model, -1, filter_indices = None, see
d_input=img, backprop_modifier='guided')
    row = math.floor(i/2)
    col = i % 2
    if row == 0: title = 'CNV'
    elif row == 1: title = 'DME'
    elif row == 2: title = 'DRUSEN'
    elif row == 3: title = 'NORMAL'
    # visualize grads as heatmap
    ax[row,col].imshow(grads, cmap='jet')
    ax[row,col].set_title(title)
    ax[row,col].get_xaxis().set_visible(False)
    ax[row,col].get_yaxis().set_visible(False)
```

```
Processing image 1...  
Processing image 2...  
Processing image 3...  
Processing image 4...  
Processing image 5...  
Processing image 6...  
Processing image 7...  
Processing image 8...
```

```
<matplotlib.figure.Figure at 0x7f918b3cdda0>
```

VGG19 Saliency Maps



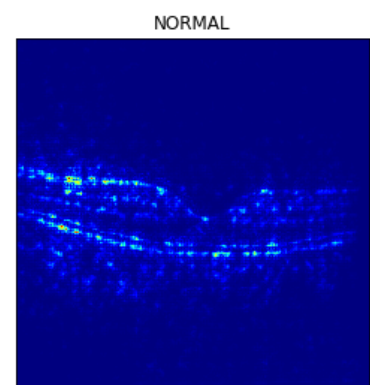
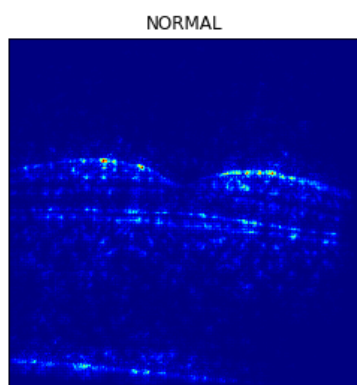
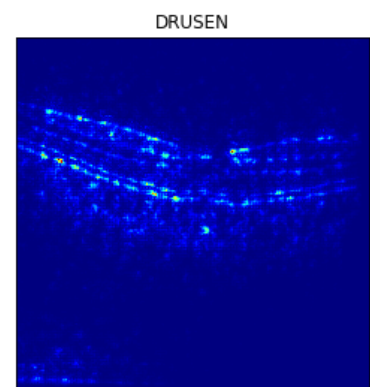
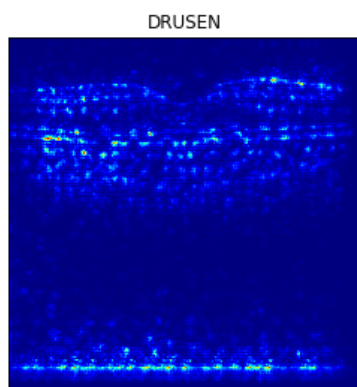
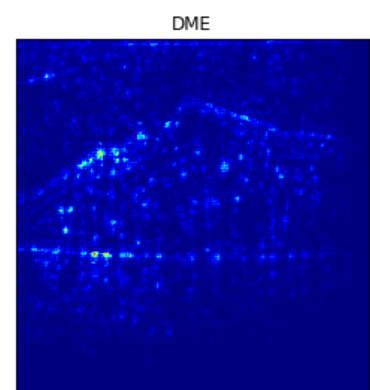
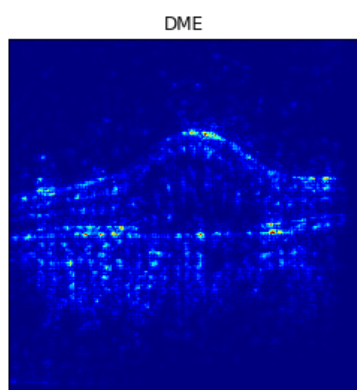
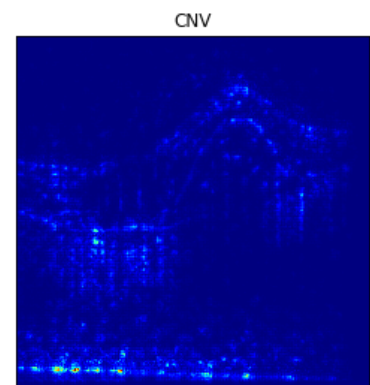
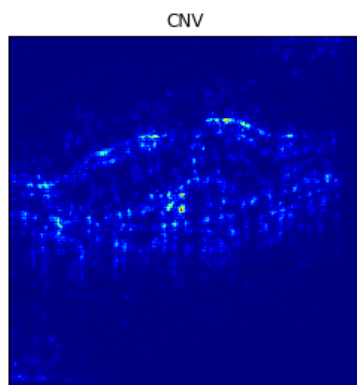
Visualize saliency for InceptionV3

```
In [34]: plt.figure()
f, ax = plt.subplots(4, 2)
plt.suptitle('InceptionV3 Saliency Maps', fontsize = 24)
for i, img in enumerate([img1, img2, img3, img4, img5, img6, img7, img8
]):
    print('Processing image %d...' % (i+1))
    grads = visualize_saliency(inceptionv3_model, -1, filter_indices = None,
seed_input=img, backprop_modifier='guided')
    row = math.floor(i/2)
    col = i % 2
    if row == 0: title = 'CNV'
    elif row == 1: title = 'DME'
    elif row == 2: title = 'DRUSEN'
    elif row == 3: title = 'NORMAL'
    # visualize grads as heatmap
    ax[row,col].imshow(grads, cmap='jet')
    ax[row,col].set_title(title)
    ax[row,col].get_xaxis().set_visible(False)
    ax[row,col].get_yaxis().set_visible(False)
```

```
Processing image 1...  
Processing image 2...  
Processing image 3...  
Processing image 4...  
Processing image 5...  
Processing image 6...  
Processing image 7...  
Processing image 8...
```

```
<matplotlib.figure.Figure at 0x7f918b46d2b0>
```


InceptionV3 Saliency Maps



Visualize class activation maps (CAM) for VGG19 and InceptionV3

```

In [76]: plt.figure()
f, ax = plt.subplots(4, 2)
plt.suptitle('InceptionV3 Class Activation Maps', fontsize = 24)
for i, img in enumerate([img1, img2, img3, img4, img5, img6, img7, img8
]):
    print('Processing image %d...' % (i+1))
    grads = visualize_cam(incep_v3_model, -1, filter_indices=None,
                        seed_input=img, backprop_modifier='guided')
    row = math.floor(i/2)
    col = i % 2
    if row == 0: title = 'CNV'
    elif row == 1: title = 'DME'
    elif row == 2: title = 'DRUSEN'
    elif row == 3: title = 'NORMAL'

    img = np.concatenate((img[:, :, 1][:, :, np.newaxis], img[:, :, 1][:, :, np.n
ewaxis], img[:, :, 1][:, :, np.newaxis]), axis=2)

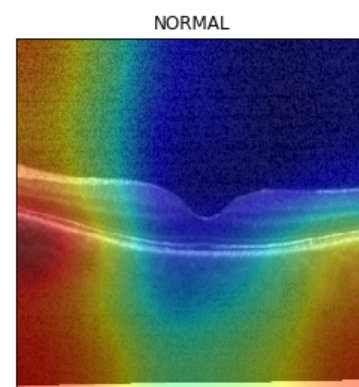
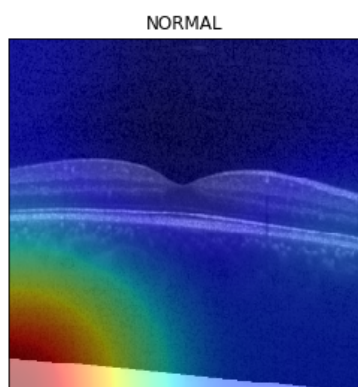
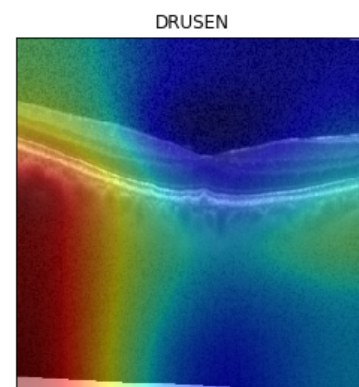
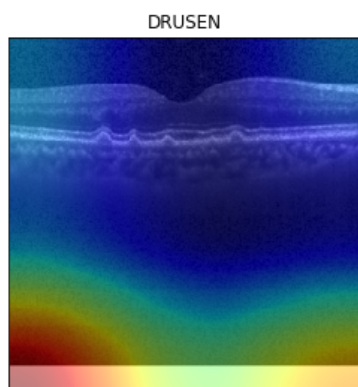
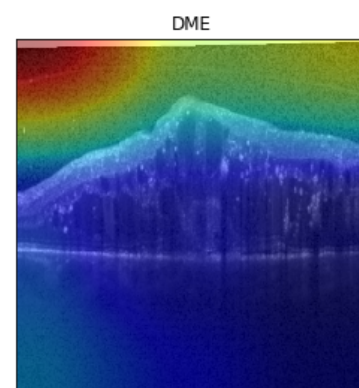
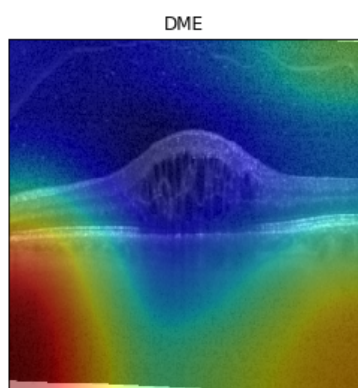
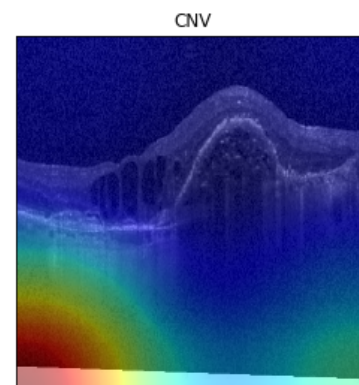
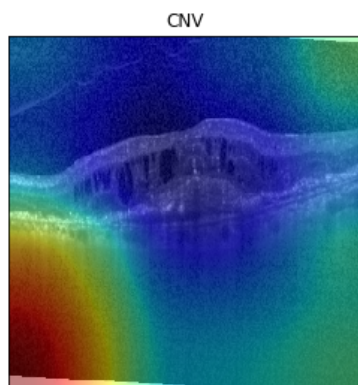
    # Lets overlay the heatmap onto original image.
    jet_heatmap = grads
    ax[row, col].imshow(overlay(jet_heatmap, img))
    ax[row, col].set_title(title)
    ax[row, col].get_xaxis().set_visible(False)
    ax[row, col].get_yaxis().set_visible(False)

```

```
Processing image 1...  
Processing image 2...  
Processing image 3...  
Processing image 4...  
Processing image 5...  
Processing image 6...  
Processing image 7...  
Processing image 8...
```

```
<matplotlib.figure.Figure at 0x7f91518ebcf8>
```

InceptionV3 Class Activation Maps



```

In [71]: plt.figure()
f, ax = plt.subplots(4, 2)
plt.suptitle('VGG19 Class Activation Maps', fontsize = 24)
for i, img in enumerate([img1, img2, img3, img4, img5, img6, img7, img8
]):
    print('Processing image %d...' % (i+1))
    grads = visualize_cam(vgg_model, -1, filter_indices=None,
                        seed_input=img, backprop_modifier='guided')
    row = math.floor(i/2)
    col = i % 2
    if row == 0: title = 'CNV'
    elif row == 1: title = 'DME'
    elif row == 2: title = 'DRUSEN'
    elif row == 3: title = 'NORMAL'

    img = np.concatenate((img[:, :, 1][:, :, np.newaxis], img[:, :, 1][:, :, np.n
ewaxis], img[:, :, 1][:, :, np.newaxis]), axis=2)

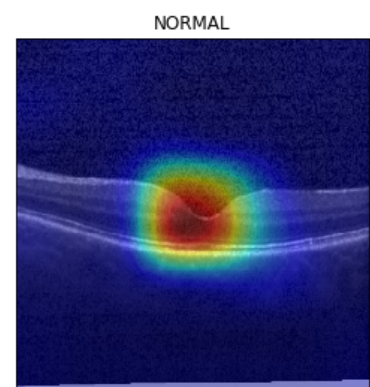
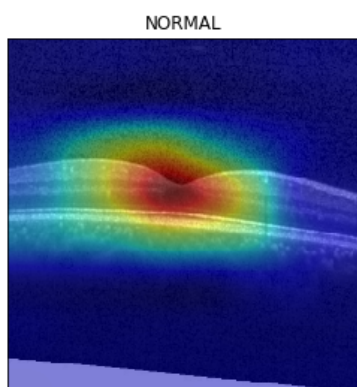
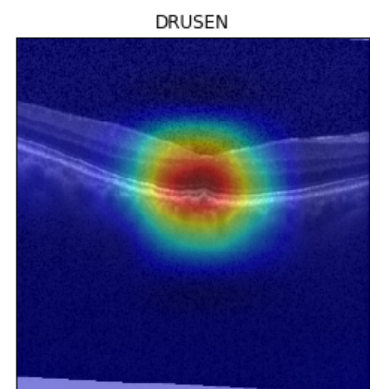
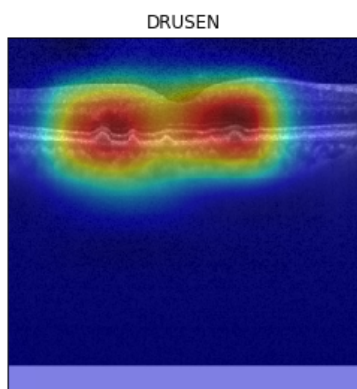
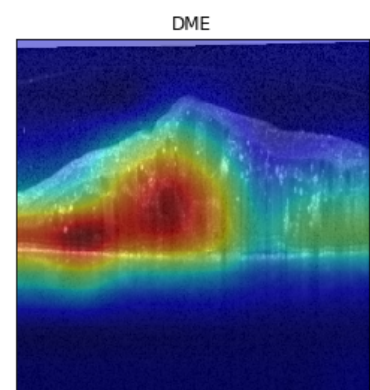
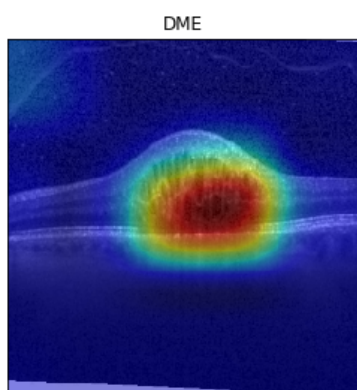
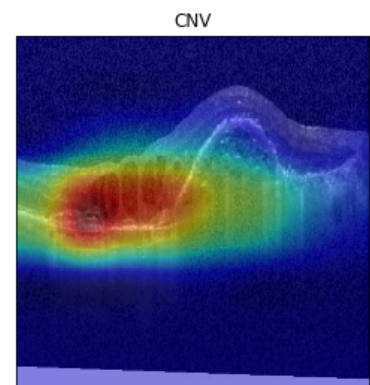
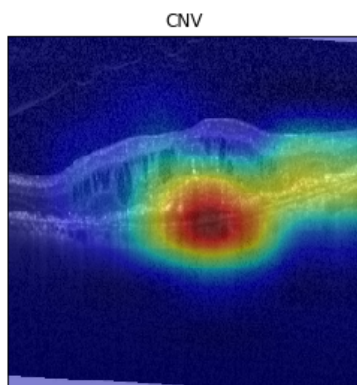
    # Lets overlay the heatmap onto original image.
    jet_heatmap = grads
    ax[row, col].imshow(overlay(jet_heatmap, img))
    ax[row, col].set_title(title)
    ax[row, col].get_xaxis().set_visible(False)
    ax[row, col].get_yaxis().set_visible(False)

```

```
Processing image 1...  
Processing image 2...  
Processing image 3...  
Processing image 4...  
Processing image 5...  
Processing image 6...  
Processing image 7...  
Processing image 8...
```

```
<matplotlib.figure.Figure at 0x7f915223b550>
```

VGG19 Class Activation Maps



ROC curves and Confusion Matrices

Load VGG

```
In [4]: # Load the VGG model
vgg = load_model('vgg16_1.h5', custom_objects={'tf':tf})
print('Done.')
```

Done.

Test VGG

```
In [6]: # Constants
img_width, img_height = 256, 256
train_data_dir = "data/train"
validation_data_dir = "data/val"
test_data_dir = "data/test"

test_datagen = ImageDataGenerator(rescale = 1./255,
                                   horizontal_flip = True,
                                   fill_mode = "nearest",
                                   zoom_range = 0.3,
                                   width_shift_range = 0.3,
                                   height_shift_range=0.3,
                                   rotation_range=30)

# Test the network!! First initiate test generator
test_generator = test_datagen.flow_from_directory(test_data_dir,
                                                  target_size = (img_height, img_width),
                                                  class_mode = "categorical",
                                                  shuffle = False)

# Now test model
y_pred = vgg.predict_generator(test_generator)
print('Done.')
```

Found 968 images belonging to 4 classes.
Done.

```
In [11]: y_pred_real = np.argmax(y_pred,axis=1)
y_test = test_generator.classes
```

Confusion matrix

```

In [13]: 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`.
    """
    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
        print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')

    print(cm)

    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)

    fmt = '.2f' if normalize else 'd'
    thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j, i, format(cm[i, j], fmt),
                 horizontalalignment="center",
                 color="white" if cm[i, j] > thresh else "black")

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

```

```
In [26]: class_names = list(test_generator.class_indices.keys())
# Compute confusion matrix
cnf_matrix = confusion_matrix(y_test, y_pred_real)
np.set_printoptions(precision=2)

# Plot non-normalized confusion matrix
plt.figure()
plot_confusion_matrix(cnf_matrix, classes=class_names,
                      title='VGG Confusion Matrix')

# Plot normalized confusion matrix
plt.figure()
plot_confusion_matrix(cnf_matrix, classes=class_names, normalize=True,
                      title='Normalized confusion matrix')

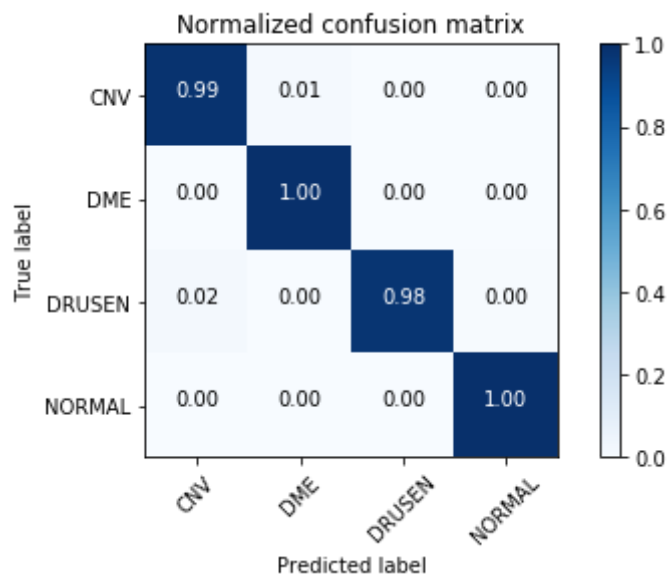
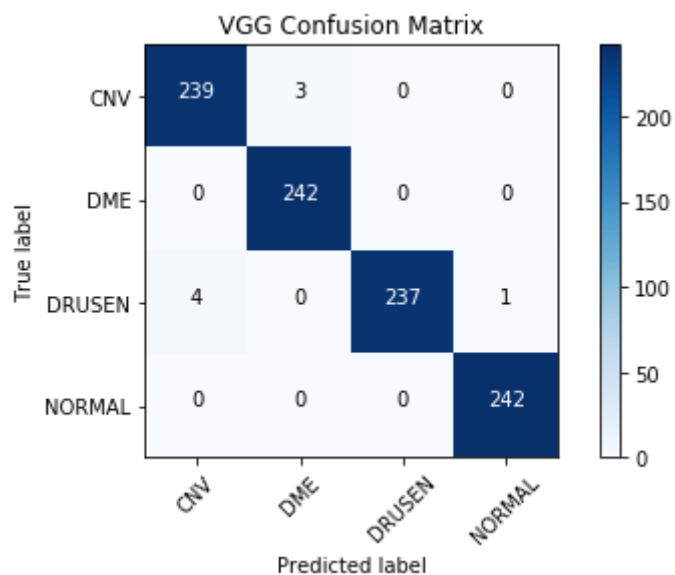
plt.show()
```

Confusion matrix, without normalization

```
[[239  3  0  0]
 [  0 242  0  0]
 [  4  0 237  1]
 [  0  0  0 242]]
```

Normalized confusion matrix

```
[[0.99 0.01 0.  0.  ]
 [0.   1.   0.  0.  ]
 [0.02 0.   0.98 0.  ]
 [0.   0.   0.   1.  ]]
```



ROC curve

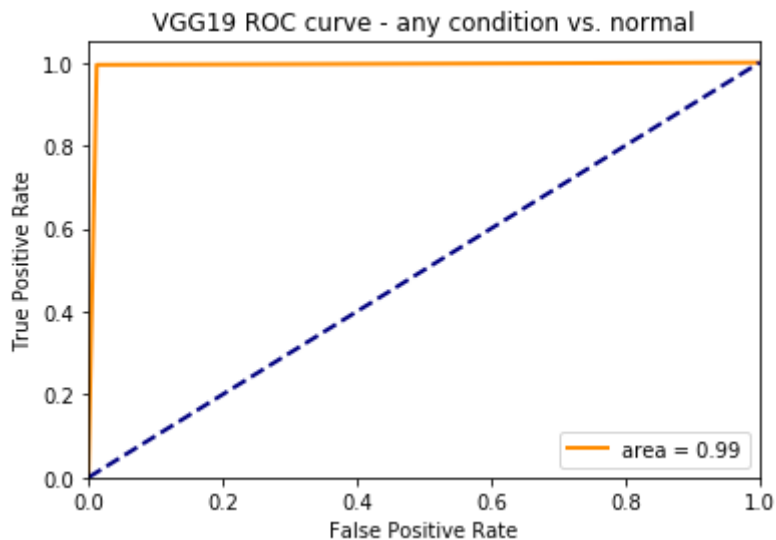
```

In [25]: # Recreate labels - 1 if not normal, 0 if normal
y_test_roc = [bool(y) for y in y_test]
y_pred_roc = [bool(y) for y in y_pred_real]

fpr, tpr, _ = roc_curve(y_test_roc, y_pred_roc)
roc_auc = auc(fpr, tpr)

plt.figure()
lw = 2
plt.plot(fpr, tpr, color='darkorange',
         lw=lw, label='area = %0.2f' % roc_auc)
plt.plot([0, 1], [0, 1], color='navy', lw=lw, linestyle='--')
plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.05])
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('VGG19 ROC curve - any condition vs. normal')
plt.legend(loc="lower right")
plt.show()

```



Load IV3

```

In [36]: # Load the VGG model
iv3 = load_model('InceptionV3.h5', custom_objects={'tf':tf})
print('Done.')

```

Done.

Test IV3

```

In [37]: # Constants
img_width, img_height = 256, 256
train_data_dir = "data/train"
validation_data_dir = "data/val"
test_data_dir = "data/test"

test_datagen = ImageDataGenerator(rescale = 1./255,
                                   horizontal_flip = True,
                                   fill_mode = "nearest",
                                   zoom_range = 0.3,
                                   width_shift_range = 0.3,
                                   height_shift_range=0.3,
                                   rotation_range=30)

# Test the network!! First initiate test generator
test_generator = test_datagen.flow_from_directory(test_data_dir,
                                                  target_size = (img_height, img_width),
                                                  class_mode = "categorical",
                                                  shuffle = False)

# Now test model
y_pred = iv3.predict_generator(test_generator)
print('Done.')

```

Found 968 images belonging to 4 classes.
Done.

```

In [38]: y_pred_real = np.argmax(y_pred,axis=1)
y_test = test_generator.classes

```

Confusion matrix

```

In [39]: 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`.
    """
    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
        print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')

    print(cm)

    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)

    fmt = '.2f' if normalize else 'd'
    thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j, i, format(cm[i, j], fmt),
                 horizontalalignment="center",
                 color="white" if cm[i, j] > thresh else "black")

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

```

```
In [40]: class_names = list(test_generator.class_indices.keys())
# Compute confusion matrix
cnf_matrix = confusion_matrix(y_test, y_pred_real)
np.set_printoptions(precision=2)

# Plot non-normalized confusion matrix
plt.figure()
plot_confusion_matrix(cnf_matrix, classes=class_names,
                      title='IV3 Confusion Matrix')

# Plot normalized confusion matrix
plt.figure()
plot_confusion_matrix(cnf_matrix, classes=class_names, normalize=True,
                      title='Normalized confusion matrix')

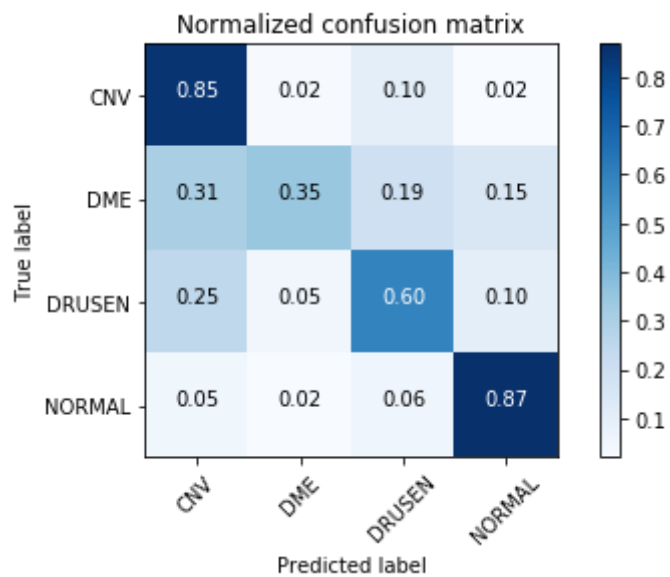
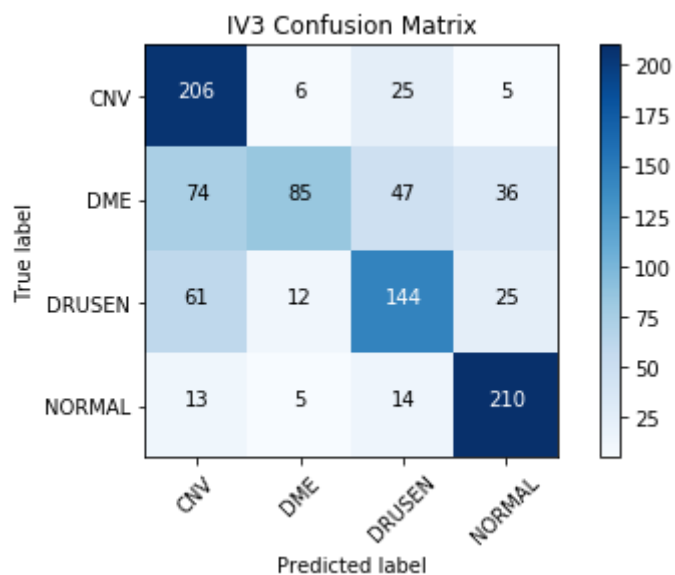
plt.show()
```


Confusion matrix, without normalization

```
[[206  6  25  5]
 [ 74 85 47 36]
 [ 61 12 144 25]
 [ 13  5  14 210]]
```

Normalized confusion matrix

```
[[0.85 0.02 0.1  0.02]
 [0.31 0.35 0.19 0.15]
 [0.25 0.05 0.6  0.1 ]
 [0.05 0.02 0.06 0.87]]
```



ROC curve

```
In [41]: # Recreate labels - 1 if not normal, 0 if normal
y_test_roc = [bool(y) for y in y_test]
y_pred_roc = [bool(y) for y in y_pred_real]

fpr, tpr, _ = roc_curve(y_test_roc, y_pred_roc)
roc_auc = auc(fpr, tpr)

plt.figure()
lw = 2
plt.plot(fpr, tpr, color='darkorange',
         lw=lw, label='area = %0.2f' % roc_auc)
plt.plot([0, 1], [0, 1], color='navy', lw=lw, linestyle='--')
plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.05])
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('IV3 ROC curve - any condition vs. normal')
plt.legend(loc="lower right")
plt.show()
```

