

VGG16 Notes:

- On first pass, validation is better than training from the start. Something feels wrong about this.
- Shuffling the data still results in the same problems
- No luck with reducing the alterations (rotation, shearing, zoom and so on) either
- Lowering the learning rate and adding momentum - works! Overfitting starts happening around epoch 5, so we'll train for 5 epochs in the final model.

InceptionV3 Notes:

- Overfit right away, even with slow learning rate, slow momentum and lots of training data manipulation (shearing, zoom, crop, and so on).
- Adding dropout makes it difficult to learn (accuracy below 90%)
- Even allowing a few of the later layers of the inception model to be trainable results in a really awful accuracy + loss curve (see below).
- Since we were able to achieve approx. 95% accuracy with the vgg16-based model, it doesn't seem worth it to continue fine-tuning the inception model, especially at 30 minutes per epoch!

```
In [11]: # Standard modules
from IPython.display import display
import itertools
import random

# Third-party modules
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from pathlib import Path
import pydot_ng as pydot
import shutil
from sklearn.metrics import confusion_matrix
from tensorflow.keras import optimizers
from tensorflow.keras.applications import VGG16, InceptionV3
from tensorflow.keras.layers import Flatten, Dense, Dropout
from tensorflow.keras.models import Sequential, load_model
from tensorflow.keras.preprocessing.image import ImageDataGenerator, array_to_img
from tensorflow.keras.utils import plot_model
```

```
In [5]: random.seed(42)
plt.xkcd()
# The following is a very ugly hack to get the plot_model function to work properly
import os
os.environ['PATH'] = os.environ['PATH']+';' + os.environ['CONDA_PREFIX'] + r"\Library\bin\graphviz"
```

```
In [3]: img_attribute_file = (r'D:\Google_Drive_Sync\Data\Documents\School\UL\Independent_Study_AI_ML'
                                r'\Celeb_Photos_20191114\list_attr_celeba_csv_correction_headers.txt')
attributes = pd.read_csv(img_attribute_file).sample(frac=1).reset_index(drop=True)

# Use all images for this binary classification problem.
males = attributes[['FileName']][attributes['Male']==1]
females = attributes[['FileName']][attributes['Male']==-1]

training_count = round(0.6 * len(males)) + round(0.6 * len(females))
val_count = round(0.2 * len(males)) + round(0.2 * len(females))
```

```
In [4]: src_root = Path(r'D:\Google_Drive_Sync\Data\Documents\School\UL\Independent_Study_AI_ML'
                        r'\Celeb_Photos_20191114\img_align_celeba')

dst_root = Path(r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                r'\Assignment_2\Images')
shutil.rmtree(dst_root)
```

Let's use a 60%, 20%, 20% split for training, validation, and test. That's roughly 120k, 40k, and 40k respectively.

```
In [5]: train_male_pth = Path(r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                              r'\Assignment_2\Images\train\male')
train_male_pth.mkdir(parents=True, exist_ok=True)
for img in np.array(males[:round(0.6 * len(males))]):
    shutil.copy(src_root / img[0], train_male_pth)
```

```
In [6]: train_female_pth = Path(r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                                 r'\Assignment_2\Images\train\female')
train_female_pth.mkdir(parents=True, exist_ok=True)
for img in np.array(females[:round(0.6*len(females))]):
    shutil.copy(src_root / img[0], train_female_pth)
```

```
In [7]: validation_male_pth = Path(r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                                r'\Assignment_2\Images\validation\male')
validation_male_pth.mkdir(parents=True, exist_ok=True)
for img in np.array(males[round(0.6 * len(males)):round(0.8 * len(males))]):
    shutil.copy(src_root / img[0], validation_male_pth)
```

```
In [8]: validation_female_pth = Path(r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learnin
g'
                                    r'\Assignment_2\Images\validation\female')
validation_female_pth.mkdir(parents=True, exist_ok=True)
for img in np.array(females[round(0.6*len(females)):round(0.8*len(females))]):
    shutil.copy(src_root / img[0], validation_female_pth)
```

```
In [9]: test_male_pth = Path(r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                             r'\Assignment_2\Images\test\male')
test_male_pth.mkdir(parents=True, exist_ok=True)
for img in np.array(males[round(0.8*len(males)):]):
    shutil.copy(src_root / img[0], test_male_pth)
```

```
In [10]: test_female_pth = Path(r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                                r'\Assignment_2\Images\test\female')
test_female_pth.mkdir(parents=True, exist_ok=True)
for img in np.array(females[round(0.8*len(females)):]):
    shutil.copy(src_root / img[0], test_female_pth)
```

```
In [6]: vgg16_conv_base = VGG16(weights='imagenet',
include_top=False,
input_shape=(178, 218, 3))

vgg16_md1 = Sequential()
vgg16_md1.add(vgg16_conv_base)
vgg16_md1.add(Flatten())
vgg16_md1.add(Dense(256, activation='relu'))
vgg16_md1.add(Dense(1, activation='sigmoid'))

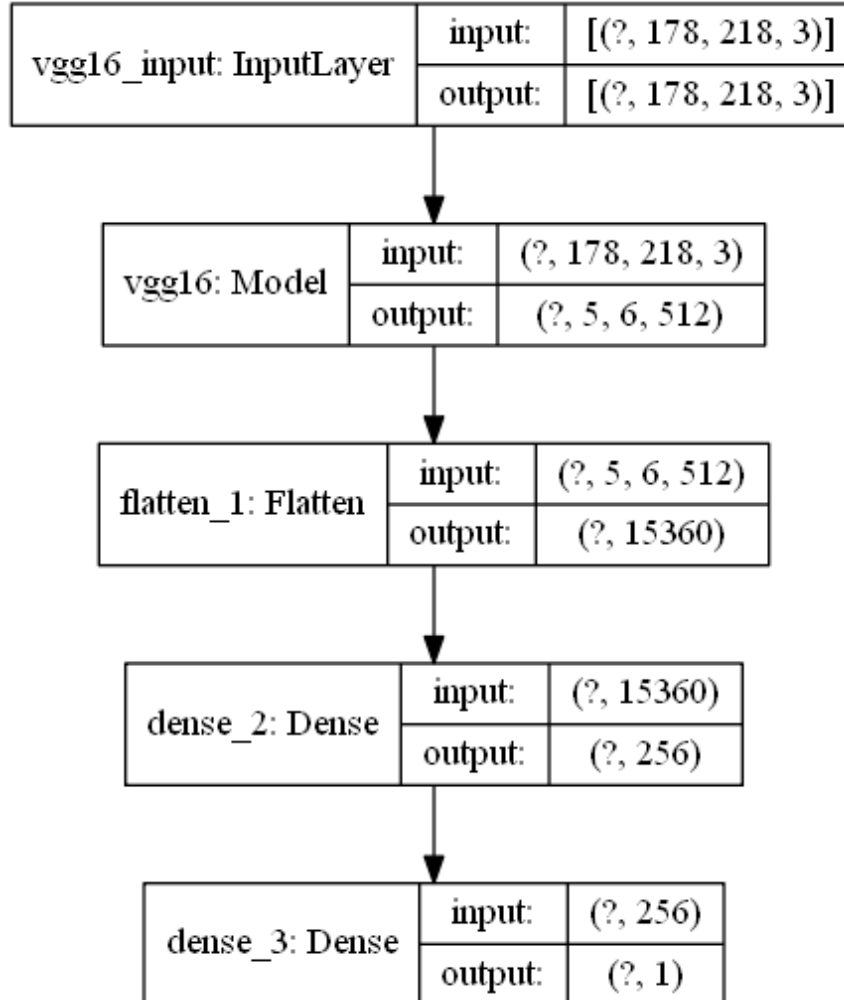
vgg16_conv_base.trainable = False
```

```
In [12]: vgg16_md1.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
vgg16 (Model)	(None, 5, 6, 512)	14714688
flatten (Flatten)	(None, 15360)	0
dense (Dense)	(None, 256)	3932416
dense_1 (Dense)	(None, 1)	257
=====		
Total params: 18,647,361		
Trainable params: 3,932,673		
Non-trainable params: 14,714,688		
=====		

```
In [10]: plot_model(vgg16_md1, to_file='model_plot.png', show_shapes=True, show_layer_names=True)
```



```
In [13]: train_dir = (r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                    r'\Assignment_2\Images\train')
validation_dir = (r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                  r'\Assignment_2\Images\validation')

train_datagen = ImageDataGenerator(
```

```
        rescale=1./255,
        #rotation_range=40,
        #width_shift_range=0.2,
        #height_shift_range=0.2,
        #shear_range=0.2,
        #zoom_range=0.2,
        #horizontal_flip=True,
        fill_mode='nearest')

val_datagen = ImageDataGenerator(rescale=1./255)

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(178, 218),
    batch_size=20,
    class_mode='binary')

validation_generator = val_datagen.flow_from_directory(
    validation_dir,
    target_size=(178, 218),
    batch_size=20,
    class_mode='binary')

vgg16_md1.compile(loss='binary_crossentropy',
    optimizer=optimizers.RMSprop(learning_rate=0.000005, rho=0.9),
    metrics=['acc'])

history = vgg16_md1.fit_generator(
    train_generator,
    steps_per_epoch=round(training_count/20),
    epochs=20,
    validation_data=validation_generator,
    validation_steps= round(val_count/20))
```




```
Found 121559 images belonging to 2 classes.
Found 40520 images belonging to 2 classes.
WARNING:tensorflow:From <ipython-input-13-be1cb70322cd>:39: Model.fit_generator (from tensorflow.python.keras.engine.training) is deprecated and will be removed in a future version.
Instructions for updating:
Please use Model.fit, which supports generators.
WARNING:tensorflow:sample_weight modes were coerced from
...
to
['...']
WARNING:tensorflow:sample_weight modes were coerced from
...
to
['...']
Train for 6078 steps, validate for 2026 steps
Epoch 1/20
6078/6078 [=====] - 2644s 435ms/step - loss: 0.2062 - acc: 0.9188 - val_loss: 0.1680 - val_acc: 0.9352
Epoch 2/20
6078/6078 [=====] - 1890s 311ms/step - loss: 0.1529 - acc: 0.9409 - val_loss: 0.1689 - val_acc: 0.9331
Epoch 3/20
6078/6078 [=====] - 1891s 311ms/step - loss: 0.1399 - acc: 0.9462 - val_loss: 0.1449 - val_acc: 0.9451
Epoch 4/20
6078/6078 [=====] - 1888s 311ms/step - loss: 0.1321 - acc: 0.9495 - val_loss: 0.1411 - val_acc: 0.9464
Epoch 5/20
6078/6078 [=====] - 1890s 311ms/step - loss: 0.1269 - acc: 0.9516 - val_loss: 0.1419 - val_acc: 0.9463
Epoch 6/20
6078/6078 [=====] - 1887s 311ms/step - loss: 0.1223 - acc: 0.9537 - val_loss: 0.1364 - val_acc: 0.9493
Epoch 7/20
6078/6078 [=====] - 1887s 311ms/step - loss: 0.1192 - acc: 0.9548 - val_loss: 0.1354 - val_acc: 0.9498
Epoch 8/20
6078/6078 [=====] - 1888s 311ms/step - loss: 0.1162 - acc: 0.9561 - val_loss: 0.1370 - val_acc: 0.9486
Epoch 9/20
6078/6078 [=====] - 1889s 311ms/step - loss: 0.1136 - acc: 0.9575 - val_loss: 0.1341 - val_acc: 0.9510
Epoch 10/20
6078/6078 [=====] - 1888s 311ms/step - loss: 0.1116 - acc: 0.9584 - val_loss: 0.1345 - val_acc: 0.9501
Epoch 11/20
6078/6078 [=====] - 1888s 311ms/step - loss: 0.1094 - acc: 0.9593 - val_loss: 0.1341 - val_acc: 0.9510
Epoch 12/20
6078/6078 [=====] - 1888s 311ms/step - loss: 0.1076 - acc: 0.9600 - val_loss: 0.1333 - val_acc: 0.9514
Epoch 13/20
6078/6078 [=====] - 1888s 311ms/step - loss: 0.1060 - acc: 0.9612 - val_loss: 0.1345 - val_acc: 0.9513
Epoch 14/20
6078/6078 [=====] - 1888s 311ms/step - loss: 0.1043 - acc: 0.9619 - val_loss: 0.1340 - val_acc: 0.9517
```

Epoch 15/20

6078/6078 [=====] - 1888s 311ms/step - loss: 0.1029 - acc: 0.9622 - val_loss: 0.1336 - val_acc: 0.9520

Epoch 16/20

6078/6078 [=====] - 1887s 310ms/step - loss: 0.1017 - acc: 0.9631 - val_loss: 0.1346 - val_acc: 0.9519

Epoch 17/20

6078/6078 [=====] - 1886s 310ms/step - loss: 0.1002 - acc: 0.9636 - val_loss: 0.1392 - val_acc: 0.9501

Epoch 18/20

6078/6078 [=====] - 1886s 310ms/step - loss: 0.0989 - acc: 0.9644 - val_loss: 0.1363 - val_acc: 0.9521

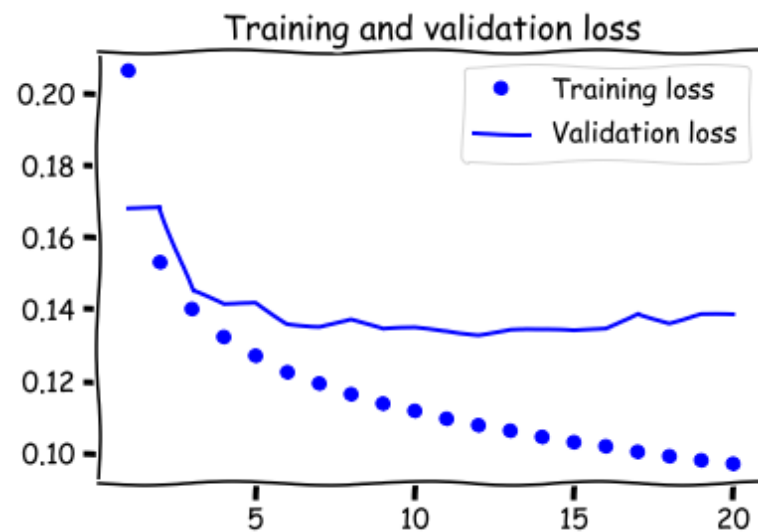
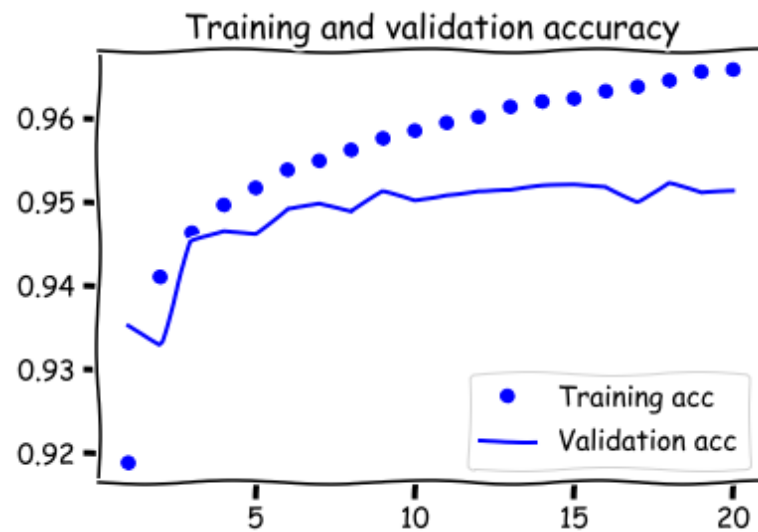
Epoch 19/20

6078/6078 [=====] - 1885s 310ms/step - loss: 0.0978 - acc: 0.9654 - val_loss: 0.1383 - val_acc: 0.9508

Epoch 20/20

6078/6078 [=====] - 1884s 310ms/step - loss: 0.0969 - acc: 0.9657 - val_loss: 0.1380 - val_acc: 0.9512

```
In [14]: acc = history.history['acc']
val_acc = history.history['val_acc']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs = range(1, len(acc) + 1)
plt.plot(epochs, acc, 'bo', label='Training acc')
plt.plot(epochs, val_acc, 'b', label='Validation acc')
plt.title('Training and validation accuracy')
plt.legend()
plt.figure()
plt.plot(epochs, loss, 'bo', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
plt.title('Training and validation loss')
plt.legend()
plt.show()
```



```
In [20]: inception_conv_base = InceptionV3(weights='imagenet',  
include_top=False,  
input_shape=(178, 218, 3))
```

```
In [21]: inception_conv_base.summary()
```

Model: "inception_v3"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_5 (InputLayer)	[(None, 178, 218, 3)]	0	
conv2d_376 (Conv2D)	(None, 88, 108, 32)	864	input_5[0][0]
batch_normalization_376 (Batch Normalization)	(None, 88, 108, 32)	96	conv2d_376[0][0]
activation_376 (Activation)	(None, 88, 108, 32)	0	batch_normalization_376[0][0]
conv2d_377 (Conv2D)	(None, 86, 106, 32)	9216	activation_376[0][0]
batch_normalization_377 (Batch Normalization)	(None, 86, 106, 32)	96	conv2d_377[0][0]
activation_377 (Activation)	(None, 86, 106, 32)	0	batch_normalization_377[0][0]
conv2d_378 (Conv2D)	(None, 86, 106, 64)	18432	activation_377[0][0]
batch_normalization_378 (Batch Normalization)	(None, 86, 106, 64)	192	conv2d_378[0][0]
activation_378 (Activation)	(None, 86, 106, 64)	0	batch_normalization_378[0][0]
max_pooling2d_16 (MaxPooling2D)	(None, 42, 52, 64)	0	activation_378[0][0]
conv2d_379 (Conv2D)	(None, 42, 52, 80)	5120	max_pooling2d_16[0][0]
batch_normalization_379 (Batch Normalization)	(None, 42, 52, 80)	240	conv2d_379[0][0]
activation_379 (Activation)	(None, 42, 52, 80)	0	batch_normalization_379[0][0]
conv2d_380 (Conv2D)	(None, 40, 50, 192)	138240	activation_379[0][0]
batch_normalization_380 (Batch Normalization)	(None, 40, 50, 192)	576	conv2d_380[0][0]
activation_380 (Activation)	(None, 40, 50, 192)	0	batch_normalization_380[0][0]
max_pooling2d_17 (MaxPooling2D)	(None, 19, 24, 192)	0	activation_380[0][0]
conv2d_384 (Conv2D)	(None, 19, 24, 64)	12288	max_pooling2d_17[0][0]
batch_normalization_384 (Batch Normalization)	(None, 19, 24, 64)	192	conv2d_384[0][0]

activation_384 (Activation)	(None, 19, 24, 64)	0	batch_normalization_384[0][0]
conv2d_382 (Conv2D)	(None, 19, 24, 48)	9216	max_pooling2d_17[0][0]
conv2d_385 (Conv2D)	(None, 19, 24, 96)	55296	activation_384[0][0]
batch_normalization_382 (BatchN	(None, 19, 24, 48)	144	conv2d_382[0][0]
batch_normalization_385 (BatchN	(None, 19, 24, 96)	288	conv2d_385[0][0]
activation_382 (Activation)	(None, 19, 24, 48)	0	batch_normalization_382[0][0]
activation_385 (Activation)	(None, 19, 24, 96)	0	batch_normalization_385[0][0]
average_pooling2d_36 (AveragePo	(None, 19, 24, 192)	0	max_pooling2d_17[0][0]
conv2d_381 (Conv2D)	(None, 19, 24, 64)	12288	max_pooling2d_17[0][0]
conv2d_383 (Conv2D)	(None, 19, 24, 64)	76800	activation_382[0][0]
conv2d_386 (Conv2D)	(None, 19, 24, 96)	82944	activation_385[0][0]
conv2d_387 (Conv2D)	(None, 19, 24, 32)	6144	average_pooling2d_36[0][0]
batch_normalization_381 (BatchN	(None, 19, 24, 64)	192	conv2d_381[0][0]
batch_normalization_383 (BatchN	(None, 19, 24, 64)	192	conv2d_383[0][0]
batch_normalization_386 (BatchN	(None, 19, 24, 96)	288	conv2d_386[0][0]
batch_normalization_387 (BatchN	(None, 19, 24, 32)	96	conv2d_387[0][0]
activation_381 (Activation)	(None, 19, 24, 64)	0	batch_normalization_381[0][0]
activation_383 (Activation)	(None, 19, 24, 64)	0	batch_normalization_383[0][0]
activation_386 (Activation)	(None, 19, 24, 96)	0	batch_normalization_386[0][0]
activation_387 (Activation)	(None, 19, 24, 32)	0	batch_normalization_387[0][0]
mixed0 (Concatenate)	(None, 19, 24, 256)	0	activation_381[0][0] activation_383[0][0]

			activation_386[0][0]
			activation_387[0][0]
conv2d_391 (Conv2D)	(None, 19, 24, 64)	16384	mixed0[0][0]
batch_normalization_391 (BatchN	(None, 19, 24, 64)	192	conv2d_391[0][0]
activation_391 (Activation)	(None, 19, 24, 64)	0	batch_normalization_391[0][0]
conv2d_389 (Conv2D)	(None, 19, 24, 48)	12288	mixed0[0][0]
conv2d_392 (Conv2D)	(None, 19, 24, 96)	55296	activation_391[0][0]
batch_normalization_389 (BatchN	(None, 19, 24, 48)	144	conv2d_389[0][0]
batch_normalization_392 (BatchN	(None, 19, 24, 96)	288	conv2d_392[0][0]
activation_389 (Activation)	(None, 19, 24, 48)	0	batch_normalization_389[0][0]
activation_392 (Activation)	(None, 19, 24, 96)	0	batch_normalization_392[0][0]
average_pooling2d_37 (AveragePo	(None, 19, 24, 256)	0	mixed0[0][0]
conv2d_388 (Conv2D)	(None, 19, 24, 64)	16384	mixed0[0][0]
conv2d_390 (Conv2D)	(None, 19, 24, 64)	76800	activation_389[0][0]
conv2d_393 (Conv2D)	(None, 19, 24, 96)	82944	activation_392[0][0]
conv2d_394 (Conv2D)	(None, 19, 24, 64)	16384	average_pooling2d_37[0][0]
batch_normalization_388 (BatchN	(None, 19, 24, 64)	192	conv2d_388[0][0]
batch_normalization_390 (BatchN	(None, 19, 24, 64)	192	conv2d_390[0][0]
batch_normalization_393 (BatchN	(None, 19, 24, 96)	288	conv2d_393[0][0]
batch_normalization_394 (BatchN	(None, 19, 24, 64)	192	conv2d_394[0][0]
activation_388 (Activation)	(None, 19, 24, 64)	0	batch_normalization_388[0][0]
activation_390 (Activation)	(None, 19, 24, 64)	0	batch_normalization_390[0][0]

activation_393 (Activation)	(None, 19, 24, 96)	0	batch_normalization_393[0][0]
activation_394 (Activation)	(None, 19, 24, 64)	0	batch_normalization_394[0][0]
mixed1 (Concatenate)	(None, 19, 24, 288)	0	activation_388[0][0] activation_390[0][0] activation_393[0][0] activation_394[0][0]
conv2d_398 (Conv2D)	(None, 19, 24, 64)	18432	mixed1[0][0]
batch_normalization_398 (BatchN	(None, 19, 24, 64)	192	conv2d_398[0][0]
activation_398 (Activation)	(None, 19, 24, 64)	0	batch_normalization_398[0][0]
conv2d_396 (Conv2D)	(None, 19, 24, 48)	13824	mixed1[0][0]
conv2d_399 (Conv2D)	(None, 19, 24, 96)	55296	activation_398[0][0]
batch_normalization_396 (BatchN	(None, 19, 24, 48)	144	conv2d_396[0][0]
batch_normalization_399 (BatchN	(None, 19, 24, 96)	288	conv2d_399[0][0]
activation_396 (Activation)	(None, 19, 24, 48)	0	batch_normalization_396[0][0]
activation_399 (Activation)	(None, 19, 24, 96)	0	batch_normalization_399[0][0]
average_pooling2d_38 (AveragePo	(None, 19, 24, 288)	0	mixed1[0][0]
conv2d_395 (Conv2D)	(None, 19, 24, 64)	18432	mixed1[0][0]
conv2d_397 (Conv2D)	(None, 19, 24, 64)	76800	activation_396[0][0]
conv2d_400 (Conv2D)	(None, 19, 24, 96)	82944	activation_399[0][0]
conv2d_401 (Conv2D)	(None, 19, 24, 64)	18432	average_pooling2d_38[0][0]
batch_normalization_395 (BatchN	(None, 19, 24, 64)	192	conv2d_395[0][0]
batch_normalization_397 (BatchN	(None, 19, 24, 64)	192	conv2d_397[0][0]
batch_normalization_400 (BatchN	(None, 19, 24, 96)	288	conv2d_400[0][0]

batch_normalization_401 (BatchN	(None, 19, 24, 64)	192	conv2d_401[0][0]
activation_395 (Activation)	(None, 19, 24, 64)	0	batch_normalization_395[0][0]
activation_397 (Activation)	(None, 19, 24, 64)	0	batch_normalization_397[0][0]
activation_400 (Activation)	(None, 19, 24, 96)	0	batch_normalization_400[0][0]
activation_401 (Activation)	(None, 19, 24, 64)	0	batch_normalization_401[0][0]
mixed2 (Concatenate)	(None, 19, 24, 288)	0	activation_395[0][0] activation_397[0][0] activation_400[0][0] activation_401[0][0]
conv2d_403 (Conv2D)	(None, 19, 24, 64)	18432	mixed2[0][0]
batch_normalization_403 (BatchN	(None, 19, 24, 64)	192	conv2d_403[0][0]
activation_403 (Activation)	(None, 19, 24, 64)	0	batch_normalization_403[0][0]
conv2d_404 (Conv2D)	(None, 19, 24, 96)	55296	activation_403[0][0]
batch_normalization_404 (BatchN	(None, 19, 24, 96)	288	conv2d_404[0][0]
activation_404 (Activation)	(None, 19, 24, 96)	0	batch_normalization_404[0][0]
conv2d_402 (Conv2D)	(None, 9, 11, 384)	995328	mixed2[0][0]
conv2d_405 (Conv2D)	(None, 9, 11, 96)	82944	activation_404[0][0]
batch_normalization_402 (BatchN	(None, 9, 11, 384)	1152	conv2d_402[0][0]
batch_normalization_405 (BatchN	(None, 9, 11, 96)	288	conv2d_405[0][0]
activation_402 (Activation)	(None, 9, 11, 384)	0	batch_normalization_402[0][0]
activation_405 (Activation)	(None, 9, 11, 96)	0	batch_normalization_405[0][0]
max_pooling2d_18 (MaxPooling2D)	(None, 9, 11, 288)	0	mixed2[0][0]
mixed3 (Concatenate)	(None, 9, 11, 768)	0	activation_402[0][0] activation_405[0][0]

			max_pooling2d_18[0][0]
conv2d_410 (Conv2D)	(None, 9, 11, 128)	98304	mixed3[0][0]
batch_normalization_410 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_410[0][0]
activation_410 (Activation)	(None, 9, 11, 128)	0	batch_normalization_410[0][0]
conv2d_411 (Conv2D)	(None, 9, 11, 128)	114688	activation_410[0][0]
batch_normalization_411 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_411[0][0]
activation_411 (Activation)	(None, 9, 11, 128)	0	batch_normalization_411[0][0]
conv2d_407 (Conv2D)	(None, 9, 11, 128)	98304	mixed3[0][0]
conv2d_412 (Conv2D)	(None, 9, 11, 128)	114688	activation_411[0][0]
batch_normalization_407 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_407[0][0]
batch_normalization_412 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_412[0][0]
activation_407 (Activation)	(None, 9, 11, 128)	0	batch_normalization_407[0][0]
activation_412 (Activation)	(None, 9, 11, 128)	0	batch_normalization_412[0][0]
conv2d_408 (Conv2D)	(None, 9, 11, 128)	114688	activation_407[0][0]
conv2d_413 (Conv2D)	(None, 9, 11, 128)	114688	activation_412[0][0]
batch_normalization_408 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_408[0][0]
batch_normalization_413 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_413[0][0]
activation_408 (Activation)	(None, 9, 11, 128)	0	batch_normalization_408[0][0]
activation_413 (Activation)	(None, 9, 11, 128)	0	batch_normalization_413[0][0]
average_pooling2d_39 (Average Pooling)	(None, 9, 11, 768)	0	mixed3[0][0]
conv2d_406 (Conv2D)	(None, 9, 11, 192)	147456	mixed3[0][0]
conv2d_409 (Conv2D)	(None, 9, 11, 192)	172032	activation_408[0][0]

conv2d_414 (Conv2D)	(None, 9, 11, 192)	172032	activation_413[0][0]
conv2d_415 (Conv2D)	(None, 9, 11, 192)	147456	average_pooling2d_39[0][0]
batch_normalization_406 (BatchN	(None, 9, 11, 192)	576	conv2d_406[0][0]
batch_normalization_409 (BatchN	(None, 9, 11, 192)	576	conv2d_409[0][0]
batch_normalization_414 (BatchN	(None, 9, 11, 192)	576	conv2d_414[0][0]
batch_normalization_415 (BatchN	(None, 9, 11, 192)	576	conv2d_415[0][0]
activation_406 (Activation)	(None, 9, 11, 192)	0	batch_normalization_406[0][0]
activation_409 (Activation)	(None, 9, 11, 192)	0	batch_normalization_409[0][0]
activation_414 (Activation)	(None, 9, 11, 192)	0	batch_normalization_414[0][0]
activation_415 (Activation)	(None, 9, 11, 192)	0	batch_normalization_415[0][0]
mixed4 (Concatenate)	(None, 9, 11, 768)	0	activation_406[0][0] activation_409[0][0] activation_414[0][0] activation_415[0][0]
conv2d_420 (Conv2D)	(None, 9, 11, 160)	122880	mixed4[0][0]
batch_normalization_420 (BatchN	(None, 9, 11, 160)	480	conv2d_420[0][0]
activation_420 (Activation)	(None, 9, 11, 160)	0	batch_normalization_420[0][0]
conv2d_421 (Conv2D)	(None, 9, 11, 160)	179200	activation_420[0][0]
batch_normalization_421 (BatchN	(None, 9, 11, 160)	480	conv2d_421[0][0]
activation_421 (Activation)	(None, 9, 11, 160)	0	batch_normalization_421[0][0]
conv2d_417 (Conv2D)	(None, 9, 11, 160)	122880	mixed4[0][0]
conv2d_422 (Conv2D)	(None, 9, 11, 160)	179200	activation_421[0][0]
batch_normalization_417 (BatchN	(None, 9, 11, 160)	480	conv2d_417[0][0]

batch_normalization_422 (BatchN	(None, 9, 11, 160)	480	conv2d_422[0][0]
activation_417 (Activation)	(None, 9, 11, 160)	0	batch_normalization_417[0][0]
activation_422 (Activation)	(None, 9, 11, 160)	0	batch_normalization_422[0][0]
conv2d_418 (Conv2D)	(None, 9, 11, 160)	179200	activation_417[0][0]
conv2d_423 (Conv2D)	(None, 9, 11, 160)	179200	activation_422[0][0]
batch_normalization_418 (BatchN	(None, 9, 11, 160)	480	conv2d_418[0][0]
batch_normalization_423 (BatchN	(None, 9, 11, 160)	480	conv2d_423[0][0]
activation_418 (Activation)	(None, 9, 11, 160)	0	batch_normalization_418[0][0]
activation_423 (Activation)	(None, 9, 11, 160)	0	batch_normalization_423[0][0]
average_pooling2d_40 (AveragePo	(None, 9, 11, 768)	0	mixed4[0][0]
conv2d_416 (Conv2D)	(None, 9, 11, 192)	147456	mixed4[0][0]
conv2d_419 (Conv2D)	(None, 9, 11, 192)	215040	activation_418[0][0]
conv2d_424 (Conv2D)	(None, 9, 11, 192)	215040	activation_423[0][0]
conv2d_425 (Conv2D)	(None, 9, 11, 192)	147456	average_pooling2d_40[0][0]
batch_normalization_416 (BatchN	(None, 9, 11, 192)	576	conv2d_416[0][0]
batch_normalization_419 (BatchN	(None, 9, 11, 192)	576	conv2d_419[0][0]
batch_normalization_424 (BatchN	(None, 9, 11, 192)	576	conv2d_424[0][0]
batch_normalization_425 (BatchN	(None, 9, 11, 192)	576	conv2d_425[0][0]
activation_416 (Activation)	(None, 9, 11, 192)	0	batch_normalization_416[0][0]
activation_419 (Activation)	(None, 9, 11, 192)	0	batch_normalization_419[0][0]
activation_424 (Activation)	(None, 9, 11, 192)	0	batch_normalization_424[0][0]

activation_425 (Activation)	(None, 9, 11, 192)	0	batch_normalization_425[0][0]
mixed5 (Concatenate)	(None, 9, 11, 768)	0	activation_416[0][0]
			activation_419[0][0]
			activation_424[0][0]
			activation_425[0][0]
conv2d_430 (Conv2D)	(None, 9, 11, 160)	122880	mixed5[0][0]
batch_normalization_430 (Batch Normalization)	(None, 9, 11, 160)	480	conv2d_430[0][0]
activation_430 (Activation)	(None, 9, 11, 160)	0	batch_normalization_430[0][0]
conv2d_431 (Conv2D)	(None, 9, 11, 160)	179200	activation_430[0][0]
batch_normalization_431 (Batch Normalization)	(None, 9, 11, 160)	480	conv2d_431[0][0]
activation_431 (Activation)	(None, 9, 11, 160)	0	batch_normalization_431[0][0]
conv2d_427 (Conv2D)	(None, 9, 11, 160)	122880	mixed5[0][0]
conv2d_432 (Conv2D)	(None, 9, 11, 160)	179200	activation_431[0][0]
batch_normalization_427 (Batch Normalization)	(None, 9, 11, 160)	480	conv2d_427[0][0]
batch_normalization_432 (Batch Normalization)	(None, 9, 11, 160)	480	conv2d_432[0][0]
activation_427 (Activation)	(None, 9, 11, 160)	0	batch_normalization_427[0][0]
activation_432 (Activation)	(None, 9, 11, 160)	0	batch_normalization_432[0][0]
conv2d_428 (Conv2D)	(None, 9, 11, 160)	179200	activation_427[0][0]
conv2d_433 (Conv2D)	(None, 9, 11, 160)	179200	activation_432[0][0]
batch_normalization_428 (Batch Normalization)	(None, 9, 11, 160)	480	conv2d_428[0][0]
batch_normalization_433 (Batch Normalization)	(None, 9, 11, 160)	480	conv2d_433[0][0]
activation_428 (Activation)	(None, 9, 11, 160)	0	batch_normalization_428[0][0]
activation_433 (Activation)	(None, 9, 11, 160)	0	batch_normalization_433[0][0]

average_pooling2d_41 (AveragePo	(None, 9, 11, 768)	0	mixed5[0][0]
conv2d_426 (Conv2D)	(None, 9, 11, 192)	147456	mixed5[0][0]
conv2d_429 (Conv2D)	(None, 9, 11, 192)	215040	activation_428[0][0]
conv2d_434 (Conv2D)	(None, 9, 11, 192)	215040	activation_433[0][0]
conv2d_435 (Conv2D)	(None, 9, 11, 192)	147456	average_pooling2d_41[0][0]
batch_normalization_426 (BatchN	(None, 9, 11, 192)	576	conv2d_426[0][0]
batch_normalization_429 (BatchN	(None, 9, 11, 192)	576	conv2d_429[0][0]
batch_normalization_434 (BatchN	(None, 9, 11, 192)	576	conv2d_434[0][0]
batch_normalization_435 (BatchN	(None, 9, 11, 192)	576	conv2d_435[0][0]
activation_426 (Activation)	(None, 9, 11, 192)	0	batch_normalization_426[0][0]
activation_429 (Activation)	(None, 9, 11, 192)	0	batch_normalization_429[0][0]
activation_434 (Activation)	(None, 9, 11, 192)	0	batch_normalization_434[0][0]
activation_435 (Activation)	(None, 9, 11, 192)	0	batch_normalization_435[0][0]
mixed6 (Concatenate)	(None, 9, 11, 768)	0	activation_426[0][0] activation_429[0][0] activation_434[0][0] activation_435[0][0]
conv2d_440 (Conv2D)	(None, 9, 11, 192)	147456	mixed6[0][0]
batch_normalization_440 (BatchN	(None, 9, 11, 192)	576	conv2d_440[0][0]
activation_440 (Activation)	(None, 9, 11, 192)	0	batch_normalization_440[0][0]
conv2d_441 (Conv2D)	(None, 9, 11, 192)	258048	activation_440[0][0]
batch_normalization_441 (BatchN	(None, 9, 11, 192)	576	conv2d_441[0][0]
activation_441 (Activation)	(None, 9, 11, 192)	0	batch_normalization_441[0][0]

conv2d_437 (Conv2D)	(None, 9, 11, 192)	147456	mixed6[0][0]
conv2d_442 (Conv2D)	(None, 9, 11, 192)	258048	activation_441[0][0]
batch_normalization_437 (BatchN	(None, 9, 11, 192)	576	conv2d_437[0][0]
batch_normalization_442 (BatchN	(None, 9, 11, 192)	576	conv2d_442[0][0]
activation_437 (Activation)	(None, 9, 11, 192)	0	batch_normalization_437[0][0]
activation_442 (Activation)	(None, 9, 11, 192)	0	batch_normalization_442[0][0]
conv2d_438 (Conv2D)	(None, 9, 11, 192)	258048	activation_437[0][0]
conv2d_443 (Conv2D)	(None, 9, 11, 192)	258048	activation_442[0][0]
batch_normalization_438 (BatchN	(None, 9, 11, 192)	576	conv2d_438[0][0]
batch_normalization_443 (BatchN	(None, 9, 11, 192)	576	conv2d_443[0][0]
activation_438 (Activation)	(None, 9, 11, 192)	0	batch_normalization_438[0][0]
activation_443 (Activation)	(None, 9, 11, 192)	0	batch_normalization_443[0][0]
average_pooling2d_42 (AveragePo	(None, 9, 11, 768)	0	mixed6[0][0]
conv2d_436 (Conv2D)	(None, 9, 11, 192)	147456	mixed6[0][0]
conv2d_439 (Conv2D)	(None, 9, 11, 192)	258048	activation_438[0][0]
conv2d_444 (Conv2D)	(None, 9, 11, 192)	258048	activation_443[0][0]
conv2d_445 (Conv2D)	(None, 9, 11, 192)	147456	average_pooling2d_42[0][0]
batch_normalization_436 (BatchN	(None, 9, 11, 192)	576	conv2d_436[0][0]
batch_normalization_439 (BatchN	(None, 9, 11, 192)	576	conv2d_439[0][0]
batch_normalization_444 (BatchN	(None, 9, 11, 192)	576	conv2d_444[0][0]
batch_normalization_445 (BatchN	(None, 9, 11, 192)	576	conv2d_445[0][0]
activation_436 (Activation)	(None, 9, 11, 192)	0	batch_normalization_436[0][0]

activation_439 (Activation)	(None, 9, 11, 192)	0	batch_normalization_439[0][0]
activation_444 (Activation)	(None, 9, 11, 192)	0	batch_normalization_444[0][0]
activation_445 (Activation)	(None, 9, 11, 192)	0	batch_normalization_445[0][0]
mixed7 (Concatenate)	(None, 9, 11, 768)	0	activation_436[0][0] activation_439[0][0] activation_444[0][0] activation_445[0][0]
conv2d_448 (Conv2D)	(None, 9, 11, 192)	147456	mixed7[0][0]
batch_normalization_448 (BatchN	(None, 9, 11, 192)	576	conv2d_448[0][0]
activation_448 (Activation)	(None, 9, 11, 192)	0	batch_normalization_448[0][0]
conv2d_449 (Conv2D)	(None, 9, 11, 192)	258048	activation_448[0][0]
batch_normalization_449 (BatchN	(None, 9, 11, 192)	576	conv2d_449[0][0]
activation_449 (Activation)	(None, 9, 11, 192)	0	batch_normalization_449[0][0]
conv2d_446 (Conv2D)	(None, 9, 11, 192)	147456	mixed7[0][0]
conv2d_450 (Conv2D)	(None, 9, 11, 192)	258048	activation_449[0][0]
batch_normalization_446 (BatchN	(None, 9, 11, 192)	576	conv2d_446[0][0]
batch_normalization_450 (BatchN	(None, 9, 11, 192)	576	conv2d_450[0][0]
activation_446 (Activation)	(None, 9, 11, 192)	0	batch_normalization_446[0][0]
activation_450 (Activation)	(None, 9, 11, 192)	0	batch_normalization_450[0][0]
conv2d_447 (Conv2D)	(None, 4, 5, 320)	552960	activation_446[0][0]
conv2d_451 (Conv2D)	(None, 4, 5, 192)	331776	activation_450[0][0]
batch_normalization_447 (BatchN	(None, 4, 5, 320)	960	conv2d_447[0][0]
batch_normalization_451 (BatchN	(None, 4, 5, 192)	576	conv2d_451[0][0]

activation_447 (Activation)	(None, 4, 5, 320)	0	batch_normalization_447[0][0]
activation_451 (Activation)	(None, 4, 5, 192)	0	batch_normalization_451[0][0]
max_pooling2d_19 (MaxPooling2D)	(None, 4, 5, 768)	0	mixed7[0][0]
mixed8 (Concatenate)	(None, 4, 5, 1280)	0	activation_447[0][0] activation_451[0][0] max_pooling2d_19[0][0]
conv2d_456 (Conv2D)	(None, 4, 5, 448)	573440	mixed8[0][0]
batch_normalization_456 (Batch Normalization)	(None, 4, 5, 448)	1344	conv2d_456[0][0]
activation_456 (Activation)	(None, 4, 5, 448)	0	batch_normalization_456[0][0]
conv2d_453 (Conv2D)	(None, 4, 5, 384)	491520	mixed8[0][0]
conv2d_457 (Conv2D)	(None, 4, 5, 384)	1548288	activation_456[0][0]
batch_normalization_453 (Batch Normalization)	(None, 4, 5, 384)	1152	conv2d_453[0][0]
batch_normalization_457 (Batch Normalization)	(None, 4, 5, 384)	1152	conv2d_457[0][0]
activation_453 (Activation)	(None, 4, 5, 384)	0	batch_normalization_453[0][0]
activation_457 (Activation)	(None, 4, 5, 384)	0	batch_normalization_457[0][0]
conv2d_454 (Conv2D)	(None, 4, 5, 384)	442368	activation_453[0][0]
conv2d_455 (Conv2D)	(None, 4, 5, 384)	442368	activation_453[0][0]
conv2d_458 (Conv2D)	(None, 4, 5, 384)	442368	activation_457[0][0]
conv2d_459 (Conv2D)	(None, 4, 5, 384)	442368	activation_457[0][0]
average_pooling2d_43 (AveragePooling2D)	(None, 4, 5, 1280)	0	mixed8[0][0]
conv2d_452 (Conv2D)	(None, 4, 5, 320)	409600	mixed8[0][0]
batch_normalization_454 (Batch Normalization)	(None, 4, 5, 384)	1152	conv2d_454[0][0]

batch_normalization_455 (BatchN	(None, 4, 5, 384)	1152	conv2d_455[0][0]
batch_normalization_458 (BatchN	(None, 4, 5, 384)	1152	conv2d_458[0][0]
batch_normalization_459 (BatchN	(None, 4, 5, 384)	1152	conv2d_459[0][0]
conv2d_460 (Conv2D)	(None, 4, 5, 192)	245760	average_pooling2d_43[0][0]
batch_normalization_452 (BatchN	(None, 4, 5, 320)	960	conv2d_452[0][0]
activation_454 (Activation)	(None, 4, 5, 384)	0	batch_normalization_454[0][0]
activation_455 (Activation)	(None, 4, 5, 384)	0	batch_normalization_455[0][0]
activation_458 (Activation)	(None, 4, 5, 384)	0	batch_normalization_458[0][0]
activation_459 (Activation)	(None, 4, 5, 384)	0	batch_normalization_459[0][0]
batch_normalization_460 (BatchN	(None, 4, 5, 192)	576	conv2d_460[0][0]
activation_452 (Activation)	(None, 4, 5, 320)	0	batch_normalization_452[0][0]
mixed9_0 (Concatenate)	(None, 4, 5, 768)	0	activation_454[0][0] activation_455[0][0]
concatenate_8 (Concatenate)	(None, 4, 5, 768)	0	activation_458[0][0] activation_459[0][0]
activation_460 (Activation)	(None, 4, 5, 192)	0	batch_normalization_460[0][0]
mixed9 (Concatenate)	(None, 4, 5, 2048)	0	activation_452[0][0] mixed9_0[0][0] concatenate_8[0][0] activation_460[0][0]
conv2d_465 (Conv2D)	(None, 4, 5, 448)	917504	mixed9[0][0]
batch_normalization_465 (BatchN	(None, 4, 5, 448)	1344	conv2d_465[0][0]
activation_465 (Activation)	(None, 4, 5, 448)	0	batch_normalization_465[0][0]
conv2d_462 (Conv2D)	(None, 4, 5, 384)	786432	mixed9[0][0]

conv2d_466 (Conv2D)	(None, 4, 5, 384)	1548288	activation_465[0][0]
batch_normalization_462 (BatchN	(None, 4, 5, 384)	1152	conv2d_462[0][0]
batch_normalization_466 (BatchN	(None, 4, 5, 384)	1152	conv2d_466[0][0]
activation_462 (Activation)	(None, 4, 5, 384)	0	batch_normalization_462[0][0]
activation_466 (Activation)	(None, 4, 5, 384)	0	batch_normalization_466[0][0]
conv2d_463 (Conv2D)	(None, 4, 5, 384)	442368	activation_462[0][0]
conv2d_464 (Conv2D)	(None, 4, 5, 384)	442368	activation_462[0][0]
conv2d_467 (Conv2D)	(None, 4, 5, 384)	442368	activation_466[0][0]
conv2d_468 (Conv2D)	(None, 4, 5, 384)	442368	activation_466[0][0]
average_pooling2d_44 (AveragePo	(None, 4, 5, 2048)	0	mixed9[0][0]
conv2d_461 (Conv2D)	(None, 4, 5, 320)	655360	mixed9[0][0]
batch_normalization_463 (BatchN	(None, 4, 5, 384)	1152	conv2d_463[0][0]
batch_normalization_464 (BatchN	(None, 4, 5, 384)	1152	conv2d_464[0][0]
batch_normalization_467 (BatchN	(None, 4, 5, 384)	1152	conv2d_467[0][0]
batch_normalization_468 (BatchN	(None, 4, 5, 384)	1152	conv2d_468[0][0]
conv2d_469 (Conv2D)	(None, 4, 5, 192)	393216	average_pooling2d_44[0][0]
batch_normalization_461 (BatchN	(None, 4, 5, 320)	960	conv2d_461[0][0]
activation_463 (Activation)	(None, 4, 5, 384)	0	batch_normalization_463[0][0]
activation_464 (Activation)	(None, 4, 5, 384)	0	batch_normalization_464[0][0]
activation_467 (Activation)	(None, 4, 5, 384)	0	batch_normalization_467[0][0]
activation_468 (Activation)	(None, 4, 5, 384)	0	batch_normalization_468[0][0]
batch_normalization_469 (BatchN	(None, 4, 5, 192)	576	conv2d_469[0][0]

activation_461 (Activation)	(None, 4, 5, 320)	0	batch_normalization_461[0][0]
mixed9_1 (Concatenate)	(None, 4, 5, 768)	0	activation_463[0][0] activation_464[0][0]
concatenate_9 (Concatenate)	(None, 4, 5, 768)	0	activation_467[0][0] activation_468[0][0]
activation_469 (Activation)	(None, 4, 5, 192)	0	batch_normalization_469[0][0]
mixed10 (Concatenate)	(None, 4, 5, 2048)	0	activation_461[0][0] mixed9_1[0][0] concatenate_9[0][0] activation_469[0][0]

=====

Total params: 21,802,784
Trainable params: 21,768,352
Non-trainable params: 34,432

```
In [27]: inception_conv_base.trainable = True
set_trainable = False
for layer in inception_conv_base.layers:
    if layer.name == 'conv2d_461':
        print('Found it!')
        set_trainable = True
    if set_trainable:
        layer.trainable = True
    else:
        layer.trainable = False
```

Found it!

```
In [28]: inception_md1 = Sequential()
inception_md1.add(inception_conv_base)
inception_md1.add(Flatten())
inception_md1.add(Dense(256, activation='relu'))
inception_md1.add(Dropout(0.5))
inception_md1.add(Dense(1, activation='sigmoid'))
```

```
In [26]: inception_conv_base.summary()
```


Model: "inception_v3"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_5 (InputLayer)	[(None, 178, 218, 3)]	0	
conv2d_376 (Conv2D)	(None, 88, 108, 32)	864	input_5[0][0]
batch_normalization_376 (Batch Normalization)	(None, 88, 108, 32)	96	conv2d_376[0][0]
activation_376 (Activation)	(None, 88, 108, 32)	0	batch_normalization_376[0][0]
conv2d_377 (Conv2D)	(None, 86, 106, 32)	9216	activation_376[0][0]
batch_normalization_377 (Batch Normalization)	(None, 86, 106, 32)	96	conv2d_377[0][0]
activation_377 (Activation)	(None, 86, 106, 32)	0	batch_normalization_377[0][0]
conv2d_378 (Conv2D)	(None, 86, 106, 64)	18432	activation_377[0][0]
batch_normalization_378 (Batch Normalization)	(None, 86, 106, 64)	192	conv2d_378[0][0]
activation_378 (Activation)	(None, 86, 106, 64)	0	batch_normalization_378[0][0]
max_pooling2d_16 (MaxPooling2D)	(None, 42, 52, 64)	0	activation_378[0][0]
conv2d_379 (Conv2D)	(None, 42, 52, 80)	5120	max_pooling2d_16[0][0]
batch_normalization_379 (Batch Normalization)	(None, 42, 52, 80)	240	conv2d_379[0][0]
activation_379 (Activation)	(None, 42, 52, 80)	0	batch_normalization_379[0][0]
conv2d_380 (Conv2D)	(None, 40, 50, 192)	138240	activation_379[0][0]
batch_normalization_380 (Batch Normalization)	(None, 40, 50, 192)	576	conv2d_380[0][0]
activation_380 (Activation)	(None, 40, 50, 192)	0	batch_normalization_380[0][0]
max_pooling2d_17 (MaxPooling2D)	(None, 19, 24, 192)	0	activation_380[0][0]
conv2d_384 (Conv2D)	(None, 19, 24, 64)	12288	max_pooling2d_17[0][0]
batch_normalization_384 (Batch Normalization)	(None, 19, 24, 64)	192	conv2d_384[0][0]

activation_384 (Activation)	(None, 19, 24, 64)	0	batch_normalization_384[0][0]
conv2d_382 (Conv2D)	(None, 19, 24, 48)	9216	max_pooling2d_17[0][0]
conv2d_385 (Conv2D)	(None, 19, 24, 96)	55296	activation_384[0][0]
batch_normalization_382 (BatchN	(None, 19, 24, 48)	144	conv2d_382[0][0]
batch_normalization_385 (BatchN	(None, 19, 24, 96)	288	conv2d_385[0][0]
activation_382 (Activation)	(None, 19, 24, 48)	0	batch_normalization_382[0][0]
activation_385 (Activation)	(None, 19, 24, 96)	0	batch_normalization_385[0][0]
average_pooling2d_36 (AveragePo	(None, 19, 24, 192)	0	max_pooling2d_17[0][0]
conv2d_381 (Conv2D)	(None, 19, 24, 64)	12288	max_pooling2d_17[0][0]
conv2d_383 (Conv2D)	(None, 19, 24, 64)	76800	activation_382[0][0]
conv2d_386 (Conv2D)	(None, 19, 24, 96)	82944	activation_385[0][0]
conv2d_387 (Conv2D)	(None, 19, 24, 32)	6144	average_pooling2d_36[0][0]
batch_normalization_381 (BatchN	(None, 19, 24, 64)	192	conv2d_381[0][0]
batch_normalization_383 (BatchN	(None, 19, 24, 64)	192	conv2d_383[0][0]
batch_normalization_386 (BatchN	(None, 19, 24, 96)	288	conv2d_386[0][0]
batch_normalization_387 (BatchN	(None, 19, 24, 32)	96	conv2d_387[0][0]
activation_381 (Activation)	(None, 19, 24, 64)	0	batch_normalization_381[0][0]
activation_383 (Activation)	(None, 19, 24, 64)	0	batch_normalization_383[0][0]
activation_386 (Activation)	(None, 19, 24, 96)	0	batch_normalization_386[0][0]
activation_387 (Activation)	(None, 19, 24, 32)	0	batch_normalization_387[0][0]
mixed0 (Concatenate)	(None, 19, 24, 256)	0	activation_381[0][0] activation_383[0][0]

			activation_386[0][0]
			activation_387[0][0]
conv2d_391 (Conv2D)	(None, 19, 24, 64)	16384	mixed0[0][0]
batch_normalization_391 (BatchN	(None, 19, 24, 64)	192	conv2d_391[0][0]
activation_391 (Activation)	(None, 19, 24, 64)	0	batch_normalization_391[0][0]
conv2d_389 (Conv2D)	(None, 19, 24, 48)	12288	mixed0[0][0]
conv2d_392 (Conv2D)	(None, 19, 24, 96)	55296	activation_391[0][0]
batch_normalization_389 (BatchN	(None, 19, 24, 48)	144	conv2d_389[0][0]
batch_normalization_392 (BatchN	(None, 19, 24, 96)	288	conv2d_392[0][0]
activation_389 (Activation)	(None, 19, 24, 48)	0	batch_normalization_389[0][0]
activation_392 (Activation)	(None, 19, 24, 96)	0	batch_normalization_392[0][0]
average_pooling2d_37 (AveragePo	(None, 19, 24, 256)	0	mixed0[0][0]
conv2d_388 (Conv2D)	(None, 19, 24, 64)	16384	mixed0[0][0]
conv2d_390 (Conv2D)	(None, 19, 24, 64)	76800	activation_389[0][0]
conv2d_393 (Conv2D)	(None, 19, 24, 96)	82944	activation_392[0][0]
conv2d_394 (Conv2D)	(None, 19, 24, 64)	16384	average_pooling2d_37[0][0]
batch_normalization_388 (BatchN	(None, 19, 24, 64)	192	conv2d_388[0][0]
batch_normalization_390 (BatchN	(None, 19, 24, 64)	192	conv2d_390[0][0]
batch_normalization_393 (BatchN	(None, 19, 24, 96)	288	conv2d_393[0][0]
batch_normalization_394 (BatchN	(None, 19, 24, 64)	192	conv2d_394[0][0]
activation_388 (Activation)	(None, 19, 24, 64)	0	batch_normalization_388[0][0]
activation_390 (Activation)	(None, 19, 24, 64)	0	batch_normalization_390[0][0]

activation_393 (Activation)	(None, 19, 24, 96)	0	batch_normalization_393[0][0]
activation_394 (Activation)	(None, 19, 24, 64)	0	batch_normalization_394[0][0]
mixed1 (Concatenate)	(None, 19, 24, 288)	0	activation_388[0][0]
			activation_390[0][0]
			activation_393[0][0]
			activation_394[0][0]
conv2d_398 (Conv2D)	(None, 19, 24, 64)	18432	mixed1[0][0]
batch_normalization_398 (BatchN	(None, 19, 24, 64)	192	conv2d_398[0][0]
activation_398 (Activation)	(None, 19, 24, 64)	0	batch_normalization_398[0][0]
conv2d_396 (Conv2D)	(None, 19, 24, 48)	13824	mixed1[0][0]
conv2d_399 (Conv2D)	(None, 19, 24, 96)	55296	activation_398[0][0]
batch_normalization_396 (BatchN	(None, 19, 24, 48)	144	conv2d_396[0][0]
batch_normalization_399 (BatchN	(None, 19, 24, 96)	288	conv2d_399[0][0]
activation_396 (Activation)	(None, 19, 24, 48)	0	batch_normalization_396[0][0]
activation_399 (Activation)	(None, 19, 24, 96)	0	batch_normalization_399[0][0]
average_pooling2d_38 (AveragePo	(None, 19, 24, 288)	0	mixed1[0][0]
conv2d_395 (Conv2D)	(None, 19, 24, 64)	18432	mixed1[0][0]
conv2d_397 (Conv2D)	(None, 19, 24, 64)	76800	activation_396[0][0]
conv2d_400 (Conv2D)	(None, 19, 24, 96)	82944	activation_399[0][0]
conv2d_401 (Conv2D)	(None, 19, 24, 64)	18432	average_pooling2d_38[0][0]
batch_normalization_395 (BatchN	(None, 19, 24, 64)	192	conv2d_395[0][0]
batch_normalization_397 (BatchN	(None, 19, 24, 64)	192	conv2d_397[0][0]
batch_normalization_400 (BatchN	(None, 19, 24, 96)	288	conv2d_400[0][0]

batch_normalization_401 (BatchN	(None, 19, 24, 64)	192	conv2d_401[0][0]
activation_395 (Activation)	(None, 19, 24, 64)	0	batch_normalization_395[0][0]
activation_397 (Activation)	(None, 19, 24, 64)	0	batch_normalization_397[0][0]
activation_400 (Activation)	(None, 19, 24, 96)	0	batch_normalization_400[0][0]
activation_401 (Activation)	(None, 19, 24, 64)	0	batch_normalization_401[0][0]
mixed2 (Concatenate)	(None, 19, 24, 288)	0	activation_395[0][0] activation_397[0][0] activation_400[0][0] activation_401[0][0]
conv2d_403 (Conv2D)	(None, 19, 24, 64)	18432	mixed2[0][0]
batch_normalization_403 (BatchN	(None, 19, 24, 64)	192	conv2d_403[0][0]
activation_403 (Activation)	(None, 19, 24, 64)	0	batch_normalization_403[0][0]
conv2d_404 (Conv2D)	(None, 19, 24, 96)	55296	activation_403[0][0]
batch_normalization_404 (BatchN	(None, 19, 24, 96)	288	conv2d_404[0][0]
activation_404 (Activation)	(None, 19, 24, 96)	0	batch_normalization_404[0][0]
conv2d_402 (Conv2D)	(None, 9, 11, 384)	995328	mixed2[0][0]
conv2d_405 (Conv2D)	(None, 9, 11, 96)	82944	activation_404[0][0]
batch_normalization_402 (BatchN	(None, 9, 11, 384)	1152	conv2d_402[0][0]
batch_normalization_405 (BatchN	(None, 9, 11, 96)	288	conv2d_405[0][0]
activation_402 (Activation)	(None, 9, 11, 384)	0	batch_normalization_402[0][0]
activation_405 (Activation)	(None, 9, 11, 96)	0	batch_normalization_405[0][0]
max_pooling2d_18 (MaxPooling2D)	(None, 9, 11, 288)	0	mixed2[0][0]
mixed3 (Concatenate)	(None, 9, 11, 768)	0	activation_402[0][0] activation_405[0][0]

			max_pooling2d_18[0][0]
conv2d_410 (Conv2D)	(None, 9, 11, 128)	98304	mixed3[0][0]
batch_normalization_410 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_410[0][0]
activation_410 (Activation)	(None, 9, 11, 128)	0	batch_normalization_410[0][0]
conv2d_411 (Conv2D)	(None, 9, 11, 128)	114688	activation_410[0][0]
batch_normalization_411 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_411[0][0]
activation_411 (Activation)	(None, 9, 11, 128)	0	batch_normalization_411[0][0]
conv2d_407 (Conv2D)	(None, 9, 11, 128)	98304	mixed3[0][0]
conv2d_412 (Conv2D)	(None, 9, 11, 128)	114688	activation_411[0][0]
batch_normalization_407 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_407[0][0]
batch_normalization_412 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_412[0][0]
activation_407 (Activation)	(None, 9, 11, 128)	0	batch_normalization_407[0][0]
activation_412 (Activation)	(None, 9, 11, 128)	0	batch_normalization_412[0][0]
conv2d_408 (Conv2D)	(None, 9, 11, 128)	114688	activation_407[0][0]
conv2d_413 (Conv2D)	(None, 9, 11, 128)	114688	activation_412[0][0]
batch_normalization_408 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_408[0][0]
batch_normalization_413 (Batch Normalization)	(None, 9, 11, 128)	384	conv2d_413[0][0]
activation_408 (Activation)	(None, 9, 11, 128)	0	batch_normalization_408[0][0]
activation_413 (Activation)	(None, 9, 11, 128)	0	batch_normalization_413[0][0]
average_pooling2d_39 (Average Pooling)	(None, 9, 11, 768)	0	mixed3[0][0]
conv2d_406 (Conv2D)	(None, 9, 11, 192)	147456	mixed3[0][0]
conv2d_409 (Conv2D)	(None, 9, 11, 192)	172032	activation_408[0][0]

conv2d_414 (Conv2D)	(None, 9, 11, 192)	172032	activation_413[0][0]
conv2d_415 (Conv2D)	(None, 9, 11, 192)	147456	average_pooling2d_39[0][0]
batch_normalization_406 (BatchN	(None, 9, 11, 192)	576	conv2d_406[0][0]
batch_normalization_409 (BatchN	(None, 9, 11, 192)	576	conv2d_409[0][0]
batch_normalization_414 (BatchN	(None, 9, 11, 192)	576	conv2d_414[0][0]
batch_normalization_415 (BatchN	(None, 9, 11, 192)	576	conv2d_415[0][0]
activation_406 (Activation)	(None, 9, 11, 192)	0	batch_normalization_406[0][0]
activation_409 (Activation)	(None, 9, 11, 192)	0	batch_normalization_409[0][0]
activation_414 (Activation)	(None, 9, 11, 192)	0	batch_normalization_414[0][0]
activation_415 (Activation)	(None, 9, 11, 192)	0	batch_normalization_415[0][0]
mixed4 (Concatenate)	(None, 9, 11, 768)	0	activation_406[0][0] activation_409[0][0] activation_414[0][0] activation_415[0][0]
conv2d_420 (Conv2D)	(None, 9, 11, 160)	122880	mixed4[0][0]
batch_normalization_420 (BatchN	(None, 9, 11, 160)	480	conv2d_420[0][0]
activation_420 (Activation)	(None, 9, 11, 160)	0	batch_normalization_420[0][0]
conv2d_421 (Conv2D)	(None, 9, 11, 160)	179200	activation_420[0][0]
batch_normalization_421 (BatchN	(None, 9, 11, 160)	480	conv2d_421[0][0]
activation_421 (Activation)	(None, 9, 11, 160)	0	batch_normalization_421[0][0]
conv2d_417 (Conv2D)	(None, 9, 11, 160)	122880	mixed4[0][0]
conv2d_422 (Conv2D)	(None, 9, 11, 160)	179200	activation_421[0][0]
batch_normalization_417 (BatchN	(None, 9, 11, 160)	480	conv2d_417[0][0]

batch_normalization_422 (BatchN	(None, 9, 11, 160)	480	conv2d_422[0][0]
activation_417 (Activation)	(None, 9, 11, 160)	0	batch_normalization_417[0][0]
activation_422 (Activation)	(None, 9, 11, 160)	0	batch_normalization_422[0][0]
conv2d_418 (Conv2D)	(None, 9, 11, 160)	179200	activation_417[0][0]
conv2d_423 (Conv2D)	(None, 9, 11, 160)	179200	activation_422[0][0]
batch_normalization_418 (BatchN	(None, 9, 11, 160)	480	conv2d_418[0][0]
batch_normalization_423 (BatchN	(None, 9, 11, 160)	480	conv2d_423[0][0]
activation_418 (Activation)	(None, 9, 11, 160)	0	batch_normalization_418[0][0]
activation_423 (Activation)	(None, 9, 11, 160)	0	batch_normalization_423[0][0]
average_pooling2d_40 (AveragePo	(None, 9, 11, 768)	0	mixed4[0][0]
conv2d_416 (Conv2D)	(None, 9, 11, 192)	147456	mixed4[0][0]
conv2d_419 (Conv2D)	(None, 9, 11, 192)	215040	activation_418[0][0]
conv2d_424 (Conv2D)	(None, 9, 11, 192)	215040	activation_423[0][0]
conv2d_425 (Conv2D)	(None, 9, 11, 192)	147456	average_pooling2d_40[0][0]
batch_normalization_416 (BatchN	(None, 9, 11, 192)	576	conv2d_416[0][0]
batch_normalization_419 (BatchN	(None, 9, 11, 192)	576	conv2d_419[0][0]
batch_normalization_424 (BatchN	(None, 9, 11, 192)	576	conv2d_424[0][0]
batch_normalization_425 (BatchN	(None, 9, 11, 192)	576	conv2d_425[0][0]
activation_416 (Activation)	(None, 9, 11, 192)	0	batch_normalization_416[0][0]
activation_419 (Activation)	(None, 9, 11, 192)	0	batch_normalization_419[0][0]
activation_424 (Activation)	(None, 9, 11, 192)	0	batch_normalization_424[0][0]

activation_425 (Activation)	(None, 9, 11, 192)	0	batch_normalization_425[0][0]
mixed5 (Concatenate)	(None, 9, 11, 768)	0	activation_416[0][0]
			activation_419[0][0]
			activation_424[0][0]
			activation_425[0][0]
conv2d_430 (Conv2D)	(None, 9, 11, 160)	122880	mixed5[0][0]
batch_normalization_430 (BatchN	(None, 9, 11, 160)	480	conv2d_430[0][0]
activation_430 (Activation)	(None, 9, 11, 160)	0	batch_normalization_430[0][0]
conv2d_431 (Conv2D)	(None, 9, 11, 160)	179200	activation_430[0][0]
batch_normalization_431 (BatchN	(None, 9, 11, 160)	480	conv2d_431[0][0]
activation_431 (Activation)	(None, 9, 11, 160)	0	batch_normalization_431[0][0]
conv2d_427 (Conv2D)	(None, 9, 11, 160)	122880	mixed5[0][0]
conv2d_432 (Conv2D)	(None, 9, 11, 160)	179200	activation_431[0][0]
batch_normalization_427 (BatchN	(None, 9, 11, 160)	480	conv2d_427[0][0]
batch_normalization_432 (BatchN	(None, 9, 11, 160)	480	conv2d_432[0][0]
activation_427 (Activation)	(None, 9, 11, 160)	0	batch_normalization_427[0][0]
activation_432 (Activation)	(None, 9, 11, 160)	0	batch_normalization_432[0][0]
conv2d_428 (Conv2D)	(None, 9, 11, 160)	179200	activation_427[0][0]
conv2d_433 (Conv2D)	(None, 9, 11, 160)	179200	activation_432[0][0]
batch_normalization_428 (BatchN	(None, 9, 11, 160)	480	conv2d_428[0][0]
batch_normalization_433 (BatchN	(None, 9, 11, 160)	480	conv2d_433[0][0]
activation_428 (Activation)	(None, 9, 11, 160)	0	batch_normalization_428[0][0]
activation_433 (Activation)	(None, 9, 11, 160)	0	batch_normalization_433[0][0]

average_pooling2d_41 (AveragePo	(None, 9, 11, 768)	0	mixed5[0][0]
conv2d_426 (Conv2D)	(None, 9, 11, 192)	147456	mixed5[0][0]
conv2d_429 (Conv2D)	(None, 9, 11, 192)	215040	activation_428[0][0]
conv2d_434 (Conv2D)	(None, 9, 11, 192)	215040	activation_433[0][0]
conv2d_435 (Conv2D)	(None, 9, 11, 192)	147456	average_pooling2d_41[0][0]
batch_normalization_426 (BatchN	(None, 9, 11, 192)	576	conv2d_426[0][0]
batch_normalization_429 (BatchN	(None, 9, 11, 192)	576	conv2d_429[0][0]
batch_normalization_434 (BatchN	(None, 9, 11, 192)	576	conv2d_434[0][0]
batch_normalization_435 (BatchN	(None, 9, 11, 192)	576	conv2d_435[0][0]
activation_426 (Activation)	(None, 9, 11, 192)	0	batch_normalization_426[0][0]
activation_429 (Activation)	(None, 9, 11, 192)	0	batch_normalization_429[0][0]
activation_434 (Activation)	(None, 9, 11, 192)	0	batch_normalization_434[0][0]
activation_435 (Activation)	(None, 9, 11, 192)	0	batch_normalization_435[0][0]
mixed6 (Concatenate)	(None, 9, 11, 768)	0	activation_426[0][0] activation_429[0][0] activation_434[0][0] activation_435[0][0]
conv2d_440 (Conv2D)	(None, 9, 11, 192)	147456	mixed6[0][0]
batch_normalization_440 (BatchN	(None, 9, 11, 192)	576	conv2d_440[0][0]
activation_440 (Activation)	(None, 9, 11, 192)	0	batch_normalization_440[0][0]
conv2d_441 (Conv2D)	(None, 9, 11, 192)	258048	activation_440[0][0]
batch_normalization_441 (BatchN	(None, 9, 11, 192)	576	conv2d_441[0][0]
activation_441 (Activation)	(None, 9, 11, 192)	0	batch_normalization_441[0][0]

conv2d_437 (Conv2D)	(None, 9, 11, 192)	147456	mixed6[0][0]
conv2d_442 (Conv2D)	(None, 9, 11, 192)	258048	activation_441[0][0]
batch_normalization_437 (BatchN	(None, 9, 11, 192)	576	conv2d_437[0][0]
batch_normalization_442 (BatchN	(None, 9, 11, 192)	576	conv2d_442[0][0]
activation_437 (Activation)	(None, 9, 11, 192)	0	batch_normalization_437[0][0]
activation_442 (Activation)	(None, 9, 11, 192)	0	batch_normalization_442[0][0]
conv2d_438 (Conv2D)	(None, 9, 11, 192)	258048	activation_437[0][0]
conv2d_443 (Conv2D)	(None, 9, 11, 192)	258048	activation_442[0][0]
batch_normalization_438 (BatchN	(None, 9, 11, 192)	576	conv2d_438[0][0]
batch_normalization_443 (BatchN	(None, 9, 11, 192)	576	conv2d_443[0][0]
activation_438 (Activation)	(None, 9, 11, 192)	0	batch_normalization_438[0][0]
activation_443 (Activation)	(None, 9, 11, 192)	0	batch_normalization_443[0][0]
average_pooling2d_42 (AveragePo	(None, 9, 11, 768)	0	mixed6[0][0]
conv2d_436 (Conv2D)	(None, 9, 11, 192)	147456	mixed6[0][0]
conv2d_439 (Conv2D)	(None, 9, 11, 192)	258048	activation_438[0][0]
conv2d_444 (Conv2D)	(None, 9, 11, 192)	258048	activation_443[0][0]
conv2d_445 (Conv2D)	(None, 9, 11, 192)	147456	average_pooling2d_42[0][0]
batch_normalization_436 (BatchN	(None, 9, 11, 192)	576	conv2d_436[0][0]
batch_normalization_439 (BatchN	(None, 9, 11, 192)	576	conv2d_439[0][0]
batch_normalization_444 (BatchN	(None, 9, 11, 192)	576	conv2d_444[0][0]
batch_normalization_445 (BatchN	(None, 9, 11, 192)	576	conv2d_445[0][0]
activation_436 (Activation)	(None, 9, 11, 192)	0	batch_normalization_436[0][0]

activation_439 (Activation)	(None, 9, 11, 192)	0	batch_normalization_439[0][0]
activation_444 (Activation)	(None, 9, 11, 192)	0	batch_normalization_444[0][0]
activation_445 (Activation)	(None, 9, 11, 192)	0	batch_normalization_445[0][0]
mixed7 (Concatenate)	(None, 9, 11, 768)	0	activation_436[0][0] activation_439[0][0] activation_444[0][0] activation_445[0][0]
conv2d_448 (Conv2D)	(None, 9, 11, 192)	147456	mixed7[0][0]
batch_normalization_448 (BatchN	(None, 9, 11, 192)	576	conv2d_448[0][0]
activation_448 (Activation)	(None, 9, 11, 192)	0	batch_normalization_448[0][0]
conv2d_449 (Conv2D)	(None, 9, 11, 192)	258048	activation_448[0][0]
batch_normalization_449 (BatchN	(None, 9, 11, 192)	576	conv2d_449[0][0]
activation_449 (Activation)	(None, 9, 11, 192)	0	batch_normalization_449[0][0]
conv2d_446 (Conv2D)	(None, 9, 11, 192)	147456	mixed7[0][0]
conv2d_450 (Conv2D)	(None, 9, 11, 192)	258048	activation_449[0][0]
batch_normalization_446 (BatchN	(None, 9, 11, 192)	576	conv2d_446[0][0]
batch_normalization_450 (BatchN	(None, 9, 11, 192)	576	conv2d_450[0][0]
activation_446 (Activation)	(None, 9, 11, 192)	0	batch_normalization_446[0][0]
activation_450 (Activation)	(None, 9, 11, 192)	0	batch_normalization_450[0][0]
conv2d_447 (Conv2D)	(None, 4, 5, 320)	552960	activation_446[0][0]
conv2d_451 (Conv2D)	(None, 4, 5, 192)	331776	activation_450[0][0]
batch_normalization_447 (BatchN	(None, 4, 5, 320)	960	conv2d_447[0][0]
batch_normalization_451 (BatchN	(None, 4, 5, 192)	576	conv2d_451[0][0]

activation_447 (Activation)	(None, 4, 5, 320)	0	batch_normalization_447[0][0]
activation_451 (Activation)	(None, 4, 5, 192)	0	batch_normalization_451[0][0]
max_pooling2d_19 (MaxPooling2D)	(None, 4, 5, 768)	0	mixed7[0][0]
mixed8 (Concatenate)	(None, 4, 5, 1280)	0	activation_447[0][0] activation_451[0][0] max_pooling2d_19[0][0]
conv2d_456 (Conv2D)	(None, 4, 5, 448)	573440	mixed8[0][0]
batch_normalization_456 (Batch Normalization)	(None, 4, 5, 448)	1344	conv2d_456[0][0]
activation_456 (Activation)	(None, 4, 5, 448)	0	batch_normalization_456[0][0]
conv2d_453 (Conv2D)	(None, 4, 5, 384)	491520	mixed8[0][0]
conv2d_457 (Conv2D)	(None, 4, 5, 384)	1548288	activation_456[0][0]
batch_normalization_453 (Batch Normalization)	(None, 4, 5, 384)	1152	conv2d_453[0][0]
batch_normalization_457 (Batch Normalization)	(None, 4, 5, 384)	1152	conv2d_457[0][0]
activation_453 (Activation)	(None, 4, 5, 384)	0	batch_normalization_453[0][0]
activation_457 (Activation)	(None, 4, 5, 384)	0	batch_normalization_457[0][0]
conv2d_454 (Conv2D)	(None, 4, 5, 384)	442368	activation_453[0][0]
conv2d_455 (Conv2D)	(None, 4, 5, 384)	442368	activation_453[0][0]
conv2d_458 (Conv2D)	(None, 4, 5, 384)	442368	activation_457[0][0]
conv2d_459 (Conv2D)	(None, 4, 5, 384)	442368	activation_457[0][0]
average_pooling2d_43 (AveragePooling2D)	(None, 4, 5, 1280)	0	mixed8[0][0]
conv2d_452 (Conv2D)	(None, 4, 5, 320)	409600	mixed8[0][0]
batch_normalization_454 (Batch Normalization)	(None, 4, 5, 384)	1152	conv2d_454[0][0]

batch_normalization_455 (BatchN	(None, 4, 5, 384)	1152	conv2d_455[0][0]
batch_normalization_458 (BatchN	(None, 4, 5, 384)	1152	conv2d_458[0][0]
batch_normalization_459 (BatchN	(None, 4, 5, 384)	1152	conv2d_459[0][0]
conv2d_460 (Conv2D)	(None, 4, 5, 192)	245760	average_pooling2d_43[0][0]
batch_normalization_452 (BatchN	(None, 4, 5, 320)	960	conv2d_452[0][0]
activation_454 (Activation)	(None, 4, 5, 384)	0	batch_normalization_454[0][0]
activation_455 (Activation)	(None, 4, 5, 384)	0	batch_normalization_455[0][0]
activation_458 (Activation)	(None, 4, 5, 384)	0	batch_normalization_458[0][0]
activation_459 (Activation)	(None, 4, 5, 384)	0	batch_normalization_459[0][0]
batch_normalization_460 (BatchN	(None, 4, 5, 192)	576	conv2d_460[0][0]
activation_452 (Activation)	(None, 4, 5, 320)	0	batch_normalization_452[0][0]
mixed9_0 (Concatenate)	(None, 4, 5, 768)	0	activation_454[0][0] activation_455[0][0]
concatenate_8 (Concatenate)	(None, 4, 5, 768)	0	activation_458[0][0] activation_459[0][0]
activation_460 (Activation)	(None, 4, 5, 192)	0	batch_normalization_460[0][0]
mixed9 (Concatenate)	(None, 4, 5, 2048)	0	activation_452[0][0] mixed9_0[0][0] concatenate_8[0][0] activation_460[0][0]
conv2d_465 (Conv2D)	(None, 4, 5, 448)	917504	mixed9[0][0]
batch_normalization_465 (BatchN	(None, 4, 5, 448)	1344	conv2d_465[0][0]
activation_465 (Activation)	(None, 4, 5, 448)	0	batch_normalization_465[0][0]
conv2d_462 (Conv2D)	(None, 4, 5, 384)	786432	mixed9[0][0]

conv2d_466 (Conv2D)	(None, 4, 5, 384)	1548288	activation_465[0][0]
batch_normalization_462 (BatchN	(None, 4, 5, 384)	1152	conv2d_462[0][0]
batch_normalization_466 (BatchN	(None, 4, 5, 384)	1152	conv2d_466[0][0]
activation_462 (Activation)	(None, 4, 5, 384)	0	batch_normalization_462[0][0]
activation_466 (Activation)	(None, 4, 5, 384)	0	batch_normalization_466[0][0]
conv2d_463 (Conv2D)	(None, 4, 5, 384)	442368	activation_462[0][0]
conv2d_464 (Conv2D)	(None, 4, 5, 384)	442368	activation_462[0][0]
conv2d_467 (Conv2D)	(None, 4, 5, 384)	442368	activation_466[0][0]
conv2d_468 (Conv2D)	(None, 4, 5, 384)	442368	activation_466[0][0]
average_pooling2d_44 (AveragePo	(None, 4, 5, 2048)	0	mixed9[0][0]
conv2d_461 (Conv2D)	(None, 4, 5, 320)	655360	mixed9[0][0]
batch_normalization_463 (BatchN	(None, 4, 5, 384)	1152	conv2d_463[0][0]
batch_normalization_464 (BatchN	(None, 4, 5, 384)	1152	conv2d_464[0][0]
batch_normalization_467 (BatchN	(None, 4, 5, 384)	1152	conv2d_467[0][0]
batch_normalization_468 (BatchN	(None, 4, 5, 384)	1152	conv2d_468[0][0]
conv2d_469 (Conv2D)	(None, 4, 5, 192)	393216	average_pooling2d_44[0][0]
batch_normalization_461 (BatchN	(None, 4, 5, 320)	960	conv2d_461[0][0]
activation_463 (Activation)	(None, 4, 5, 384)	0	batch_normalization_463[0][0]
activation_464 (Activation)	(None, 4, 5, 384)	0	batch_normalization_464[0][0]
activation_467 (Activation)	(None, 4, 5, 384)	0	batch_normalization_467[0][0]
activation_468 (Activation)	(None, 4, 5, 384)	0	batch_normalization_468[0][0]
batch_normalization_469 (BatchN	(None, 4, 5, 192)	576	conv2d_469[0][0]

activation_461 (Activation)	(None, 4, 5, 320)	0	batch_normalization_461[0][0]
mixed9_1 (Concatenate)	(None, 4, 5, 768)	0	activation_463[0][0] activation_464[0][0]
concatenate_9 (Concatenate)	(None, 4, 5, 768)	0	activation_467[0][0] activation_468[0][0]
activation_469 (Activation)	(None, 4, 5, 192)	0	batch_normalization_469[0][0]
mixed10 (Concatenate)	(None, 4, 5, 2048)	0	activation_461[0][0] mixed9_1[0][0] concatenate_9[0][0] activation_469[0][0]

=====

Total params: 21,802,784
Trainable params: 192
Non-trainable params: 21,802,592

```
In [29]: inception_md1.summary()
```

```
Model: "sequential_6"
```

Layer (type)	Output Shape	Param #
inception_v3 (Model)	(None, 4, 5, 2048)	21802784
flatten_6 (Flatten)	(None, 40960)	0
dense_13 (Dense)	(None, 256)	10486016
dropout_7 (Dropout)	(None, 256)	0
dense_14 (Dense)	(None, 1)	257

=====

Total params: 32,289,057
Trainable params: 11,536,897
Non-trainable params: 20,752,160


```
In [33]: train_dir = (r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
            r'\Assignment_2\Images\train')
validation_dir = (r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                  r'\Assignment_2\Images\validation')

train_datagen = ImageDataGenerator(
    rescale=1./255,
    rotation_range=40,
    width_shift_range=0.2,
    height_shift_range=0.2,
    shear_range=0.2,
    zoom_range=0.2,
    horizontal_flip=True,
    fill_mode='nearest')

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(178, 218),
    batch_size=20,
    class_mode='binary')

val_datagen = ImageDataGenerator(rescale=1./255)
validation_generator = val_datagen.flow_from_directory(
    validation_dir,
    target_size=(178, 218),
    batch_size=20,
    class_mode='binary')

inception_md1.compile(loss='binary_crossentropy',
    optimizer=optimizers.RMSprop(learning_rate=0.000005, rho=0.9),
    metrics=['acc'])

history = inception_md1.fit_generator(
```

```
train_generator,  
steps_per_epoch=round(training_count/20),  
epochs=20,  
validation_data=validation_generator,  
validation_steps=round(val_count/20))
```

```
Found 121559 images belonging to 2 classes.
Found 40520 images belonging to 2 classes.
WARNING:tensorflow:From <ipython-input-33-c0b1d441167d>:38: Model.fit_generator (from tensorflow.python.keras.engine.training) is deprecated and will be removed in a future version.
Instructions for updating:
Please use Model.fit, which supports generators.
WARNING:tensorflow:sample_weight modes were coerced from
...
to
['...']
WARNING:tensorflow:sample_weight modes were coerced from
...
to
['...']
Train for 6078 steps, validate for 2026 steps
Epoch 1/20
6078/6078 [=====] - 1452s 239ms/step - loss: 0.3544 - acc: 0.8432 - val_loss: 1.7566 - val_acc: 0.7972
Epoch 2/20
6078/6078 [=====] - 1419s 233ms/step - loss: 0.2946 - acc: 0.8799 - val_loss: 2.5109 - val_acc: 0.7993
Epoch 3/20
6078/6078 [=====] - 1379s 227ms/step - loss: 0.2817 - acc: 0.8901 - val_loss: 3.1236 - val_acc: 0.7859
Epoch 4/20
6078/6078 [=====] - 1399s 230ms/step - loss: 0.2782 - acc: 0.8927 - val_loss: 3.5366 - val_acc: 0.7776
Epoch 5/20
6078/6078 [=====] - 1403s 231ms/step - loss: 0.2722 - acc: 0.8971 - val_loss: 4.0775 - val_acc: 0.7614
Epoch 6/20
6078/6078 [=====] - 1401s 231ms/step - loss: 0.2693 - acc: 0.8990 - val_loss: 4.9030 - val_acc: 0.7404
Epoch 7/20
6078/6078 [=====] - 1428s 235ms/step - loss: 0.2640 - acc: 0.9015 - val_loss: 6.1425 - val_acc: 0.7106
Epoch 8/20
6078/6078 [=====] - 1420s 234ms/step - loss: 0.2618 - acc: 0.9031 - val_loss: 6.2003 - val_acc: 0.7088
Epoch 9/20
6078/6078 [=====] - 1414s 233ms/step - loss: 0.2593 - acc: 0.9044 - val_loss: 6.6808 - val_acc: 0.6973
Epoch 10/20
6078/6078 [=====] - 1519s 250ms/step - loss: 0.2502 - acc: 0.9072 - val_loss: 7.4777 - val_acc: 0.6880
Epoch 11/20
6078/6078 [=====] - 1761s 290ms/step - loss: 0.2494 - acc: 0.9077 - val_loss: 7.1592 - val_acc: 0.6945
Epoch 12/20
6078/6078 [=====] - 1375s 226ms/step - loss: 0.2456 - acc: 0.9088 - val_loss: 7.9718 - val_acc: 0.6788
Epoch 13/20
6078/6078 [=====] - 1369s 225ms/step - loss: 0.2458 - acc: 0.9090 - val_loss: 6.8948 - val_acc: 0.7016
Epoch 14/20
6078/6078 [=====] - 1369s 225ms/step - loss: 0.2401 - acc: 0.9108 - val_loss: 9.1274 - val_acc: 0.6609
```

Epoch 15/20

6078/6078 [=====] - 1366s 225ms/step - loss: 0.2384 - acc: 0.9111 - val_loss: 8.3881 - val_acc: 0.6797

Epoch 16/20

6078/6078 [=====] - 1368s 225ms/step - loss: 0.2350 - acc: 0.9129 - val_loss: 8.8567 - val_acc: 0.6768

Epoch 17/20

6078/6078 [=====] - 1363s 224ms/step - loss: 0.2359 - acc: 0.9126 - val_loss: 10.3694 - val_acc: 0.6518

Epoch 18/20

6078/6078 [=====] - 1362s 224ms/step - loss: 0.2331 - acc: 0.9130 - val_loss: 8.5357 - val_acc: 0.6800

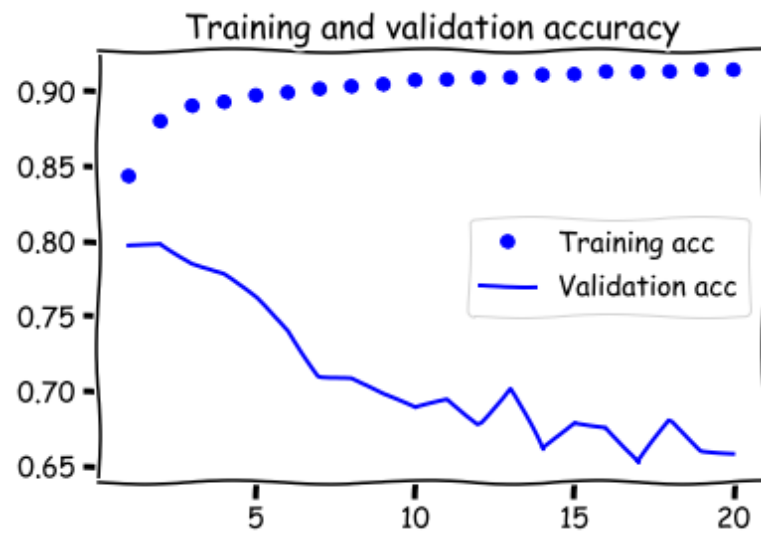
Epoch 19/20

6078/6078 [=====] - 1383s 228ms/step - loss: 0.2335 - acc: 0.9142 - val_loss: 9.8357 - val_acc: 0.6606

Epoch 20/20

6078/6078 [=====] - 1539s 253ms/step - loss: 0.2323 - acc: 0.9141 - val_loss: 10.0375 - val_acc: 0.6594

```
In [34]: acc = history.history['acc']
val_acc = history.history['val_acc']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs = range(1, len(acc) + 1)
plt.plot(epochs, acc, 'bo', label='Training acc')
plt.plot(epochs, val_acc, 'b', label='Validation acc')
plt.title('Training and validation accuracy')
plt.legend()
plt.figure()
plt.plot(epochs, loss, 'bo', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
plt.title('Training and validation loss')
plt.legend()
plt.show()
```



```
In [15]: vgg16_conv_base = VGG16(weights='imagenet',
      include_top=False,
      input_shape=(178, 218, 3))

      vgg16_md1 = Sequential()
      vgg16_md1.add(vgg16_conv_base)
      vgg16_md1.add(Flatten())
      vgg16_md1.add(Dense(256, activation='relu'))
      vgg16_md1.add(Dense(1, activation='sigmoid'))

      vgg16_conv_base.trainable = False
```



```
In [16]: train_dir = (r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                r'\Assignment_2\Images\train')
validation_dir = (r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                  r'\Assignment_2\Images\validation')

train_datagen = ImageDataGenerator(
    rescale=1./255,
    #rotation_range=40,
    #width_shift_range=0.2,
    #height_shift_range=0.2,
    #shear_range=0.2,
    #zoom_range=0.2,
    #horizontal_flip=True,
    fill_mode='nearest')

val_datagen = ImageDataGenerator(rescale=1./255)

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(178, 218),
```

```
        batch_size=5,
        class_mode='binary')

validation_generator = val_datagen.flow_from_directory(
    validation_dir,
    target_size=(178, 218),
    batch_size=5,
    class_mode='binary')

vgg16_md1.compile(loss='binary_crossentropy',
    optimizer=optimizers.RMSprop(learning_rate=0.000005, rho=0.9),
    metrics=['acc'])

history = vgg16_md1.fit_generator(
    train_generator,
    steps_per_epoch=round(training_count/5),
    epochs=5,
    validation_data=validation_generator,
    validation_steps= round(val_count/5))
```

Found 121559 images belonging to 2 classes.

Found 40520 images belonging to 2 classes.

WARNING:tensorflow:sample_weight modes were coerced from

...

to

['...']

WARNING:tensorflow:sample_weight modes were coerced from

...

to

['...']

Train for 24312 steps, validate for 8104 steps

Epoch 1/5

24312/24312 [=====] - 2451s 101ms/step - loss: 0.1931 - acc: 0.9236 - val_loss: 0.1660 - val_acc: 0.9374

Epoch 2/5

24312/24312 [=====] - 2460s 101ms/step - loss: 0.1543 - acc: 0.9414 - val_loss: 0.1582 - val_acc: 0.9424

Epoch 3/5

24312/24312 [=====] - 2460s 101ms/step - loss: 0.1480 - acc: 0.9458 - val_loss: 0.1591 - val_acc: 0.9439

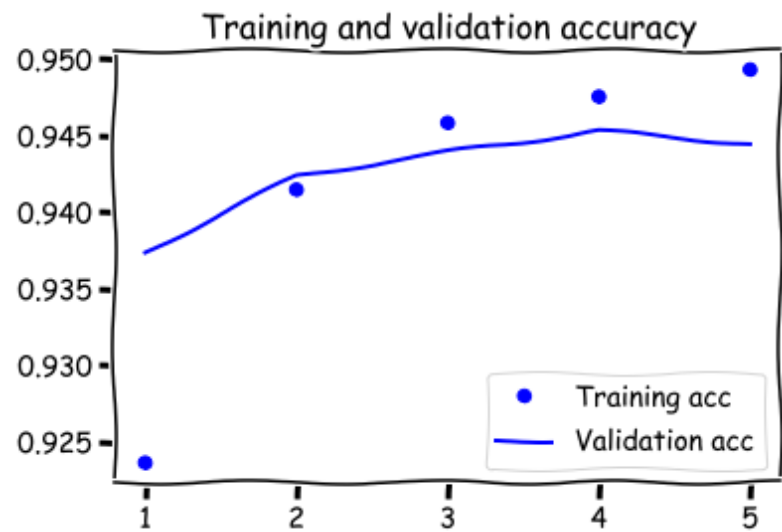
Epoch 4/5

24312/24312 [=====] - 2469s 102ms/step - loss: 0.1458 - acc: 0.9475 - val_loss: 0.1610 - val_acc: 0.9453

Epoch 5/5

24312/24312 [=====] - 2470s 102ms/step - loss: 0.1446 - acc: 0.9493 - val_loss: 0.1661 - val_acc: 0.9445

```
In [17]: acc = history.history['acc']
val_acc = history.history['val_acc']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs = range(1, len(acc) + 1)
plt.plot(epochs, acc, 'bo', label='Training acc')
plt.plot(epochs, val_acc, 'b', label='Validation acc')
plt.title('Training and validation accuracy')
plt.legend()
plt.figure()
plt.plot(epochs, loss, 'bo', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
plt.title('Training and validation loss')
plt.legend()
plt.show()
```



```
In [18]: vgg16_md1.save('male_female_classification.h5')
```

```
In [12]: mdl = load_model('male_female_classification.h5')
```

```
In [28]: test_dir = Path(r'D:\Google_Drive_Sync\Data\Documents\School\UL\CECS590_Deep_Learning'
                        r'\Assignment_2\Images\test')
test_datagen = ImageDataGenerator(rescale=1./255)
test_generator = test_datagen.flow_from_directory(
    test_dir,
    target_size=(178, 218),
    batch_size=1,
    class_mode='binary')
```

Found 40520 images belonging to 2 classes.

```
In [29]: test_truth = []
test_pred = []
for idx, x in enumerate(test_generator):
    if idx == 40520: break
    img_array = x[0]
    arr = img_array[0]
    img_class = mdl.predict_classes(img_array)

    test_truth.append(x[1][0])
    test_pred.append(img_class[0][0])
    #img = array_to_img(arr)
    #display(img)
    #if img_class[0][0]:
    #    print("Male")
    #else:
    #    print("Female")
    if (1 + idx) % 1000 == 0:
        print('Processed %i images...' % (1 + idx))
```

Processed 1000 images...
Processed 2000 images...
Processed 3000 images...
Processed 4000 images...
Processed 5000 images...
Processed 6000 images...
Processed 7000 images...
Processed 8000 images...
Processed 9000 images...
Processed 10000 images...
Processed 11000 images...
Processed 12000 images...
Processed 13000 images...
Processed 14000 images...
Processed 15000 images...
Processed 16000 images...
Processed 17000 images...
Processed 18000 images...
Processed 19000 images...
Processed 20000 images...
Processed 21000 images...
Processed 22000 images...
Processed 23000 images...
Processed 24000 images...
Processed 25000 images...
Processed 26000 images...
Processed 27000 images...
Processed 28000 images...
Processed 29000 images...
Processed 30000 images...
Processed 31000 images...
Processed 32000 images...
Processed 33000 images...
Processed 34000 images...
Processed 35000 images...
Processed 36000 images...
Processed 37000 images...
Processed 38000 images...
Processed 39000 images...
Processed 40000 images...


```
In [30]: class_names = ['Female', 'Male']
```

```

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="#555555" if cm[i, j] > thresh else "black")

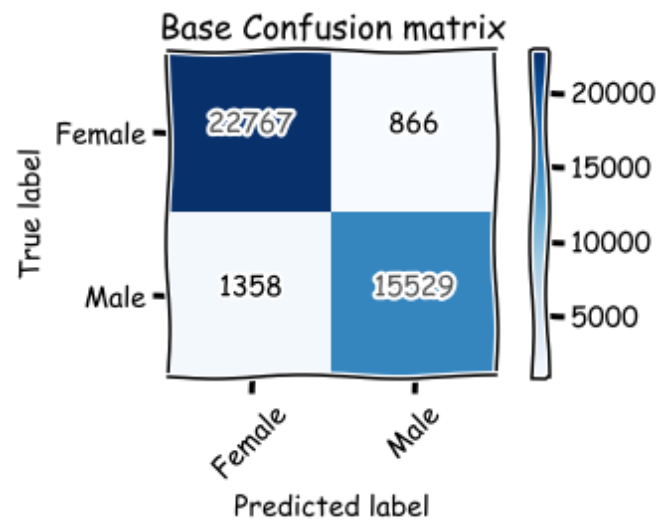
    plt.ylabel('True label')
    plt.xlabel('Predicted label')
    plt.tight_layout()

```

```
In [31]: cnf = confusion_matrix(test_truth, test_pred)
```

```
In [34]: plot_confusion_matrix(cnf, classes=class_names, title='Base Confusion matrix')
```

Confusion matrix, without normalization
[[22767 866]
 [1358 15529]]

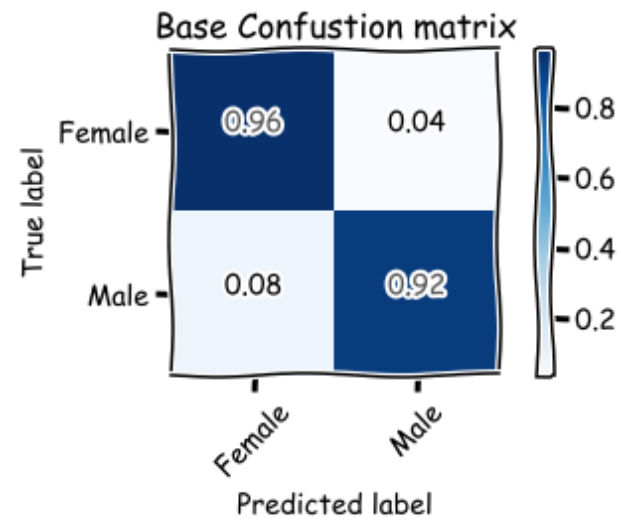


```
In [35]: plot_confusion_matrix(cnf, normalize=True, classes=class_names, title='Base Confustion matrix')
```

Normalized confusion matrix

```
[[0.96335632 0.03664368]
```

```
 [0.08041689 0.91958311]]
```



```
In [37]: for idx, x in enumerate(test_generator):
        if idx == 1000: break
        img_array = x[0]
        arr = img_array[0]
        img_class = mdl.predict_classes(img_array)
        img = array_to_img(arr)
        if x[1][0] != img_class[0][0]:
            display(img)
            if img_class[0][0]:
                print("Male")
            else:
                print("Female")
        if (1 + idx) % 1000 == 0:
            print('Processed %i images...' % (1 + idx))
```



Male



Male



Female



Male



Female



Male



Male



Female



Female



Female



Female



Female



Male



Male



Female



Male



Male



Male



Female



Male



Male



Female



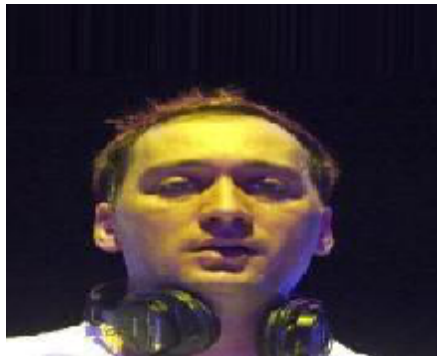
Female



Male



Female



Female



Female



Female



Female



Male



Female



Female



Female



Male



Female



Male



Male



Male



Male



Female



Female



Female



Female



Male



Female



Female



Male



Female



Female



Male



Female



Female



Male



Female



Female

Processed 1000 images...

In []: