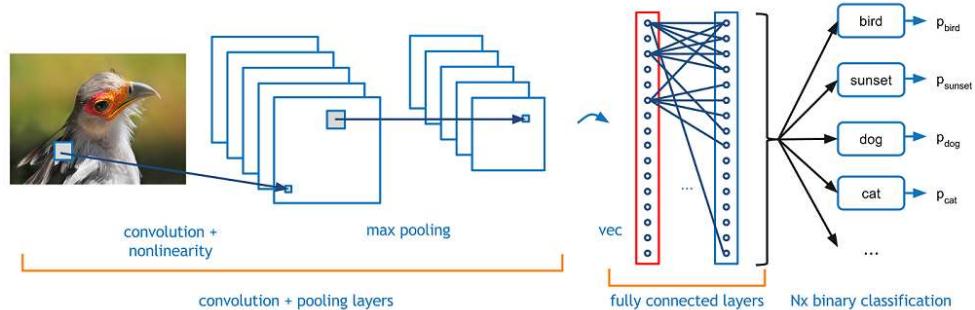


Type *Markdown* and *LaTeX*:  $\alpha^2$

# CNN for lesion classification

## 1. Brief CNN theory

A Convolutional Neural Network (CNN, or ConvNet) is a type of **feed-forward** artificial neural network in which the connectivity pattern between its neurons is inspired by the organization of the animal visual cortex.



source: [\(https://flickrccode.files.wordpress.com/2014/10/conv-net2.png\)](https://flickrccode.files.wordpress.com/2014/10/conv-net2.png)

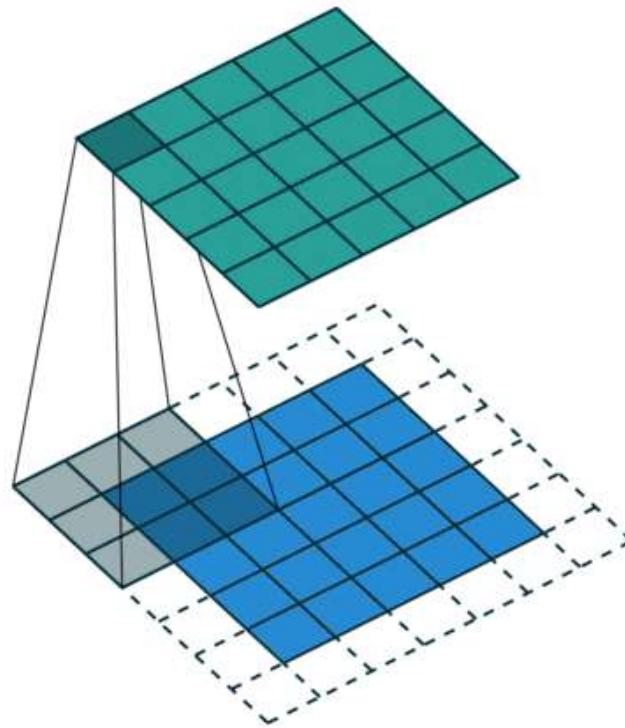
### 1.1 Structure of a CNN

A more detailed overview of what CNNs do would be that you take the image, pass it through a series of convolutional, nonlinear, pooling (downsampling), and fully connected layers, and get an output. As we said earlier, the output can be a single class or a probability of classes that best describes the image.

source: [1]

#### Convolutional Layer

The first layer in a CNN is always a **Convolutional Layer**.



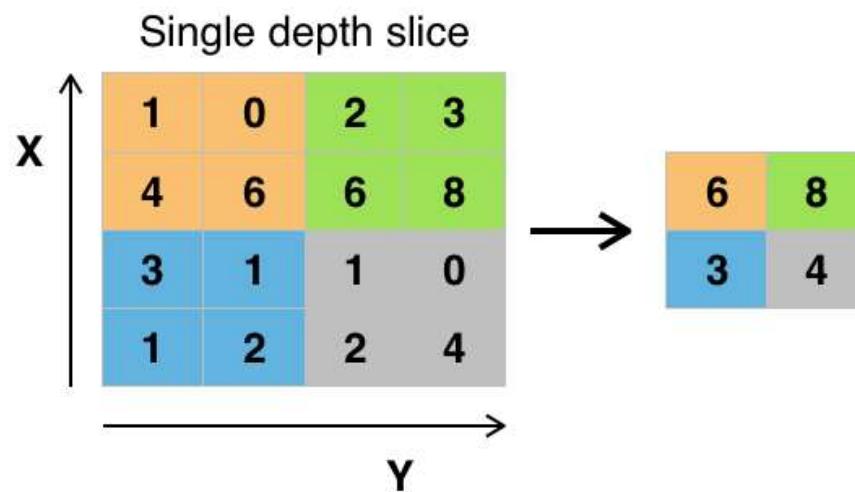
### Typical CNN Structure

A traditional CNN architecture consists of other layers interspaced between convolution layers

Input -> Conv -> ReLU -> Conv -> ReLU -> Pool -> ReLU -> Conv -> ReLU -> Pool -> Fully Connected

### Pooling layer

After some ReLU layers, **pooling layer** is typically applied.



Example of Maxpool with a  $2 \times 2$  filter and a stride of 2

Pooling reduces the amount of parameters (helping with computational efficiency) and controls overfitting

## 2. We build one using keras and tensorflow

## 2.1 Preparation

In [13]:

```
# setup code for this notebook
import numpy as np
import matplotlib.pyplot as plt
from functions import data, Timer
timer = Timer()

# This makes matplotlib figures appear inline in the notebook
# rather than in a new window.
%matplotlib inline
plt.rcParams['figure.figsize'] = (10.0, 8.0) # set default size of plots
plt.rcParams['image.interpolation'] = 'nearest'
plt.rcParams['image.cmap'] = 'gray'

# Make the notebook reload external python modules;
# see http://stackoverflow.com/questions/1907993/autoreload-of-modules-in-ipython
%load_ext autoreload
%autoreload 2
```

The autoreload extension is already loaded. To reload it, use:

```
%reload_ext autoreload
```

### The ultra sound scan data

- 163 scans total, clinically confirmed as having either benign or malignant (cancerous) lesions
- 100 scans for training, 63 for testing
- Training data was passed through 7 transformations to give us 800 training images total
- We balanced the training data to have half malignant and half benign
- Since the malignant cases were less than benign cases, we use only 528 images for training
- Empirically we have found this improves our overall performance
- Testing images not transformed
- Both training and testing images were resized to 224X224
- Raw pngs then converted to numpy arrays and saved

In [14]:

```
# Import keras Libraries
from keras.models import Sequential
from keras.applications.inception_v3 import InceptionV3
from keras.layers.core import Dense, Dropout, Activation, Flatten
from keras.layers.convolutional import Conv2D
from keras.layers.pooling import MaxPooling2D
from keras.layers import GlobalAveragePooling2D
from keras import backend as K
from keras.models import Model
from keras.layers import Input
from functions import data, Timer
```

In [15]:

```
img_rows, img_cols = 224, 224 # 224, 224 works resized down from 360, 528
color_channels = 3

if K.image_data_format() == 'channels_first':
    input_shape = (color_channels, img_rows, img_cols)
else:
    input_shape = (img_rows, img_cols, color_channels)

print('Input shape', input_shape)
```

Input shape (224, 224, 3)

In [16]:

```
# data Loader and generator helper methods
from keras.preprocessing.image import ImageDataGenerator
```

In [17]:

```
# data readers
base = "J:\\final year project\\code and models\\data\\augmented\\"
train_directory = base+'training'
validation_directory = base+'validation'

batch_size = 8

# normalization
train_generator = ImageDataGenerator(rescale=1./255)
validation_generator = ImageDataGenerator(rescale=1./255)

# this is a generator that will read scans found in
# the train directory, and indefinitely generate
# batches of image data
train_generator = train_generator.flow_from_directory(
    train_directory,
    target_size=(img_rows, img_cols),
    batch_size=batch_size,
    class_mode='binary')

# A similar generator, for validation data
validation_generator = validation_generator.flow_from_directory(
    validation_directory,
    target_size=(img_rows, img_cols),
    batch_size=batch_size,
    class_mode='binary')
```

Found 400 images belonging to 2 classes.  
 Found 128 images belonging to 2 classes.

In [18]:

```
# Visualize
from IPython.display import SVG
from keras.utils.vis_utils import model_to_dot

def visualize(model):
    model.summary()
    SVG(model_to_dot(model).create(prog='dot', format='svg'))
```

In [19]:

```
def getModelMemoryUsage(batch_size, model):
    shapes_mem_count = 0
    for l in model.layers:
        single_layer_mem = 1
        for s in l.output_shape:
            if s is None:
                continue
            single_layer_mem *= s
        shapes_mem_count += single_layer_mem

    trainable_count = np.sum([K.count_params(p) for p in set(model.trainable_weights)])
    non_trainable_count = np.sum([K.count_params(p) for p in set(model.non_trainable_weights)])

    total_memory = 4.0*batch_size*(shapes_mem_count + trainable_count + non_trainable_count)
    gbytes = np.round(total_memory / (1024.0 ** 3), 3)
    return str(gbytes) + " GB", str(gbytes*2)
```

## 2.2 Retraining InceptionV3 CNN

In [20]:

```
# We define the custom inception based model
def buildBaseModel():
    input_tensor = Input(shape=input_shape)
    base_model = InceptionV3(input_tensor=input_tensor, weights='imagenet', include_top=False)

    # we add a global spatial average pooling layer
    x = base_model.output
    x = GlobalAveragePooling2D()(x)

    # we add a fully-connected layer
    x = Dense(256, activation='relu')(x)

    # we add a logistic layer for our 2 classes
    predictions = Dense(1, activation='softmax')(x)

    # this is our model, a hybrid inceptionv3
    model = Model(inputs=base_model.input, outputs=predictions)

    # first we train our custom top layer
    # we freeze all convolutional inceptionv3 layers
    for layer in base_model.layers:
        layer.trainable = False

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

    return model
```

In [21]:

```
model = buildBaseModel()
```

In [22]:

```
# helpers for checkpointing and early stopping
from keras.callbacks import ModelCheckpoint, EarlyStopping

def trainModel(model, epochs=5, text="Re training inception model",
              file_name='best_cnn_inc_model.h5'):
    best_model_file = file_name
    early_stop = EarlyStopping(monitor='val_loss', patience=3, verbose=True)
    best_model = ModelCheckpoint(best_model_file, verbose=True, save_best_only=True)

    timer.start()
    network_history = model.fit_generator(
        train_generator,
        steps_per_epoch=800,
        epochs=epochs,
        validation_data = validation_generator,
        validation_steps=256,
        verbose=True,
        callbacks=[best_model])
    timer.stop(text)
    return network_history
```

In [23]:

```
# visualize(model)
model.summary()
gpu, ram = getModelMemoryUsage(batch_size, model)
print("GPU Memory:" + gpu + "RAM:" + ram)
```

Layer (type)	Output Shape	Param #	Connected to
0			
input_2 (InputLayer)	(None, 224, 224, 3)	0	
conv2d_95 (Conv2D)	(None, 111, 111, 32)	864	input_2[0]
[0]			
batch_normalization_95 (BatchNormalizat	(None, 111, 111, 32)	96	conv2d_95
on)[0][0]			
activation_95 (Activation)	(None, 111, 111, 32)	0	batch_norma
lization_95[0][0]			
conv2d_96 (Conv2D)	(None, 109, 109, 32)	9216	activation_
95[0][0]			
batch_normalization_96 (BatchNormalizat	(None, 109, 109, 32)	96	conv2d_96
on)[0][0]			
activation_96 (Activation)	(None, 109, 109, 32)	0	batch_norma
lization_96[0][0]			
conv2d_97 (Conv2D)	(None, 109, 109, 64)	18432	activation_
96[0][0]			
batch_normalization_97 (BatchNormalizat	(None, 109, 109, 64)	192	conv2d_97
on)[0][0]			
activation_97 (Activation)	(None, 109, 109, 64)	0	batch_norma
lization_97[0][0]			
max_pooling2d_5 (MaxPooling2D)	(None, 54, 54, 64)	0	activation_
97[0][0]			
conv2d_98 (Conv2D)	(None, 54, 54, 80)	5120	max_pooling
2d_5[0][0]			

batch_normalization_98 (BatchNo [0][0]	(None, 54, 54, 80)	240	conv2d_98
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activation_98 (Activation) [0][0]	(None, 54, 54, 80)	0	batch_norma
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conv2d_99 (Conv2D) [0][0]	(None, 52, 52, 192)	138240	activation_
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batch_normalization_99 [0][0]	(BatchNo (None, 52, 52, 192)	576	conv2d_99
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activation_99 (Activation) [0][0]	(None, 52, 52, 192)	0	batch_norma
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max_pooling2d_6 (MaxPooling2D) [0][0]	(None, 25, 25, 192)	0	activation_
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conv2d_103 (Conv2D) [0][0]	(None, 25, 25, 64)	12288	max_pooling
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batch_normalization_103 [0][0]	(BatchN (None, 25, 25, 64)	192	conv2d_103
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activation_103 (Activation) [0][0]	(None, 25, 25, 64)	0	batch_norma
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conv2d_101 (Conv2D) [0][0]	(None, 25, 25, 48)	9216	max_pooling
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conv2d_104 (Conv2D) [0][0]	(None, 25, 25, 96)	55296	activation_
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batch_normalization_101 [0][0]	(BatchN (None, 25, 25, 48)	144	conv2d_101
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batch_normalization_104 [0][0]	(BatchN (None, 25, 25, 96)	288	conv2d_104
--------------------------------	----------------------------	-----	------------

activation_101 (Activation) [0][0]	(None, 25, 25, 48)	0	batch_norma
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activation_104 (Activation) [0][0]	(None, 25, 25, 96)	0	batch_norma
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average_pooling2d_10 (AveragePo [0][0]	(None, 25, 25, 192)	0	max_pooling
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2d\_6[0][0]

conv2d_100 (Conv2D) 2d_6[0][0]	(None, 25, 25, 64)	12288	max_pooling
conv2d_102 (Conv2D) 101[0][0]	(None, 25, 25, 64)	76800	activation_
conv2d_105 (Conv2D) 104[0][0]	(None, 25, 25, 96)	82944	activation_
conv2d_106 (Conv2D) 105[0][0]	(None, 25, 25, 32)	6144	average_poo
batch_normalization_100 (BatchN) [0][0]	(None, 25, 25, 64)	192	conv2d_100
batch_normalization_102 (BatchN) [0][0]	(None, 25, 25, 64)	192	conv2d_102
batch_normalization_105 (BatchN) [0][0]	(None, 25, 25, 96)	288	conv2d_105
batch_normalization_106 (BatchN) [0][0]	(None, 25, 25, 32)	96	conv2d_106
activation_100 (Activation) batch_normalization_100[0][0]	(None, 25, 25, 64)	0	batch_norma
activation_102 (Activation) batch_normalization_102[0][0]	(None, 25, 25, 64)	0	batch_norma
activation_105 (Activation) batch_normalization_105[0][0]	(None, 25, 25, 96)	0	batch_norma
activation_106 (Activation) batch_normalization_106[0][0]	(None, 25, 25, 32)	0	batch_norma
mixed0 (Concatenate) 100[0][0]	(None, 25, 25, 256)	0	activation_
102[0][0]			activation_
105[0][0]			activation_
106[0][0]			activation_

conv2d_110 (Conv2D) [0]	(None, 25, 25, 64)	16384	mixed0[0]
batch_normalization_110 (BatchN [0][0])	(None, 25, 25, 64)	192	conv2d_110 [0][0]
activation_110 (Activation) activation_110[0][0]	(None, 25, 25, 64)	0	batch_norma lization_110[0][0]
conv2d_108 (Conv2D) [0]	(None, 25, 25, 48)	12288	mixed0[0]
conv2d_111 (Conv2D) 110[0][0]	(None, 25, 25, 96)	55296	activation_
batch_normalization_108 (BatchN [0][0])	(None, 25, 25, 48)	144	conv2d_108 [0][0]
batch_normalization_111 (BatchN [0][0])	(None, 25, 25, 96)	288	conv2d_111 [0][0]
activation_108 (Activation) activation_108[0][0]	(None, 25, 25, 48)	0	batch_norma lization_108[0][0]
activation_111 (Activation) activation_111[0][0]	(None, 25, 25, 96)	0	batch_norma lization_111[0][0]
average_pooling2d_11 (AveragePo [0])	(None, 25, 25, 256)	0	mixed0[0]
conv2d_107 (Conv2D) [0]	(None, 25, 25, 64)	16384	mixed0[0]
conv2d_109 (Conv2D) 108[0][0]	(None, 25, 25, 64)	76800	activation_
conv2d_112 (Conv2D) 111[0][0]	(None, 25, 25, 96)	82944	activation_
conv2d_113 (Conv2D) 111[0][0]	(None, 25, 25, 64)	16384	average_poo ling2d_11[0][0]
batch_normalization_107 (BatchN [0][0])	(None, 25, 25, 64)	192	conv2d_107 [0][0]
batch_normalization_109 (BatchN [0][0])	(None, 25, 25, 64)	192	conv2d_109

[0][0]

batch_normalization_112 (BatchN (None, 25, 25, 96)	288	conv2d_112
[0][0]		
batch_normalization_113 (BatchN (None, 25, 25, 64)	192	conv2d_113
[0][0]		
activation_107 (Activation) (None, 25, 25, 64)	0	batch_normalization_107[0][0]
activation_109 (Activation) (None, 25, 25, 64)	0	batch_normalization_109[0][0]
activation_112 (Activation) (None, 25, 25, 96)	0	batch_normalization_112[0][0]
activation_113 (Activation) (None, 25, 25, 64)	0	batch_normalization_113[0][0]
mixed1 (Concatenate) (None, 25, 25, 288)	0	activation_107[0][0]
activation_109[0][0]		
activation_112[0][0]		
activation_113[0][0]		
conv2d_117 (Conv2D) (None, 25, 25, 64)	18432	mixed1[0]
[0]		
batch_normalization_117 (BatchN (None, 25, 25, 64)	192	conv2d_117
[0][0]		
activation_117 (Activation) (None, 25, 25, 64)	0	batch_normalization_117[0][0]
conv2d_115 (Conv2D) (None, 25, 25, 48)	13824	mixed1[0]
[0]		
conv2d_118 (Conv2D) (None, 25, 25, 96)	55296	activation_117[0][0]
batch_normalization_115 (BatchN (None, 25, 25, 48)	144	conv2d_115
[0][0]		

batch_normalization_118 (BatchN (None, 25, 25, 96) [0][0]	288	conv2d_118
activation_115 (Activation) lization_115[0][0]	0	batch_norma
activation_118 (Activation) lization_118[0][0]	0	batch_norma
average_pooling2d_12 (AveragePo [0]	0	mixed1[0]
conv2d_114 (Conv2D) [0]	18432	mixed1[0]
conv2d_116 (Conv2D) 115[0][0]	76800	activation_
conv2d_119 (Conv2D) 118[0][0]	82944	activation_
conv2d_120 (Conv2D) ling2d_12[0][0]	18432	average_poo
batch_normalization_114 (BatchN [0][0]	192	conv2d_114
batch_normalization_116 (BatchN [0][0]	192	conv2d_116
batch_normalization_119 (BatchN [0][0]	288	conv2d_119
batch_normalization_120 (BatchN [0][0]	192	conv2d_120
activation_114 (Activation) lization_114[0][0]	0	batch_norma
activation_116 (Activation) lization_116[0][0]	0	batch_norma
activation_119 (Activation) lization_119[0][0]	0	batch_norma
activation_120 (Activation)	0	batch_norma

lization\_120[0][0]

---

mixed2 (Concatenate) 114[0][0]	(None, 25, 25, 288) 0	activation_ activation_
116[0][0]		activation_
119[0][0]		activation_
120[0][0]		activation_

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conv2d_122 (Conv2D) [0]	(None, 25, 25, 64) 18432	mixed2[0]
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batch_normalization_122 (BatchN) [0][0]	(None, 25, 25, 64) 192	conv2d_122
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activation_122 (Activation) lization_122[0][0]	(None, 25, 25, 64) 0	batch_norma
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conv2d_123 (Conv2D) 122[0][0]	(None, 25, 25, 96) 55296	activation_
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batch_normalization_123 (BatchN) [0][0]	(None, 25, 25, 96) 288	conv2d_123
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activation_123 (Activation) lization_123[0][0]	(None, 25, 25, 96) 0	batch_norma
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conv2d_121 (Conv2D) [0]	(None, 12, 12, 384) 995328	mixed2[0]
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conv2d_124 (Conv2D) 123[0][0]	(None, 12, 12, 96) 82944	activation_
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batch_normalization_121 (BatchN) [0][0]	(None, 12, 12, 384) 1152	conv2d_121
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batch_normalization_124 (BatchN) [0][0]	(None, 12, 12, 96) 288	conv2d_124
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---

activation_121 (Activation) lization_121[0][0]	(None, 12, 12, 384) 0	batch_norma
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activation_124 (Activation) lization_124[0][0]	(None, 12, 12, 96) 0	batch_norma
---	----------------------	-------------

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max_pooling2d_7 (MaxPooling2D)	(None, 12, 12, 288)	0	mixed2[0]
[0]			

---

mixed3 (Concatenate)	(None, 12, 12, 768)	0	activation_
121[0][0]			activation_
124[0][0]			max_pooling
2d_7[0][0]			

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conv2d_129 (Conv2D)	(None, 12, 12, 128)	98304	mixed3[0]
[0]			

---

batch_normalization_129 (BatchN)	(None, 12, 12, 128)	384	conv2d_129
[0][0]			

---

activation_129 (Activation)	(None, 12, 12, 128)	0	batch_norma
lization_129[0][0]			lization_

---

conv2d_130 (Conv2D)	(None, 12, 12, 128)	114688	activation_
129[0][0]			activation_

---

batch_normalization_130 (BatchN)	(None, 12, 12, 128)	384	conv2d_130
[0][0]			

---

activation_130 (Activation)	(None, 12, 12, 128)	0	batch_norma
lization_130[0][0]			lization_

---

conv2d_126 (Conv2D)	(None, 12, 12, 128)	98304	mixed3[0]
[0]			

---

conv2d_131 (Conv2D)	(None, 12, 12, 128)	114688	activation_
130[0][0]			activation_

---

batch_normalization_126 (BatchN)	(None, 12, 12, 128)	384	conv2d_126
[0][0]			

---

batch_normalization_131 (BatchN)	(None, 12, 12, 128)	384	conv2d_131
[0][0]			

---

activation_126 (Activation)	(None, 12, 12, 128)	0	batch_norma
lization_126[0][0]			lization_

---

activation_131 (Activation)	(None, 12, 12, 128)	0	batch_norma
lization_131[0][0]			lization_

---

conv2d_127 (Conv2D)	(None, 12, 12, 128)	114688	activation_

---

126[0][0]

conv2d_132 (Conv2D) 131[0][0]	(None, 12, 12, 128)	114688	activation_
batch_normalization_127 (BatchN [0][0]	(None, 12, 12, 128)	384	conv2d_127
batch_normalization_132 (BatchN [0][0]	(None, 12, 12, 128)	384	conv2d_132
activation_127 (Activation) lization_127[0][0]	(None, 12, 12, 128)	0	batch_norma
activation_132 (Activation) lization_132[0][0]	(None, 12, 12, 128)	0	batch_norma
average_pooling2d_13 (AveragePo [0]	(None, 12, 12, 768)	0	mixed3[0]
conv2d_125 (Conv2D) [0]	(None, 12, 12, 192)	147456	mixed3[0]
conv2d_128 (Conv2D) 127[0][0]	(None, 12, 12, 192)	172032	activation_
conv2d_133 (Conv2D) 132[0][0]	(None, 12, 12, 192)	172032	activation_
conv2d_134 (Conv2D) ling2d_13[0][0]	(None, 12, 12, 192)	147456	average_poo
batch_normalization_125 (BatchN [0][0]	(None, 12, 12, 192)	576	conv2d_125
batch_normalization_128 (BatchN [0][0]	(None, 12, 12, 192)	576	conv2d_128
batch_normalization_133 (BatchN [0][0]	(None, 12, 12, 192)	576	conv2d_133
batch_normalization_134 (BatchN [0][0]	(None, 12, 12, 192)	576	conv2d_134
activation_125 (Activation) lization_125[0][0]	(None, 12, 12, 192)	0	batch_norma

activation_128 (Activation) lization_128[0][0]	(None, 12, 12, 192) 0	batch_norma
activation_133 (Activation) lization_133[0][0]	(None, 12, 12, 192) 0	batch_norma
activation_134 (Activation) lization_134[0][0]	(None, 12, 12, 192) 0	batch_norma
mixed4 (Concatenate) 125[0][0]	(None, 12, 12, 768) 0	activation_
128[0][0]		activation_
133[0][0]		activation_
134[0][0]		activation_
conv2d_139 (Conv2D) [0]	(None, 12, 12, 160) 122880	mixed4[0]
batch_normalization_139 (BatchN) [0][0]	(None, 12, 12, 160) 480	conv2d_139
activation_139 (Activation) lization_139[0][0]	(None, 12, 12, 160) 0	batch_norma
conv2d_140 (Conv2D) 139[0][0]	(None, 12, 12, 160) 179200	activation_
batch_normalization_140 (BatchN) [0][0]	(None, 12, 12, 160) 480	conv2d_140
activation_140 (Activation) lization_140[0][0]	(None, 12, 12, 160) 0	batch_norma
conv2d_136 (Conv2D) [0]	(None, 12, 12, 160) 122880	mixed4[0]
conv2d_141 (Conv2D) 140[0][0]	(None, 12, 12, 160) 179200	activation_
batch_normalization_136 (BatchN) [0][0]	(None, 12, 12, 160) 480	conv2d_136
batch_normalization_141 (BatchN) [0][0]	(None, 12, 12, 160) 480	conv2d_141

[0][0]

activation_136 (Activation) lization_136[0][0]	(None, 12, 12, 160) 0	batch_norma
activation_141 (Activation) lization_141[0][0]	(None, 12, 12, 160) 0	batch_norma
conv2d_137 (Conv2D) 136[0][0]	(None, 12, 12, 160) 179200	activation_
conv2d_142 (Conv2D) 141[0][0]	(None, 12, 12, 160) 179200	activation_
batch_normalization_137 (BatchN [0][0])	(None, 12, 12, 160) 480	conv2d_137
batch_normalization_142 (BatchN [0][0])	(None, 12, 12, 160) 480	conv2d_142
activation_137 (Activation) lization_137[0][0]	(None, 12, 12, 160) 0	batch_norma
activation_142 (Activation) lization_142[0][0]	(None, 12, 12, 160) 0	batch_norma
average_pooling2d_14 (AveragePo [0])	(None, 12, 12, 768) 0	mixed4[0]
conv2d_135 (Conv2D) [0]	(None, 12, 12, 192) 147456	mixed4[0]
conv2d_138 (Conv2D) 137[0][0]	(None, 12, 12, 192) 215040	activation_
conv2d_143 (Conv2D) 142[0][0]	(None, 12, 12, 192) 215040	activation_
conv2d_144 (Conv2D) ling2d_14[0][0]	(None, 12, 12, 192) 147456	average_poo
batch_normalization_135 (BatchN [0][0])	(None, 12, 12, 192) 576	conv2d_135
batch_normalization_138 (BatchN [0][0])	(None, 12, 12, 192) 576	conv2d_138

batch_normalization_143 (BatchN (None, 12, 12, 192) 576 [0][0]		conv2d_143
batch_normalization_144 (BatchN (None, 12, 12, 192) 576 [0][0]		conv2d_144
activation_135 (Activation) (None, 12, 12, 192) 0 batch_normalization_135[0][0]		batch_norma
activation_138 (Activation) (None, 12, 12, 192) 0 batch_normalization_138[0][0]		batch_norma
activation_143 (Activation) (None, 12, 12, 192) 0 batch_normalization_143[0][0]		batch_norma
activation_144 (Activation) (None, 12, 12, 192) 0 batch_normalization_144[0][0]		batch_norma
mixed5 (Concatenate) (None, 12, 12, 768) 0 activation_135[0][0]		activation_
activation_138[0][0]		activation_
activation_143[0][0]		activation_
activation_144[0][0]		activation_
conv2d_149 (Conv2D) (None, 12, 12, 160) 122880 mixed5[0][0]		mixed5[0]
batch_normalization_149 (BatchN (None, 12, 12, 160) 480 [0][0]		conv2d_149
activation_149 (Activation) (None, 12, 12, 160) 0 batch_normalization_149[0][0]		batch_norma
conv2d_150 (Conv2D) (None, 12, 12, 160) 179200 activation_149[0][0]		activation_
batch_normalization_150 (BatchN (None, 12, 12, 160) 480 [0][0]		conv2d_150
activation_150 (Activation) (None, 12, 12, 160) 0 batch_normalization_150[0][0]		batch_norma
conv2d_146 (Conv2D) (None, 12, 12, 160) 122880 mixed5[0]		mixed5[0]

[0]

conv2d_151 (Conv2D) 150[0][0]	(None, 12, 12, 160)	179200	activation_
batch_normalization_146 (BatchN [0][0]	(None, 12, 12, 160)	480	conv2d_146
batch_normalization_151 (BatchN [0][0]	(None, 12, 12, 160)	480	conv2d_151
activation_146 (Activation) lization_146[0][0]	(None, 12, 12, 160)	0	batch_norma
activation_151 (Activation) lization_151[0][0]	(None, 12, 12, 160)	0	batch_norma
conv2d_147 (Conv2D) 146[0][0]	(None, 12, 12, 160)	179200	activation_
conv2d_152 (Conv2D) 151[0][0]	(None, 12, 12, 160)	179200	activation_
batch_normalization_147 (BatchN [0][0]	(None, 12, 12, 160)	480	conv2d_147
batch_normalization_152 (BatchN [0][0]	(None, 12, 12, 160)	480	conv2d_152
activation_147 (Activation) lization_147[0][0]	(None, 12, 12, 160)	0	batch_norma
activation_152 (Activation) lization_152[0][0]	(None, 12, 12, 160)	0	batch_norma
average_pooling2d_15 (AveragePo [0]	(None, 12, 12, 768)	0	mixed5[0]
conv2d_145 (Conv2D) [0]	(None, 12, 12, 192)	147456	mixed5[0]
conv2d_148 (Conv2D) 147[0][0]	(None, 12, 12, 192)	215040	activation_
conv2d_153 (Conv2D) 152[0][0]	(None, 12, 12, 192)	215040	activation_

conv2d_154 (Conv2D) ling2d_15[0][0]	(None, 12, 12, 192) 147456	average_poo
batch_normalization_145 (BatchN [0][0])	(None, 12, 12, 192) 576	conv2d_145
batch_normalization_148 (BatchN [0][0])	(None, 12, 12, 192) 576	conv2d_148
batch_normalization_153 (BatchN [0][0])	(None, 12, 12, 192) 576	conv2d_153
batch_normalization_154 (BatchN [0][0])	(None, 12, 12, 192) 576	conv2d_154
activation_145 (Activation) lization_145[0][0]	(None, 12, 12, 192) 0	batch_norma
activation_148 (Activation) lization_148[0][0]	(None, 12, 12, 192) 0	batch_norma
activation_153 (Activation) lization_153[0][0]	(None, 12, 12, 192) 0	batch_norma
activation_154 (Activation) lization_154[0][0]	(None, 12, 12, 192) 0	batch_norma
mixed6 (Concatenate) 145[0][0]	(None, 12, 12, 768) 0	activation_
148[0][0]		activation_
153[0][0]		activation_
154[0][0]		activation_
conv2d_159 (Conv2D) [0]	(None, 12, 12, 192) 147456	mixed6[0]
batch_normalization_159 (BatchN [0][0])	(None, 12, 12, 192) 576	conv2d_159
activation_159 (Activation) lization_159[0][0]	(None, 12, 12, 192) 0	batch_norma
conv2d_160 (Conv2D)	(None, 12, 12, 192) 258048	activation_

159[0][0]

batch_normalization_160 (BatchN (None, 12, 12, 192) 576 [0][0]		conv2d_160
activation_160 (Activation) (None, 12, 12, 192) 0 batch_normalization_160[0][0]		
conv2d_156 (Conv2D) (None, 12, 12, 192) 147456 [0]		mixed6[0]
conv2d_161 (Conv2D) (None, 12, 12, 192) 258048 160[0][0]		activation_
batch_normalization_156 (BatchN (None, 12, 12, 192) 576 [0][0]		conv2d_156
batch_normalization_161 (BatchN (None, 12, 12, 192) 576 [0][0]		conv2d_161
activation_156 (Activation) (None, 12, 12, 192) 0 batch_normalization_156[0][0]		
activation_161 (Activation) (None, 12, 12, 192) 0 batch_normalization_161[0][0]		
conv2d_157 (Conv2D) (None, 12, 12, 192) 258048 156[0][0]		activation_
conv2d_162 (Conv2D) (None, 12, 12, 192) 258048 161[0][0]		activation_
batch_normalization_157 (BatchN (None, 12, 12, 192) 576 [0][0]		conv2d_157
batch_normalization_162 (BatchN (None, 12, 12, 192) 576 [0][0]		conv2d_162
activation_157 (Activation) (None, 12, 12, 192) 0 batch_normalization_157[0][0]		
activation_162 (Activation) (None, 12, 12, 192) 0 batch_normalization_162[0][0]		
average_pooling2d_16 (AveragePo (None, 12, 12, 768) 0 [0]		mixed6[0]

conv2d_155 (Conv2D) [0]	(None, 12, 12, 192)	147456	mixed6[0]
conv2d_158 (Conv2D) 157[0][0]	(None, 12, 12, 192)	258048	activation_
conv2d_163 (Conv2D) 162[0][0]	(None, 12, 12, 192)	258048	activation_
conv2d_164 (Conv2D) 162[0][0]	(None, 12, 12, 192)	147456	average_poo
batch_normalization_155 (BatchN [0][0]	(None, 12, 12, 192)	576	conv2d_155
batch_normalization_158 (BatchN [0][0]	(None, 12, 12, 192)	576	conv2d_158
batch_normalization_163 (BatchN [0][0]	(None, 12, 12, 192)	576	conv2d_163
batch_normalization_164 (BatchN [0][0]	(None, 12, 12, 192)	576	conv2d_164
activation_155 (Activation) 155[0][0]	(None, 12, 12, 192)	0	batch_norma
activation_158 (Activation) 158[0][0]	(None, 12, 12, 192)	0	batch_norma
activation_163 (Activation) 163[0][0]	(None, 12, 12, 192)	0	batch_norma
activation_164 (Activation) 164[0][0]	(None, 12, 12, 192)	0	batch_norma
mixed7 (Concatenate) 155[0][0]	(None, 12, 12, 768)	0	activation_
158[0][0]			activation_
163[0][0]			activation_
164[0][0]			activation_
conv2d_167 (Conv2D)	(None, 12, 12, 192)	147456	mixed7[0]

[0]

batch_normalization_167 (BatchN (None, 12, 12, 192) [0][0]	576	conv2d_167
activation_167 (Activation) [0][0]	(None, 12, 12, 192) 0	batch_normalization_167[0][0]
conv2d_168 (Conv2D) [0][0]	(None, 12, 12, 192) 258048	activation_167[0][0]
batch_normalization_168 (BatchN (None, 12, 12, 192) [0][0]	576	conv2d_168
activation_168 (Activation) [0][0]	(None, 12, 12, 192) 0	batch_normalization_168[0][0]
conv2d_165 (Conv2D) [0]	(None, 12, 12, 192) 147456	mixed7[0]
conv2d_169 (Conv2D) [0][0]	(None, 12, 12, 192) 258048	activation_168[0][0]
batch_normalization_165 (BatchN (None, 12, 12, 192) [0][0]	576	conv2d_165
batch_normalization_169 (BatchN (None, 12, 12, 192) [0][0]	576	conv2d_169
activation_165 (Activation) [0][0]	(None, 12, 12, 192) 0	batch_normalization_165[0][0]
activation_169 (Activation) [0][0]	(None, 12, 12, 192) 0	batch_normalization_169[0][0]
conv2d_166 (Conv2D) [0][0]	(None, 5, 5, 320)	552960 activation_165[0][0]
conv2d_170 (Conv2D) [0][0]	(None, 5, 5, 192)	331776 activation_169[0][0]
batch_normalization_166 (BatchN (None, 5, 5, 320) [0][0]	960	conv2d_166
batch_normalization_170 (BatchN (None, 5, 5, 192) [0][0]	576	conv2d_170

activation_166 (Activation) lization_166[0][0]	(None, 5, 5, 320)	0	batch_norma
activation_170 (Activation) lization_170[0][0]	(None, 5, 5, 192)	0	batch_norma
max_pooling2d_8 (MaxPooling2D) [0]	(None, 5, 5, 768)	0	mixed7[0]
mixed8 (Concatenate) 166[0][0]	(None, 5, 5, 1280)	0	activation_
170[0][0]			activation_
2d_8[0][0]			max_pooling
conv2d_175 (Conv2D) [0]	(None, 5, 5, 448)	573440	mixed8[0]
batch_normalization_175 (BatchN [0][0]	(None, 5, 5, 448)	1344	conv2d_175
activation_175 (Activation) lization_175[0][0]	(None, 5, 5, 448)	0	batch_norma
conv2d_172 (Conv2D) [0]	(None, 5, 5, 384)	491520	mixed8[0]
conv2d_176 (Conv2D) 175[0][0]	(None, 5, 5, 384)	1548288	activation_
batch_normalization_172 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_172
batch_normalization_176 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_176
activation_172 (Activation) lization_172[0][0]	(None, 5, 5, 384)	0	batch_norma
activation_176 (Activation) lization_176[0][0]	(None, 5, 5, 384)	0	batch_norma
conv2d_173 (Conv2D) 172[0][0]	(None, 5, 5, 384)	442368	activation_

conv2d_174 (Conv2D) 172[0][0]	(None, 5, 5, 384)	442368	activation_
conv2d_177 (Conv2D) 176[0][0]	(None, 5, 5, 384)	442368	activation_
conv2d_178 (Conv2D) 176[0][0]	(None, 5, 5, 384)	442368	activation_
average_pooling2d_17 (AveragePo [0]	(None, 5, 5, 1280)	0	mixed8[0]
conv2d_171 (Conv2D) [0]	(None, 5, 5, 320)	409600	mixed8[0]
batch_normalization_173 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_173
batch_normalization_174 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_174
batch_normalization_177 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_177
batch_normalization_178 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_178
conv2d_179 (Conv2D) 172[0][0]	(None, 5, 5, 192)	245760	average_poo ling2d_17[0][0]
batch_normalization_171 (BatchN [0][0]	(None, 5, 5, 320)	960	conv2d_171
activation_173 (Activation) 173[0][0]	(None, 5, 5, 384)	0	batch_norma lization_173[0][0]
activation_174 (Activation) 174[0][0]	(None, 5, 5, 384)	0	batch_norma lization_174[0][0]
activation_177 (Activation) 177[0][0]	(None, 5, 5, 384)	0	batch_norma lization_177[0][0]
activation_178 (Activation) 178[0][0]	(None, 5, 5, 384)	0	batch_norma lization_178[0][0]

batch_normalization_179 (BatchN (None, 5, 5, 192) [0][0]	576	conv2d_179
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activation_171 (Activation) [0][0]	(None, 5, 5, 320)	0	batch_norma
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mixed9_0 (Concatenate) [0][0]	(None, 5, 5, 768)	0	activation_
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activation_174[0][0]			activation_
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concatenate_3 (Concatenate) [0][0]	(None, 5, 5, 768)	0	activation_
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activation_178[0][0]			activation_
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activation_179 (Activation) [0][0]	(None, 5, 5, 192)	0	batch_norma
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mixed9 (Concatenate) [0][0]	(None, 5, 5, 2048)	0	activation_
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mixed9_0[0]			mixed9_0[0]
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concatenate_3[0][0]			concatenate
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activation_179[0][0]			activation_
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conv2d_184 (Conv2D) [0]	(None, 5, 5, 448)	917504	mixed9[0]
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batch_normalization_184 (BatchN (None, 5, 5, 448) [0][0]	1344	conv2d_184
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activation_184 (Activation) [0][0]	(None, 5, 5, 448)	0	batch_norma
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conv2d_181 (Conv2D) [0]	(None, 5, 5, 384)	786432	mixed9[0]
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conv2d_185 (Conv2D) [0][0]	(None, 5, 5, 384)	1548288	activation_
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batch_normalization_181 (BatchN (None, 5, 5, 384) [0][0]	1152	conv2d_181
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batch_normalization_185 (BatchN (None, 5, 5, 384) [0][0]	1152	conv2d_185
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activation_181 (Activation) lization_181[0][0]	(None, 5, 5, 384)	0	batch_norma
activation_185 (Activation) lization_185[0][0]	(None, 5, 5, 384)	0	batch_norma
conv2d_182 (Conv2D) 181[0][0]	(None, 5, 5, 384)	442368	activation_
conv2d_183 (Conv2D) 181[0][0]	(None, 5, 5, 384)	442368	activation_
conv2d_186 (Conv2D) 185[0][0]	(None, 5, 5, 384)	442368	activation_
conv2d_187 (Conv2D) 185[0][0]	(None, 5, 5, 384)	442368	activation_
average_pooling2d_18 (AveragePo [0]	(None, 5, 5, 2048)	0	mixed9[0]
conv2d_180 (Conv2D) [0]	(None, 5, 5, 320)	655360	mixed9[0]
batch_normalization_182 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_182
batch_normalization_183 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_183
batch_normalization_186 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_186
batch_normalization_187 (BatchN [0][0]	(None, 5, 5, 384)	1152	conv2d_187
conv2d_188 (Conv2D) ling2d_18[0][0]	(None, 5, 5, 192)	393216	average_poo
batch_normalization_180 (BatchN [0][0]	(None, 5, 5, 320)	960	conv2d_180
activation_182 (Activation) lization_182[0][0]	(None, 5, 5, 384)	0	batch_norma

activation_183 (Activation) lization_183[0][0]	(None, 5, 5, 384)	0	batch_norma
activation_186 (Activation) lization_186[0][0]	(None, 5, 5, 384)	0	batch_norma
activation_187 (Activation) lization_187[0][0]	(None, 5, 5, 384)	0	batch_norma
batch_normalization_188 (BatchN [0][0]	(None, 5, 5, 192)	576	conv2d_188
activation_180 (Activation) lization_180[0][0]	(None, 5, 5, 320)	0	batch_norma
mixed9_1 (Concatenate) 182[0][0]	(None, 5, 5, 768)	0	activation_
activation_183[0][0]			activation_
concatenate_4 (Concatenate) 186[0][0]	(None, 5, 5, 768)	0	activation_
activation_187[0][0]			activation_
activation_188 (Activation) lization_188[0][0]	(None, 5, 5, 192)	0	batch_norma
mixed10 (Concatenate) 180[0][0]	(None, 5, 5, 2048)	0	activation_
[0]			mixed9_1[0]
concatenate_4[0][0]			concatenate
activation_188[0][0]			activation_
global_average_pooling2d_2 (Glo [0]	(None, 2048)	0	mixed10[0]
dense_3 (Dense) age_pooling2d_2[0][0]	(None, 256)	524544	global_aver
dense_4 (Dense) [0]	(None, 1)	257	dense_3[0]
<hr/>			
<hr/>			
Total params: 22,327,585			
Trainable params: 524,801			
Non-trainable params: 21,802,784			

---

GPU Memory:1.164 GBRAM:2.328

In [24]:

```
history = trainModel(model,
                      1,
                      "Re training on our data with frozen inception layers", "5 lower.h5")
```

Epoch 1/1  
800/800 [=====] - 242s 302ms/step - loss: 7.9712 -  
acc: 0.5000 - val\_loss: 7.9712 - val\_acc: 0.5000

Epoch 00001: val\_loss improved from inf to 7.97119, saving model to 5 lower.  
h5

Timing:: took 4 minutes Re training on our data with frozen inception layers

---

In [26]:

```
# building the rest of the model

# here we choose to retrain the top 2 inception blocks
for layer in model.layers[:249]:
    layer.trainable = False
for layer in model.layers[249:]:
    layer.trainable = True

# we now recompile the model for our layer modifications to take effect
from keras.optimizers import SGD
model.compile(optimizer='adagrad',
              loss='binary_crossentropy',
              metrics=['accuracy'])

history_2 = trainModel(model,
                       1,
                       "Re training the full inception model",
                       '5 upper.h5')
```

Epoch 1/1  
800/800 [=====] - 316s 395ms/step - loss: 7.9712 -  
acc: 0.5000 - val\_loss: 7.9712 - val\_acc: 0.5000

Epoch 00001: val\_loss improved from inf to 7.97119, saving model to 5 upper.  
h5

Timing:: took 5 minutes Re training the full inception model

---

In [27]:

```
def plot(network_history):
    plt.figure()
    plt.xlabel('Epochs')
    plt.ylabel('Loss')
    plt.plot(network_history.history['loss'])
    plt.plot(network_history.history['val_loss'])
    plt.legend(['Training', 'Validation'])

    plt.figure()
    plt.xlabel('Epochs')
    plt.ylabel('Accuracy')
    plt.plot(network_history.history['acc'])
    plt.plot(network_history.history['val_acc'])
    plt.legend(['Training', 'Validation'], loc='lower right')
```

## 2.3 Evaluating the CNNs performance

### Evaluating the lower frozen CNN performance

In [28]:

```
# Get test data
x_test, y_test = data.getTestData()
print('Test data shape: ', x_test.shape)
print('Test labels shape: ', y_test.shape)
```

Test data shape: (63, 224, 224, 3)

Test labels shape: (63, 1)

In [29]:

```
x_test = x_test/255
```

In [30]:

```
# Load and evaluate best model
from keras.models import load_model
best_model = load_model('5_lower.h5')
best_model.summary()
```

```
average_pooling2d_14 (AveragePooling2D)    (None, 12, 12, 768)    0          mixed4[0]
[0]
```

```
conv2d_135 (Conv2D)                      (None, 12, 12, 192)   147456     mixed4[0]
[0]
```

```
conv2d_138 (Conv2D)                      (None, 12, 12, 192)   215040     activation
n_137[0][0]
```

```
conv2d_143 (Conv2D)                      (None, 12, 12, 192)   215040     activation
n_142[0][0]
```

```
conv2d_144 (Conv2D)                      (None, 12, 12, 192)   147456     average_p
ooling2d_14[0][0]
```

In [31]:

```
best_model.predict(x_test)
```

Out[31]:

```
[ 1.],
[ 1.],
[ 1.],
[ 1.],
[ 1.],
[ 1.],
[ 1.],
[ 1.]], dtype=float32)
```

In [32]:

```
# Evaluate all
from sklearn.metrics import confusion_matrix
expected = y_test
prediction = best_model.predict(x_test)
x, y, fn, tp = confusion_matrix(y_test, prediction).ravel()
print(fn, tp)
```

0 21

The model only predicts malignant cases. This means the models performance is poor. In this case transfer learning has not been of much help because of the difference between the natural images that were used to train the Inception model and very different from breast ultrasounds but also, our dataset is not sufficiently large for newly learned features to significantly affect the hybrid models output hence the poor performance.

## Evaluating the upper retrained CNN performance

In [34]:

```
# Load and evaluate best model
from keras.models import load_model
best_model = load_model('5_upper.h5')
best_model.summary()
```

Layer (type)	Output Shape	Param #	Connected to
<hr/>			
<hr/>			
input_2 (InputLayer)	(None, 224, 224, 3)	0	
<hr/>			
conv2d_95 (Conv2D)	(None, 111, 111, 32)	864	input_2[0][0]
<hr/>			
batch_normalization_95 (BatchNormalizat	(None, 111, 111, 32)	96	conv2d_95[0][0]
<hr/>			
activation_95 (Activation)	(None, 111, 111, 32)	0	batch_norm[0][0]

In [35]:

```
best_model.predict(x_test)
```

Out[35]:

```
[ 1.],  
[ 1.],  
[ 1.],  
[ 1.],  
[ 1.],  
[ 1.],  
[ 1.],  
[ 1.],  
[ 1.]], dtype=float32)
```

In [36]:

```
expected = y_test  
prediction = best_model.predict(x_test)  
x, y, fn, tp = confusion_matrix(y_test, prediction).ravel()  
print(fn, tp)
```

0 21

The model only predicts malignant cases. This means the models performance is poor. In this case transfer learning has not been of much help because of the difference between the natural images that were used to train the Inception model and very different from breast ultrasounds but also, our dataset is not sufficiently large for newly learned features to significantly affect the hybrid models output hence the poor performance.

## References for images and some content:

- [1] [https://adeshpande3.github.io/adeshpande3.github.io/\(\)](https://adeshpande3.github.io/adeshpande3.github.io/)
- [2] "Neural Networks and Deep Learning" (<http://neuralnetworksanddeeplearning.com/>) by Michael Nielsen.
- [3] Deep learning with TensorFlow and Keras by Valerio Maggio

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