Introduction To GradCAM (Part 1) - Lecture Notebook

In this lecture notebook we'll be looking at an introduction to Grad-CAM, a powerful technique for interpreting Convolutional Neural Networks. Grad-CAM stands for Gradient-weighted Class Activation Mapping.

CNN's are very flexible models and their great predictive power comes at the cost of losing interpretability (something that is true for all Artificial Neural Networks). Grad-CAM attempts to solve this by giving us a graphical visualisation of parts of an image that are the most relevant for the CNN when predicting a particular class.

Aside from working on some Grad-CAM concepts we'll also look at how we can use Keras to access some concrete information of our model. Let's dive into it!

```
In [1]: import keras
from keras import backend as K
from util import *
```

Using TensorFlow backend.

The load_C3M3_model() function has been taken care of and its internals are out of the scope of this notebook. But if it intrigues you, you can take a look at it in util.py

```
In [2]: # Load the model we are going to be using
model = load_C3M3_model()
```

Got loss weights Loaded DenseNet Added layers Compiled Model Loaded Weights

As you may already know, we can check the architecture of our model using the summary () method.

After running the code block below we'll see that this model has a lot of layers. One advantage of Grad-CAM over previous attempts of interpreting CNN's (such as CAM) is that it is architecture agnostic. This means it can be used for CNN's with complex architectures such as this one:

```
In [3]: # Print all of the model's layers
model.summary()
```

Layer (type) onnected to	Output	_			Param #	C
input_1 (InputLayer)	= (None,					
<pre>zero_padding2d_1 (ZeroPadding2D nput_1[0][0]</pre>	(None,	None,	None,	3	0	i
conv1/conv (Conv2D) ero_padding2d_1[0][0]	(None,	None,	None,	6	9408	z
<pre>conv1/bn (BatchNormalization) onv1/conv[0][0]</pre>	(None,	None,	None,	6	256	c
<pre>conv1/relu (Activation) onv1/bn[0][0]</pre>	_ (None,	None,	None,	6	0	С
<pre>zero_padding2d_2 (ZeroPadding2D onv1/relu[0][0]</pre>	(None,	None,	None,	6	0	С
<pre>pool1 (MaxPooling2D) ero_padding2d_2[0][0]</pre>	(None,	None,	None,	6	0	z
<pre>conv2_block1_0_bn (BatchNormali ool1[0][0]</pre>	(None,	None,	None,	6	256	р
conv2_block1_0_relu (Activation onv2_block1_0_bn[0][0]	(None,	None,	None,	6	0	c
conv2_block1_1_conv (Conv2D) onv2_block1_0_relu[0][0]	(None,	None,	None,	1	8192	c
conv2_block1_1_bn (BatchNormali onv2_block1_1_conv[0][0]	(None,	None,	None,	1	512	С
conv2_block1_1_relu (Activation onv2_block1_1_bn[0][0]	(None,	None,	None,	1	0	С
conv2_block1_2_conv (Conv2D)	(None,	None,	None,	3	36864	С

onv2_block1_1_relu[0][0]

<pre>conv2_block1_concat (Concatenat ool1[0][0]</pre>	(None,	None,	None,	9	0	р
onv2_block1_2_conv[0][0]						С
<pre>conv2_block2_0_bn (BatchNormali onv2_block1_concat[0][0]</pre>	(None,	None,	None,	9	384	С
conv2_block2_0_relu (Activation onv2_block2_0_bn[0][0]	(None,	None,	None,	9	0	С
conv2_block2_1_conv (Conv2D) onv2_block2_0_relu[0][0]	(None,	None,	None,	1	12288	c
conv2_block2_1_bn (BatchNormali onv2_block2_1_conv[0][0]	(None,	None,	None,	1	512	c
conv2_block2_1_relu (Activation onv2_block2_1_bn[0][0]	(None,	None,	None,	1	0	С
<pre>conv2_block2_2_conv (Conv2D) onv2_block2_1_relu[0][0]</pre>	(None,	None,	None,	3	36864	С
conv2_block2_concat (Concatenat onv2_block1_concat[0][0]	(None,	None,	None,	1	0	c
onv2_block2_2_conv[0][0]						_
<pre>conv2_block3_0_bn (BatchNormali onv2_block2_concat[0][0]</pre>	(None,	None,	None,	1	512	С
<pre>conv2_block3_0_relu (Activation onv2_block3_0_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv2_block3_1_conv (Conv2D) onv2_block3_0_relu[0][0]	(None,	None,	None,	1	16384	С
<pre>conv2_block3_1_bn (BatchNormali onv2_block3_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С

(None,	None,	None,	1	0	С
(None,	None,	None,	3	36864	С
(None,	None,	None,	1	0	c
(None,	None,	None,	1	640	С
(None,	None,	None,	1	0	С
(None,	None,	None,	1	20480	С
(None,	None,	None,	1	512	С
(None,	None,	None,	1	0	С
(None,	None,	None,	3	36864	С
(None,	None,	None,	1	0	С
					С
(None,	None,	None,	1	768	С
(None,	None,	None,	1	0	С
	(None, (None, (None, (None, (None, (None, (None, (None, (None, (None,	(None, None, (None, None,	(None, None, None, (None, None, None,	(None, None, None, 3 (None, None, None, 1 (None, None, None, 1	(None, None, None, 1 0 (None, None, None, 3 36864 (None, None, None, 1 0 (None, None, None, 1 0 (None, None, None, 1 20480 (None, None, None, 1 512 (None, None, None, 1 0 (None, None, None, 1 768 (None, None, None, 1 768

<pre>conv2_block5_1_conv (Conv2D) onv2_block5_0_relu[0][0]</pre>	(None,	None,	None,	1	24576	С
conv2_block5_1_bn (BatchNormali onv2_block5_1_conv[0][0]	(None,	None,	None,	1	512	С
conv2_block5_1_relu (Activation onv2_block5_1_bn[0][0]	(None,	None,	None,	1	0	С
conv2_block5_2_conv (Conv2D) onv2_block5_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv2_block5_concat (Concatenat onv2_block4_concat[0][0]	(None,	None,	None,	2	0	С
onv2_block5_2_conv[0][0]						
conv2_block6_0_bn (BatchNormali onv2_block5_concat[0][0]	(None,	None,	None,	2	896	С
<pre>conv2_block6_0_relu (Activation onv2_block6_0_bn[0][0]</pre>	(None,	None,	None,	2	0	С
conv2_block6_1_conv (Conv2D) onv2_block6_0_relu[0][0]	(None,	None,	None,	1	28672	С
<pre>conv2_block6_1_bn (BatchNormali onv2_block6_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
<pre>conv2_block6_1_relu (Activation onv2_block6_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv2_block6_2_conv (Conv2D) onv2_block6_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv2_block6_concat (Concatenat onv2_block5_concat[0][0]	(None,	None,	None,	2	0	c
onv2_block6_2_conv[0][0]						-

<pre>pool2_bn (BatchNormalization) onv2_block6_concat[0][0]</pre>	(None,	None,	None,	2	1024	С
<pre>pool2_relu (Activation) ool2_bn[0][0]</pre>	(None,	None,	None,	2	0	р
pool2_conv (Conv2D) ool2_relu[0][0]	(None,	None,	None,	1	32768	р
<pre>pool2_pool (AveragePooling2D) ool2_conv[0][0]</pre>	(None,	None,	None,	1	0	р
<pre>conv3_block1_0_bn (BatchNormali ool2_pool[0][0]</pre>	(None,	None,	None,	1	512	р
<pre>conv3_block1_0_relu (Activation onv3_block1_0_bn[0][0]</pre>	_ (None,	None,	None,	1	0	С
conv3_block1_1_conv (Conv2D) onv3_block1_0_relu[0][0]	(None,	None,	None,	1	16384	С
<pre>conv3_block1_1_bn (BatchNormali onv3_block1_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
<pre>conv3_block1_1_relu (Activation onv3_block1_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
<pre>conv3_block1_2_conv (Conv2D) onv3_block1_1_relu[0][0]</pre>	(None,	None,	None,	3	36864	С
<pre>conv3_block1_concat (Concatenat ool2_pool[0][0]</pre>	(None,	None,	None,	1	0	p
onv3_block1_2_conv[0][0]						С
<pre>conv3_block2_0_bn (BatchNormali onv3_block1_concat[0][0]</pre>	(None,	None,	None,	1	640	С
conv3_block2_0_relu (Activation onv3_block2_0_bn[0][0]	(None,	None,	None,	1	0	С

conv3_block2_1_conv (Conv2D) onv3_block2_0_relu[0][0]	(None,	None,	None,	1	20480	С
conv3_block2_1_bn (BatchNormali onv3_block2_1_conv[0][0]	(None,	None,	None,	1	512	С
conv3_block2_1_relu (Activation onv3_block2_1_bn[0][0]	(None,	None,	None,	1	0	С
conv3_block2_2_conv (Conv2D) onv3_block2_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv3_block2_concat (Concatenat onv3_block1_concat[0][0]</pre>	(None,	None,	None,	1	0	c
onv3_block2_2_conv[0][0]						
conv3_block3_0_bn (BatchNormali onv3_block2_concat[0][0]	(None,	None,	None,	1	768	С
conv3_block3_0_relu (Activation onv3_block3_0_bn[0][0]	(None,	None,	None,	1	0	c
conv3_block3_1_conv (Conv2D) onv3_block3_0_relu[0][0]	(None,	None,	None,	1	24576	С
conv3_block3_1_bn (BatchNormali onv3_block3_1_conv[0][0]	(None,	None,	None,	1	512	С
conv3_block3_1_relu (Activation onv3_block3_1_bn[0][0]	(None,	None,	None,	1	0	С
conv3_block3_2_conv (Conv2D) onv3_block3_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv3_block3_concat (Concatenat onv3_block2_concat[0][0]	(None,	None,	None,	2	0	С
onv3_block3_2_conv[0][0]						•
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<pre>conv3_block4_0_bn (BatchNormali onv3_block3_concat[0][0]</pre>	(None,	None,	None,	2	896	С
conv3_block4_0_relu (Activation onv3_block4_0_bn[0][0]	(None,	None,	None,	2	0	С
conv3_block4_1_conv (Conv2D) onv3_block4_0_relu[0][0]	(None,	None,	None,	1	28672	С
conv3_block4_1_bn (BatchNormali onv3_block4_1_conv[0][0]	(None,	None,	None,	1	512	С
conv3_block4_1_relu (Activation onv3_block4_1_bn[0][0]	(None,	None,	None,	1	0	С
conv3_block4_2_conv (Conv2D) onv3_block4_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv3_block4_concat (Concatenat onv3_block3_concat[0][0]	_ (None,	None,	None,	2	0	С
onv3_block4_2_conv[0][0]						
<pre>conv3_block5_0_bn (BatchNormali onv3_block4_concat[0][0]</pre>	(None,	None,	None,	2	1024	С
<pre>conv3_block5_0_relu (Activation onv3_block5_0_bn[0][0]</pre>	(None,	None,	None,	2	0	С
conv3_block5_1_conv (Conv2D) onv3_block5_0_relu[0][0]	(None,	None,	None,	1	32768	С
conv3_block5_1_bn (BatchNormali onv3_block5_1_conv[0][0]	(None,	None,	None,	1	512	С
conv3_block5_1_relu (Activation onv3_block5_1_bn[0][0]	(None,	None,	None,	1	0	С
conv3_block5_2_conv (Conv2D) onv3_block5_1_relu[0][0]	(None,	None,	None,	3	36864	С

conv3_block5_concat (Concatenat onv3_block4_concat[0][0]	(None,	None,	None,	2	0	С
onv3_block5_2_conv[0][0]						C
conv3_block6_0_bn (BatchNormalionv3_block5_concat[0][0]	(None,	None,	None,	2	1152	С
conv3_block6_0_relu (Activation onv3_block6_0_bn[0][0]	(None,	None,	None,	2	0	c
conv3_block6_1_conv (Conv2D) onv3_block6_0_relu[0][0]	(None,	None,	None,	1	36864	c
conv3_block6_1_bn (BatchNormalionv3_block6_1_conv[0][0]	(None,	None,	None,	1	512	c
conv3_block6_1_relu (Activation onv3_block6_1_bn[0][0]	(None,	None,	None,	1	0	c
conv3_block6_2_conv (Conv2D) onv3_block6_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv3_block6_concat (Concatenat onv3_block5_concat[0][0]</pre>	(None,	None,	None,	3	0	c
onv3_block6_2_conv[0][0]						
<pre>conv3_block7_0_bn (BatchNormali onv3_block6_concat[0][0]</pre>	(None,	None,	None,	3	1280	С
conv3_block7_0_relu (Activation onv3_block7_0_bn[0][0]	(None,	None,	None,	3	0	С
conv3_block7_1_conv (Conv2D) onv3_block7_0_relu[0][0]	(None,	None,	None,	1	40960	С
conv3_block7_1_bn (BatchNormali onv3_block7_1_conv[0][0]	(None,	None,	None,	1	512	С
	_					

<pre>conv3_block7_1_relu (Activation onv3_block7_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv3_block7_2_conv (Conv2D) onv3_block7_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv3_block7_concat (Concatenat onv3_block6_concat[0][0]	(None,	None,	None,	3	0	c
onv3_block7_2_conv[0][0]						С
<pre>conv3_block8_0_bn (BatchNormali onv3_block7_concat[0][0]</pre>	(None,	None,	None,	3	1408	С
<pre>conv3_block8_0_relu (Activation onv3_block8_0_bn[0][0]</pre>	(None,	None,	None,	3	0	С
conv3_block8_1_conv (Conv2D) onv3_block8_0_relu[0][0]	(None,	None,	None,	1	45056	С
conv3_block8_1_bn (BatchNormali onv3_block8_1_conv[0][0]	(None,	None,	None,	1	512	С
conv3_block8_1_relu (Activation onv3_block8_1_bn[0][0]	(None,	None,	None,	1	0	c
conv3_block8_2_conv (Conv2D) onv3_block8_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv3_block8_concat (Concatenat onv3_block7_concat[0][0]	(None,	None,	None,	3	0	c
onv3_block8_2_conv[0][0]						
<pre>conv3_block9_0_bn (BatchNormali onv3_block8_concat[0][0]</pre>	(None,	None,	None,	3	1536	С
conv3_block9_0_relu (Activation onv3_block9_0_bn[0][0]	(None,	None,	None,	3	0	С
conv3_block9_1_conv (Conv2D)	(None,	None,	None,	1	49152	c

onv3_block9	0	relu	[0]	[0]
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conv3_block9_1_bn (BatchNormali onv3_block9_1_conv[0][0]	(None,	None,	None,	1	512	С
<pre>conv3_block9_1_relu (Activation onv3_block9_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv3_block9_2_conv (Conv2D) onv3_block9_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv3_block9_concat (Concatenat onv3_block8_concat[0][0]</pre>	(None,	None,	None,	4	0	c
onv3_block9_2_conv[0][0]						
conv3_block10_0_bn (BatchNormal onv3_block9_concat[0][0]	(None,	None,	None,	4	1664	С
<pre>conv3_block10_0_relu (Activatio onv3_block10_0_bn[0][0]</pre>	(None,	None,	None,	4	0	С
conv3_block10_1_conv (Conv2D) onv3_block10_0_relu[0][0]	(None,	None,	None,	1	53248	С
conv3_block10_1_bn (BatchNormal onv3_block10_1_conv[0][0]	(None,	None,	None,	1	512	С
<pre>conv3_block10_1_relu (Activatio onv3_block10_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv3_block10_2_conv (Conv2D) onv3_block10_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv3_block10_concat (Concatena onv3_block9_concat[0][0]	(None,	None,	None,	4	0	c
onv3_block10_2_conv[0][0]						С
<pre>conv3_block11_0_bn (BatchNormal onv3_block10_concat[0][0]</pre>	(None,	None,	None,	4	1792	c

conv3_block11_0_relu (Activatio onv3_block11_0_bn[0][0]	(None,	None,	None,	4	0	С
conv3_block11_1_conv (Conv2D) onv3_block11_0_relu[0][0]	(None,	None,	None,	1	57344	С
conv3_block11_1_bn (BatchNormal onv3_block11_1_conv[0][0]	(None,	None,	None,	1	512	С
conv3_block11_1_relu (Activatio onv3_block11_1_bn[0][0]	(None,	None,	None,	1	0	С
conv3_block11_2_conv (Conv2D) onv3_block11_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv3_block11_concat (Concatena onv3_block10_concat[0][0]	(None,	None,	None,	4	0	c
onv3_block11_2_conv[0][0]						С
<pre>conv3_block12_0_bn (BatchNormal onv3_block11_concat[0][0]</pre>	(None,	None,	None,	4	1920	С
conv3_block12_0_relu (Activatio onv3_block12_0_bn[0][0]	(None,	None,	None,	4	0	С
conv3_block12_1_conv (Conv2D) onv3_block12_0_relu[0][0]	(None,	None,	None,	1	61440	С
conv3_block12_1_bn (BatchNormal onv3_block12_1_conv[0][0]	(None,	None,	None,	1	512	С
conv3_block12_1_relu (Activatio onv3_block12_1_bn[0][0]	(None,	None,	None,	1	0	С
conv3_block12_2_conv (Conv2D) onv3_block12_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv3_block12_concat (Concatena	(None,	None,	None,	5	0	c

onv3_block11_concat[0][0]						-
onv3_block12_2_conv[0][0]						С
pool3_bn (BatchNormalization) onv3_block12_concat[0][0]	(None,	None,	None,	5	2048	c
<pre>pool3_relu (Activation) ool3_bn[0][0]</pre>	(None,	None,	None,	5	0	p
pool3_conv (Conv2D) ool3_relu[0][0]	(None,	None,	None,	2	131072	p
<pre>pool3_pool (AveragePooling2D) ool3_conv[0][0]</pre>	(None,	None,	None,	2	0	р
<pre>conv4_block1_0_bn (BatchNormali ool3_pool[0][0]</pre>	(None,	None,	None,	2	1024	р
<pre>conv4_block1_0_relu (Activation onv4_block1_0_bn[0][0]</pre>	(None,	None,	None,	2	0	С
conv4_block1_1_conv (Conv2D) onv4_block1_0_relu[0][0]	(None,	None,	None,	1	32768	С
<pre>conv4_block1_1_bn (BatchNormali onv4_block1_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
conv4_block1_1_relu (Activation onv4_block1_1_bn[0][0]	(None,	None,	None,	1	0	c
conv4_block1_2_conv (Conv2D) onv4_block1_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv4_block1_concat (Concatenat ool3_pool[0][0] onv4_block1_2_conv[0][0]</pre>	(None,	None,	None,	2	0	p
<pre>conv4_block2_0_bn (BatchNormali onv4_block1_concat[0][0]</pre>	(None,	None,	None,	2	1152	c

conv4_block2_0_relu (Activation onv4_block2_0_bn[0][0]	(None,	None,	None,	2	0	С
conv4_block2_1_conv (Conv2D) onv4_block2_0_relu[0][0]	(None,	None,	None,	1	36864	С
conv4_block2_1_bn (BatchNormalionv4_block2_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block2_1_relu (Activation onv4_block2_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block2_2_conv (Conv2D) onv4_block2_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block2_concat (Concatenat onv4_block1_concat[0][0]	(None,	None,	None,	3	0	c
onv4_block2_2_conv[0][0]						
<pre>conv4_block3_0_bn (BatchNormali onv4_block2_concat[0][0]</pre>	(None,	None,	None,	3	1280	С
<pre>conv4_block3_0_relu (Activation onv4_block3_0_bn[0][0]</pre>	(None,	None,	None,	3	0	С
conv4_block3_1_conv (Conv2D) onv4_block3_0_relu[0][0]	(None,	None,	None,	1	40960	С
conv4_block3_1_bn (BatchNormalionv4_block3_1_conv[0][0]	(None,	None,	None,	1	512	C
conv4_block3_1_relu (Activation onv4_block3_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block3_2_conv (Conv2D) onv4_block3_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block3_concat (Concatenat	(None,	None,	None,	3	0	c

<pre>onv4_block2_concat[0][0]</pre>						~
onv4_block3_2_conv[0][0]						с
conv4_block4_0_bn (BatchNormali onv4_block3_concat[0][0]	(None,	None,	None,	3	1408	c
conv4_block4_0_relu (Activation onv4_block4_0_bn[0][0]	(None,	None,	None,	3	0	c
conv4_block4_1_conv (Conv2D) onv4_block4_0_relu[0][0]	(None,	None,	None,	1	45056	С
conv4_block4_1_bn (BatchNormalionv4_block4_1_conv[0][0]	(None,	None,	None,	1	512	c
conv4_block4_1_relu (Activation onv4_block4_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block4_2_conv (Conv2D) onv4_block4_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv4_block4_concat (Concatenat onv4_block3_concat[0][0]</pre>	None,	None,	None,	3	0	c
onv4_block4_2_conv[0][0]						
conv4_block5_0_bn (BatchNormalionv4_block4_concat[0][0]	(None,	None,	None,	3	1536	С
conv4_block5_0_relu (Activation onv4_block5_0_bn[0][0]	(None,	None,	None,	3	0	c
conv4_block5_1_conv (Conv2D) onv4_block5_0_relu[0][0]	(None,	None,	None,	1	49152	С
conv4_block5_1_bn (BatchNormali onv4_block5_1_conv[0][0]	(None,	None,	None,	1	512	c
conv4_block5_1_relu (Activation onv4_block5_1_bn[0][0]	(None,	None,	None,	1	0	С

conv4_block5_2_conv (Conv2D) onv4_block5_1_relu[0][0]	_ (None,	None,	None,	3	36864	С
<pre>conv4_block5_concat (Concatenat onv4_block4_concat[0][0] onv4_block5_2_conv[0][0]</pre>	(None,	None,	None,	4	0	c
<pre>conv4_block6_0_bn (BatchNormali onv4_block5_concat[0][0]</pre>	(None,	None,	None,	4	1664	C
conv4_block6_0_relu (Activation onv4_block6_0_bn[0][0]	(None,	None,	None,	4	0	c
conv4_block6_1_conv (Conv2D) onv4_block6_0_relu[0][0]	(None,	None,	None,	1	53248	С
<pre>conv4_block6_1_bn (BatchNormali onv4_block6_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
<pre>conv4_block6_1_relu (Activation onv4_block6_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv4_block6_2_conv (Conv2D) onv4_block6_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv4_block6_concat (Concatenat onv4_block5_concat[0][0]</pre>	(None,	None,	None,	4	0	С
onv4_block6_2_conv[0][0]						
conv4_block7_0_bn (BatchNormali onv4_block6_concat[0][0]	(None,	None,	None,	4	1792	С
conv4_block7_0_relu (Activation onv4_block7_0_bn[0][0]	(None,	None,	None,	4	0	С
conv4_block7_1_conv (Conv2D) onv4_block7_0_relu[0][0]	(None,	None,	None,	1	57344	С

<pre>conv4_block7_1_bn (BatchNormali onv4_block7_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
conv4_block7_1_relu (Activation onv4_block7_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block7_2_conv (Conv2D) onv4_block7_1_relu[0][0]	(None,	None,	None,	3	36864	c
<pre>conv4_block7_concat (Concatenat onv4_block6_concat[0][0]</pre>	None,	None,	None,	4	0	С
onv4_block7_2_conv[0][0]						С
conv4_block8_0_bn (BatchNormalionv4_block7_concat[0][0]	(None,	None,	None,	4	1920	c
conv4_block8_0_relu (Activation onv4_block8_0_bn[0][0]	(None,	None,	None,	4	0	С
conv4_block8_1_conv (Conv2D) onv4_block8_0_relu[0][0]	(None,	None,	None,	1	61440	c
conv4_block8_1_bn (BatchNormali onv4_block8_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block8_1_relu (Activation onv4_block8_1_bn[0][0]	(None,	None,	None,	1	0	c
conv4_block8_2_conv (Conv2D) onv4_block8_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv4_block8_concat (Concatenat onv4_block7_concat[0][0]</pre>	(None,	None,	None,	5	0	С
onv4_block8_2_conv[0][0]						С
conv4_block9_0_bn (BatchNormali onv4_block8_concat[0][0]	(None,	None,	None,	5	2048	С
	_					

<pre>conv4_block9_0_relu (Activation onv4_block9_0_bn[0][0]</pre>	(None,	None,	None,	5	0	С
conv4_block9_1_conv (Conv2D) onv4_block9_0_relu[0][0]	(None,	None,	None,	1	65536	С
<pre>conv4_block9_1_bn (BatchNormali onv4_block9_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
<pre>conv4_block9_1_relu (Activation onv4_block9_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv4_block9_2_conv (Conv2D) onv4_block9_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv4_block9_concat (Concatenat onv4_block8_concat[0][0]</pre>	(None,	None,	None,	5	0	C
onv4_block9_2_conv[0][0]						С
<pre>conv4_block10_0_bn (BatchNormal onv4_block9_concat[0][0]</pre>	(None,	None,	None,	5	2176	С
<pre>conv4_block10_0_relu (Activatio onv4_block10_0_bn[0][0]</pre>	(None,	None,	None,	5	0	С
conv4_block10_1_conv (Conv2D) onv4_block10_0_relu[0][0]	(None,	None,	None,	1	69632	С
<pre>conv4_block10_1_bn (BatchNormal onv4_block10_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
<pre>conv4_block10_1_relu (Activatio onv4_block10_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv4_block10_2_conv (Conv2D) onv4_block10_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block10_concat (Concatena onv4_block9_concat[0][0]	(None,	None,	None,	5	0	c

onv4_block10_2_conv[0][0]

<pre>conv4_block11_0_bn (BatchNormal onv4_block10_concat[0][0]</pre>	(None,	None,	None,	5	2304	С
conv4_block11_0_relu (Activatio onv4_block11_0_bn[0][0]	(None,	None,	None,	5	0	С
conv4_block11_1_conv (Conv2D) onv4_block11_0_relu[0][0]	(None,	None,	None,	1	73728	С
conv4_block11_1_bn (BatchNormal onv4_block11_1_conv[0][0]	(None,	None,	None,	1	512	С
<pre>conv4_block11_1_relu (Activatio onv4_block11_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv4_block11_2_conv (Conv2D) onv4_block11_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block11_concat (Concatena onv4_block10_concat[0][0] onv4_block11_2_conv[0][0]	(None,	None,	None,	6	0	c
conv4_block12_0_bn (BatchNormal onv4_block11_concat[0][0]	- (None,	None,	None,	6	2432	c
conv4_block12_0_relu (Activatio onv4_block12_0_bn[0][0]	(None,	None,	None,	6	0	С
conv4_block12_1_conv (Conv2D) onv4_block12_0_relu[0][0]	(None,	None,	None,	1	77824	С
conv4_block12_1_bn (BatchNormal onv4_block12_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block12_1_relu (Activatio onv4_block12_1_bn[0][0]	(None,	None,	None,	1	0	С

<pre>conv4_block12_2_conv (Conv2D) onv4_block12_1_relu[0][0]</pre>	(None,	None,	None,	3	36864	С
conv4_block12_concat (Concatena onv4_block11_concat[0][0]	(None,	None,	None,	6	0	С
onv4_block12_2_conv[0][0]						с
conv4_block13_0_bn (BatchNormal onv4_block12_concat[0][0]	(None,	None,	None,	6	2560	С
<pre>conv4_block13_0_relu (Activatio onv4_block13_0_bn[0][0]</pre>	(None,	None,	None,	6	0	С
conv4_block13_1_conv (Conv2D) onv4_block13_0_relu[0][0]	(None,	None,	None,	1	81920	С
conv4_block13_1_bn (BatchNormal onv4_block13_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block13_1_relu (Activatio onv4_block13_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block13_2_conv (Conv2D) onv4_block13_1_relu[0][0]	_ (None,	None,	None,	3	36864	С
conv4_block13_concat (Concatena onv4_block12_concat[0][0]	(None,	None,	None,	6	0	c
onv4_block13_2_conv[0][0]						с
conv4_block14_0_bn (BatchNormal onv4_block13_concat[0][0]	(None,	None,	None,	6	2688	С
conv4_block14_0_relu (Activatio onv4_block14_0_bn[0][0]	(None,	None,	None,	6	0	С
conv4_block14_1_conv (Conv2D) onv4_block14_0_relu[0][0]	(None,	None,	None,	1	86016	С
conv4_block14_1_bn (BatchNormal	(None,	None,	None,	1	512	С

onv4_block14	1	conv[0]	[0]
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conv4_block14_1_relu (Activatio onv4_block14_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block14_2_conv (Conv2D) onv4_block14_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block14_concat (Concatena onv4_block13_concat[0][0]	(None,	None,	None,	7	0	c
onv4_block14_2_conv[0][0]						
conv4_block15_0_bn (BatchNormal onv4_block14_concat[0][0]	(None,	None,	None,	7	2816	С
conv4_block15_0_relu (Activatio onv4_block15_0_bn[0][0]	(None,	None,	None,	7	0	С
conv4_block15_1_conv (Conv2D) onv4_block15_0_relu[0][0]	(None,	None,	None,	1	90112	С
conv4_block15_1_bn (BatchNormal onv4_block15_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block15_1_relu (Activatio onv4_block15_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block15_2_conv (Conv2D) onv4_block15_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block15_concat (Concatena onv4_block14_concat[0][0]	(None,	None,	None,	7	0	c
onv4_block15_2_conv[0][0]						J
conv4_block16_0_bn (BatchNormal onv4_block15_concat[0][0]	(None,	None,	None,	7	2944	c
conv4_block16_0_relu (Activatio onv4_block16_0_bn[0][0]	(None,	None,	None,	7	0	С

conv4_block16_1_conv (Conv2D) onv4_block16_0_relu[0][0]	_ (None,	None,	None,	1	94208	С
conv4_block16_1_bn (BatchNormal onv4_block16_1_conv[0][0]	_ (None,	None,	None,	1	512	С
conv4_block16_1_relu (Activatio onv4_block16_1_bn[0][0]	_ (None,	None,	None,	1	0	c
conv4_block16_2_conv (Conv2D) onv4_block16_1_relu[0][0]	_ (None,	None,	None,	3	36864	c
conv4_block16_concat (Concatena onv4_block15_concat[0][0] onv4_block16_2_conv[0][0]	(None,	None,	None,	7	0	c
conv4_block17_0_bn (BatchNormal onv4_block16_concat[0][0]	_ (None,	None,	None,	7	3072	c
conv4_block17_0_relu (Activatio onv4_block17_0_bn[0][0]	_ (None,	None,	None,	7	0	c
conv4_block17_1_conv (Conv2D) onv4_block17_0_relu[0][0]	_ (None,	None,	None,	1	98304	c
conv4_block17_1_bn (BatchNormal onv4_block17_1_conv[0][0]	_ (None,	None,	None,	1	512	c
conv4_block17_1_relu (Activatio onv4_block17_1_bn[0][0]	(None,	None,	None,	1	0	c
conv4_block17_2_conv (Conv2D) onv4_block17_1_relu[0][0]	(None,	None,	None,	3	36864	C
conv4_block17_concat (Concatena onv4_block16_concat[0][0]	(None,	None,	None,	8	0	c
onv4_block17_2_conv[0][0]						C

<pre>conv4_block18_0_bn (BatchNormal onv4_block17_concat[0][0]</pre>	(None,	None,	None,	8	3200	c
<pre>conv4_block18_0_relu (Activatio onv4_block18_0_bn[0][0]</pre>	(None,	None,	None,	8	0	С
conv4_block18_1_conv (Conv2D) onv4_block18_0_relu[0][0]	(None,	None,	None,	1	102400	С
conv4_block18_1_bn (BatchNormal onv4_block18_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block18_1_relu (Activatio onv4_block18_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block18_2_conv (Conv2D) onv4_block18_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv4_block18_concat (Concatena onv4_block17_concat[0][0] onv4_block18_2_conv[0][0]</pre>	(None,	None,	None,	8	0	c c
conv4_block19_0_bn (BatchNormal onv4_block18_concat[0][0]	(None,	None,	None,	8	3328	С
conv4_block19_0_relu (Activatio onv4_block19_0_bn[0][0]	(None,	None,	None,	8	0	c
conv4_block19_1_conv (Conv2D) onv4_block19_0_relu[0][0]	(None,	None,	None,	1	106496	С
conv4_block19_1_bn (BatchNormal onv4_block19_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block19_1_relu (Activatio onv4_block19_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block19_2_conv (Conv2D) onv4_block19_1_relu[0][0]	(None,	None,	None,	3	36864	С

<pre>conv4_block19_concat (Concatena onv4_block18_concat[0][0]</pre>	(None,	None,	None,	8	0	c
onv4_block19_2_conv[0][0]						
<pre>conv4_block20_0_bn (BatchNormal onv4_block19_concat[0][0]</pre>	(None,	None,	None,	8	3456	c
conv4_block20_0_relu (Activatio onv4_block20_0_bn[0][0]	(None,	None,	None,	8	0	c
conv4_block20_1_conv (Conv2D) onv4_block20_0_relu[0][0]	(None,	None,	None,	1	110592	c
conv4_block20_1_bn (BatchNormal onv4_block20_1_conv[0][0]	(None,	None,	None,	1	512	c
conv4_block20_1_relu (Activatio onv4_block20_1_bn[0][0]	(None,	None,	None,	1	0	c
conv4_block20_2_conv (Conv2D) onv4_block20_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block20_concat (Concatena onv4_block19_concat[0][0]	(None,	None,	None,	8	0	c
onv4_block20_2_conv[0][0]						
conv4_block21_0_bn (BatchNormal onv4_block20_concat[0][0]	(None,	None,	None,	8	3584	c
conv4_block21_0_relu (Activatio onv4_block21_0_bn[0][0]	(None,	None,	None,	8	0	С
conv4_block21_1_conv (Conv2D) onv4_block21_0_relu[0][0]	(None,	None,	None,	1	114688	С
<pre>conv4_block21_1_bn (BatchNormal onv4_block21_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С

conv4_block21_1_relu (Activatio onv4_block21_1_bn[0][0]	(None,	None,	None,	1	0	c
conv4_block21_2_conv (Conv2D) onv4_block21_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block21_concat (Concatena onv4_block20_concat[0][0]	(None,	None,	None,	9	0	c
onv4_block21_2_conv[0][0]						
<pre>conv4_block22_0_bn (BatchNormal onv4_block21_concat[0][0]</pre>	(None,	None,	None,	9	3712	С
<pre>conv4_block22_0_relu (Activatio onv4_block22_0_bn[0][0]</pre>	(None,	None,	None,	9	0	С
conv4_block22_1_conv (Conv2D) onv4_block22_0_relu[0][0]	(None,	None,	None,	1	118784	С
conv4_block22_1_bn (BatchNormal onv4_block22_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block22_1_relu (Activatio onv4_block22_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block22_2_conv (Conv2D) onv4_block22_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block22_concat (Concatena onv4_block21_concat[0][0]	(None,	None,	None,	9	0	c
onv4_block22_2_conv[0][0]						
conv4_block23_0_bn (BatchNormal onv4_block22_concat[0][0]	(None,	None,	None,	9	3840	c
<pre>conv4_block23_0_relu (Activatio onv4_block23_0_bn[0][0]</pre>	(None,	None,	None,	9	0	С

<pre>conv4_block23_1_conv (Conv2D) onv4_block23_0_relu[0][0]</pre>	(None,	None,	None,	1	122880	С
conv4_block23_1_bn (BatchNormal onv4_block23_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block23_1_relu (Activatio onv4_block23_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block23_2_conv (Conv2D) onv4_block23_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block23_concat (Concatena onv4_block22_concat[0][0]	(None,	None,	None,	9	0	c
onv4_block23_2_conv[0][0]						
conv4_block24_0_bn (BatchNormal onv4_block23_concat[0][0]	(None,	None,	None,	9	3968	С
conv4_block24_0_relu (Activatio onv4_block24_0_bn[0][0]	(None,	None,	None,	9	0	С
conv4_block24_1_conv (Conv2D) onv4_block24_0_relu[0][0]	(None,	None,	None,	1	126976	С
conv4_block24_1_bn (BatchNormal onv4_block24_1_conv[0][0]	(None,	None,	None,	1	512	С
conv4_block24_1_relu (Activatio onv4_block24_1_bn[0][0]	(None,	None,	None,	1	0	С
conv4_block24_2_conv (Conv2D) onv4_block24_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv4_block24_concat (Concatena onv4_block23_concat[0][0]	(None,	None,	None,	1	0	c
onv4_block24_2_conv[0][0]						С
pool4_bn (BatchNormalization)	(None,	None,	None,	1	4096	С

onv4_block24_concat[0][0]

pool4_relu (Activation) ool4_bn[0][0]	(None,	None,	None,	1	0	p
pool4_conv (Conv2D) ool4_relu[0][0]	(None,	None,	None,	5	524288	p
<pre>pool4_pool (AveragePooling2D) ool4_conv[0][0]</pre>	(None,	None,	None,	5	0	р
<pre>conv5_block1_0_bn (BatchNormali ool4_pool[0][0]</pre>	(None,	None,	None,	5	2048	p
conv5_block1_0_relu (Activation onv5_block1_0_bn[0][0]	(None,	None,	None,	5	0	С
conv5_block1_1_conv (Conv2D) onv5_block1_0_relu[0][0]	(None,	None,	None,	1	65536	С
<pre>conv5_block1_1_bn (BatchNormali onv5_block1_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
conv5_block1_1_relu (Activation onv5_block1_1_bn[0][0]	(None,	None,	None,	1	0	c
conv5_block1_2_conv (Conv2D) onv5_block1_1_relu[0][0]	(None,	None,	None,	3	36864	c
<pre>conv5_block1_concat (Concatenat ool4_pool[0][0] onv5 block1 2 conv[0][0]</pre>	(None,	None,	None,	5	0	p c
conv5_block2_0_bn (BatchNormalionv5_block1_concat[0][0]	(None,	None,	None,	5	2176	c
conv5_block2_0_relu (Activation onv5_block2_0_bn[0][0]	(None,	None,	None,	5	0	С

<pre>conv5_block2_1_conv (Conv2D) onv5_block2_0_relu[0][0]</pre>	(None,	None,	None,	1	69632	С
<pre>conv5_block2_1_bn (BatchNormali onv5_block2_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
conv5_block2_1_relu (Activation onv5_block2_1_bn[0][0]	(None,	None,	None,	1	0	С
conv5_block2_2_conv (Conv2D) onv5_block2_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv5_block2_concat (Concatenat onv5_block1_concat[0][0]</pre>	(None,	None,	None,	5	0	c
onv5_block2_2_conv[0][0]						
<pre>conv5_block3_0_bn (BatchNormali onv5_block2_concat[0][0]</pre>	(None,	None,	None,	5	2304	С
conv5_block3_0_relu (Activation onv5_block3_0_bn[0][0]	(None,	None,	None,	5	0	c
conv5_block3_1_conv (Conv2D) onv5_block3_0_relu[0][0]	(None,	None,	None,	1	73728	С
<pre>conv5_block3_1_bn (BatchNormali onv5_block3_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
conv5_block3_1_relu (Activation onv5_block3_1_bn[0][0]	(None,	None,	None,	1	0	С
conv5_block3_2_conv (Conv2D) onv5_block3_1_relu[0][0]	(None,	None,	None,	3	36864	c
<pre>conv5_block3_concat (Concatenat onv5_block2_concat[0][0]</pre>	(None,	None,	None,	6	0	С
onv5_block3_2_conv[0][0]						С
conv5_block4_0_bn (BatchNormali	(None,	None,	None,	6	2432	C

onv5 block3	_concat[0][0]
-------------	---------------

conv5_block4_0_relu (Activation onv5_block4_0_bn[0][0]	(None,	None,	None,	6	0	С
conv5_block4_1_conv (Conv2D) onv5_block4_0_relu[0][0]	(None,	None,	None,	1	77824	С
<pre>conv5_block4_1_bn (BatchNormali onv5_block4_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
conv5_block4_1_relu (Activation onv5_block4_1_bn[0][0]	(None,	None,	None,	1	0	С
conv5_block4_2_conv (Conv2D) onv5_block4_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv5_block4_concat (Concatenat onv5_block3_concat[0][0]</pre>	(None,	None,	None,	6	0	c
onv5_block4_2_conv[0][0]						
<pre>conv5_block5_0_bn (BatchNormali onv5_block4_concat[0][0]</pre>	(None,	None,	None,	6	2560	C
conv5_block5_0_relu (Activation onv5_block5_0_bn[0][0]	(None,	None,	None,	6	0	С
conv5_block5_1_conv (Conv2D) onv5_block5_0_relu[0][0]	(None,	None,	None,	1	81920	С
<pre>conv5_block5_1_bn (BatchNormali onv5_block5_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
conv5_block5_1_relu (Activation onv5_block5_1_bn[0][0]	(None,	None,	None,	1	0	С
conv5_block5_2_conv (Conv2D) onv5_block5_1_relu[0][0]	(None,	None,	None,	3	36864	С

<pre>conv5_block5_concat (Concatenat onv5_block4_concat[0][0]</pre>	(None,	None,	None,	6	0	С
onv5_block5_2_conv[0][0]						С
conv5_block6_0_bn (BatchNormalionv5_block5_concat[0][0]	(None,	None,	None,	6	2688	С
conv5_block6_0_relu (Activation onv5_block6_0_bn[0][0]	(None,	None,	None,	6	0	c
conv5_block6_1_conv (Conv2D) onv5_block6_0_relu[0][0]	(None,	None,	None,	1	86016	c
conv5_block6_1_bn (BatchNormalionv5_block6_1_conv[0][0]	(None,	None,	None,	1	512	c
conv5_block6_1_relu (Activation onv5_block6_1_bn[0][0]	(None,	None,	None,	1	0	c
conv5_block6_2_conv (Conv2D) onv5_block6_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv5_block6_concat (Concatenat onv5_block5_concat[0][0]</pre>	(None,	None,	None,	7	0	C
onv5_block6_2_conv[0][0]						_
<pre>conv5_block7_0_bn (BatchNormali onv5_block6_concat[0][0]</pre>	(None,	None,	None,	7	2816	С
<pre>conv5_block7_0_relu (Activation onv5_block7_0_bn[0][0]</pre>	(None,	None,	None,	7	0	С
conv5_block7_1_conv (Conv2D) onv5_block7_0_relu[0][0]	(None,	None,	None,	1	90112	С
conv5_block7_1_bn (BatchNormalionv5_block7_1_conv[0][0]	(None,	None,	None,	1	512	С
conv5_block7_1_relu (Activation	None,	None,	None,	1	0	С

onv5_	bloc	k7	1 k	on[0]	0]

conv5_block7_2_conv (Conv2D) onv5_block7_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv5_block7_concat (Concatenat onv5_block6_concat[0][0]</pre>	(None,	None,	None,	7	0	С
onv5_block7_2_conv[0][0]						С
<pre>conv5_block8_0_bn (BatchNormali onv5_block7_concat[0][0]</pre>	(None,	None,	None,	7	2944	С
conv5_block8_0_relu (Activation onv5_block8_0_bn[0][0]	(None,	None,	None,	7	0	С
conv5_block8_1_conv (Conv2D) onv5_block8_0_relu[0][0]	(None,	None,	None,	1	94208	С
<pre>conv5_block8_1_bn (BatchNormali onv5_block8_1_conv[0][0]</pre>	(None,	None,	None,	1	512	c
conv5_block8_1_relu (Activation onv5_block8_1_bn[0][0]	(None,	None,	None,	1	0	С
conv5_block8_2_conv (Conv2D) onv5_block8_1_relu[0][0]	(None,	None,	None,	3	36864	c
<pre>conv5_block8_concat (Concatenat onv5_block7_concat[0][0]</pre>	(None,	None,	None,	7	0	c
onv5_block8_2_conv[0][0]						_
<pre>conv5_block9_0_bn (BatchNormali onv5_block8_concat[0][0]</pre>	(None,	None,	None,	7	3072	C
conv5_block9_0_relu (Activation onv5_block9_0_bn[0][0]	(None,	None,	None,	7	0	c
conv5_block9_1_conv (Conv2D) onv5_block9_0_relu[0][0]	(None,	None,	None,	1	98304	С

conv5_block9_1_bn (BatchNormalionv5_block9_1_conv[0][0]	(None,	None,	None,	1	512	c
conv5_block9_1_relu (Activation onv5_block9_1_bn[0][0]	(None,	None,	None,	1	0	С
conv5_block9_2_conv (Conv2D) onv5_block9_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv5_block9_concat (Concatenat onv5_block8_concat[0][0]	(None,	None,	None,	8	0	c
onv5_block9_2_conv[0][0]						
conv5_block10_0_bn (BatchNormal onv5_block9_concat[0][0]	(None,	None,	None,	8	3200	С
conv5_block10_0_relu (Activatio onv5_block10_0_bn[0][0]	(None,	None,	None,	8	0	С
conv5_block10_1_conv (Conv2D) onv5_block10_0_relu[0][0]	(None,	None,	None,	1	102400	С
conv5_block10_1_bn (BatchNormal onv5_block10_1_conv[0][0]	(None,	None,	None,	1	512	С
conv5_block10_1_relu (Activatio onv5_block10_1_bn[0][0]	(None,	None,	None,	1	0	С
conv5_block10_2_conv (Conv2D) onv5_block10_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv5_block10_concat (Concatena onv5_block9_concat[0][0]	(None,	None,	None,	8	0	c
onv5_block10_2_conv[0][0]						С
conv5_block11_0_bn (BatchNormal onv5_block10_concat[0][0]	(None,	None,	None,	8	3328	c

<pre>conv5_block11_0_relu (Activatio onv5_block11_0_bn[0][0]</pre>	(None,	None,	None,	8	0	С
conv5_block11_1_conv (Conv2D) onv5_block11_0_relu[0][0]	(None,	None,	None,	1	106496	С
<pre>conv5_block11_1_bn (BatchNormal onv5_block11_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
<pre>conv5_block11_1_relu (Activatio onv5_block11_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv5_block11_2_conv (Conv2D) onv5_block11_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv5_block11_concat (Concatena onv5_block10_concat[0][0]	(None,	None,	None,	8	0	c
onv5_block11_2_conv[0][0]						
conv5_block12_0_bn (BatchNormal onv5_block11_concat[0][0]	(None,	None,	None,	8	3456	С
conv5_block12_0_relu (Activatio onv5_block12_0_bn[0][0]	(None,	None,	None,	8	0	С
conv5_block12_1_conv (Conv2D) onv5_block12_0_relu[0][0]	(None,	None,	None,	1	110592	С
conv5_block12_1_bn (BatchNormal onv5_block12_1_conv[0][0]	(None,	None,	None,	1	512	С
<pre>conv5_block12_1_relu (Activatio onv5_block12_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv5_block12_2_conv (Conv2D) onv5_block12_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv5_block12_concat (Concatena onv5_block11_concat[0][0]	(None,	None,	None,	8	0	С

onv5_block12_2_conv[0][0]						C
conv5_block13_0_bn (BatchNormal onv5_block12_concat[0][0]	(None,	None,	None,	8	3584	С
<pre>conv5_block13_0_relu (Activatio onv5_block13_0_bn[0][0]</pre>	(None,	None,	None,	8	0	С
conv5_block13_1_conv (Conv2D) onv5_block13_0_relu[0][0]	(None,	None,	None,	1	114688	С
<pre>conv5_block13_1_bn (BatchNormal onv5_block13_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
<pre>conv5_block13_1_relu (Activatio onv5_block13_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С
conv5_block13_2_conv (Conv2D) onv5_block13_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv5_block13_concat (Concatena onv5_block12_concat[0][0] onv5 block13 2 conv[0][0]</pre>	(None,	None,	None,	9	0	c
conv5_block14_0_bn (BatchNormal	(None,	None,	None,	9	3712	
onv5_block13_concat[0][0]	- (None	None	None		0	
<pre>conv5_block14_0_relu (Activatio onv5_block14_0_bn[0][0]</pre>	(None,	None,	none,	9		с —
conv5_block14_1_conv (Conv2D) onv5_block14_0_relu[0][0]	(None,	None,	None,	1	118784	С
<pre>conv5_block14_1_bn (BatchNormal onv5_block14_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
<pre>conv5_block14_1_relu (Activatio onv5_block14_1_bn[0][0]</pre>	(None,	None,	None,	1	0	С

conv5_block14_2_conv (Conv2D) onv5_block14_1_relu[0][0]	(None,	None,	None,	3	36864	С
conv5_block14_concat (Concatena onv5_block13_concat[0][0]	(None,	None,	None,	9	0	С
onv5_block14_2_conv[0][0]						_
conv5_block15_0_bn (BatchNormal onv5_block14_concat[0][0]	(None,	None,	None,	9	3840	C
conv5_block15_0_relu (Activatio onv5_block15_0_bn[0][0]	(None,	None,	None,	9	0	С
conv5_block15_1_conv (Conv2D) onv5_block15_0_relu[0][0]	(None,	None,	None,	1	122880	С
conv5_block15_1_bn (BatchNormal onv5_block15_1_conv[0][0]	(None,	None,	None,	1	512	С
conv5_block15_1_relu (Activatio onv5_block15_1_bn[0][0]	(None,	None,	None,	1	0	С
conv5_block15_2_conv (Conv2D) onv5_block15_1_relu[0][0]	(None,	None,	None,	3	36864	c
conv5_block15_concat (Concatena onv5_block14_concat[0][0]	(None,	None,	None,	9	0	С
onv5_block15_2_conv[0][0]						С
<pre>conv5_block16_0_bn (BatchNormal onv5_block15_concat[0][0]</pre>	(None,	None,	None,	9	3968	С
<pre>conv5_block16_0_relu (Activatio onv5_block16_0_bn[0][0]</pre>	(None,	None,	None,	9	0	С
conv5_block16_1_conv (Conv2D) onv5_block16_0_relu[0][0]	(None,	None,	None,	1	126976	C

<pre>conv5_block16_1_bn (BatchNormal onv5_block16_1_conv[0][0]</pre>	(None,	None,	None,	1	512	С
conv5_block16_1_relu (Activatio onv5_block16_1_bn[0][0]	(None,	None,	None,	1	0	С
conv5_block16_2_conv (Conv2D) onv5_block16_1_relu[0][0]	(None,	None,	None,	3	36864	С
<pre>conv5_block16_concat (Concatena onv5_block15_concat[0][0] onv5_block16_2_conv[0][0]</pre>	(None,	None,	None,	1	0	c
bn (BatchNormalization) onv5_block16_concat[0][0]	(None,	None,	None,	1	4096	c
<pre>global_average_pooling2d_1 (Glo n[0][0]</pre>	(None,	1024)			0	b
dense_1 (Dense) lobal_average_pooling2d_1[0][0]	(None,	14)		==:	14350	g ===
Total params: 7,051,854 Trainable params: 6,968,206 Non-trainable params: 83,648	=					

Keras models include abundant information about the elements that make them up. You can check all of the available methods and attributes of this class by using the dir() method:

```
In [4]: # Printing out methods and attributes for Keras model
    print(f"Keras' models have the following methods and attributes: \n
    \n{dir(model)}")
```

Keras' models have the following methods and attributes:

['__call__', '__class__', '__delattr__', '__dict__', '__dir__' ', '__reduce_e asshook__', '__weakref__', '_add_inbound_node', '_built', '_check_ num_samples', '_check_trainable_weights_consistency', '_collected_ trainable_weights', '_container_nodes', '_feed_input_names', '_fee d_input_shapes', '_feed_inputs', '_feed_loss_fns', '_feed_output_n ames', 'feed output shapes', 'feed outputs', 'feed sample weigh t_modes', '_feed_sample_weights', '_feed_targets', '_fit_loop', '_ function_kwargs', '_get_node_attribute_at_index', '_inbound_nodes' , '_internal_input_shapes', '_internal_output_shapes', '_make_pred ict_function', '_make_test_function', '_make_train_function', '_no de_key', '_nodes_by_depth', '_outbound_nodes', '_output_mask_cache _output_shape_cache', '_output_tensor_cache', '_per_input_loss, '_per_input_updates', '_predict_loop', '_standardize_user_dat a', '_test_loop', '_updated_config', 'add_loss', 'add_update', 'ad d_weight', 'assert_input_compatibility', 'build', 'built', 'call', 'compile', 'compute mask', 'compute output shape', 'count params', 'evaluate', 'evaluate_generator', 'fit', 'fit_generator', 'from_config', 'get_config', 'get_input_at', 'get_input_mask_at', 'get_inp ut_shape_at', 'get_layer', 'get_losses_for', 'get_output_at', 'get _output_mask_at', 'get_output_shape_at', 'get_updates for', 'get w eights', 'input', 'input_layers', 'input_layers_node_indices', 'in put layers tensor indices', 'input mask', 'input names', 'input sh ape', 'input_spec', 'inputs', 'layers', 'layers_by_depth', 'load_w
eights', 'loss', 'loss_functions', 'loss_weights', 'losses', 'metr ics', 'metrics_names', 'metrics_tensors', 'metrics_updates', 'name ', 'non_trainable_weights', 'optimizer', 'output', 'output_layers' , 'output_layers_node_indices', 'output_layers_tensor indices', 'o utput_mask', 'output_names', 'output_shape', 'outputs', 'predict', 'predict function', 'predict generator', 'predict on batch', 'rese t states', 'run internal graph', 'sample weight mode', 'sample wei ght_modes', 'sample_weights', 'save', 'save_weights', 'set_weights ', 'state_updates', 'stateful', 'stateful_metric_functions', 'stat eful_metric_names', 'summary', 'supports_masking', 'targets', 'tes t function', 'test on batch', 'to json', 'to yaml', 'total loss', 'train_function', 'train_on_batch', 'trainable', 'trainable_weight s', 'updates', 'uses learning phase', 'weighted metrics', 'weights ']

Wow, this certainly is a lot! These models are indeed very complex.

What we are interested in are the layers of the model which can be easily accessed as an attribute using the dot notation. They are a list of layers, which can be confirmed by checking its type:

```
In [5]: # Check the type of the model's layers
        type(model.layers)
Out[5]: list
In [6]: # Print 5 first layers along with their names
        for i in range(5):
             l = model.layers[i]
            print(f"Layer number {i}: \n{1} \nWith name: {1.name} \n")
        Layer number 0:
        <keras.engine.topology.InputLayer object at 0x7f57f3873160>
        With name: input 1
        Layer number 1:
        <keras.layers.convolutional.ZeroPadding2D object at 0x7f57f35c2c18</pre>
        With name: zero padding2d 1
        Layer number 2:
        <keras.layers.convolutional.Conv2D object at 0x7f57f3873668>
        With name: conv1/conv
        Layer number 3:
        <keras.layers.normalization.BatchNormalization object at 0x7f57f35</pre>
        eb7f0>
        With name: conv1/bn
        Layer number 4:
        <keras.layers.core.Activation object at 0x7f57f3581908>
        With name: conv1/relu
```

Let's check how many layers our model has:

```
In [7]: # Print number of layers in our model
print(f"The model has {len(model.layers)} layers")
```

The model has 428 layers

Our main goal is interpreting the representations which the neural net is creating for classifying our images. But as you can see this architecture has many layers.

Actually we are really interested in the representations that the convolutional layers produce because these are the layers that (hopefully) recognize concrete elements within the images. We are also interested in the "concatenate" layers because in our model's arquitecture they concatenate convolutional layers.

Let's check how many of those we have:

This number is still very big to try to interpret each one of these layers individually.

One characteristic of CNN's is that the earlier layers capture low-level features such as edges in an image while the deeper layers capture high-level concepts such as physical features of a "Cat".

Because of this **Grad-CAM** usually focuses on the last layers, as they provide a better picture of what the network is paying attention to when classifying a particular class. Let's grab the last concatenate layer of our model. Luckily Keras API makes this quite easy:

```
In [9]: # Save the desired layer in a variable
layer = model.layers[424]

# Print layer
layer
```

Out[9]: <keras.layers.merge.Concatenate at 0x7f570778cd68>

This approach is not the best since we will need to know the exact index of the desired layer. Luckily we can use the get layer() method in conjunction with the layer's name to get the same result.

Remember you can get the name from the information displayed earlier with the summary() method.

```
In [10]: # Save the desired layer in a variable
    layer = model.get_layer("conv5_block16_concat")

# Print layer
    layer
```

Out[10]: <keras.layers.merge.Concatenate at 0x7f570778cd68>

Let's check what methods and attributes we have available when working with this layer:

```
In [11]: # Printing out methods and attributes for Keras' layer
print(f"Keras' layers have the following methods and attributes: \n
\n{dir(layer)}")
```

Keras' layers have the following methods and attributes:

```
['_call__', '_class__', '_delattr__', '_dict__', '_dir__', '_doc__', '_eq__', '_format__', '_ge__', '_getattribute__', '_gt__', '_hash__', '_init__', '_init_subclass__', '_le__', '_lt__', '_lt__', '_new__', '_reduce__', '_reduce_e x__', '_repr__', '_setattr__', '_sizeof__', '_str__', '_subclasshook__', '_weakref__', '_add_inbound_node', '_built', '_compute_elemwise_op_output_shape', '_get_node_attribute_at_index', '_inbound_nodes', '_initial_weights', '_losses', '_merge_function', '_node_key', '_non_trainable_weights', '_outbound_nodes', '_per_input_losses', '_per_input_updates', '_reshape_required', '_trainable_weights', '_updates', 'add_loss', 'add_update', 'add_weight', 'assert_input_compatibility', 'axis', 'build', 'built', 'call', 'compute_mask', 'compute_output_shape', 'count_params', 'from_config', 'get_config', 'get_input_at', 'get_input_mask_at', 'get_input_shape_at', 'get_losses_for', 'get_output_at', 'get_output_mask_at', 'get_output_shape_at', 'get_updates_for', 'get_weights', 'input', 'input_mask', 'input_shape', 'input_spec', 'losses', 'name', 'non_trainable_weights', 'output', 'output_mask', 'output_shape', 'set_weights', 'stateful', 'supports_masking', 'trainable', 'trainable_weights', 'updates', 'weights']
```

Since we want to know the representations which this layer is abstracting from the images we should be interested in the output from this layer. Luckily we have this attribute available:

```
In [12]: # Print layer's output layer.output
```

Do you notice something odd? The shape of this tensor is undefined for some dimensions. This is because this tensor is just a placeholder and it doesn't really contain information about the activations that occurred in this layer.

To compute the actual activation values given an input we will need to use a **Keras function**.

This function accepts lists of input and output placeholders and can be used with an actual input to compute the respective output of the layer associated to the placeholder for that given input.

Before jumping onto the Keras function we should rewind a little bit to get the placeholder tensor associated with the input. You can get this from the model's input:

```
In [13]: # Print model's input tensor placeholder
model.input

Out[13]: <tf.Tensor 'input 1:0' shape=(?, ?, ?, 3) dtype=float32>
```

We can see that this is a placeholder as well. Now let's instantiate our Keras function using Keras backend. Please be aware that this **function expects its arguments as lists or tuples**:

Out[14]: <keras.backend.tensorflow backend.Function at 0x7f56fe4e6b38>

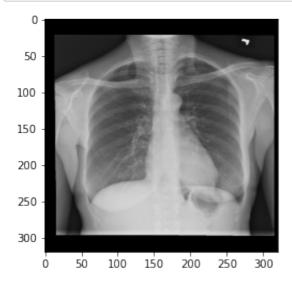
Let's test the functions for computing the last layer activation which we just defined on a particular image. Don't worry about the code to load the image, this has been taken care of for you. You should only care that an image ready to be processed will be saved in the x variable:

```
In [15]: # Load dataframe that contains information about the dataset of ima
    ges
    df = pd.read_csv("nih_new/train-small.csv")

# Path to the actual image
    im_path = 'nih_new/images-small/00000599_000.png'

# Load the image and save it to a variable
    x = load_image(im_path, df, preprocess=False)

# Display the image
    plt.imshow(x, cmap = 'gray')
    plt.show()
```



We should normalize this image before going forward, this has also been taken care of:

```
In [16]: # Calculate mean and standard deviation of a batch of images
    mean, std = get_mean_std_per_batch(df)

# Normalize image
    x = load_image_normalize(im_path, mean, std)
```

Now we have everything we need to compute the actual values of the last layer activations. In this case we should also **provide the input as a list or tuple**:

An important intermediary step is to trim the batch dimension which can be done like this. This is necessary because we are applying Grad-CAM to a single image rather than to a batch of images:

```
In [18]: # Remove batch dimension
         actual activations = actual activations[0][0, :]
In [19]: # Print shape of the activation array
         print(f"Activations of last convolutional layer have shape: {actual
         _activations.shape}")
         # Print activation array
         actual activations
         Activations of last convolutional layer have shape: (10, 10, 1024)
Out[19]: array([[[-2.34528378e-01, 1.03888601e-01, -1.02932416e-01, ...,
                   1.95436418e-01, -7.53498077e-02, 2.19688460e-01],
                 [-3.57181996e-01, -5.06232321e-01, -9.07948315e-01, ...,
                   2.82368094e-01, -8.69844109e-02,
                                                    3.30677748e-01],
                 [-2.64806449e-01, -2.77638644e-01, -9.12588179e-01, ...,
                   2.36321479e-01, -7.62053505e-02, 2.71659851e-01],
                 [-3.64358276e-01, -3.05198252e-01, -8.99509788e-01, ...,
                   1.32572979e-01, -8.87970179e-02,
                                                    1.68741941e-01],
                 [-2.63873279e-01, -7.16017410e-02, -5.18461406e-01, ...,
                   2.81986654e-01, -1.17035367e-01, 3.22839558e-01],
                 [-2.43374333e-01, 2.20213965e-01, -8.40832591e-02, ...,
                   1.46577284e-01, -7.16667622e-02, 2.15044528e-01]],
                [-3.85911673e-01, 3.94917391e-02, -4.89642799e-01, ...,
                   2.48766661e-01, -1.19054504e-01, 3.17032784e-01],
                 [-1.75690576e-01, -4.20813233e-01, -2.65263885e-01, ...,
                   4.28565294e-01, -1.49629205e-01,
                                                    4.95896250e-01],
                 [-4.91427928e-01, -5.02470613e-01, -4.33604479e-01, ...,
                   2.54681349e-01, -1.14488915e-01, 3.30634028e-01],
                 [-3.48210931e-01, -6.88572645e-01, -5.79565823e-01, ...,
                                                    1.86690331e-011,
                   1.57981083e-01, -1.03492454e-01,
                 [-5.49307585e-01, -1.02473652e+00, -5.52783251e-01, ...,
                   3.22046787e-01, -1.36609375e-01, 3.27671915e-01],
                 [-5.58423519e-01, -2.21409231e-01, 1.06662512e-03, ...,
                   1.82196707e-01, -6.83000088e-02, 2.14448273e-01]],
                [[-4.71532464e-01, -2.07039714e-02, -4.41627920e-01, ...,
                   2.74705708e-01, -1.13615274e-01, 2.43266091e-01],
                 [-1.25446916e+00, -1.07412672e+00, -2.99405575e-01, ...,
                   4.57139075e-01, -1.50618494e-01, 3.55173171e-01],
                 [-8.81016254e-01, -6.38861358e-01, -5.44111133e-01, ...,
                   2.96926886e-01, -9.45852175e-02, 2.02520594e-01],
                 [-1.35452890e+00, 1.39726698e-01, 2.32537150e-01, ...,
                   6.18949272e-02, -5.50149716e-02,
                                                    8.78603086e-02],
                 [-8.59520912e-01, -2.54668146e-02, 1.16123629e+00, ...,
                   1.92437738e-01, -6.40552044e-02, 1.55031338e-01],
                 [-7.34347522e-01, -1.96773887e-01, 9.74288881e-01, ...,
                   8.12446550e-02, -2.90456042e-02, 8.25736523e-02]],
```

[[-5.93337536e-01, -1.82666220e-02, 1.42468229e-01, ...,3.61059010e-01, -1.51442572e-01, 4.83031034e-01], [-3.63586664e-01, 7.16439605e-01, -3.02047431e-02, ...,4.02755529e-01, -1.18740864e-01, 7.14342237e-01, [-2.05164719e+00, -1.17773581e+00, -7.96093047e-01, ...,1.89675331e-01, 1.75441876e-01, 8.65436077e-01], [-2.01013637e+00, -7.83838212e-01, -5.93690693e-01, ...,-5.11500984e-03, 1.18342265e-01, 1.41481924e+00], [-1.60671687e+00, 1.13282204e-01, -5.64168513e-01, ..., 1.52177155e-01, -6.78298026e-02, 1.50779235e+00], [-8.80541980e-01, 7.00929314e-02, 3.32438052e-01, ..., -1.33873001e-01, 1.02796428e-01, 1.05246699e+00]], [[-6.13768876e-01, -3.93467359e-02, -4.98745412e-01, ...,2.92770028e-01, -1.57525450e-01, 4.59775925e-01], [-5.27344882e-01, 1.30507693e-01, -4.39863145e-01, ...,4.43635941e-01, -2.30797246e-01, 6.37336552e-01], [-4.47706252e-01, 6.52831733e-01, -1.53416073e+00, ..., 3.40943336e-01, -1.47531837e-01, 5.14184713e-01], [-7.80842781e-01, -4.47639078e-02, -1.02739370e+00, ...,7.32436106e-02, -7.68425167e-02, 7.55837739e-01], [-2.56010771e-01, 4.37092394e-01, -9.21702981e-01, ...,2.91409433e-01, -2.16206715e-01, 8.89219224e-01], [-5.01822472e-01, -1.27244368e-01, -5.96898496e-01, ...,3.73074859e-02, -8.50129500e-02, 7.00261950e-01]], [-8.94521356e-01, 2.00330630e-01, 4.90441620e-01, ...,2.61645913e-01, -1.10321872e-01, 2.45080888e-01], [-9.90234494e-01, -1.79898918e-01, 3.64397943e-01, ...,

3.40280712e-01, -1.21333607e-01, 3.34408909e-01], [-6.39139056e-01, 2.23485455e-02, -2.95223475e-01, ..., 2.67534673e-01, -9.55819562e-02, 2.79736072e-01],

[-5.00944674e-01, 7.57797956e-01, -8.73069167e-02, ..., 1.57085389e-01, -8.08104724e-02, 2.48510525e-01], [-5.79098403e-01, 3.12583297e-01, 2.19042763e-01, ..., 2.47366339e-01, -1.15144797e-01, 3.04065764e-01], [-6.10948980e-01, 2.20546141e-01, 5.39675653e-01, ..., 1.70019135e-01, -5.60478754e-02, 2.50791490e-01]]],

Looks like everything worked out nicely! This is all for this lecture notebook (Grad-CAM Part 1). In Part 2 we will see how to calculate the gradients of the model's output with respect to the activations in this layer. This is the "Grad" part of Grad-CAM.

dtype=float32)

Congratulations on finishing this lecture notebook! Hopefully you will now have a better understanding of how to leverage Keras's API power for computing activations in specific layers. Keep it up!