

# bg2502-take-at-home

October 27, 2023

## 1 Problem Set 1 (Take Home) - 40 points

bg2502: Kindly use the Outline to navigate through this notebook

### 1.1 PS 1.A - 20 points

In [this video](#) the author explains how to extract various visualizations of what CNNs learn. [Your course site](#) also covers the topic.

Using the [CIFAR-10 dataset](#), train a ResNet-50 based CNN on the classification task of  $K = 9$  classes (filter out the class `ship`) and create the following visualizations for first, middle and last blocks of ResNet-50. You are free to select a class to showcase such visualizations.

- Visualizing intermediate convnet outputs (“intermediate activations”). This is useful to understand how successive convnet layers transform their input.
- Visualizing convnets filters. This is useful to understand precisely what visual pattern or concept each filter in a convnet is receptive to.
- Visualizing heatmaps of class activation in an image. This is useful to understand which part of an image where identified as belonging to a given class, and thus allows to localize objects in images.

```
[ ]: # Insert your code here
import numpy as np
import tensorflow as tf
from keras.models import Sequential, Model, load_model
from keras.layers import Flatten, Input, Dense, Add, Activation, Dropout,
    ↳ Conv2D, MaxPooling2D, GlobalAveragePooling2D, GlobalMaxPooling2D,
    ↳ AveragePooling2D, Resizing, ZeroPadding2D, BatchNormalization
from keras.optimizers import Adam, SGD
from keras.regularizers import l2
from keras.initializers import glorot_uniform
from keras.preprocessing import image
from keras.utils import to_categorical
from keras.preprocessing.image import ImageDataGenerator
from keras.preprocessing import image

from keras.applications import ResNet50
```

```
# from keras.applications.resnet50 import preprocess_input

from keras import backend as K

from keras.datasets import cifar10

from sklearn.preprocessing import LabelEncoder , OneHotEncoder

from matplotlib.pyplot import imshow
import matplotlib.pyplot as plt
```

### 1.1.1 PS 1.A Data preprocessing

```
[ ]: # Load the CIFAR-10 dataset
(x_train, y_train), (x_test, y_test) = cifar10.load_data()

classes = len(np.unique(y_train))
classes
```

```
[ ]: 10
```

we know the number of classes and index associated from her [https://www.tensorflow.org/api\\_docs/python/tf/keras/datasets/cifar10/load\\_data](https://www.tensorflow.org/api_docs/python/tf/keras/datasets/cifar10/load_data)

The classes are: | Label | Description | | 0 | airplane | | 1 | automobile | | 2 | bird | | 3 | cat | | 4 | deer | | 5 | dog | | 6 | frog | | 7 | horse | | 8 | ship | | 9 | truck |

We have to exclude the class “ship with index 8”

```
[ ]: # preprocess the CIFAR-10 dataset
index = np.where(y_train == 8)
# Filter the dataset for 9 out of 10 classes (exclude "ship index 8" class)
class_to_exclude = 8 # Adjust this class to exclude "ship"
x_train = x_train[y_train.flatten() != class_to_exclude] / 255.0
y_train = y_train[y_train != class_to_exclude].astype(int)
x_test = x_test[y_test.flatten() != class_to_exclude] / 255.0
y_test = y_test[y_test != class_to_exclude].astype(int)

# number of classes K=9
k = 9
```

```
[ ]: print(len(np.unique(y_train)), len(np.unique(x_train)), len(np.unique(y_test)),
        len(np.unique(x_test)))
print(x_train.shape, y_train.shape, x_test.shape, y_test.shape)
```

```
9 256 9 256
(45000, 32, 32, 3) (45000,) (9000, 32, 32, 3) (9000,)
```

```
[ ]: # visualize data by plotting images
fig, ax = plt.subplots(5, 5)
k = 5

for i in range(5):
    for j in range(5):
        ax[i][j].imshow(x_test[k], aspect='auto')
        k += 1

plt.show()
```



```
[ ]: # A ResNet-50 model expects 224 × 224-pixel images
im_size = 224

x_train_resize = Resizing(height=im_size, width=im_size,
    ↳crop_to_aspect_ratio=True)(x_train)
x_test_resize = Resizing(height=im_size, width=im_size,
    ↳crop_to_aspect_ratio=True)(x_test)

[ ]: print(x_train_resize.shape, y_train.shape, x_test_resize.shape, y_test.shape)
```

```
(45000, 224, 224, 3) (45000,) (9000, 224, 224, 3) (9000,)
```

```
[ ]: # for y
y_labelencoder = LabelEncoder ()
y = y_labelencoder.fit_transform (y_train)
#print (y)

ytr=y_train.reshape(-1,1)
onehotencoder = OneHotEncoder() #Converted scalar output into vector output
    →where the correct class will be 1 and other will be 0
y_train_norm= onehotencoder.fit_transform(ytr).toarray()

yts=y_test.reshape(-1,1)
# onehotencoder = OneHotEncoder() #Converted scalar output into vector output
    →where the correct class will be 1 and other will be 0
y_test_norm= onehotencoder.transform(yts).toarray()
```

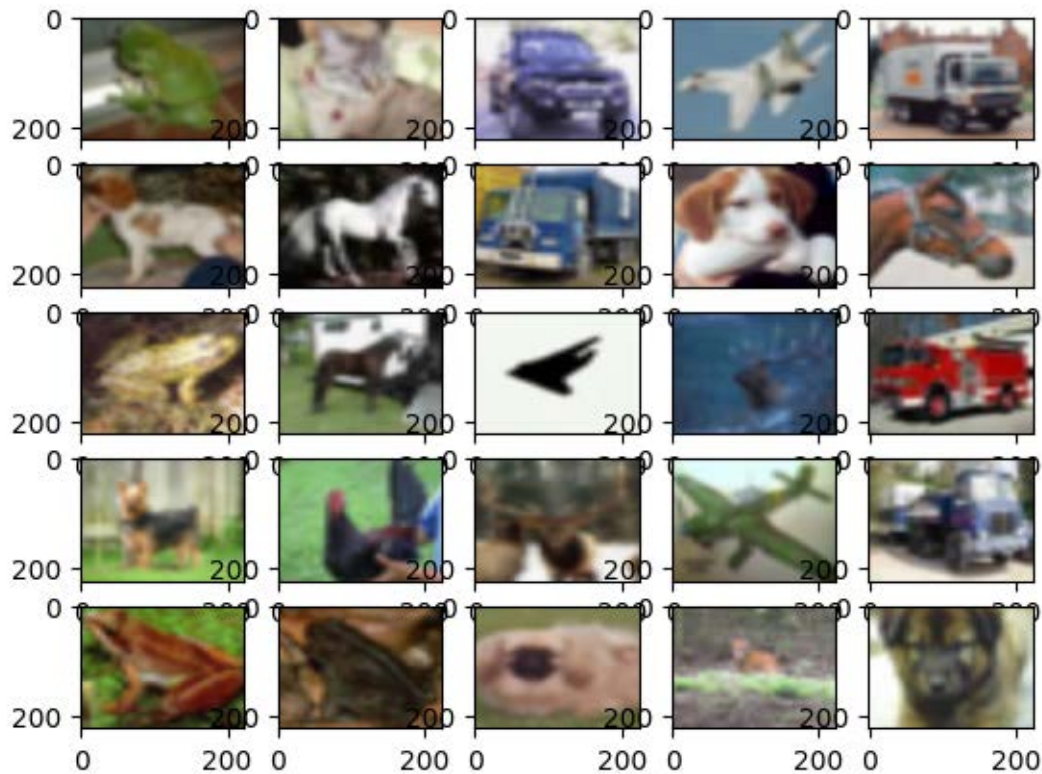
```
[ ]: print(x_train_resize.shape, y_train_norm.shape, x_test_resize.shape,
    →y_test_norm.shape)
```

(45000, 224, 224, 3) (45000, 9) (9000, 224, 224, 3) (9000, 9)

```
[ ]: # visualize data by plotting images
fig, ax = plt.subplots(5, 5)
k = 5

for i in range(5):
    for j in range(5):
        ax[i][j].imshow(x_test_resize[k], aspect='auto')
        k += 1

plt.show()
```



### 1.1.2 PS 1A. Let's build the ResNet50 architecture

### Identity or Skip path Block

```
[ ]: # x is input,  $y=F(x)$ 
      # identity block simply means input should be equal to output.
      #  $y = x + F(x)$  the layers in a traditional network are learning the true ↵
      ↵ output  $H(x)$ 
      #  $F(x) = y - x$  the layers in a residual network are learning the residual  $F(x)$ 
      # Hence, the name: Residual Block.
```

```
def identity_block(X, f, filters, stage, block):
```

///

### Arguments:

$X$  -- input of shape (m, height, width, channel)

*f* -- shape of the middle CONV's window for the main path

`filters` -- python list of integers, defining the number of filters in the

→ CONV layers of the main path

stage -- integer, used to name the layers, depending on their position in

→ the network

```

    block -- string/character, used to name the layers, depending on their
    ↪ position in the network

Returns:
X -- output of the identity block, tensor of shape (n_H, n_W, n_C)
"""

# defining name basis
# conv_name_base = 'res' + str(stage) + block + '_branch'
# bn_name_base = 'bn' + str(stage) + block + '_branch'
conv_name_base = 'conv' + str(stage) + block + '_branch'
bn_name_base = 'bn' + str(stage) + block + '_branch'

# Retrieve Filters
F1, F2, F3 = filters

# Saving the input value. we need this later to add to the output.
X_shortcut = X

# First component of main path
X = Conv2D(filters = F1, kernel_size = (1, 1), strides = (1,1), padding =
    ↪ 'valid', name = conv_name_base + '2a')(X)
X = BatchNormalization(axis = 3, name = bn_name_base + '2a')(X)
X = Activation('relu')(X)

# Second component of main path (3 lines)
X = Conv2D(filters = F2, kernel_size = (f, f), strides = (1,1), padding =
    ↪ 'same', name = conv_name_base + '2b')(X)
X = BatchNormalization(axis = 3, name = bn_name_base + '2b')(X)
X = Activation('relu')(X)

# Third component of main path (2 lines)
X = Conv2D(filters = F3, kernel_size = (1, 1), strides = (1,1), padding =
    ↪ 'valid', name = conv_name_base + '2c')(X)
X = BatchNormalization(axis = 3, name = bn_name_base + '2c')(X)

# Final step: Add shortcut value to main path, and pass it through a RELU
    ↪ activation
X = Add()(X, X_shortcut)
X = Activation('relu')(X)

return X

```

## Convolutional BLock

```
[ ]: def convolutional_block(X, f, filters, stage, block, s = 2):

    # defining name basis
    # conv_name_base = 'res' + str(stage) + block + '_branch'
    # bn_name_base = 'bn' + str(stage) + block + '_branch'

    conv_name_base = 'conv' + str(stage) + block + '_branch'
    bn_name_base = 'bn' + str(stage) + block + '_branch'

    # Retrieve Filters
    F1, F2, F3 = filters

    # Save the input value
    X_shortcut = X

    # First layer
    X = Conv2D(F1, (1, 1), strides = (s,s), name = conv_name_base + '2a')(X) #
    ↪ 1,1 is filter size
    X = BatchNormalization(axis = 3, name = bn_name_base + '2a')(X) #
    ↪ normalization on channels
    X = Activation('relu')(X)

    # Second layer (f,f)=3*3 filter by default
    X = Conv2D(filters = F2, kernel_size = (f, f), strides = (1,1), padding =
    ↪ 'same', name = conv_name_base + '2b')(X)
    X = BatchNormalization(axis = 3, name = bn_name_base + '2b')(X)
    X = Activation('relu')(X)

    # Third layer
    X = Conv2D(filters = F3, kernel_size = (1, 1), strides = (1,1), padding =
    ↪ 'valid', name = conv_name_base + '2c')(X)
    X = BatchNormalization(axis = 3, name = bn_name_base + '2c')(X)

    ##### SHORTCUT PATH #####
    X_shortcut = Conv2D(filters = F3, kernel_size = (1, 1), strides = (s,s),
    ↪ padding = 'valid', name = conv_name_base + '1')(X_shortcut)
    X_shortcut = BatchNormalization(axis = 3, name = bn_name_base +
    ↪ '1')(X_shortcut)

    # Final step: Add shortcut value here, and pass it through a RELU activation
    X = Add()(X, X_shortcut)
    X = Activation('relu')(X)
```



```
return X
```

```
[ ]: #Each ResNet block is either 2 layer deep
def ResNet50custom(input_shape=(64, 64, 3), classes=3):
    """
    Implementation of the ResNet50 architecture:
    CONV2D -> BATCHNORM -> RELU -> MAXPOOL -> CONVBLOCK -> IDBLOCK*2 ->
    ↪CONVBLOCK -> IDBLOCK*3
    -> CONVBLOCK -> IDBLOCK*5 -> CONVBLOCK -> IDBLOCK*2 -> AVGPPOOL -> TOPLAYER

    """

    # Define the input as a tensor with shape input_shape
    X_input = Input(input_shape)

    # Zero-Padding
    X = ZeroPadding2D((3, 3))(X_input) #3,3 padding

    # Stage 1
    X = Conv2D(64, (7, 7), strides=(2, 2), name='conv1')(X) #64 filters of 7*7
    X = BatchNormalization(axis=3, name='bn_conv1')(X) #batchnorm applied on
    ↪channels
    X = Activation('relu')(X)
    X = MaxPooling2D((3, 3), strides=(2, 2))(X) #window size is 3*3

    # Stage 2
    X = convolutional_block(X, f=3, filters=[64, 64, 256], stage=2, block='a',
    ↪s=1)
    # convolutional_block is a function defined above. Convolutional_block have
    ↪3 layers.
    #filters=[64, 64, 256] first 64 is for 1st layer and 2nd 64 is for 2nd
    ↪layer and 256 is for 3rd layer of convultional block
    # below are the conv layers from convolutional_block function
    #X = Conv2D(F1, (1, 1), strides = (s,s), name = conv_name_base + '2a')(X)
    #X = Conv2D(F2, kernel_size = (f, f), strides = (1,1), padding = 'same',
    ↪name = conv_name_base + '2b')(X)
    #X = Conv2D(F3, (1, 1), strides = (s,s), name = conv_name_base + '2a')(X)

    X = identity_block(X, 3, [64, 64, 256], stage=2, block='b')
    #X = Conv2D(filters = F1, kernel_size = (1, 1), strides = (1,1), padding =
    ↪'valid', name = conv_name_base + '2a')(X)
    #X = Conv2D(filters = F2, kernel_size = (f, f), strides = (1,1), padding =
    ↪'same', name = conv_name_base + '2b')(X)
    #X = Conv2D(filters = F3, kernel_size = (1, 1), strides = (1,1), padding =
    ↪'valid', name = conv_name_base + '2c')(X)
```



```

X = identity_block(X, 3, [64, 64, 256], stage=2, block='c')
#X = Conv2D(filters = F1, kernel_size = (1, 1), strides = (1,1), padding =
↳ 'valid', name = conv_name_base + '2a')(X)
#X = Conv2D(filters = F2, kernel_size = (f, f), strides = (1,1), padding =
↳ 'same', name = conv_name_base + '2b')(X)
#X = Conv2D(filters = F3, kernel_size = (1, 1), strides = (1,1), padding =
↳ 'valid', name = conv_name_base + '2c')(X)

### START CODE HERE ###

# Stage 3
X = convolutional_block(X, f = 3, filters = [128, 128, 512], stage = 3,
↳ block='a', s = 2)
X = identity_block(X, 3, [128, 128, 512], stage=3, block='b')
X = identity_block(X, 3, [128, 128, 512], stage=3, block='c')
X = identity_block(X, 3, [128, 128, 512], stage=3, block='d')

# Stage 4
X = convolutional_block(X, f = 3, filters = [256, 256, 1024], stage = 4,
↳ block='a', s = 2)
X = identity_block(X, 3, [256, 256, 1024], stage=4, block='b')
X = identity_block(X, 3, [256, 256, 1024], stage=4, block='c')
X = identity_block(X, 3, [256, 256, 1024], stage=4, block='d')
X = identity_block(X, 3, [256, 256, 1024], stage=4, block='e')
X = identity_block(X, 3, [256, 256, 1024], stage=4, block='f')

# Stage 5
X = convolutional_block(X, f = 3, filters = [512, 512, 2048], stage = 5,
↳ block='a', s = 2)
X = identity_block(X, 3, [512, 512, 2048], stage=5, block='b')
X = identity_block(X, 3, [512, 512, 2048], stage=5, block='c')

# AVGPOOL
X = AveragePooling2D((2,2), name="avg_pool")(X)

### END CODE HERE ###

# output layer
X = Flatten()(X)
X = Dense(classes, activation='softmax', name='fc' + str(classes),
↳ kernel_initializer = glorot_uniform(seed=0))(X)

# Create model
model = Model(inputs = X_input, outputs = X, name='ResNet50')

```

```
return model
```

```
[ ]: im_size = 224
      k = 9
      model = ResNet50custom(input_shape = (im_size, im_size, 3), classes = k)
```

```
[ ]: model.summary()
```

Model: "ResNet50"

```
-----
-----
Layer (type)                 Output Shape              Param #   Connected to
=====
input_1 (InputLayer)        [(None, 224, 224, 3)]     0         []

zero_padding2d (ZeroPaddin  (None, 230, 230, 3)       0         ['input_1[0][0]']
g2D)

conv1 (Conv2D)               (None, 112, 112, 64)      9472      ['zero_padding2d[0][0]']

bn_conv1 (BatchNormalizati  (None, 112, 112, 64)      256      ['conv1[0][0]']
on)

activation (Activation)      (None, 112, 112, 64)      0         ['bn_conv1[0][0]']

max_pooling2d (MaxPooling2  (None, 55, 55, 64)        0         ['activation[0][0]']
D)

conv2a_branch2a (Conv2D)     (None, 55, 55, 64)        4160      ['max_pooling2d[0][0]']

bn2a_branch2a (BatchNormal  (None, 55, 55, 64)        256      ['conv2a_branch2a[0][0]']
ization)

activation_1 (Activation)    (None, 55, 55, 64)        0         ['bn2a_branch2a[0][0]']

conv2a_branch2b (Conv2D)     (None, 55, 55, 64)        36928     ['activation_1[0][0]']
```

bn2a_branch2b (BatchNormal ['conv2a_branch2b[0][0]' ization)	(None, 55, 55, 64)	256
activation_2 (Activation) ['bn2a_branch2b[0][0]']	(None, 55, 55, 64)	0
conv2a_branch2c (Conv2D) ['activation_2[0][0]']	(None, 55, 55, 256)	16640
conv2a_branch1 (Conv2D) ['max_pooling2d[0][0]']	(None, 55, 55, 256)	16640
bn2a_branch2c (BatchNormal ['conv2a_branch2c[0][0]' ization)	(None, 55, 55, 256)	1024
bn2a_branch1 (BatchNormali ['conv2a_branch1[0][0]' ization)	(None, 55, 55, 256)	1024
add (Add) ['bn2a_branch2c[0][0]', 'bn2a_branch1[0][0]']	(None, 55, 55, 256)	0
activation_3 (Activation) ['add[0][0]']	(None, 55, 55, 256)	0
conv2b_branch2a (Conv2D) ['activation_3[0][0]']	(None, 55, 55, 64)	16448
bn2b_branch2a (BatchNormal ['conv2b_branch2a[0][0]' ization)	(None, 55, 55, 64)	256
activation_4 (Activation) ['bn2b_branch2a[0][0]']	(None, 55, 55, 64)	0
conv2b_branch2b (Conv2D) ['activation_4[0][0]']	(None, 55, 55, 64)	36928
bn2b_branch2b (BatchNormal ['conv2b_branch2b[0][0]' ization)	(None, 55, 55, 64)	256
activation_5 (Activation) ['bn2b_branch2b[0][0]']	(None, 55, 55, 64)	0

conv2b_branch2c (Conv2D) ['activation_5[0][0]']	(None, 55, 55, 256)	16640
bn2b_branch2c (BatchNormal ization) ['conv2b_branch2c[0][0]']	(None, 55, 55, 256)	1024
add_1 (Add) ['bn2b_branch2c[0][0]', 'activation_3[0][0]']	(None, 55, 55, 256)	0
activation_6 (Activation) ['add_1[0][0]']	(None, 55, 55, 256)	0
conv2c_branch2a (Conv2D) ['activation_6[0][0]']	(None, 55, 55, 64)	16448
bn2c_branch2a (BatchNormal ization) ['conv2c_branch2a[0][0]']	(None, 55, 55, 64)	256
activation_7 (Activation) ['bn2c_branch2a[0][0]']	(None, 55, 55, 64)	0
conv2c_branch2b (Conv2D) ['activation_7[0][0]']	(None, 55, 55, 64)	36928
bn2c_branch2b (BatchNormal ization) ['conv2c_branch2b[0][0]']	(None, 55, 55, 64)	256
activation_8 (Activation) ['bn2c_branch2b[0][0]']	(None, 55, 55, 64)	0
conv2c_branch2c (Conv2D) ['activation_8[0][0]']	(None, 55, 55, 256)	16640
bn2c_branch2c (BatchNormal ization) ['conv2c_branch2c[0][0]']	(None, 55, 55, 256)	1024
add_2 (Add) ['bn2c_branch2c[0][0]', 'activation_6[0][0]']	(None, 55, 55, 256)	0
activation_9 (Activation) ['add_2[0][0]']	(None, 55, 55, 256)	0

conv3a_branch2a (Conv2D) ['activation_9[0][0]']	(None, 28, 28, 128)	32896
bn3a_branch2a (BatchNormal ization) ['conv3a_branch2a[0][0]']	(None, 28, 28, 128)	512
activation_10 (Activation) ['bn3a_branch2a[0][0]']	(None, 28, 28, 128)	0
conv3a_branch2b (Conv2D) ['activation_10[0][0]']	(None, 28, 28, 128)	147584
bn3a_branch2b (BatchNormal ization) ['conv3a_branch2b[0][0]']	(None, 28, 28, 128)	512
activation_11 (Activation) ['bn3a_branch2b[0][0]']	(None, 28, 28, 128)	0
conv3a_branch2c (Conv2D) ['activation_11[0][0]']	(None, 28, 28, 512)	66048
conv3a_branch1 (Conv2D) ['activation_9[0][0]']	(None, 28, 28, 512)	131584
bn3a_branch2c (BatchNormal ization) ['conv3a_branch2c[0][0]']	(None, 28, 28, 512)	2048
bn3a_branch1 (BatchNormali zation) ['conv3a_branch1[0][0]']	(None, 28, 28, 512)	2048
add_3 (Add) ['bn3a_branch2c[0][0]', 'bn3a_branch1[0][0]']	(None, 28, 28, 512)	0
activation_12 (Activation) ['add_3[0][0]']	(None, 28, 28, 512)	0
conv3b_branch2a (Conv2D) ['activation_12[0][0]']	(None, 28, 28, 128)	65664
bn3b_branch2a (BatchNormal ization) ['conv3b_branch2a[0][0]']	(None, 28, 28, 128)	512

activation_13 (Activation)	(None, 28, 28, 128)	0
['bn3b_branch2a[0][0]']		
conv3b_branch2b (Conv2D)	(None, 28, 28, 128)	147584
['activation_13[0][0]']		
bn3b_branch2b (BatchNormal	(None, 28, 28, 128)	512
ization)		
['conv3b_branch2b[0][0]']		
activation_14 (Activation)	(None, 28, 28, 128)	0
['bn3b_branch2b[0][0]']		
conv3b_branch2c (Conv2D)	(None, 28, 28, 512)	66048
['activation_14[0][0]']		
bn3b_branch2c (BatchNormal	(None, 28, 28, 512)	2048
ization)		
['conv3b_branch2c[0][0]']		
add_4 (Add)	(None, 28, 28, 512)	0
['bn3b_branch2c[0][0]',		
'activation_12[0][0]']		
activation_15 (Activation)	(None, 28, 28, 512)	0
['add_4[0][0]']		
conv3c_branch2a (Conv2D)	(None, 28, 28, 128)	65664
['activation_15[0][0]']		
bn3c_branch2a (BatchNormal	(None, 28, 28, 128)	512
ization)		
['conv3c_branch2a[0][0]']		
activation_16 (Activation)	(None, 28, 28, 128)	0
['bn3c_branch2a[0][0]']		
conv3c_branch2b (Conv2D)	(None, 28, 28, 128)	147584
['activation_16[0][0]']		
bn3c_branch2b (BatchNormal	(None, 28, 28, 128)	512
ization)		
['conv3c_branch2b[0][0]']		
activation_17 (Activation)	(None, 28, 28, 128)	0
['bn3c_branch2b[0][0]']		

conv3c_branch2c (Conv2D) ['activation_17[0][0]']	(None, 28, 28, 512)	66048
bn3c_branch2c (BatchNormal ization) ['conv3c_branch2c[0][0]']	(None, 28, 28, 512)	2048
add_5 (Add) ['bn3c_branch2c[0][0]', 'activation_15[0][0]']	(None, 28, 28, 512)	0
activation_18 (Activation) ['add_5[0][0]']	(None, 28, 28, 512)	0
conv3d_branch2a (Conv2D) ['activation_18[0][0]']	(None, 28, 28, 128)	65664
bn3d_branch2a (BatchNormal ization) ['conv3d_branch2a[0][0]']	(None, 28, 28, 128)	512
activation_19 (Activation) ['bn3d_branch2a[0][0]']	(None, 28, 28, 128)	0
conv3d_branch2b (Conv2D) ['activation_19[0][0]']	(None, 28, 28, 128)	147584
bn3d_branch2b (BatchNormal ization) ['conv3d_branch2b[0][0]']	(None, 28, 28, 128)	512
activation_20 (Activation) ['bn3d_branch2b[0][0]']	(None, 28, 28, 128)	0
conv3d_branch2c (Conv2D) ['activation_20[0][0]']	(None, 28, 28, 512)	66048
bn3d_branch2c (BatchNormal ization) ['conv3d_branch2c[0][0]']	(None, 28, 28, 512)	2048
add_6 (Add) ['bn3d_branch2c[0][0]', 'activation_18[0][0]']	(None, 28, 28, 512)	0
activation_21 (Activation) ['add_6[0][0]']	(None, 28, 28, 512)	0



conv4a_branch2a (Conv2D) ['activation_21[0][0]']	(None, 14, 14, 256)	131328
bn4a_branch2a (BatchNormal ization) ['conv4a_branch2a[0][0]']	(None, 14, 14, 256)	1024
activation_22 (Activation) ['bn4a_branch2a[0][0]']	(None, 14, 14, 256)	0
conv4a_branch2b (Conv2D) ['activation_22[0][0]']	(None, 14, 14, 256)	590080
bn4a_branch2b (BatchNormal ization) ['conv4a_branch2b[0][0]']	(None, 14, 14, 256)	1024
activation_23 (Activation) ['bn4a_branch2b[0][0]']	(None, 14, 14, 256)	0
conv4a_branch2c (Conv2D) ['activation_23[0][0]']	(None, 14, 14, 1024)	263168
conv4a_branch1 (Conv2D) ['activation_21[0][0]']	(None, 14, 14, 1024)	525312
bn4a_branch2c (BatchNormal ization) ['conv4a_branch2c[0][0]']	(None, 14, 14, 1024)	4096
bn4a_branch1 (BatchNormali zation) ['conv4a_branch1[0][0]']	(None, 14, 14, 1024)	4096
add_7 (Add) ['bn4a_branch2c[0][0]', 'bn4a_branch1[0][0]']	(None, 14, 14, 1024)	0
activation_24 (Activation) ['add_7[0][0]']	(None, 14, 14, 1024)	0
conv4b_branch2a (Conv2D) ['activation_24[0][0]']	(None, 14, 14, 256)	262400
bn4b_branch2a (BatchNormal ization) ['conv4b_branch2a[0][0]']	(None, 14, 14, 256)	1024

activation_25 (Activation)	(None, 14, 14, 256)	0
['bn4b_branch2a[0][0]']		
conv4b_branch2b (Conv2D)	(None, 14, 14, 256)	590080
['activation_25[0][0]']		
bn4b_branch2b (BatchNormal	(None, 14, 14, 256)	1024
['conv4b_branch2b[0][0]']		
ization)		
activation_26 (Activation)	(None, 14, 14, 256)	0
['bn4b_branch2b[0][0]']		
conv4b_branch2c (Conv2D)	(None, 14, 14, 1024)	263168
['activation_26[0][0]']		
bn4b_branch2c (BatchNormal	(None, 14, 14, 1024)	4096
['conv4b_branch2c[0][0]']		
ization)		
add_8 (Add)	(None, 14, 14, 1024)	0
['bn4b_branch2c[0][0]',		
'activation_24[0][0]']		
activation_27 (Activation)	(None, 14, 14, 1024)	0
['add_8[0][0]']		
conv4c_branch2a (Conv2D)	(None, 14, 14, 256)	262400
['activation_27[0][0]']		
bn4c_branch2a (BatchNormal	(None, 14, 14, 256)	1024
['conv4c_branch2a[0][0]']		
ization)		
activation_28 (Activation)	(None, 14, 14, 256)	0
['bn4c_branch2a[0][0]']		
conv4c_branch2b (Conv2D)	(None, 14, 14, 256)	590080
['activation_28[0][0]']		
bn4c_branch2b (BatchNormal	(None, 14, 14, 256)	1024
['conv4c_branch2b[0][0]']		
ization)		
activation_29 (Activation)	(None, 14, 14, 256)	0
['bn4c_branch2b[0][0]']		
conv4c_branch2c (Conv2D)	(None, 14, 14, 1024)	263168

['activation_29[0][0]']		
bn4c_branch2c (BatchNormal ['conv4c_branch2c[0][0]'] ization)	(None, 14, 14, 1024)	4096
add_9 (Add) ['bn4c_branch2c[0][0]', 'activation_27[0][0]']	(None, 14, 14, 1024)	0
activation_30 (Activation) ['add_9[0][0]']	(None, 14, 14, 1024)	0
conv4d_branch2a (Conv2D) ['activation_30[0][0]']	(None, 14, 14, 256)	262400
bn4d_branch2a (BatchNormal ['conv4d_branch2a[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_31 (Activation) ['bn4d_branch2a[0][0]']	(None, 14, 14, 256)	0
conv4d_branch2b (Conv2D) ['activation_31[0][0]']	(None, 14, 14, 256)	590080
bn4d_branch2b (BatchNormal ['conv4d_branch2b[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_32 (Activation) ['bn4d_branch2b[0][0]']	(None, 14, 14, 256)	0
conv4d_branch2c (Conv2D) ['activation_32[0][0]']	(None, 14, 14, 1024)	263168
bn4d_branch2c (BatchNormal ['conv4d_branch2c[0][0]'] ization)	(None, 14, 14, 1024)	4096
add_10 (Add) ['bn4d_branch2c[0][0]', 'activation_30[0][0]']	(None, 14, 14, 1024)	0
activation_33 (Activation) ['add_10[0][0]']	(None, 14, 14, 1024)	0
conv4e_branch2a (Conv2D)	(None, 14, 14, 256)	262400

['activation_33[0][0]']		
bn4e_branch2a (BatchNormal ['conv4e_branch2a[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_34 (Activation) ['bn4e_branch2a[0][0]']	(None, 14, 14, 256)	0
conv4e_branch2b (Conv2D) ['activation_34[0][0]']	(None, 14, 14, 256)	590080
bn4e_branch2b (BatchNormal ['conv4e_branch2b[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_35 (Activation) ['bn4e_branch2b[0][0]']	(None, 14, 14, 256)	0
conv4e_branch2c (Conv2D) ['activation_35[0][0]']	(None, 14, 14, 1024)	263168
bn4e_branch2c (BatchNormal ['conv4e_branch2c[0][0]'] ization)	(None, 14, 14, 1024)	4096
add_11 (Add) ['bn4e_branch2c[0][0]', 'activation_33[0][0]']	(None, 14, 14, 1024)	0
activation_36 (Activation) ['add_11[0][0]']	(None, 14, 14, 1024)	0
conv4f_branch2a (Conv2D) ['activation_36[0][0]']	(None, 14, 14, 256)	262400
bn4f_branch2a (BatchNormal ['conv4f_branch2a[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_37 (Activation) ['bn4f_branch2a[0][0]']	(None, 14, 14, 256)	0
conv4f_branch2b (Conv2D) ['activation_37[0][0]']	(None, 14, 14, 256)	590080
bn4f_branch2b (BatchNormal ['conv4f_branch2b[0][0]']	(None, 14, 14, 256)	1024

ization)		
activation_38 (Activation) ['bn4f_branch2b[0][0]']	(None, 14, 14, 256)	0
conv4f_branch2c (Conv2D) ['activation_38[0][0]']	(None, 14, 14, 1024)	263168
bn4f_branch2c (BatchNormal ['conv4f_branch2c[0][0]'] ization)	(None, 14, 14, 1024)	4096
add_12 (Add) ['bn4f_branch2c[0][0]', 'activation_36[0][0]']	(None, 14, 14, 1024)	0
activation_39 (Activation) ['add_12[0][0]']	(None, 14, 14, 1024)	0
conv5a_branch2a (Conv2D) ['activation_39[0][0]']	(None, 7, 7, 512)	524800
bn5a_branch2a (BatchNormal ['conv5a_branch2a[0][0]'] ization)	(None, 7, 7, 512)	2048
activation_40 (Activation) ['bn5a_branch2a[0][0]']	(None, 7, 7, 512)	0
conv5a_branch2b (Conv2D) ['activation_40[0][0]']	(None, 7, 7, 512)	2359808
bn5a_branch2b (BatchNormal ['conv5a_branch2b[0][0]'] ization)	(None, 7, 7, 512)	2048
activation_41 (Activation) ['bn5a_branch2b[0][0]']	(None, 7, 7, 512)	0
conv5a_branch2c (Conv2D) ['activation_41[0][0]']	(None, 7, 7, 2048)	1050624
conv5a_branch1 (Conv2D) ['activation_39[0][0]']	(None, 7, 7, 2048)	2099200
bn5a_branch2c (BatchNormal ['conv5a_branch2c[0][0]'] ization)	(None, 7, 7, 2048)	8192

bn5a_branch1 (BatchNormali ['conv5a_branch1[0][0]' zation)	(None, 7, 7, 2048)	8192
add_13 (Add) ['bn5a_branch2c[0][0]' 'bn5a_branch1[0][0]']	(None, 7, 7, 2048)	0
activation_42 (Activation) ['add_13[0][0]']	(None, 7, 7, 2048)	0
conv5b_branch2a (Conv2D) ['activation_42[0][0]']	(None, 7, 7, 512)	1049088
bn5b_branch2a (BatchNormal ['conv5b_branch2a[0][0]' ization)	(None, 7, 7, 512)	2048
activation_43 (Activation) ['bn5b_branch2a[0][0]']	(None, 7, 7, 512)	0
conv5b_branch2b (Conv2D) ['activation_43[0][0]']	(None, 7, 7, 512)	2359808
bn5b_branch2b (BatchNormal ['conv5b_branch2b[0][0]' ization)	(None, 7, 7, 512)	2048
activation_44 (Activation) ['bn5b_branch2b[0][0]']	(None, 7, 7, 512)	0
conv5b_branch2c (Conv2D) ['activation_44[0][0]']	(None, 7, 7, 2048)	1050624
bn5b_branch2c (BatchNormal ['conv5b_branch2c[0][0]' ization)	(None, 7, 7, 2048)	8192
add_14 (Add) ['bn5b_branch2c[0][0]' 'activation_42[0][0]']	(None, 7, 7, 2048)	0
activation_45 (Activation) ['add_14[0][0]']	(None, 7, 7, 2048)	0
conv5c_branch2a (Conv2D) ['activation_45[0][0]']	(None, 7, 7, 512)	1049088

bn5c_branch2a (BatchNormal ['conv5c_branch2a[0][0]' ization)	(None, 7, 7, 512)	2048
activation_46 (Activation) ['bn5c_branch2a[0][0]']	(None, 7, 7, 512)	0
conv5c_branch2b (Conv2D) ['activation_46[0][0]']	(None, 7, 7, 512)	2359808
bn5c_branch2b (BatchNormal ['conv5c_branch2b[0][0]' ization)	(None, 7, 7, 512)	2048
activation_47 (Activation) ['bn5c_branch2b[0][0]']	(None, 7, 7, 512)	0
conv5c_branch2c (Conv2D) ['activation_47[0][0]']	(None, 7, 7, 2048)	1050624
bn5c_branch2c (BatchNormal ['conv5c_branch2c[0][0]' ization)	(None, 7, 7, 2048)	8192
add_15 (Add) ['bn5c_branch2c[0][0] ', 'activation_45[0][0]']	(None, 7, 7, 2048)	0
activation_48 (Activation) ['add_15[0][0]']	(None, 7, 7, 2048)	0
avg_pool (AveragePooling2D ['activation_48[0][0]'] )	(None, 3, 3, 2048)	0
flatten (Flatten) ['avg_pool[0][0]']	(None, 18432)	0
fc9 (Dense) ['flatten[0][0]']	(None, 9)	165897

```
=====
=====
Total params: 23753609 (90.61 MB)
Trainable params: 23700489 (90.41 MB)
Non-trainable params: 53120 (207.50 KB)
```



### 1.1.3 PS1A. ResNet50 non custom

```
[ ]: # Load ResNet-50 with custom input size
im_size = 224

base_model = ResNet50(input_shape=(im_size, im_size, 3), include_top=False,
    weights=None)

# Adjust the feature map size before average pooling
x = base_model.layers[-1].output
x = AveragePooling2D(pool_size=(2, 2))(x) # Reduce feature map size to 1x1
x = Flatten()(x)
output = Dense(9, activation='softmax')(x)

# Create a new model
model = Model(inputs=base_model.input, outputs=output)
```

```
[ ]: model.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	[]
conv1_pad (ZeroPadding2D) ['input_1[0][0]']	(None, 230, 230, 3)	0	
conv1_conv (Conv2D) ['conv1_pad[0][0]']	(None, 112, 112, 64)	9472	
conv1_bn (BatchNormalizati on) ['conv1_conv[0][0]']	(None, 112, 112, 64)	256	
conv1_relu (Activation) ['conv1_bn[0][0]']	(None, 112, 112, 64)	0	
pool1_pad (ZeroPadding2D) ['conv1_relu[0][0]']	(None, 114, 114, 64)	0	
pool1_pool (MaxPooling2D) ['pool1_pad[0][0]']	(None, 56, 56, 64)	0	

conv2_block1_1_conv (Conv2D) ['pool1_pool[0][0]']	(None, 56, 56, 64)	4160
conv2_block1_1_bn (Batch Normalization) ['conv2_block1_1_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block1_1_relu (Activation) ['conv2_block1_1_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block1_2_conv (Conv2D) ['conv2_block1_1_relu[0][0]']	(None, 56, 56, 64)	36928
conv2_block1_2_bn (Batch Normalization) ['conv2_block1_2_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block1_2_relu (Activation) ['conv2_block1_2_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block1_0_conv (Conv2D) ['pool1_pool[0][0]']	(None, 56, 56, 256)	16640
conv2_block1_3_conv (Conv2D) ['conv2_block1_2_relu[0][0]']	(None, 56, 56, 256)	16640
conv2_block1_0_bn (Batch Normalization) ['conv2_block1_0_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block1_3_bn (Batch Normalization) ['conv2_block1_3_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block1_add (Add) ['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block1_out (Activation) ['conv2_block1_add[0][0]']	(None, 56, 56, 256)	0

conv2_block2_1_conv (Conv2D) ['conv2_block1_out[0][0]']	(None, 56, 56, 64)	16448
conv2_block2_1_bn (Batch Normalization) ['conv2_block2_1_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block2_1_relu (Activation) ['conv2_block2_1_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block2_2_conv (Conv2D) ['conv2_block2_1_relu[0][0]']	(None, 56, 56, 64)	36928
conv2_block2_2_bn (Batch Normalization) ['conv2_block2_2_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block2_2_relu (Activation) ['conv2_block2_2_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block2_3_conv (Conv2D) ['conv2_block2_2_relu[0][0]']	(None, 56, 56, 256)	16640
conv2_block2_3_bn (Batch Normalization) ['conv2_block2_3_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block2_add (Add) ['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block2_out (Activation) ['conv2_block2_add[0][0]']	(None, 56, 56, 256)	0
conv2_block3_1_conv (Conv2D) ['conv2_block2_out[0][0]']	(None, 56, 56, 64)	16448
conv2_block3_1_bn (Batch Normalization) ['conv2_block3_1_conv[0][0]']	(None, 56, 56, 64)	256

conv2_block3_1_relu (Activation) ['conv2_block3_1_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block3_2_conv (Conv2D) ['conv2_block3_1_relu[0][0]']	(None, 56, 56, 64)	36928
conv2_block3_2_bn (BatchNormalization) ['conv2_block3_2_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block3_2_relu (Activation) ['conv2_block3_2_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block3_3_conv (Conv2D) ['conv2_block3_2_relu[0][0]']	(None, 56, 56, 256)	16640
conv2_block3_3_bn (BatchNormalization) ['conv2_block3_3_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block3_add (Add) ['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block3_out (Activation) ['conv2_block3_add[0][0]']	(None, 56, 56, 256)	0
conv3_block1_1_conv (Conv2D) ['conv2_block3_out[0][0]']	(None, 28, 28, 128)	32896
conv3_block1_1_bn (BatchNormalization) ['conv3_block1_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block1_1_relu (Activation) ['conv3_block1_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block1_2_conv (Conv2D) ['conv3_block1_1_relu[0][0]']	(None, 28, 28, 128)	147584

conv3_block1_2_bn (BatchNormal- ization) ['conv3_block1_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block1_2_relu (Activation) ['conv3_block1_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block1_0_conv (Conv2D) ['conv2_block3_out[0][0]']	(None, 28, 28, 512)	131584
conv3_block1_3_conv (Conv2D) ['conv3_block1_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block1_0_bn (BatchNormal- ization) ['conv3_block1_0_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block1_3_bn (BatchNormal- ization) ['conv3_block1_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block1_add (Add) ['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block1_out (Activation) ['conv3_block1_add[0][0]']	(None, 28, 28, 512)	0
conv3_block2_1_conv (Conv2D) ['conv3_block1_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block2_1_bn (BatchNormal- ization) ['conv3_block2_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block2_1_relu (Activation) ['conv3_block2_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block2_2_conv (Conv2D) ['conv3_block2_1_relu[0][0]']	(None, 28, 28, 128)	147584

conv3_block2_2_bn (BatchNormal ization)	(None, 28, 28, 128)	512
conv3_block2_2_relu (Activation)	(None, 28, 28, 128)	0
conv3_block2_3_conv (Conv2D)	(None, 28, 28, 512)	66048
conv3_block2_3_bn (BatchNormal ization)	(None, 28, 28, 512)	2048
conv3_block2_add (Add)	(None, 28, 28, 512)	0
conv3_block2_out (Activation)	(None, 28, 28, 512)	0
conv3_block3_1_conv (Conv2D)	(None, 28, 28, 128)	65664
conv3_block3_1_bn (BatchNormal ization)	(None, 28, 28, 128)	512
conv3_block3_1_relu (Activation)	(None, 28, 28, 128)	0
conv3_block3_2_conv (Conv2D)	(None, 28, 28, 128)	147584
conv3_block3_2_bn (BatchNormal ization)	(None, 28, 28, 128)	512
conv3_block3_2_relu (Activation)	(None, 28, 28, 128)	0

conv3_block3_3_conv (Conv2D) ['conv3_block3_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block3_3_bn (Batch Normalization) ['conv3_block3_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block3_add (Add) ['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block3_out (Activation) ['conv3_block3_add[0][0]']	(None, 28, 28, 512)	0
conv3_block4_1_conv (Conv2D) ['conv3_block3_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block4_1_bn (Batch Normalization) ['conv3_block4_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_1_relu (Activation) ['conv3_block4_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_2_conv (Conv2D) ['conv3_block4_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block4_2_bn (Batch Normalization) ['conv3_block4_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_2_relu (Activation) ['conv3_block4_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_3_conv (Conv2D) ['conv3_block4_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block4_3_bn (Batch Normalization) ['conv3_block4_3_conv[0][0]']	(None, 28, 28, 512)	2048



conv3_block4_add (Add)	(None, 28, 28, 512)	0
['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']		
conv3_block4_out (Activation)	(None, 28, 28, 512)	0
['conv3_block4_add[0][0]']		
conv4_block1_1_conv (Conv2D)	(None, 14, 14, 256)	131328
['conv3_block4_out[0][0]']		
conv4_block1_1_bn (Batch Normalization)	(None, 14, 14, 256)	1024
['conv4_block1_1_conv[0][0]']		
conv4_block1_1_relu (Activation)	(None, 14, 14, 256)	0
['conv4_block1_1_bn[0][0]']		
conv4_block1_2_conv (Conv2D)	(None, 14, 14, 256)	590080
['conv4_block1_1_relu[0][0]']		
conv4_block1_2_bn (Batch Normalization)	(None, 14, 14, 256)	1024
['conv4_block1_2_conv[0][0]']		
conv4_block1_2_relu (Activation)	(None, 14, 14, 256)	0
['conv4_block1_2_bn[0][0]']		
conv4_block1_0_conv (Conv2D)	(None, 14, 14, 1024)	525312
['conv3_block4_out[0][0]']		
conv4_block1_3_conv (Conv2D)	(None, 14, 14, 1024)	263168
['conv4_block1_2_relu[0][0]']		
conv4_block1_0_bn (Batch Normalization)	(None, 14, 14, 1024)	4096
['conv4_block1_0_conv[0][0]']		
conv4_block1_3_bn (Batch Normalization)	(None, 14, 14, 1024)	4096
['conv4_block1_3_conv[0][0]']		

conv4_block1_add (Add) ['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block1_out (Activation) ['conv4_block1_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block2_1_conv (Conv2D) ['conv4_block1_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block2_1_bn (Batch Normalization) ['conv4_block2_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block2_1_relu (Activation) ['conv4_block2_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block2_2_conv (Conv2D) ['conv4_block2_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block2_2_bn (Batch Normalization) ['conv4_block2_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block2_2_relu (Activation) ['conv4_block2_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block2_3_conv (Conv2D) ['conv4_block2_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block2_3_bn (Batch Normalization) ['conv4_block2_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block2_add (Add) ['conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block2_out (Activation) ['conv4_block2_add[0][0]']	(None, 14, 14, 1024)	0

conv4_block3_1_conv (Conv2D) ['conv4_block2_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block3_1_bn (Batch Normalization) ['conv4_block3_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block3_1_relu (Activation) ['conv4_block3_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block3_2_conv (Conv2D) ['conv4_block3_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block3_2_bn (Batch Normalization) ['conv4_block3_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block3_2_relu (Activation) ['conv4_block3_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block3_3_conv (Conv2D) ['conv4_block3_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block3_3_bn (Batch Normalization) ['conv4_block3_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block3_add (Add) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block3_out (Activation) ['conv4_block3_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block4_1_conv (Conv2D) ['conv4_block3_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block4_1_bn (Batch Normalization) ['conv4_block4_1_conv[0][0]']	(None, 14, 14, 256)	1024

conv4_block4_1_relu (Activation) (None, 14, 14, 256)	0
conv4_block4_2_conv (Conv2D) (None, 14, 14, 256)	590080
conv4_block4_2_bn (Batch Normalization) (None, 14, 14, 256)	1024
conv4_block4_2_relu (Activation) (None, 14, 14, 256)	0
conv4_block4_3_conv (Conv2D) (None, 14, 14, 1024)	263168
conv4_block4_3_bn (Batch Normalization) (None, 14, 14, 1024)	4096
conv4_block4_add (Add) (None, 14, 14, 1024)	0
conv4_block4_out (Activation) (None, 14, 14, 1024)	0
conv4_block5_1_conv (Conv2D) (None, 14, 14, 256)	262400
conv4_block5_1_bn (Batch Normalization) (None, 14, 14, 256)	1024
conv4_block5_1_relu (Activation) (None, 14, 14, 256)	0
conv4_block5_2_conv (Conv2D) (None, 14, 14, 256)	590080

conv4_block5_2_bn (BatchNormal ['conv4_block5_2_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block5_2_relu (Activation ['conv4_block5_2_bn[0][0]' ation)	(None, 14, 14, 256)	0
conv4_block5_3_conv (Conv2D ['conv4_block5_2_relu[0][0]' D)	(None, 14, 14, 1024)	263168
conv4_block5_3_bn (BatchNormal ['conv4_block5_3_conv[0][0]' rmalization)	(None, 14, 14, 1024)	4096
conv4_block5_add (Add ['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]')	(None, 14, 14, 1024)	0
conv4_block5_out (Activation ['conv4_block5_add[0][0]' on)	(None, 14, 14, 1024)	0
conv4_block6_1_conv (Conv2D ['conv4_block5_out[0][0]' D)	(None, 14, 14, 256)	262400
conv4_block6_1_bn (BatchNormal ['conv4_block6_1_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block6_1_relu (Activation ['conv4_block6_1_bn[0][0]' ation)	(None, 14, 14, 256)	0
conv4_block6_2_conv (Conv2D ['conv4_block6_1_relu[0][0]' D)	(None, 14, 14, 256)	590080
conv4_block6_2_bn (BatchNormal ['conv4_block6_2_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block6_2_relu (Activation ['conv4_block6_2_bn[0][0]' ation)	(None, 14, 14, 256)	0

conv4_block6_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block6_2_relu[0][0]'] D)	263168
conv4_block6_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block6_3_conv[0][0]'] rmalization)	4096
conv4_block6_add (Add) (None, 14, 14, 1024) ['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']	0
conv4_block6_out (Activati (None, 14, 14, 1024) ['conv4_block6_add[0][0]'] on)	0
conv5_block1_1_conv (Conv2 (None, 7, 7, 512) ['conv4_block6_out[0][0]'] D)	524800
conv5_block1_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block1_1_conv[0][0]'] rmalization)	2048
conv5_block1_1_relu (Activ (None, 7, 7, 512) ['conv5_block1_1_bn[0][0]'] ation)	0
conv5_block1_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block1_1_relu[0][0]'] D)	2359808
conv5_block1_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block1_2_conv[0][0]'] rmalization)	2048
conv5_block1_2_relu (Activ (None, 7, 7, 512) ['conv5_block1_2_bn[0][0]'] ation)	0
conv5_block1_0_conv (Conv2 (None, 7, 7, 2048) ['conv4_block6_out[0][0]'] D)	2099200
conv5_block1_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block1_2_relu[0][0]'] D)	1050624

conv5_block1_0_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_0_conv[0][0]'] rmalization)	8192
conv5_block1_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_3_conv[0][0]'] rmalization)	8192
conv5_block1_add (Add) (None, 7, 7, 2048) ['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']	0
conv5_block1_out (Activati (None, 7, 7, 2048) ['conv5_block1_add[0][0]'] on)	0
conv5_block2_1_conv (Conv2 (None, 7, 7, 512) ['conv5_block1_out[0][0]'] D)	1049088
conv5_block2_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_1_conv[0][0]'] rmalization)	2048
conv5_block2_1_relu (Activ (None, 7, 7, 512) ['conv5_block2_1_bn[0][0]'] ation)	0
conv5_block2_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block2_1_relu[0][0]'] D)	2359808
conv5_block2_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_2_conv[0][0]'] rmalization)	2048
conv5_block2_2_relu (Activ (None, 7, 7, 512) ['conv5_block2_2_bn[0][0]'] ation)	0
conv5_block2_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block2_2_relu[0][0]'] D)	1050624
conv5_block2_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block2_3_conv[0][0]'] rmalization)	8192



conv5_block2_add (Add) ['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']	(None, 7, 7, 2048)	0
conv5_block2_out (Activation) ['conv5_block2_add[0][0]']	(None, 7, 7, 2048)	0
conv5_block3_1_conv (Conv2D) ['conv5_block2_out[0][0]']	(None, 7, 7, 512)	1049088
conv5_block3_1_bn (Batch Normalization) ['conv5_block3_1_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block3_1_relu (Activation) ['conv5_block3_1_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block3_2_conv (Conv2D) ['conv5_block3_1_relu[0][0]']	(None, 7, 7, 512)	2359808
conv5_block3_2_bn (Batch Normalization) ['conv5_block3_2_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block3_2_relu (Activation) ['conv5_block3_2_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block3_3_conv (Conv2D) ['conv5_block3_2_relu[0][0]']	(None, 7, 7, 2048)	1050624
conv5_block3_3_bn (Batch Normalization) ['conv5_block3_3_conv[0][0]']	(None, 7, 7, 2048)	8192
conv5_block3_add (Add) ['conv5_block2_out[0][0]', 'conv5_block3_3_bn[0][0]']	(None, 7, 7, 2048)	0
conv5_block3_out (Activation) ['conv5_block3_add[0][0]']	(None, 7, 7, 2048)	0

```

average_pooling2d (Average Pooling2D) (None, 3, 3, 2048) 0
['conv5_block3_out[0][0]']

flatten (Flatten) (None, 18432) 0
['average_pooling2d[0][0]']

dense (Dense) (None, 9) 165897
['flatten[0][0]']

```

```

=====
Total params: 23753609 (90.61 MB)
Trainable params: 23700489 (90.41 MB)
Non-trainable params: 53120 (207.50 KB)
-----

```

```

[ ]: # Set hyperparameters
learning_rate = 0.001
beta1 = 0.9
beta2 = 0.999
epsilon = 1e-8
num_iterations = 1000
# optimizer = Adam(learning_rate=learning_rate, beta_1=beta1, beta_2=beta2,
    ↪epsilon=1e-8)
optimizer = SGD(learning_rate=learning_rate, momentum=0.9, nesterov=True)
model.compile(optimizer=optimizer, loss='categorical_crossentropy',
    ↪metrics=['accuracy'])

```

WARNING:absl:At this time, the v2.11+ optimizer `tf.keras.optimizers.SGD` runs slowly on M1/M2 Macs, please use the legacy Keras optimizer instead, located at `tf.keras.optimizers.legacy.SGD`.

```

[ ]: # Define a custom callback to track validation loss
class ValidationLoss(tf.keras.callbacks.Callback):
    def __init__(self):
        super(ValidationLoss, self).__init__()
        self.validation_losses = []

    def on_epoch_end(self, epoch, logs=None):
        self.validation_losses.append(logs['val_loss'])

# Define the custom learning rate scheduler
class CustomLRScheduler(tf.keras.callbacks.Callback):
    def on_epoch_begin(self, epoch, logs=None):
        new_lr = 0.001 / np.sqrt(epoch+1)

```

```
K.set_value(self.model.optimizer.lr, new_lr)
print(f'Epoch {epoch + 1}: Learning Rate = {new_lr}')
```

```
[ ]: # x_train_resize.shape, y_train_norm.shape, x_test_resize.shape, y_test_norm.
      ↪shape
# model.fit(x_train_resize, x_train_resize, epochs = 10, batch_size = 128)
# model.fit(x_train_resize, y_train_norm, epochs=45, batch_size = 128,
      ↪validation_data=(x_test_resize, x_test_resize), verbose=0)
# Create a custom callback to track validation loss
val_loss_callback = ValidationLoss()
lr_scheduler = CustomLRScheduler()

model.fit(
    x=x_train_resize, # Your training data
    y=y_train_norm, # Your training labels
    batch_size=64,
    epochs=5,
    steps_per_epoch=10,
    validation_data=(x_test_resize, y_test_norm),
    callbacks=[val_loss_callback, lr_scheduler]
    # Other parameters like callbacks, verbose, etc.
)

model.save('model_my_custom_resnet50_aug2.h5')
```

```
Epoch 1: Learning Rate = 0.001
Epoch 1/5
10/10 [=====] - 245s 26s/step - loss: 2.6025 -
accuracy: 0.3016 - val_loss: 10.6501 - val_accuracy: 0.1303
Epoch 2: Learning Rate = 0.0007071067811865475
Epoch 2/5
10/10 [=====] - 249s 27s/step - loss: 2.8719 -
accuracy: 0.2797 - val_loss: 4.2490 - val_accuracy: 0.1624
Epoch 3: Learning Rate = 0.0005773502691896258
Epoch 3/5
10/10 [=====] - 244s 26s/step - loss: 2.9668 -
accuracy: 0.2578 - val_loss: 3.3308 - val_accuracy: 0.1773
Epoch 4: Learning Rate = 0.0005
Epoch 4/5
10/10 [=====] - 254s 27s/step - loss: 2.1763 -
accuracy: 0.3219 - val_loss: 5.5174 - val_accuracy: 0.1897
Epoch 5: Learning Rate = 0.0004472135954999579
Epoch 5/5
10/10 [=====] - 245s 26s/step - loss: 2.5679 -
accuracy: 0.2703 - val_loss: 4.3503 - val_accuracy: 0.1356
```

/Users/banani/Library/Python/3.9/lib/python/site-

```
packages/keras/src/engine/training.py:3079: UserWarning: You are saving your
model as an HDF5 file via `model.save()`. This file format is considered legacy.
We recommend using instead the native Keras format, e.g.
`model.save('my_model.keras')`.
    saving_api.save_model(
```

```
[ ]: model.save('model_resnet50_lr3.h5')
```

```
[ ]: ## Data augmentation

batch_size = 32
data_generator = tf.keras.preprocessing.image.ImageDataGenerator(
    width_shift_range=0.1, height_shift_range=0.1, horizontal_flip=True)

train_generator = data_generator.flow(x_train_resize,
                                     y_train_norm,
                                     batch_size)

steps_per_epoch = x_train_resize.shape[0] // batch_size
```

```
[ ]: r = model.fit_generator(train_generator,
                           validation_data=(x_test_resize, y_test_norm),
                           batch_size=64,
                           epochs=5,
                           steps_per_epoch=10,
                           callbacks=[val_loss_callback, lr_scheduler]
                           )

model.save('model_resnet50_aug4.h5')
```

```
Epoch 1: Learning Rate = 0.001
Epoch 1/5
10/10 [=====] - 219s 24s/step - loss: 3.6517 -
accuracy: 0.2594 - val_loss: 4.6119 - val_accuracy: 0.1868
Epoch 2: Learning Rate = 0.0007071067811865475
Epoch 2/5
10/10 [=====] - 214s 23s/step - loss: 4.8421 -
accuracy: 0.2062 - val_loss: 2.7708 - val_accuracy: 0.1488
Epoch 3: Learning Rate = 0.0005773502691896258
Epoch 3/5
10/10 [=====] - 217s 24s/step - loss: 4.4187 -
accuracy: 0.2562 - val_loss: 2.0583 - val_accuracy: 0.2217
Epoch 4: Learning Rate = 0.0005
Epoch 4/5
10/10 [=====] - 219s 24s/step - loss: 3.4729 -
accuracy: 0.2562 - val_loss: 2.2416 - val_accuracy: 0.1591
Epoch 5: Learning Rate = 0.0004472135954999579
Epoch 5/5
10/10 [=====] - 217s 24s/step - loss: 2.9785 -
```

accuracy: 0.3063 - val\_loss: 2.1482 - val\_accuracy: 0.1852

```
[ ]: r = model.fit(train_generator,
                  validation_data=(x_test_resize, y_test_norm),
                  batch_size=64,
                  epochs=10,
                  steps_per_epoch=10,
                  callbacks=[val_loss_callback, lr_scheduler]
                  )

model.save('model_resnet50_aug3.h5')
```

```
Epoch 1/10
10/10 [=====] - 314s 35s/step - loss: 2.0968 -
accuracy: 0.2250 - val_loss: 2.2848 - val_accuracy: 0.2092
Epoch 2/10
10/10 [=====] - 201s 22s/step - loss: 2.0053 -
accuracy: 0.2594 - val_loss: 2.4074 - val_accuracy: 0.2038
Epoch 3/10
10/10 [=====] - 202s 22s/step - loss: 2.1640 -
accuracy: 0.2188 - val_loss: 2.2850 - val_accuracy: 0.2204
Epoch 4/10
10/10 [=====] - 205s 22s/step - loss: 2.2046 -
accuracy: 0.2281 - val_loss: 2.3327 - val_accuracy: 0.2210
Epoch 5/10
10/10 [=====] - 213s 23s/step - loss: 2.0892 -
accuracy: 0.2594 - val_loss: 2.2663 - val_accuracy: 0.2052
Epoch 6/10
10/10 [=====] - 199s 22s/step - loss: 2.1547 -
accuracy: 0.2313 - val_loss: 2.2064 - val_accuracy: 0.2120
Epoch 7/10
10/10 [=====] - 200s 22s/step - loss: 2.0995 -
accuracy: 0.2781 - val_loss: 2.2011 - val_accuracy: 0.2298
Epoch 8/10
10/10 [=====] - 198s 22s/step - loss: 2.0862 -
accuracy: 0.2281 - val_loss: 2.2361 - val_accuracy: 0.2410
Epoch 9/10
10/10 [=====] - 201s 22s/step - loss: 1.9848 -
accuracy: 0.2812 - val_loss: 2.3769 - val_accuracy: 0.2199
Epoch 10/10
10/10 [=====] - 202s 22s/step - loss: 2.1150 -
accuracy: 0.2031 - val_loss: 2.3380 - val_accuracy: 0.2232
```

#### 1.1.4 Predict the architecture

```
[ ]: predictions = model.predict(x_test_resize)
```

```
282/282 [=====] - 190s 671ms/step
```

```
[ ]: # Visualize predictions on test data
predicted_labels = (predictions > 0.5).astype(int)
sample_images = x_test_resize[:10]
sample_labels = y_test_norm[:10]

[ ]: for i in range(10):
    plt.subplot(2, 5, i + 1)
    plt.imshow(sample_images[i])
    plt.title(f'Predicted: {predicted_labels[i][0]}\nActual: {sample_labels[i][0]}')
    plt.axis('off')

plt.show()
```

Predicted: 0 Predicted: 0 Predicted: 0 Predicted: 0 Predicted: 0  
 Actual: 0.0 Actual: 1.0 Actual: 0.0 Actual: 0.0 Actual: 0.0



Predicted: 0 Predicted: 0 Predicted: 0 Predicted: 0 Predicted: 0  
 Actual: 0.0 Actual: 0.0 Actual: 0.0 Actual: 1.0 Actual: 0.0



### 1.1.5 PS1A. Visualize Intermediate Activations

```
[ ]: img_path='refs/dog_1.jpg'
# img_path='refs/dog.215.jpg'

# We preprocess the image into a 4D tensor
from keras.preprocessing import image
import numpy as np

im_size = 224
```

```

img = tf.keras.utils.load_img(img_path, target_size=(im_size, im_size))
img_tensor = tf.keras.utils.img_to_array(img)
img_tensor = np.expand_dims(img_tensor, axis=0)
# the model was trained on inputs
# that were preprocessed in the following way:
img_tensor /= 255.

# Its shape is (1, 224, 224, 3)
print(img_tensor.shape)

```

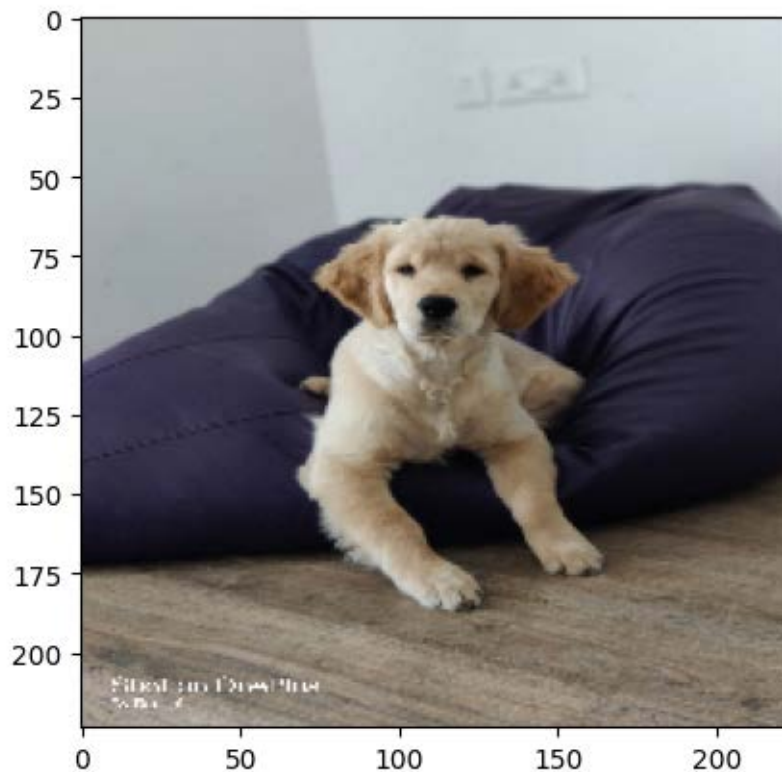
(1, 224, 224, 3)

```

[ ]: import matplotlib.pyplot as plt

plt.imshow(img_tensor[0])
plt.show()

```



### Chosen layers

```

[ ]: layers_to_visualize = [model.get_layer('conv1_relu'),
                             model.get_layer('conv3_block2_2_relu'),
                             model.get_layer('conv5_block3_2_relu')]

```

```
for layer in layers_to_visualize:
    print(layer.name)
```

```
conv1_relu
conv3_block2_2_relu
conv5_block3_2_relu
```

```
[ ]: act_input1 = model.input
# Extracts the outputs of the top 50 layers:
# layer_outputs = [layer.output for layer in model.layers[1:51]]
layer_outputs1 = [layer.output for layer in layers_to_visualize]

# First block : conv1
# Middle block: conv2c_branch2b
# Last block: conv5c_branch2c

# layer_outputs = [layer.output for layer in layers_to_visualize]
# Creates a model that will return these outputs, given the model input:
activation_model1 = Model(inputs=act_input1, outputs=layer_outputs1)
activation_model1.summary()
```

Model: "model\_1"

```
-----
Layer (type)                Output Shape              Param #   Connected to
=====
input_1 (InputLayer)        [(None, 224, 224, 3)]    0         []
conv1_pad (ZeroPadding2D)   (None, 230, 230, 3)      0         ['input_1[0][0]']
conv1_conv (Conv2D)         (None, 112, 112, 64)     9472      ['conv1_pad[0][0]']
conv1_bn (BatchNormalizati (None, 112, 112, 64)     256       ['conv1_conv[0][0]']
on)
conv1_relu (Activation)     (None, 112, 112, 64)     0         ['conv1_bn[0][0]']
pool1_pad (ZeroPadding2D)   (None, 114, 114, 64)     0         ['conv1_relu[0][0]']
pool1_pool (MaxPooling2D)   (None, 56, 56, 64)       0         ['pool1_pad[0][0]']
```



conv2_block1_1_conv (Conv2 (None, 56, 56, 64) ['pool1_pool[0][0]'] D)	4160
conv2_block1_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block1_1_conv[0][0]'] rmalization)	256
conv2_block1_1_relu (Activ (None, 56, 56, 64) ['conv2_block1_1_bn[0][0]'] ation)	0
conv2_block1_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block1_1_relu[0][0]'] D)	36928
conv2_block1_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block1_2_conv[0][0]'] rmalization)	256
conv2_block1_2_relu (Activ (None, 56, 56, 64) ['conv2_block1_2_bn[0][0]'] ation)	0
conv2_block1_0_conv (Conv2 (None, 56, 56, 256) ['pool1_pool[0][0]'] D)	16640
conv2_block1_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block1_2_relu[0][0]'] D)	16640
conv2_block1_0_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_0_conv[0][0]'] rmalization)	1024
conv2_block1_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_3_conv[0][0]'] rmalization)	1024
conv2_block1_add (Add) (None, 56, 56, 256) ['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']	0
conv2_block1_out (Activati (None, 56, 56, 256) ['conv2_block1_add[0][0]'] on)	0

conv2_block2_1_conv (Conv2D) ['conv2_block1_out[0][0]']	(None, 56, 56, 64)	16448
conv2_block2_1_bn (Batch Normalization) ['conv2_block2_1_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block2_1_relu (Activation) ['conv2_block2_1_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block2_2_conv (Conv2D) ['conv2_block2_1_relu[0][0]']	(None, 56, 56, 64)	36928
conv2_block2_2_bn (Batch Normalization) ['conv2_block2_2_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block2_2_relu (Activation) ['conv2_block2_2_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block2_3_conv (Conv2D) ['conv2_block2_2_relu[0][0]']	(None, 56, 56, 256)	16640
conv2_block2_3_bn (Batch Normalization) ['conv2_block2_3_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block2_add (Add) ['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block2_out (Activation) ['conv2_block2_add[0][0]']	(None, 56, 56, 256)	0
conv2_block3_1_conv (Conv2D) ['conv2_block2_out[0][0]']	(None, 56, 56, 64)	16448
conv2_block3_1_bn (Batch Normalization) ['conv2_block3_1_conv[0][0]']	(None, 56, 56, 64)	256

conv2_block3_1_relu (Activation) (None, 56, 56, 64) ['conv2_block3_1_bn[0][0]'])	0
conv2_block3_2_conv (Conv2D) (None, 56, 56, 64) ['conv2_block3_1_relu[0][0]'])	36928
conv2_block3_2_bn (Batch Normalization) (None, 56, 56, 64) ['conv2_block3_2_conv[0][0]'])	256
conv2_block3_2_relu (Activation) (None, 56, 56, 64) ['conv2_block3_2_bn[0][0]'])	0
conv2_block3_3_conv (Conv2D) (None, 56, 56, 256) ['conv2_block3_2_relu[0][0]'])	16640
conv2_block3_3_bn (Batch Normalization) (None, 56, 56, 256) ['conv2_block3_3_conv[0][0]'])	1024
conv2_block3_add (Add) (None, 56, 56, 256) ['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']	0
conv2_block3_out (Activation) (None, 56, 56, 256) ['conv2_block3_add[0][0]'])	0
conv3_block1_1_conv (Conv2D) (None, 28, 28, 128) ['conv2_block3_out[0][0]'])	32896
conv3_block1_1_bn (Batch Normalization) (None, 28, 28, 128) ['conv3_block1_1_conv[0][0]'])	512
conv3_block1_1_relu (Activation) (None, 28, 28, 128) ['conv3_block1_1_bn[0][0]'])	0
conv3_block1_2_conv (Conv2D) (None, 28, 28, 128) ['conv3_block1_1_relu[0][0]'])	147584

conv3_block1_2_bn (BatchNormal- ization) ['conv3_block1_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block1_2_relu (Activation) ['conv3_block1_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block1_0_conv (Conv2D) ['conv2_block3_out[0][0]']	(None, 28, 28, 512)	131584
conv3_block1_3_conv (Conv2D) ['conv3_block1_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block1_0_bn (BatchNormal- ization) ['conv3_block1_0_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block1_3_bn (BatchNormal- ization) ['conv3_block1_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block1_add (Add) ['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block1_out (Activation) ['conv3_block1_add[0][0]']	(None, 28, 28, 512)	0
conv3_block2_1_conv (Conv2D) ['conv3_block1_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block2_1_bn (BatchNormal- ization) ['conv3_block2_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block2_1_relu (Activation) ['conv3_block2_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block2_2_conv (Conv2D) ['conv3_block2_1_relu[0][0]']	(None, 28, 28, 128)	147584

conv3_block2_2_bn (BatchNormal- alization) ['conv3_block2_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block2_2_relu (Activation) ['conv3_block2_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block2_3_conv (Conv2D) ['conv3_block2_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block2_3_bn (BatchNormal- alization) ['conv3_block2_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block2_add (Add) ['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block2_out (Activation) ['conv3_block2_add[0][0]']	(None, 28, 28, 512)	0
conv3_block3_1_conv (Conv2D) ['conv3_block2_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block3_1_bn (BatchNormal- alization) ['conv3_block3_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block3_1_relu (Activation) ['conv3_block3_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block3_2_conv (Conv2D) ['conv3_block3_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block3_2_bn (BatchNormal- alization) ['conv3_block3_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block3_2_relu (Activation) ['conv3_block3_2_bn[0][0]']	(None, 28, 28, 128)	0

conv3_block3_3_conv (Conv2 (None, 28, 28, 512) ['conv3_block3_2_relu[0][0]'] D)	66048
conv3_block3_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block3_3_conv[0][0]'] rmalization)	2048
conv3_block3_add (Add) (None, 28, 28, 512) ['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']	0
conv3_block3_out (Activati (None, 28, 28, 512) ['conv3_block3_add[0][0]'] on)	0
conv3_block4_1_conv (Conv2 (None, 28, 28, 128) ['conv3_block3_out[0][0]'] D)	65664
conv3_block4_1_bn (BatchNo (None, 28, 28, 128) ['conv3_block4_1_conv[0][0]'] rmalization)	512
conv3_block4_1_relu (Activ (None, 28, 28, 128) ['conv3_block4_1_bn[0][0]'] ation)	0
conv3_block4_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block4_1_relu[0][0]'] D)	147584
conv3_block4_2_bn (BatchNo (None, 28, 28, 128) ['conv3_block4_2_conv[0][0]'] rmalization)	512
conv3_block4_2_relu (Activ (None, 28, 28, 128) ['conv3_block4_2_bn[0][0]'] ation)	0
conv3_block4_3_conv (Conv2 (None, 28, 28, 512) ['conv3_block4_2_relu[0][0]'] D)	66048
conv3_block4_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block4_3_conv[0][0]'] rmalization)	2048

conv3_block4_add (Add) ['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block4_out (Activation) ['conv3_block4_add[0][0]']	(None, 28, 28, 512)	0
conv4_block1_1_conv (Conv2D) ['conv3_block4_out[0][0]']	(None, 14, 14, 256)	131328
conv4_block1_1_bn (Batch Normalization) ['conv4_block1_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block1_1_relu (Activation) ['conv4_block1_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block1_2_conv (Conv2D) ['conv4_block1_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block1_2_bn (Batch Normalization) ['conv4_block1_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block1_2_relu (Activation) ['conv4_block1_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block1_0_conv (Conv2D) ['conv3_block4_out[0][0]']	(None, 14, 14, 1024)	525312
conv4_block1_3_conv (Conv2D) ['conv4_block1_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block1_0_bn (Batch Normalization) ['conv4_block1_0_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block1_3_bn (Batch Normalization) ['conv4_block1_3_conv[0][0]']	(None, 14, 14, 1024)	4096

conv4_block1_add (Add) ['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block1_out (Activation) ['conv4_block1_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block2_1_conv (Conv2D) ['conv4_block1_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block2_1_bn (Batch Normalization) ['conv4_block2_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block2_1_relu (Activation) ['conv4_block2_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block2_2_conv (Conv2D) ['conv4_block2_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block2_2_bn (Batch Normalization) ['conv4_block2_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block2_2_relu (Activation) ['conv4_block2_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block2_3_conv (Conv2D) ['conv4_block2_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block2_3_bn (Batch Normalization) ['conv4_block2_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block2_add (Add) ['conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block2_out (Activation) ['conv4_block2_add[0][0]']	(None, 14, 14, 1024)	0



conv4_block3_1_conv (Conv2D) ['conv4_block2_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block3_1_bn (Batch Normalization) ['conv4_block3_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block3_1_relu (Activation) ['conv4_block3_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block3_2_conv (Conv2D) ['conv4_block3_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block3_2_bn (Batch Normalization) ['conv4_block3_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block3_2_relu (Activation) ['conv4_block3_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block3_3_conv (Conv2D) ['conv4_block3_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block3_3_bn (Batch Normalization) ['conv4_block3_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block3_add (Add) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block3_out (Activation) ['conv4_block3_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block4_1_conv (Conv2D) ['conv4_block3_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block4_1_bn (Batch Normalization) ['conv4_block4_1_conv[0][0]']	(None, 14, 14, 256)	1024

conv4_block4_1_relu (Activation) (None, 14, 14, 256) ['conv4_block4_1_bn[0][0]']	0
conv4_block4_2_conv (Conv2D) (None, 14, 14, 256) ['conv4_block4_1_relu[0][0]']	590080
conv4_block4_2_bn (BatchNormalization) (None, 14, 14, 256) ['conv4_block4_2_conv[0][0]']	1024
conv4_block4_2_relu (Activation) (None, 14, 14, 256) ['conv4_block4_2_bn[0][0]']	0
conv4_block4_3_conv (Conv2D) (None, 14, 14, 1024) ['conv4_block4_2_relu[0][0]']	263168
conv4_block4_3_bn (BatchNormalization) (None, 14, 14, 1024) ['conv4_block4_3_conv[0][0]']	4096
conv4_block4_add (Add) (None, 14, 14, 1024) ['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']	0
conv4_block4_out (Activation) (None, 14, 14, 1024) ['conv4_block4_add[0][0]']	0
conv4_block5_1_conv (Conv2D) (None, 14, 14, 256) ['conv4_block4_out[0][0]']	262400
conv4_block5_1_bn (BatchNormalization) (None, 14, 14, 256) ['conv4_block5_1_conv[0][0]']	1024
conv4_block5_1_relu (Activation) (None, 14, 14, 256) ['conv4_block5_1_bn[0][0]']	0
conv4_block5_2_conv (Conv2D) (None, 14, 14, 256) ['conv4_block5_1_relu[0][0]']	590080

conv4_block5_2_bn (BatchNormal ['conv4_block5_2_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block5_2_relu (Activati ['conv4_block5_2_bn[0][0]' ation)	(None, 14, 14, 256)	0
conv4_block5_3_conv (Conv2D) ['conv4_block5_2_relu[0][0]'	(None, 14, 14, 1024)	263168
conv4_block5_3_bn (BatchNormal ['conv4_block5_3_conv[0][0]' rmalization)	(None, 14, 14, 1024)	4096
conv4_block5_add (Add) ['conv4_block4_out[0][0]' 'conv4_block5_3_bn[0][0]'	(None, 14, 14, 1024)	0
conv4_block5_out (Activation) ['conv4_block5_add[0][0]'	(None, 14, 14, 1024)	0
conv4_block6_1_conv (Conv2D) ['conv4_block5_out[0][0]'	(None, 14, 14, 256)	262400
conv4_block6_1_bn (BatchNormal ['conv4_block6_1_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block6_1_relu (Activati ['conv4_block6_1_bn[0][0]' ation)	(None, 14, 14, 256)	0
conv4_block6_2_conv (Conv2D) ['conv4_block6_1_relu[0][0]'	(None, 14, 14, 256)	590080
conv4_block6_2_bn (BatchNormal ['conv4_block6_2_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block6_2_relu (Activati ['conv4_block6_2_bn[0][0]' ation)	(None, 14, 14, 256)	0

conv4_block6_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block6_2_relu[0][0]'] D)	263168
conv4_block6_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block6_3_conv[0][0]'] rmalization)	4096
conv4_block6_add (Add) (None, 14, 14, 1024) ['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']	0
conv4_block6_out (Activati (None, 14, 14, 1024) ['conv4_block6_add[0][0]'] on)	0
conv5_block1_1_conv (Conv2 (None, 7, 7, 512) ['conv4_block6_out[0][0]'] D)	524800
conv5_block1_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block1_1_conv[0][0]'] rmalization)	2048
conv5_block1_1_relu (Activ (None, 7, 7, 512) ['conv5_block1_1_bn[0][0]'] ation)	0
conv5_block1_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block1_1_relu[0][0]'] D)	2359808
conv5_block1_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block1_2_conv[0][0]'] rmalization)	2048
conv5_block1_2_relu (Activ (None, 7, 7, 512) ['conv5_block1_2_bn[0][0]'] ation)	0
conv5_block1_0_conv (Conv2 (None, 7, 7, 2048) ['conv4_block6_out[0][0]'] D)	2099200
conv5_block1_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block1_2_relu[0][0]'] D)	1050624

conv5_block1_0_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_0_conv[0][0]'] rmalization)	8192
conv5_block1_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_3_conv[0][0]'] rmalization)	8192
conv5_block1_add (Add) (None, 7, 7, 2048) ['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']	0
conv5_block1_out (Activati (None, 7, 7, 2048) ['conv5_block1_add[0][0]'] on)	0
conv5_block2_1_conv (Conv2 (None, 7, 7, 512) ['conv5_block1_out[0][0]'] D)	1049088
conv5_block2_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_1_conv[0][0]'] rmalization)	2048
conv5_block2_1_relu (Activ (None, 7, 7, 512) ['conv5_block2_1_bn[0][0]'] ation)	0
conv5_block2_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block2_1_relu[0][0]'] D)	2359808
conv5_block2_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_2_conv[0][0]'] rmalization)	2048
conv5_block2_2_relu (Activ (None, 7, 7, 512) ['conv5_block2_2_bn[0][0]'] ation)	0
conv5_block2_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block2_2_relu[0][0]'] D)	1050624
conv5_block2_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block2_3_conv[0][0]'] rmalization)	8192

conv5_block2_add (Add)	(None, 7, 7, 2048)	0
['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']		
conv5_block2_out (Activation)	(None, 7, 7, 2048)	0
['conv5_block2_add[0][0]']		
conv5_block3_1_conv (Conv2D)	(None, 7, 7, 512)	1049088
['conv5_block2_out[0][0]']		
conv5_block3_1_bn (Batch Normalization)	(None, 7, 7, 512)	2048
['conv5_block3_1_conv[0][0]']		
conv5_block3_1_relu (Activation)	(None, 7, 7, 512)	0
['conv5_block3_1_bn[0][0]']		
conv5_block3_2_conv (Conv2D)	(None, 7, 7, 512)	2359808
['conv5_block3_1_relu[0][0]']		
conv5_block3_2_bn (Batch Normalization)	(None, 7, 7, 512)	2048
['conv5_block3_2_conv[0][0]']		
conv5_block3_2_relu (Activation)	(None, 7, 7, 512)	0
['conv5_block3_2_bn[0][0]']		

=====

Total params: 22528896 (85.94 MB)  
 Trainable params: 22479872 (85.75 MB)  
 Non-trainable params: 49024 (191.50 KB)

-----

### First Layer Activation

```
[ ]: activations = activation_model1.predict([img_tensor])

first_layer_activation = activations[0]
print(first_layer_activation.shape)
```

```

for activation in activations:
    print(activation.shape)

# for i in range(3): # Loop through the channels
#     plt.matshow(first_layer_activation[0, :, :, i], cmap='viridis')
#     plt.show()
# plt.matshow(first_layer_activation[0, :, :, 0], cmap='viridis')
# plt.show()
# plt.matshow(first_layer_activation[0, :, :, 3], cmap='viridis')
# plt.show()

fig, axes = plt.subplots(1, 3, figsize=(20, 8))

# Display the first activation
axes[0].matshow(first_layer_activation[0, :, :, 6], cmap='viridis')
axes[0].set_title('Layer 1')

# Display the second activation
axes[1].matshow(first_layer_activation[0, :, :, 10], cmap='viridis')
axes[1].set_title('Layer 10')

# Display the second activation
axes[2].matshow(first_layer_activation[0, :, :, 15], cmap='viridis')
axes[2].set_title('Layer 26')

# Show the subplots
plt.show()

```

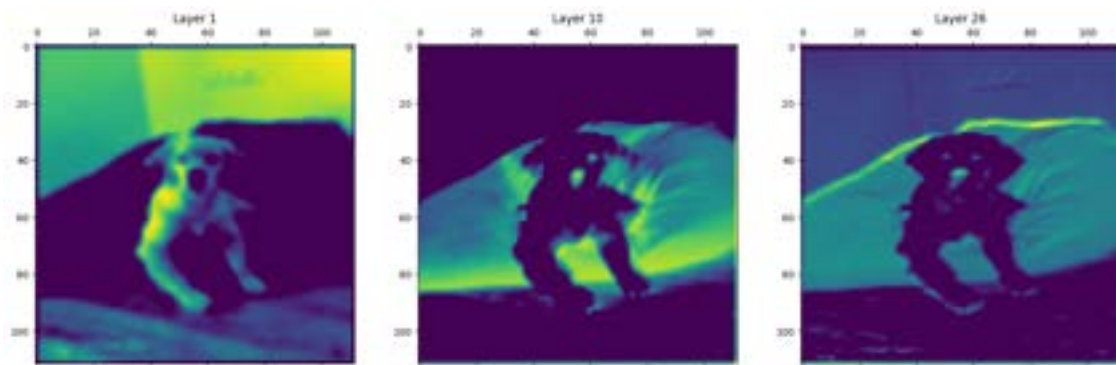
1/1 [=====] - 0s 57ms/step

(1, 112, 112, 64)

(1, 112, 112, 64)

(1, 28, 28, 128)

(1, 7, 7, 512)



This one looks like a “bright green dots” detector, useful to encode dog eyes and nose and it gets further clear as we go deeper.

At this point, let’s go and plot a complete visualization of all the activations in the network.

### Middle Layer Activation

```
[ ]: activations = activation_model1.predict([img_tensor])

middle_layer_activation = activations[1]
print(middle_layer_activation.shape)

for activation in activations:
    print(activation.shape)

fig, axes = plt.subplots(1, 3,  figsize=(20, 8))

# Display the first activation
axes[0].matshow(middle_layer_activation[0, :, :, 6], cmap='viridis')
axes[0].set_title('Layer 1')

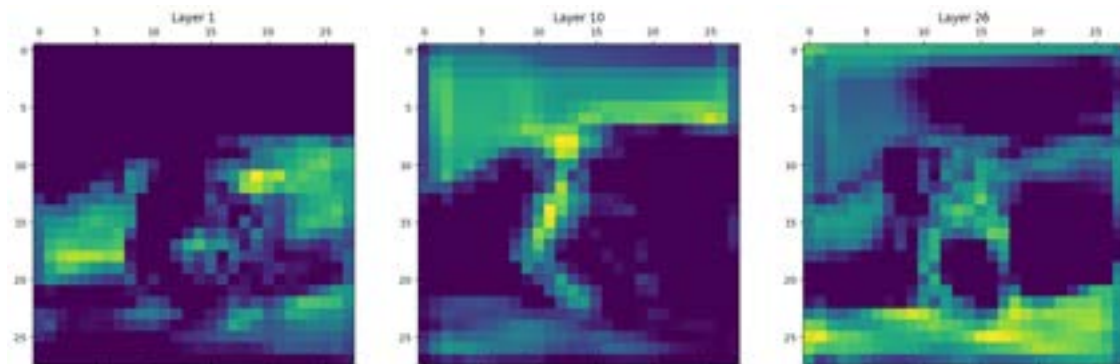
# Display the second activation
axes[1].matshow(middle_layer_activation[0, :, :, 10], cmap='viridis')
axes[1].set_title('Layer 10')

# Display the second activation
axes[2].matshow(middle_layer_activation[0, :, :, 15], cmap='viridis')
axes[2].set_title('Layer 26')

# Show the subplots
plt.show()
```

1/1 [=====] - 0s 57ms/step

(1, 28, 28, 128)  
(1, 112, 112, 64)  
(1, 28, 28, 128)  
(1, 7, 7, 512)





This one looks like the edges of the dog are more defined and some contours noticed in the deeper layers

At this point, let's go and plot a complete visualization of all the activations in the network.

### Last Layer Activation

```
[ ]: activations = activation_model1.predict([img_tensor])

last_layer_activation = activations[2]
print(first_layer_activation.shape)

for activation in activations:
    print(activation.shape)

fig, axes = plt.subplots(1, 3, figsize=(20, 8))

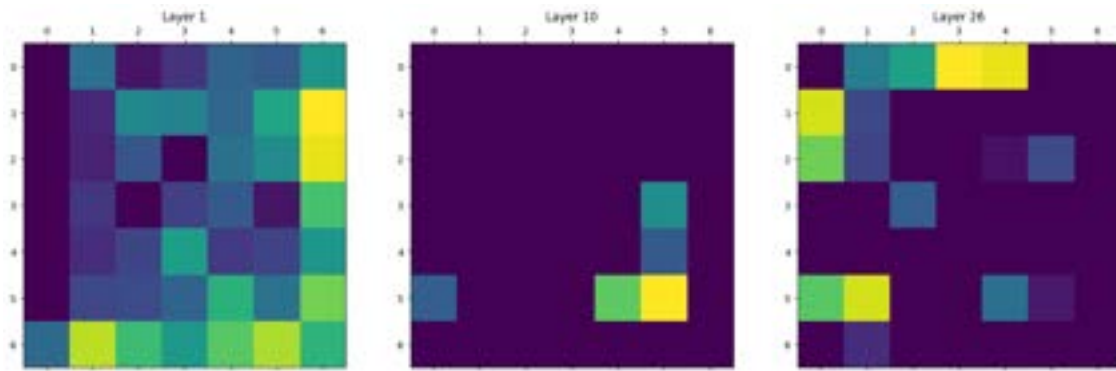
# Display the first activation
axes[0].matshow(last_layer_activation[0, :, :, 1], cmap='viridis')
axes[0].set_title('Layer 1')

# Display the second activation
axes[1].matshow(last_layer_activation[0, :, :, 11], cmap='viridis')
axes[1].set_title('Layer 10')

# Display the second activation
axes[2].matshow(last_layer_activation[0, :, :, 15], cmap='viridis')
axes[2].set_title('Layer 26')

# Show the subplots
plt.show()
```

```
1/1 [=====] - 0s 51ms/step
(1, 112, 112, 64)
(1, 112, 112, 64)
(1, 28, 28, 128)
(1, 7, 7, 512)
```



```
[ ]: # These are the names of the layers, so can have them as part of our plot
layer_names = []
# layer_names = layers_to_visualize
# for layer in model.layers[:50]:
#     layer_names.append(layer.name)
# for layer in layers_to_visualize:
#     print(layer.name)
#     layer_names.append(layer.name)

layer_names = layers_to_visualize

images_per_row = 16

# Now let's display our feature maps
for layer_name, layer_activation in zip(layer_names, activations):
    # This is the number of features in the feature map
    n_features = layer_activation.shape[-1]

    # The feature map has shape (1, size, size, n_features)
    size = layer_activation.shape[1]

    # We will tile the activation channels in this matrix
    n_cols = n_features // images_per_row
    display_grid = np.zeros((size * n_cols, images_per_row * size))

    # We'll tile each filter into this big horizontal grid
    for col in range(n_cols):
        for row in range(images_per_row):
            channel_image = layer_activation[0,
                                             :, :,
                                             col * images_per_row + row]

            # Post-process the feature to make it visually palatable
            channel_image -= channel_image.mean()
            channel_image /= channel_image.std()
```

```

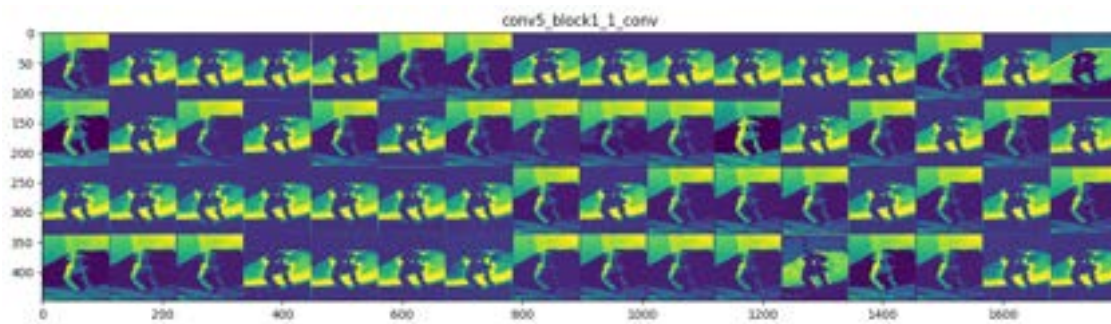
channel_image *= 64
channel_image += 128
channel_image = np.clip(channel_image, 0, 255).astype('uint8')
display_grid[col * size : (col + 1) * size,
              row * size : (row + 1) * size] = channel_image

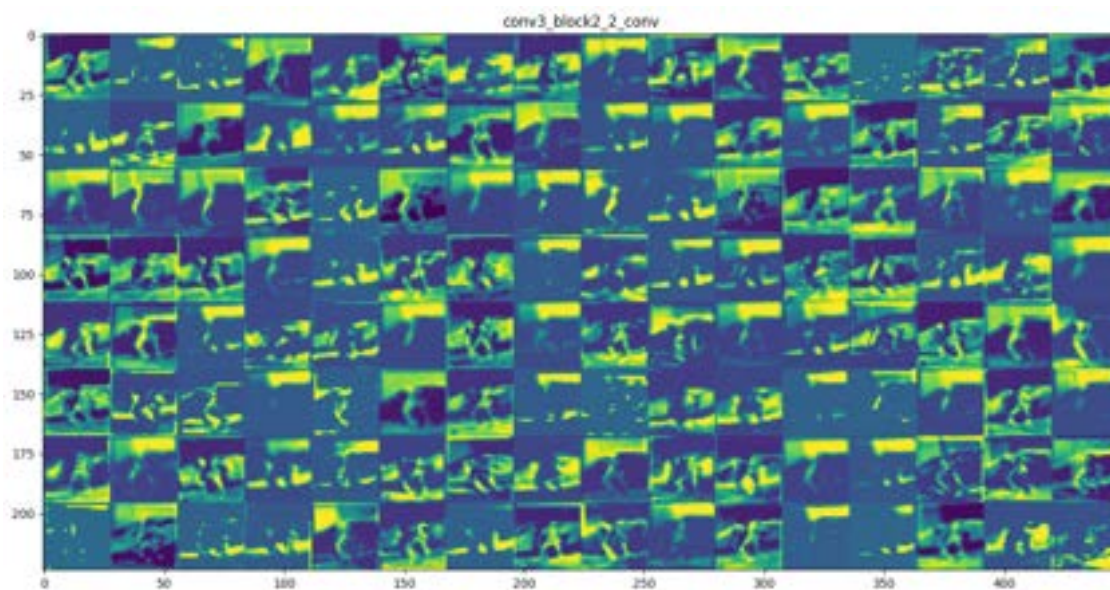
# Display the grid
scale = 1. / size
plt.figure(figsize=(scale * display_grid.shape[1],
                    scale * display_grid.shape[0]))
plt.title(layer_name)
plt.grid(False)
plt.imshow(display_grid, aspect='auto', cmap='viridis')

plt.show()

```

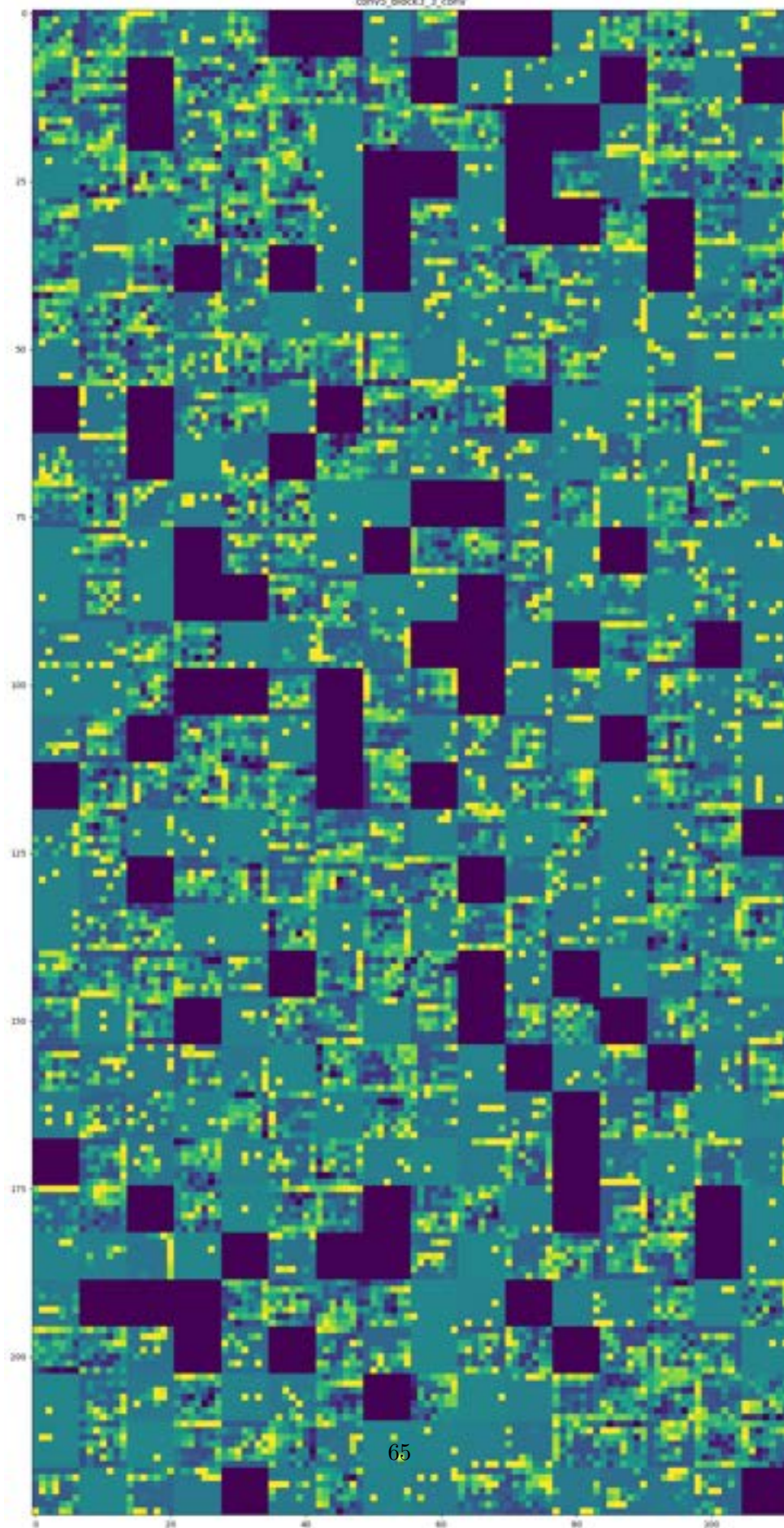
/var/folders/xy/1x8f1v1x569\_ch145p3563j80000gn/T/ipykernel\_13280/1408363524.py:3  
7: RuntimeWarning: invalid value encountered in cast  
channel\_image = np.clip(channel\_image, 0, 255).astype('uint8')







conv5\_block3\_3\_conv



Please scroll to view all the plots

### 1.1.6 Ignore - OLD\_ CustomResNET50 Visualizing intermediate activations

Chose class Dog

```
[ ]: # import tensorflow as tf
model = tf.keras.models.load_model('model_my_custom_resnet50_lr.h5')
# model.summary()
```

WARNING:absl:At this time, the v2.11+ optimizer `tf.keras.optimizers.SGD` runs slowly on M1/M2 Macs, please use the legacy Keras optimizer instead, located at `tf.keras.optimizers.legacy.SGD`.

```
[ ]: # img_path = 'dogscats/subset/test/cats/cat.1700.jpg'
img_path='refs/dog_1.jpg'
# img_path='refs/dog.215.jpg'

# We preprocess the image into a 4D tensor
from keras.preprocessing import image
import numpy as np

im_size = 224

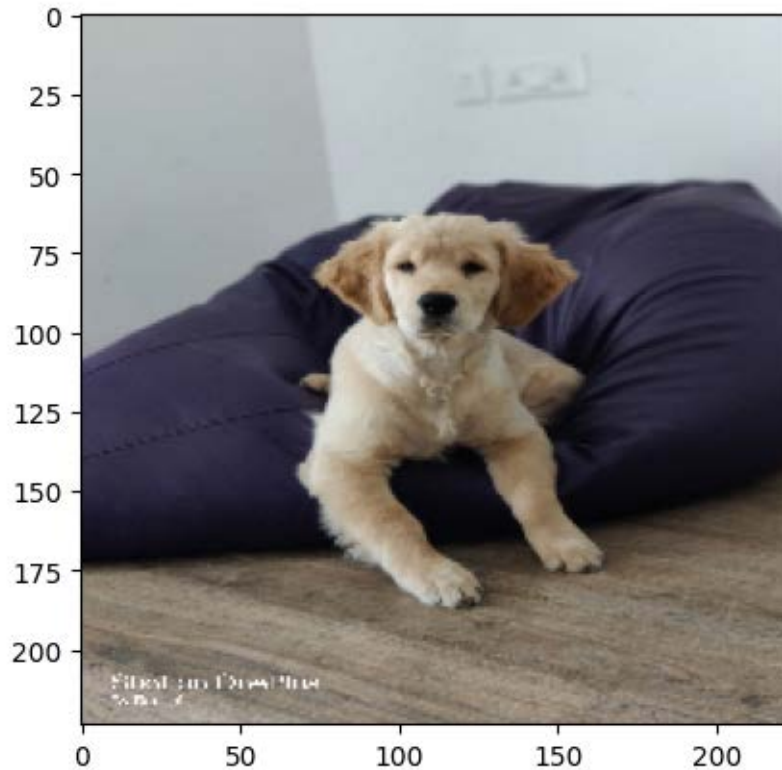
img = tf.keras.utils.load_img(img_path, target_size=(im_size, im_size))
img_tensor = tf.keras.utils.img_to_array(img)
img_tensor = np.expand_dims(img_tensor, axis=0)
# Remember that the model was trained on inputs
# that were preprocessed in the following way:
img_tensor /= 255.

# Its shape is (1, 150, 150, 3)
print(img_tensor.shape)
```

(1, 224, 224, 3)

```
[ ]: import matplotlib.pyplot as plt

plt.imshow(img_tensor[0])
plt.show()
```



```
[ ]: # layers = model.layers
# # for layer in layers:
# #     print(layer.name)
# # filters, biases = model.layers[2]._get_trainable_state
# layers_to_visualize = [model.get_layer('conv1'), model.
    ↪ get_layer('conv2c_branch2b'), model.get_layer('conv5c_branch2c')]

# layers_to_visualize = [model.get_layer('conv1_relu'),
#                         model.get_layer('conv3_block2_2_relu'),
#                         model.get_layer('conv5_block3_2_relu')]

layers_to_visualize = [model.get_layer('conv1'),
                       model.get_layer('conv3a_branch2b'),
                       model.get_layer('conv5c_branch2c')]
# # print(layers[2].name, filters.shape)
# for layer in model.layers:
#     print(layer.name)
for layer in layers_to_visualize:
    print(layer.name)
```

```
conv1
conv3a_branch2b
```

conv5c\_branch2c

```
[ ]: act_input = model.input
# Extracts the outputs of the top 50 layers:
# layer_outputs = [layer.output for layer in model.layers[1:51]]
layer_outputs = [layer.output for layer in layers_to_visualize]

# First block : conv1
# Middle block: conv2c_branch2b
# Last block: conv5c_branch2c

# layer_outputs = [layer.output for layer in layers_to_visualize]
# Creates a model that will return these outputs, given the model input:
activation_model = Model(inputs=act_input, outputs=layer_outputs)
```

```
[ ]: activation_model.summary()
```

Model: "model"

```
-----
-----
Layer (type)                 Output Shape              Param #   Connected to
=====
input_1 (InputLayer)         [(None, 224, 224, 3)]    0         []

zero_padding2d (ZeroPaddin   (None, 230, 230, 3)      0
['input_1[0][0]']
g2D)

conv1 (Conv2D)               (None, 112, 112, 64)     9472
['zero_padding2d[0][0]']

bn_conv1 (BatchNormalizati   (None, 112, 112, 64)     256
['conv1[0][0]']
on)

activation (Activation)      (None, 112, 112, 64)     0
['bn_conv1[0][0]']

max_pooling2d (MaxPooling2   (None, 55, 55, 64)      0
['activation[0][0]']
D)

conv2a_branch2a (Conv2D)     (None, 55, 55, 64)      4160
['max_pooling2d[0][0]']

bn2a_branch2a (BatchNormal   (None, 55, 55, 64)      256
['conv2a_branch2a[0][0]']
```



ization)		
activation_1 (Activation) ['bn2a_branch2a[0][0]']	(None, 55, 55, 64)	0
conv2a_branch2b (Conv2D) ['activation_1[0][0]']	(None, 55, 55, 64)	36928
bn2a_branch2b (BatchNormal ['conv2a_branch2b[0][0]'] ization)	(None, 55, 55, 64)	256
activation_2 (Activation) ['bn2a_branch2b[0][0]']	(None, 55, 55, 64)	0
conv2a_branch2c (Conv2D) ['activation_2[0][0]']	(None, 55, 55, 256)	16640
conv2a_branch1 (Conv2D) ['max_pooling2d[0][0]']	(None, 55, 55, 256)	16640
bn2a_branch2c (BatchNormal ['conv2a_branch2c[0][0]'] ization)	(None, 55, 55, 256)	1024
bn2a_branch1 (BatchNormali ['conv2a_branch1[0][0]'] ization)	(None, 55, 55, 256)	1024
add (Add) ['bn2a_branch2c[0][0]'], 'bn2a_branch1[0][0]']	(None, 55, 55, 256)	0
activation_3 (Activation) ['add[0][0]']	(None, 55, 55, 256)	0
conv2b_branch2a (Conv2D) ['activation_3[0][0]']	(None, 55, 55, 64)	16448
bn2b_branch2a (BatchNormal ['conv2b_branch2a[0][0]'] ization)	(None, 55, 55, 64)	256
activation_4 (Activation) ['bn2b_branch2a[0][0]']	(None, 55, 55, 64)	0
conv2b_branch2b (Conv2D) ['activation_4[0][0]']	(None, 55, 55, 64)	36928

bn2b_branch2b (BatchNormal ['conv2b_branch2b[0][0]' ization)	(None, 55, 55, 64)	256
activation_5 (Activation) ['bn2b_branch2b[0][0]']	(None, 55, 55, 64)	0
conv2b_branch2c (Conv2D) ['activation_5[0][0]']	(None, 55, 55, 256)	16640
bn2b_branch2c (BatchNormal ['conv2b_branch2c[0][0]' ization)	(None, 55, 55, 256)	1024
add_1 (Add) ['bn2b_branch2c[0][0]', 'activation_3[0][0]']	(None, 55, 55, 256)	0
activation_6 (Activation) ['add_1[0][0]']	(None, 55, 55, 256)	0
conv2c_branch2a (Conv2D) ['activation_6[0][0]']	(None, 55, 55, 64)	16448
bn2c_branch2a (BatchNormal ['conv2c_branch2a[0][0]' ization)	(None, 55, 55, 64)	256
activation_7 (Activation) ['bn2c_branch2a[0][0]']	(None, 55, 55, 64)	0
conv2c_branch2b (Conv2D) ['activation_7[0][0]']	(None, 55, 55, 64)	36928
bn2c_branch2b (BatchNormal ['conv2c_branch2b[0][0]' ization)	(None, 55, 55, 64)	256
activation_8 (Activation) ['bn2c_branch2b[0][0]']	(None, 55, 55, 64)	0
conv2c_branch2c (Conv2D) ['activation_8[0][0]']	(None, 55, 55, 256)	16640
bn2c_branch2c (BatchNormal ['conv2c_branch2c[0][0]' ization)	(None, 55, 55, 256)	1024

add_2 (Add) ['bn2c_branch2c[0][0]', 'activation_6[0][0]']	(None, 55, 55, 256)	0
activation_9 (Activation) ['add_2[0][0]']	(None, 55, 55, 256)	0
conv3a_branch2a (Conv2D) ['activation_9[0][0]']	(None, 28, 28, 128)	32896
bn3a_branch2a (BatchNormal ization) ['conv3a_branch2a[0][0]']	(None, 28, 28, 128)	512
activation_10 (Activation) ['bn3a_branch2a[0][0]']	(None, 28, 28, 128)	0
conv3a_branch2b (Conv2D) ['activation_10[0][0]']	(None, 28, 28, 128)	147584
bn3a_branch2b (BatchNormal ization) ['conv3a_branch2b[0][0]']	(None, 28, 28, 128)	512
activation_11 (Activation) ['bn3a_branch2b[0][0]']	(None, 28, 28, 128)	0
conv3a_branch2c (Conv2D) ['activation_11[0][0]']	(None, 28, 28, 512)	66048
conv3a_branch1 (Conv2D) ['activation_9[0][0]']	(None, 28, 28, 512)	131584
bn3a_branch2c (BatchNormal ization) ['conv3a_branch2c[0][0]']	(None, 28, 28, 512)	2048
bn3a_branch1 (BatchNormali zation) ['conv3a_branch1[0][0]']	(None, 28, 28, 512)	2048
add_3 (Add) ['bn3a_branch2c[0][0]', 'bn3a_branch1[0][0]']	(None, 28, 28, 512)	0
activation_12 (Activation) ['add_3[0][0]']	(None, 28, 28, 512)	0

conv3b_branch2a (Conv2D) ['activation_12[0][0]']	(None, 28, 28, 128)	65664
bn3b_branch2a (BatchNormal ization) ['conv3b_branch2a[0][0]']	(None, 28, 28, 128)	512
activation_13 (Activation) ['bn3b_branch2a[0][0]']	(None, 28, 28, 128)	0
conv3b_branch2b (Conv2D) ['activation_13[0][0]']	(None, 28, 28, 128)	147584
bn3b_branch2b (BatchNormal ization) ['conv3b_branch2b[0][0]']	(None, 28, 28, 128)	512
activation_14 (Activation) ['bn3b_branch2b[0][0]']	(None, 28, 28, 128)	0
conv3b_branch2c (Conv2D) ['activation_14[0][0]']	(None, 28, 28, 512)	66048
bn3b_branch2c (BatchNormal ization) ['conv3b_branch2c[0][0]']	(None, 28, 28, 512)	2048
add_4 (Add) ['bn3b_branch2c[0][0]', 'activation_12[0][0]']	(None, 28, 28, 512)	0
activation_15 (Activation) ['add_4[0][0]']	(None, 28, 28, 512)	0
conv3c_branch2a (Conv2D) ['activation_15[0][0]']	(None, 28, 28, 128)	65664
bn3c_branch2a (BatchNormal ization) ['conv3c_branch2a[0][0]']	(None, 28, 28, 128)	512
activation_16 (Activation) ['bn3c_branch2a[0][0]']	(None, 28, 28, 128)	0
conv3c_branch2b (Conv2D) ['activation_16[0][0]']	(None, 28, 28, 128)	147584

bn3c_branch2b (BatchNormal ['conv3c_branch2b[0][0]' ization)	(None, 28, 28, 128)	512
activation_17 (Activation) ['bn3c_branch2b[0][0]']	(None, 28, 28, 128)	0
conv3c_branch2c (Conv2D) ['activation_17[0][0]']	(None, 28, 28, 512)	66048
bn3c_branch2c (BatchNormal ['conv3c_branch2c[0][0]' ization)	(None, 28, 28, 512)	2048
add_5 (Add) ['bn3c_branch2c[0][0]', 'activation_15[0][0]']	(None, 28, 28, 512)	0
activation_18 (Activation) ['add_5[0][0]']	(None, 28, 28, 512)	0
conv3d_branch2a (Conv2D) ['activation_18[0][0]']	(None, 28, 28, 128)	65664
bn3d_branch2a (BatchNormal ['conv3d_branch2a[0][0]' ization)	(None, 28, 28, 128)	512
activation_19 (Activation) ['bn3d_branch2a[0][0]']	(None, 28, 28, 128)	0
conv3d_branch2b (Conv2D) ['activation_19[0][0]']	(None, 28, 28, 128)	147584
bn3d_branch2b (BatchNormal ['conv3d_branch2b[0][0]' ization)	(None, 28, 28, 128)	512
activation_20 (Activation) ['bn3d_branch2b[0][0]']	(None, 28, 28, 128)	0
conv3d_branch2c (Conv2D) ['activation_20[0][0]']	(None, 28, 28, 512)	66048
bn3d_branch2c (BatchNormal ['conv3d_branch2c[0][0]' ization)	(None, 28, 28, 512)	2048

add_6 (Add) ['bn3d_branch2c[0][0]', 'activation_18[0][0]']	(None, 28, 28, 512)	0
activation_21 (Activation) ['add_6[0][0]']	(None, 28, 28, 512)	0
conv4a_branch2a (Conv2D) ['activation_21[0][0]']	(None, 14, 14, 256)	131328
bn4a_branch2a (BatchNormal ization) ['conv4a_branch2a[0][0]']	(None, 14, 14, 256)	1024
activation_22 (Activation) ['bn4a_branch2a[0][0]']	(None, 14, 14, 256)	0
conv4a_branch2b (Conv2D) ['activation_22[0][0]']	(None, 14, 14, 256)	590080
bn4a_branch2b (BatchNormal ization) ['conv4a_branch2b[0][0]']	(None, 14, 14, 256)	1024
activation_23 (Activation) ['bn4a_branch2b[0][0]']	(None, 14, 14, 256)	0
conv4a_branch2c (Conv2D) ['activation_23[0][0]']	(None, 14, 14, 1024)	263168
conv4a_branch1 (Conv2D) ['activation_21[0][0]']	(None, 14, 14, 1024)	525312
bn4a_branch2c (BatchNormal ization) ['conv4a_branch2c[0][0]']	(None, 14, 14, 1024)	4096
bn4a_branch1 (BatchNormali zation) ['conv4a_branch1[0][0]']	(None, 14, 14, 1024)	4096
add_7 (Add) ['bn4a_branch2c[0][0]', 'bn4a_branch1[0][0]']	(None, 14, 14, 1024)	0
activation_24 (Activation) ['add_7[0][0]']	(None, 14, 14, 1024)	0

conv4b_branch2a (Conv2D) ['activation_24[0][0]']	(None, 14, 14, 256)	262400
bn4b_branch2a (BatchNormal ization) ['conv4b_branch2a[0][0]']	(None, 14, 14, 256)	1024
activation_25 (Activation) ['bn4b_branch2a[0][0]']	(None, 14, 14, 256)	0
conv4b_branch2b (Conv2D) ['activation_25[0][0]']	(None, 14, 14, 256)	590080
bn4b_branch2b (BatchNormal ization) ['conv4b_branch2b[0][0]']	(None, 14, 14, 256)	1024
activation_26 (Activation) ['bn4b_branch2b[0][0]']	(None, 14, 14, 256)	0
conv4b_branch2c (Conv2D) ['activation_26[0][0]']	(None, 14, 14, 1024)	263168
bn4b_branch2c (BatchNormal ization) ['conv4b_branch2c[0][0]']	(None, 14, 14, 1024)	4096
add_8 (Add) ['bn4b_branch2c[0][0]', 'activation_24[0][0]']	(None, 14, 14, 1024)	0
activation_27 (Activation) ['add_8[0][0]']	(None, 14, 14, 1024)	0
conv4c_branch2a (Conv2D) ['activation_27[0][0]']	(None, 14, 14, 256)	262400
bn4c_branch2a (BatchNormal ization) ['conv4c_branch2a[0][0]']	(None, 14, 14, 256)	1024
activation_28 (Activation) ['bn4c_branch2a[0][0]']	(None, 14, 14, 256)	0
conv4c_branch2b (Conv2D) ['activation_28[0][0]']	(None, 14, 14, 256)	590080
bn4c_branch2b (BatchNormal	(None, 14, 14, 256)	1024

['conv4c_branch2b[0][0]'] ization)		
activation_29 (Activation) ['bn4c_branch2b[0][0]']	(None, 14, 14, 256)	0
conv4c_branch2c (Conv2D) ['activation_29[0][0]']	(None, 14, 14, 1024)	263168
bn4c_branch2c (BatchNormal ['conv4c_branch2c[0][0]'] ization)	(None, 14, 14, 1024)	4096
add_9 (Add) ['bn4c_branch2c[0][0]', 'activation_27[0][0]']	(None, 14, 14, 1024)	0
activation_30 (Activation) ['add_9[0][0]']	(None, 14, 14, 1024)	0
conv4d_branch2a (Conv2D) ['activation_30[0][0]']	(None, 14, 14, 256)	262400
bn4d_branch2a (BatchNormal ['conv4d_branch2a[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_31 (Activation) ['bn4d_branch2a[0][0]']	(None, 14, 14, 256)	0
conv4d_branch2b (Conv2D) ['activation_31[0][0]']	(None, 14, 14, 256)	590080
bn4d_branch2b (BatchNormal ['conv4d_branch2b[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_32 (Activation) ['bn4d_branch2b[0][0]']	(None, 14, 14, 256)	0
conv4d_branch2c (Conv2D) ['activation_32[0][0]']	(None, 14, 14, 1024)	263168
bn4d_branch2c (BatchNormal ['conv4d_branch2c[0][0]'] ization)	(None, 14, 14, 1024)	4096
add_10 (Add)	(None, 14, 14, 1024)	0



['bn4d_branch2c[0][0]', 'activation_30[0][0]']		
activation_33 (Activation) ['add_10[0][0]']	(None, 14, 14, 1024)	0
conv4e_branch2a (Conv2D) ['activation_33[0][0]']	(None, 14, 14, 256)	262400
bn4e_branch2a (BatchNormal ['conv4e_branch2a[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_34 (Activation) ['bn4e_branch2a[0][0]']	(None, 14, 14, 256)	0
conv4e_branch2b (Conv2D) ['activation_34[0][0]']	(None, 14, 14, 256)	590080
bn4e_branch2b (BatchNormal ['conv4e_branch2b[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_35 (Activation) ['bn4e_branch2b[0][0]']	(None, 14, 14, 256)	0
conv4e_branch2c (Conv2D) ['activation_35[0][0]']	(None, 14, 14, 1024)	263168
bn4e_branch2c (BatchNormal ['conv4e_branch2c[0][0]'] ization)	(None, 14, 14, 1024)	4096
add_11 (Add) ['bn4e_branch2c[0][0]', 'activation_33[0][0]']	(None, 14, 14, 1024)	0
activation_36 (Activation) ['add_11[0][0]']	(None, 14, 14, 1024)	0
conv4f_branch2a (Conv2D) ['activation_36[0][0]']	(None, 14, 14, 256)	262400
bn4f_branch2a (BatchNormal ['conv4f_branch2a[0][0]'] ization)	(None, 14, 14, 256)	1024
activation_37 (Activation)	(None, 14, 14, 256)	0

['bn4f_branch2a[0][0]']		
conv4f_branch2b (Conv2D)	(None, 14, 14, 256)	590080
['activation_37[0][0]']		
bn4f_branch2b (BatchNormal	(None, 14, 14, 256)	1024
['conv4f_branch2b[0][0]']		
ization)		
activation_38 (Activation)	(None, 14, 14, 256)	0
['bn4f_branch2b[0][0]']		
conv4f_branch2c (Conv2D)	(None, 14, 14, 1024)	263168
['activation_38[0][0]']		
bn4f_branch2c (BatchNormal	(None, 14, 14, 1024)	4096
['conv4f_branch2c[0][0]']		
ization)		
add_12 (Add)	(None, 14, 14, 1024)	0
['bn4f_branch2c[0][0]',		
'activation_36[0][0]']		
activation_39 (Activation)	(None, 14, 14, 1024)	0
['add_12[0][0]']		
conv5a_branch2a (Conv2D)	(None, 7, 7, 512)	524800
['activation_39[0][0]']		
bn5a_branch2a (BatchNormal	(None, 7, 7, 512)	2048
['conv5a_branch2a[0][0]']		
ization)		
activation_40 (Activation)	(None, 7, 7, 512)	0
['bn5a_branch2a[0][0]']		
conv5a_branch2b (Conv2D)	(None, 7, 7, 512)	2359808
['activation_40[0][0]']		
bn5a_branch2b (BatchNormal	(None, 7, 7, 512)	2048
['conv5a_branch2b[0][0]']		
ization)		
activation_41 (Activation)	(None, 7, 7, 512)	0
['bn5a_branch2b[0][0]']		
conv5a_branch2c (Conv2D)	(None, 7, 7, 2048)	1050624
['activation_41[0][0]']		

conv5a_branch1 (Conv2D) ['activation_39[0][0]']	(None, 7, 7, 2048)	2099200
bn5a_branch2c (BatchNormal ization)	(None, 7, 7, 2048)	8192
bn5a_branch1 (BatchNormali ization)	(None, 7, 7, 2048)	8192
add_13 (Add) ['bn5a_branch2c[0][0]', 'bn5a_branch1[0][0]']	(None, 7, 7, 2048)	0
activation_42 (Activation) ['add_13[0][0]']	(None, 7, 7, 2048)	0
conv5b_branch2a (Conv2D) ['activation_42[0][0]']	(None, 7, 7, 512)	1049088
bn5b_branch2a (BatchNormal ization)	(None, 7, 7, 512)	2048
activation_43 (Activation) ['bn5b_branch2a[0][0]']	(None, 7, 7, 512)	0
conv5b_branch2b (Conv2D) ['activation_43[0][0]']	(None, 7, 7, 512)	2359808
bn5b_branch2b (BatchNormal ization)	(None, 7, 7, 512)	2048
activation_44 (Activation) ['bn5b_branch2b[0][0]']	(None, 7, 7, 512)	0
conv5b_branch2c (Conv2D) ['activation_44[0][0]']	(None, 7, 7, 2048)	1050624
bn5b_branch2c (BatchNormal ization)	(None, 7, 7, 2048)	8192
add_14 (Add) ['bn5b_branch2c[0][0]',	(None, 7, 7, 2048)	0

```

'activation_42[0][0]']

activation_45 (Activation) (None, 7, 7, 2048) 0
['add_14[0][0]']

conv5c_branch2a (Conv2D) (None, 7, 7, 512) 1049088
['activation_45[0][0]']

bn5c_branch2a (BatchNormal (None, 7, 7, 512) 2048
['conv5c_branch2a[0][0]']
ization)

activation_46 (Activation) (None, 7, 7, 512) 0
['bn5c_branch2a[0][0]']

conv5c_branch2b (Conv2D) (None, 7, 7, 512) 2359808
['activation_46[0][0]']

bn5c_branch2b (BatchNormal (None, 7, 7, 512) 2048
['conv5c_branch2b[0][0]']
ization)

activation_47 (Activation) (None, 7, 7, 512) 0
['bn5c_branch2b[0][0]']

conv5c_branch2c (Conv2D) (None, 7, 7, 2048) 1050624
['activation_47[0][0]']

```

```

=====
=====
Total params: 23579520 (89.95 MB)
Trainable params: 23530496 (89.76 MB)
Non-trainable params: 49024 (191.50 KB)
-----
-----

```

### First Layer activation

```

[ ]: # This will return a list of 5 Numpy arrays:
      # one array per layer activation

      activations = activation_model.predict([img_tensor])

      first_layer_activation = activations[0]
      print(first_layer_activation.shape)

      for activation in activations:
          print(activation.shape)

```

```

# for i in range(3): # Loop through the channels
#     plt.matshow(first_layer_activation[0, :, :, i], cmap='viridis')
#     plt.show()
# plt.matshow(first_layer_activation[0, :, :, 0], cmap='viridis')
# plt.show()
# plt.matshow(first_layer_activation[0, :, :, 3], cmap='viridis')
# plt.show()

fig, axes = plt.subplots(1, 3, figsize=(20, 8))

# Display the first activation
axes[0].matshow(first_layer_activation[0, :, :, 4], cmap='viridis')
axes[0].set_title('Layer 1')

# Display the second activation
axes[1].matshow(first_layer_activation[0, :, :, 13], cmap='viridis')
axes[1].set_title('Layer 10')

# Display the second activation
axes[2].matshow(first_layer_activation[0, :, :, 49], cmap='viridis')
axes[2].set_title('Layer 26')

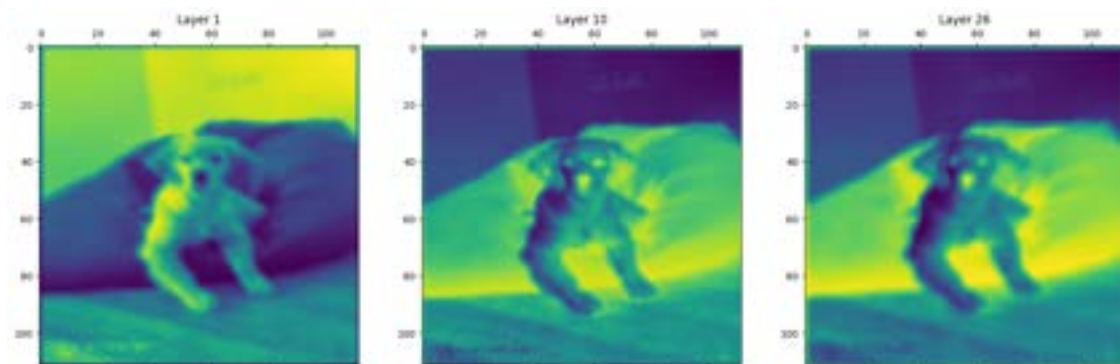
# Show the subplots
plt.show()

```

```

1/1 [=====] - 0s 49ms/step
(1, 112, 112, 64)
(1, 112, 112, 64)
(1, 28, 28, 128)
(1, 7, 7, 2048)

```



```

[ ]: # for i in range(64):

```

```
# plt.matshow(first_layer_activation[0, :, :, i], cmap='viridis')
# plt.show()
```

This one looks like a “bright green dot” detector, useful to encode dog eyes and nose.

At this point, let’s go and plot a complete visualization of all the activations in the network.

### Middle Layer activation

```
[ ]: activations = activation_model.predict([img_tensor])

middle_layer_activation = activations[1]
print(middle_layer_activation.shape)

# for activation in activations:
#     print(activation.shape)

# plt.matshow(middle_layer_activation[0, :, :, 1], cmap='viridis')
# plt.show()

fig, axes = plt.subplots(1, 3, figsize=(20, 8))

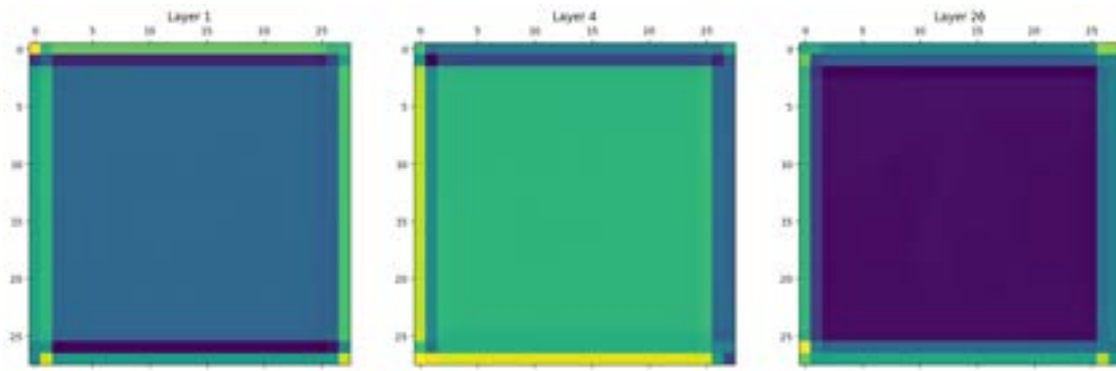
# Display the first activation
axes[0].matshow(middle_layer_activation[0, :, :, 1], cmap='viridis')
axes[0].set_title('Layer 1')

# Display the second activation
axes[1].matshow(middle_layer_activation[0, :, :, 4], cmap='viridis')
axes[1].set_title('Layer 4')

# Display the second activation
axes[2].matshow(middle_layer_activation[0, :, :, 26], cmap='viridis')
axes[2].set_title('Layer 26')

# Show the subplots
plt.show()
```

```
1/1 [=====] - 0s 86ms/step
(1, 28, 28, 128)
```



### Last Layer activation

```
[ ]: activations = activation_model.predict([img_tensor])

last_layer_activation = activations[2]
print(last_layer_activation.shape)

# plt.matshow(last_layer_activation[0, :, :, 1], cmap='viridis')
# plt.show()
# plt.matshow(last_layer_activation[0, :, :, 26], cmap='viridis')
# plt.show()

fig, axes = plt.subplots(1, 3, figsize=(20, 8))

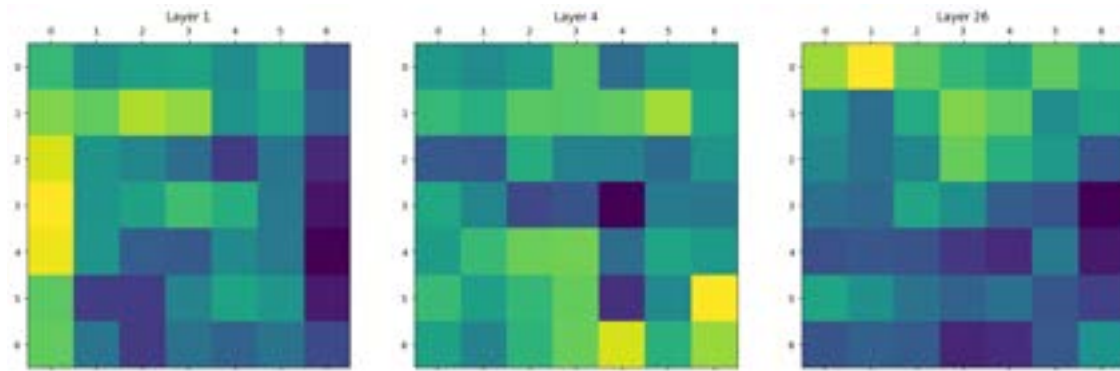
# Display the first activation
axes[0].matshow(last_layer_activation[0, :, :, 1], cmap='viridis')
axes[0].set_title('Layer 1')

# Display the second activation
axes[1].matshow(last_layer_activation[0, :, :, 4], cmap='viridis')
axes[1].set_title('Layer 4')

# Display the second activation
axes[2].matshow(last_layer_activation[0, :, :, 26], cmap='viridis')
axes[2].set_title('Layer 26')

# Show the subplots
plt.show()
```

```
1/1 [=====] - 0s 54ms/step
(1, 7, 7, 2048)
```



```
[ ]: # These are the names of the layers, so can have them as part of our plot
layer_names = []
# layer_names = layers_to_visualize
# for layer in model.layers[:50]:
#     layer_names.append(layer.name)
for layer in layers_to_visualize:
    print(layer.name)
    layer_names.append(layer.name)

images_per_row = 16

# Now let's display our feature maps
for layer_name, layer_activation in zip(layer_names, activations):
    # This is the number of features in the feature map
    n_features = layer_activation.shape[-1]

    # The feature map has shape (1, size, size, n_features)
    size = layer_activation.shape[1]

    # We will tile the activation channels in this matrix
    n_cols = n_features // images_per_row
    display_grid = np.zeros((size * n_cols, images_per_row * size))

    # We'll tile each filter into this big horizontal grid
    for col in range(n_cols):
        for row in range(images_per_row):
            channel_image = layer_activation[0,
                                             :, :,
                                             col * images_per_row + row]

            # Post-process the feature to make it visually palatable
            channel_image -= channel_image.mean()
            channel_image /= channel_image.std()
            channel_image *= 64
            channel_image += 128
```



```

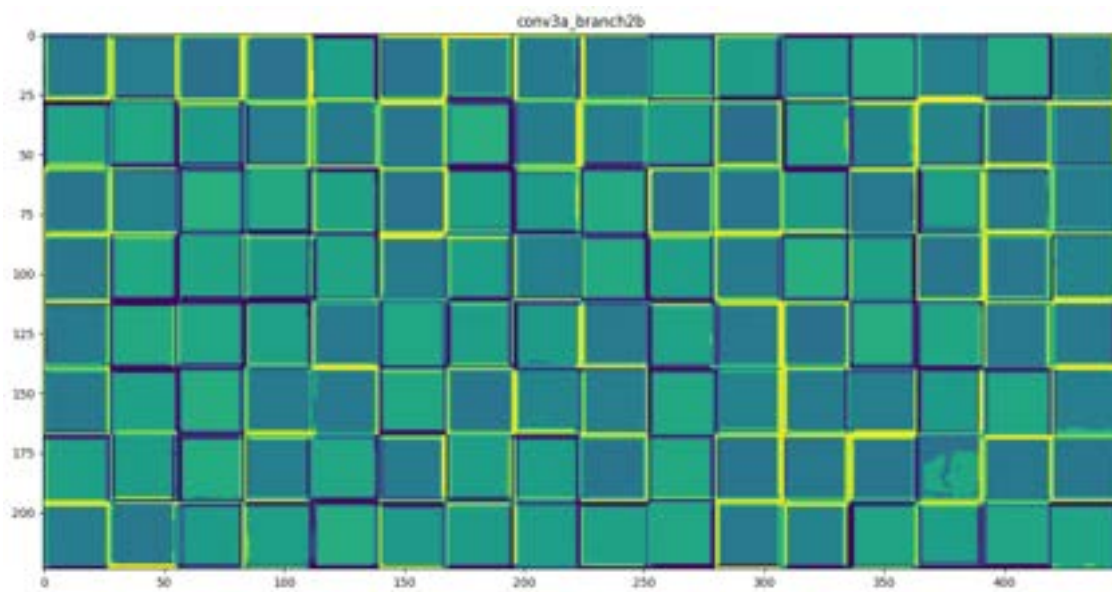
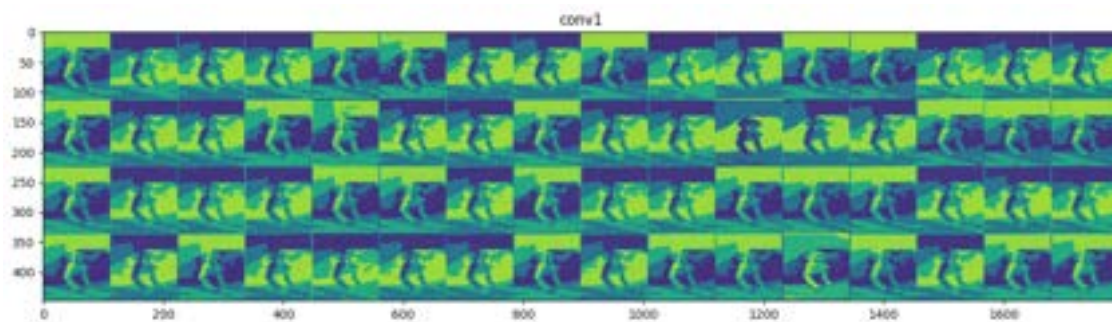
channel_image = np.clip(channel_image, 0, 255).astype('uint8')
display_grid[col * size : (col + 1) * size,
              row * size : (row + 1) * size] = channel_image

# Display the grid
scale = 1. / size
plt.figure(figsize=(scale * display_grid.shape[1],
                    scale * display_grid.shape[0]))
plt.title(layer_name)
plt.grid(False)
plt.imshow(display_grid, aspect='auto', cmap='viridis')

plt.show()

```

conv1  
conv3a\_branch2b  
conv5c\_branch2c







### 1.1.7 PS1.A - Visualizing convnet filters

Chosen blocks/ layers

```
[ ]: model = load_model('model_resnet50_aug4.h5')
```

WARNING:absl:At this time, the v2.11+ optimizer `tf.keras.optimizers.SGD` runs slowly on M1/M2 Macs, please use the legacy Keras optimizer instead, located at `tf.keras.optimizers.legacy.SGD`.

```
[ ]: layers_to_visualize = [model.get_layer('conv1_conv'),
                           model.get_layer('conv3_block2_2_conv'),
                           model.get_layer('conv5_block3_3_conv')]

for layer in layers_to_visualize:
    print(layer.name)
```

```
conv1_conv
conv3_block2_2_conv
conv5_block3_3_conv
```

```
[ ]: def deprocess_image(x):
    # normalize tensor: center on 0., ensure std is 0.1
    x -= x.mean()
    x /= (x.std() + 1e-5)
    x *= 0.1

    # clip to [0, 1]
    x += 0.5
    x = np.clip(x, 0, 1)

    # convert to RGB array
    x *= 255
    x = np.clip(x, 0, 255).astype('uint8')
    return x
```

```
[ ]: # Disable eager execution
tf.compat.v1.disable_eager_execution()
```

```
[ ]: # Modify your deprocess_image function as follows
def deprocess_image(x):
    x = x - x.mean()
    x = x / (x.std() + 1e-5)
    x = x * 0.1 + 0.5
    x = np.clip(x, 0, 1)
    x = x * 255
```

```
x = np.clip(x, 0, 255).astype('uint8')
return x
```

```
[ ]: def generate_pattern(layer_name, filter_index, size=224):

    # Load ResNet50 without top layers
    # model1 = ResNet50(weights=None, include_top=False)

    layer = model.get_layer(layer_name)
    input_img = model.input

    loss = K.mean(layer.output[:, :, :, filter_index])
    grads = K.gradients(loss, input_img)[0]
    grads /= (K.sqrt(K.mean(K.square(grads))) + 1e-5)

    iterate = K.function([input_img], [loss, grads])

    input_img_data = np.random.random((1, size, size, 3)) * 20 + 128.

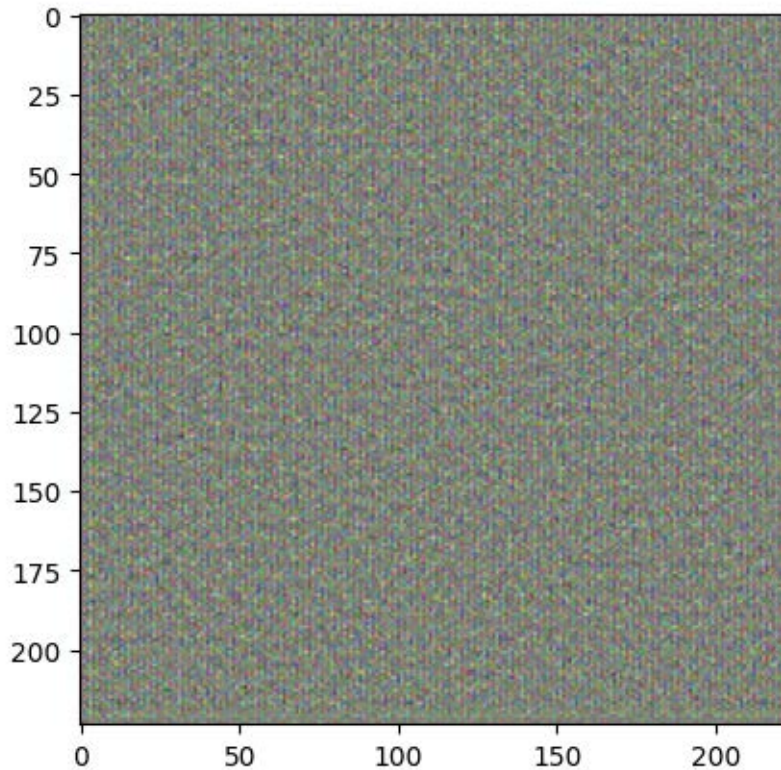
    step = 1.
    for i in range(40):
        loss_value, grads_value = iterate([input_img_data])
        input_img_data += grads_value * step

    img = input_img_data[0]
    return deprocess_image(img)
```

```
[ ]: # Now you can use generate_pattern function
layer_name = 'conv2_block1_1_conv'
filter_index = 0

pattern = generate_pattern(layer_name, filter_index)
plt.imshow(pattern)
plt.show()
```

2023-10-25 22:11:58.001917: W tensorflow/c/c\_api.cc:305] Operation '{name:'conv5\_block2\_1\_conv\_6/kernel/Assign' id:32615 op device:{requested: '', assigned: ''} def:{{{node conv5\_block2\_1\_conv\_6/kernel/Assign}}} = AssignVariableOp[\_has\_manual\_control\_dependencies=true, dtype=DT\_FLOAT, validate\_shape=false](conv5\_block2\_1\_conv\_6/kernel, conv5\_block2\_1\_conv\_6/kernel/Initializer/stateless\_random\_uniform)}}' was changed by setting attribute after it was run by a session. This mutation will have no effect, and will trigger an error in the future. Either don't modify nodes after running them or create a new session.



**First and Middle Conv Blocks** Scroll for both the images

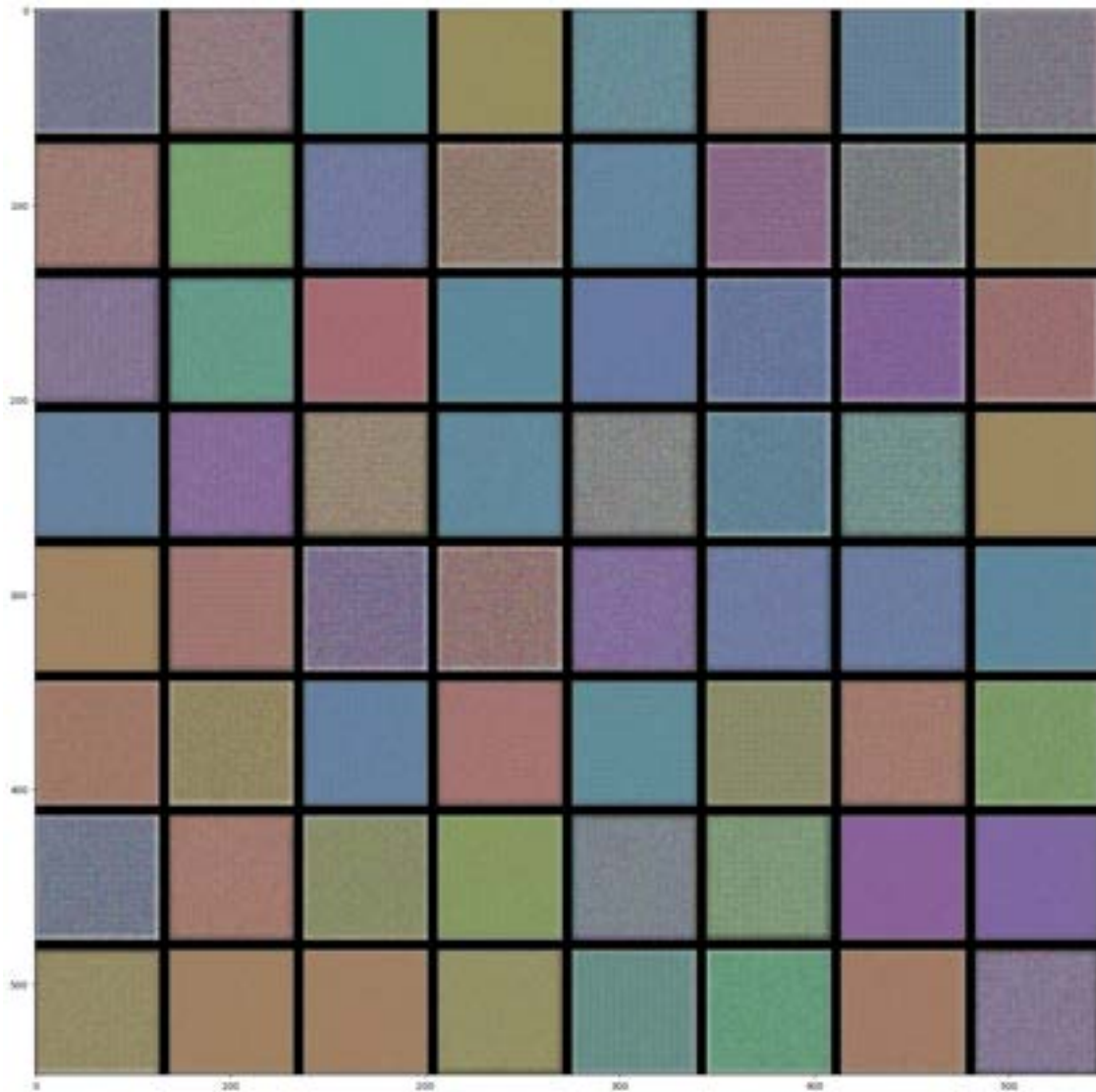
```
[ ]: for layer_name in ['conv1_conv', 'conv3_block2_2_conv', 'conv5_block3_3_conv']:
    size = 64
    margin = 5

    # This a empty (black) image where we will store our results.
    results = np.zeros((8 * size + 7 * margin, 8 * size + 7 * margin, 3))

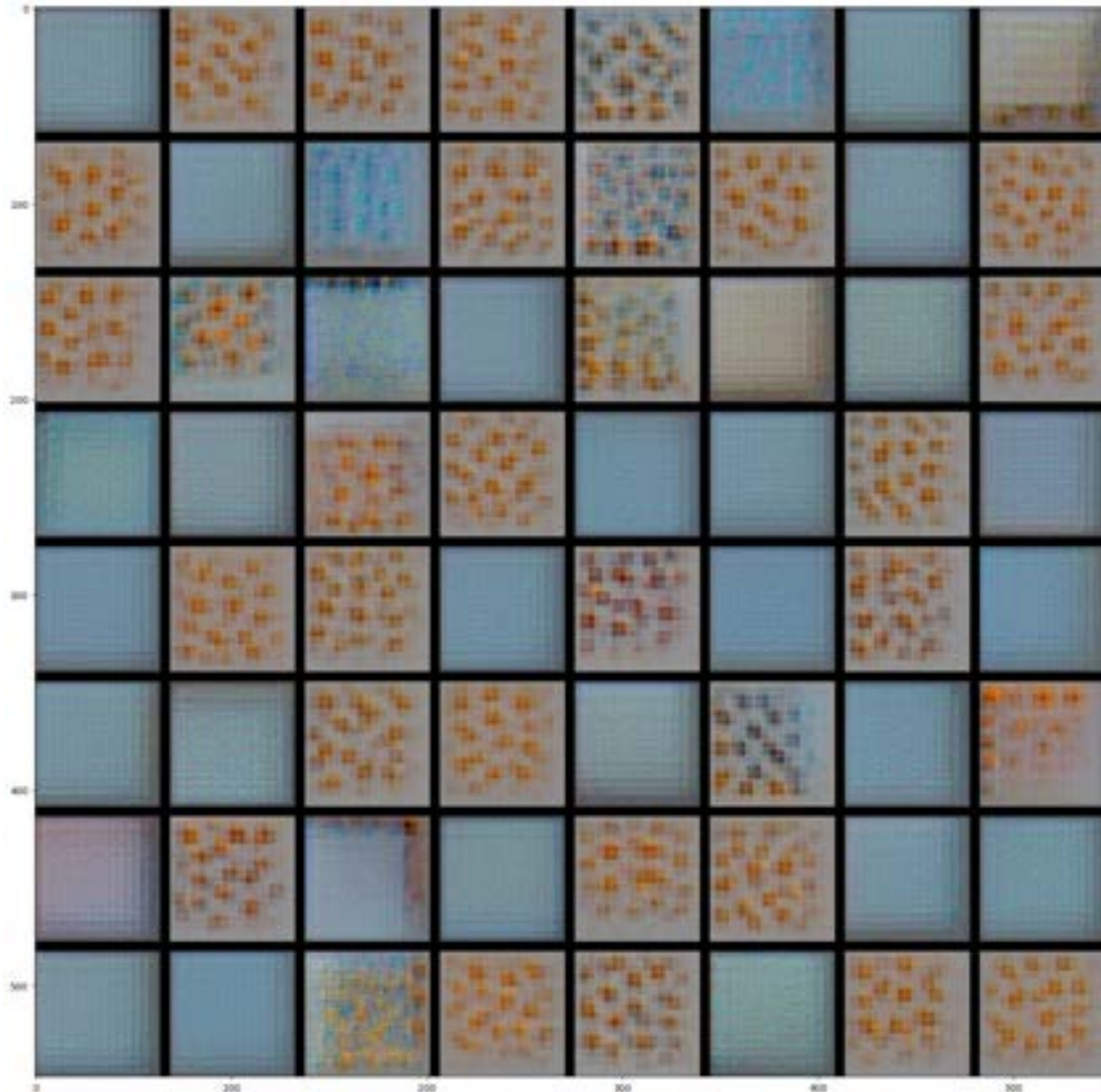
    for i in range(8): # iterate over the rows of our results grid
        for j in range(8): # iterate over the columns of our results grid
            # Generate the pattern for filter `i + (j * 8)` in `layer_name`
            filter_img = generate_pattern(layer_name, i + (j * 8), size=size)

            # Put the result in the square `(i, j)` of the results grid
            horizontal_start = i * size + i * margin
            horizontal_end = horizontal_start + size
            vertical_start = j * size + j * margin
            vertical_end = vertical_start + size
            results[horizontal_start: horizontal_end, vertical_start:
↪vertical_end, :] = filter_img
```

```
# Display the results grid  
plt.figure(figsize=(20, 20))  
plt.imshow((results * 255).astype(np.uint8))  
plt.show()
```







The Kernel crashed while executing code in the the current cell or a previous cell. Please review the code in the cell(s) to identify a possible cause of the failure. Click [here](https://aka.ms/vscodeJupyterKernelCrash) for more info. View Jupyter [log](command:jupyter.viewOutput) for further details.

```
[ ]: model = load_model('model_resnet50_aug4.h5')
model.summary()
```

WARNING:absl:At this time, the v2.11+ optimizer `tf.keras.optimizers.SGD` runs slowly on M1/M2 Macs, please use the legacy Keras optimizer instead, located at `tf.keras.optimizers.legacy.SGD`.

Model: "model"



Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	[]
conv1_pad (ZeroPadding2D) ['input_1[0][0]']	(None, 230, 230, 3)	0	
conv1_conv (Conv2D) ['conv1_pad[0][0]']	(None, 112, 112, 64)	9472	
conv1_bn (BatchNormalizati on) ['conv1_conv[0][0]']	(None, 112, 112, 64)	256	
conv1_relu (Activation) ['conv1_bn[0][0]']	(None, 112, 112, 64)	0	
pool1_pad (ZeroPadding2D) ['conv1_relu[0][0]']	(None, 114, 114, 64)	0	
pool1_pool (MaxPooling2D) ['pool1_pad[0][0]']	(None, 56, 56, 64)	0	
conv2_block1_1_conv (Conv2 D) ['pool1_pool[0][0]']	(None, 56, 56, 64)	4160	
conv2_block1_1_bn (BatchNo rmalization) ['conv2_block1_1_conv[0][0]']	(None, 56, 56, 64)	256	
conv2_block1_1_relu (Activ ation) ['conv2_block1_1_bn[0][0]']	(None, 56, 56, 64)	0	
conv2_block1_2_conv (Conv2 D) ['conv2_block1_1_relu[0][0]']	(None, 56, 56, 64)	36928	
conv2_block1_2_bn (BatchNo rmalization) ['conv2_block1_2_conv[0][0]']	(None, 56, 56, 64)	256	
conv2_block1_2_relu (Activ ation) ['conv2_block1_2_bn[0][0]']	(None, 56, 56, 64)	0	

```

ation)

conv2_block1_0_conv (Conv2 (None, 56, 56, 256)      16640
['pool1_pool[0][0]']
D)

conv2_block1_3_conv (Conv2 (None, 56, 56, 256)      16640
['conv2_block1_2_relu[0][0]']
D)

conv2_block1_0_bn (BatchNo (None, 56, 56, 256)      1024
['conv2_block1_0_conv[0][0]']
rmalization)

conv2_block1_3_bn (BatchNo (None, 56, 56, 256)      1024
['conv2_block1_3_conv[0][0]']
rmalization)

conv2_block1_add (Add)      (None, 56, 56, 256)      0
['conv2_block1_0_bn[0][0]',
'conv2_block1_3_bn[0][0]']

conv2_block1_out (Activati (None, 56, 56, 256)      0
['conv2_block1_add[0][0]']
on)

conv2_block2_1_conv (Conv2 (None, 56, 56, 64)      16448
['conv2_block1_out[0][0]']
D)

conv2_block2_1_bn (BatchNo (None, 56, 56, 64)      256
['conv2_block2_1_conv[0][0]']
rmalization)

conv2_block2_1_relu (Activ (None, 56, 56, 64)      0
['conv2_block2_1_bn[0][0]']
ation)

conv2_block2_2_conv (Conv2 (None, 56, 56, 64)      36928
['conv2_block2_1_relu[0][0]']
D)

conv2_block2_2_bn (BatchNo (None, 56, 56, 64)      256
['conv2_block2_2_conv[0][0]']
rmalization)

conv2_block2_2_relu (Activ (None, 56, 56, 64)      0
['conv2_block2_2_bn[0][0]']

```

ation)		
conv2_block2_3_conv (Conv2 (None, 56, 56, 256)		16640
['conv2_block2_2_relu[0][0]']		
D)		
conv2_block2_3_bn (BatchNo (None, 56, 56, 256)		1024
['conv2_block2_3_conv[0][0]']		
rmalization)		
conv2_block2_add (Add) (None, 56, 56, 256)		0
['conv2_block1_out[0][0]',		
'conv2_block2_3_bn[0][0]']		
conv2_block2_out (Activati (None, 56, 56, 256)		0
['conv2_block2_add[0][0]']		
on)		
conv2_block3_1_conv (Conv2 (None, 56, 56, 64)		16448
['conv2_block2_out[0][0]']		
D)		
conv2_block3_1_bn (BatchNo (None, 56, 56, 64)		256
['conv2_block3_1_conv[0][0]']		
rmalization)		
conv2_block3_1_relu (Activ (None, 56, 56, 64)		0
['conv2_block3_1_bn[0][0]']		
ation)		
conv2_block3_2_conv (Conv2 (None, 56, 56, 64)		36928
['conv2_block3_1_relu[0][0]']		
D)		
conv2_block3_2_bn (BatchNo (None, 56, 56, 64)		256
['conv2_block3_2_conv[0][0]']		
rmalization)		
conv2_block3_2_relu (Activ (None, 56, 56, 64)		0
['conv2_block3_2_bn[0][0]']		
ation)		
conv2_block3_3_conv (Conv2 (None, 56, 56, 256)		16640
['conv2_block3_2_relu[0][0]']		
D)		
conv2_block3_3_bn (BatchNo (None, 56, 56, 256)		1024
['conv2_block3_3_conv[0][0]']		

```

rmalization)

conv2_block3_add (Add)      (None, 56, 56, 256)      0
['conv2_block2_out[0][0]',
'conv2_block3_3_bn[0][0]']

conv2_block3_out (Activati (None, 56, 56, 256)      0
['conv2_block3_add[0][0]'
on)

conv3_block1_1_conv (Conv2  (None, 28, 28, 128)      32896
['conv2_block3_out[0][0]'
D)

conv3_block1_1_bn (BatchNo  (None, 28, 28, 128)      512
['conv3_block1_1_conv[0][0]'
rmalization)

conv3_block1_1_relu (Activ  (None, 28, 28, 128)      0
['conv3_block1_1_bn[0][0]'
ation)

conv3_block1_2_conv (Conv2  (None, 28, 28, 128)      147584
['conv3_block1_1_relu[0][0]'
D)

conv3_block1_2_bn (BatchNo  (None, 28, 28, 128)      512
['conv3_block1_2_conv[0][0]'
rmalization)

conv3_block1_2_relu (Activ  (None, 28, 28, 128)      0
['conv3_block1_2_bn[0][0]'
ation)

conv3_block1_0_conv (Conv2  (None, 28, 28, 512)      131584
['conv2_block3_out[0][0]'
D)

conv3_block1_3_conv (Conv2  (None, 28, 28, 512)      66048
['conv3_block1_2_relu[0][0]'
D)

conv3_block1_0_bn (BatchNo  (None, 28, 28, 512)      2048
['conv3_block1_0_conv[0][0]'
rmalization)

conv3_block1_3_bn (BatchNo  (None, 28, 28, 512)      2048
['conv3_block1_3_conv[0][0]'

```

```

rmalization)

conv3_block1_add (Add)      (None, 28, 28, 512)      0
['conv3_block1_0_bn[0][0]',
'conv3_block1_3_bn[0][0]']

conv3_block1_out (Activati (None, 28, 28, 512)      0
['conv3_block1_add[0][0]'
on)

conv3_block2_1_conv (Conv2   (None, 28, 28, 128)      65664
['conv3_block1_out[0][0]']
D)

conv3_block2_1_bn (BatchNo   (None, 28, 28, 128)      512
['conv3_block2_1_conv[0][0]']
rmalization)

conv3_block2_1_relu (Activ   (None, 28, 28, 128)      0
['conv3_block2_1_bn[0][0]']
ation)

conv3_block2_2_conv (Conv2   (None, 28, 28, 128)      147584
['conv3_block2_1_relu[0][0]']
D)

conv3_block2_2_bn (BatchNo   (None, 28, 28, 128)      512
['conv3_block2_2_conv[0][0]']
rmalization)

conv3_block2_2_relu (Activ   (None, 28, 28, 128)      0
['conv3_block2_2_bn[0][0]']
ation)

conv3_block2_3_conv (Conv2   (None, 28, 28, 512)      66048
['conv3_block2_2_relu[0][0]']
D)

conv3_block2_3_bn (BatchNo   (None, 28, 28, 512)      2048
['conv3_block2_3_conv[0][0]']
rmalization)

conv3_block2_add (Add)      (None, 28, 28, 512)      0
['conv3_block1_out[0][0]',
'conv3_block2_3_bn[0][0]']

conv3_block2_out (Activati   (None, 28, 28, 512)      0
['conv3_block2_add[0][0]']

```

```

on)

conv3_block3_1_conv (Conv2D (None, 28, 28, 128) 65664
['conv3_block2_out[0][0]'])

conv3_block3_1_bn (Batch Normalization (None, 28, 28, 128) 512
['conv3_block3_1_conv[0][0]'])

conv3_block3_1_relu (ReLU (None, 28, 28, 128) 0
['conv3_block3_1_bn[0][0]'])

conv3_block3_2_conv (Conv2D (None, 28, 28, 128) 147584
['conv3_block3_1_relu[0][0]'])

conv3_block3_2_bn (Batch Normalization (None, 28, 28, 128) 512
['conv3_block3_2_conv[0][0]'])

conv3_block3_2_relu (ReLU (None, 28, 28, 128) 0
['conv3_block3_2_bn[0][0]'])

conv3_block3_3_conv (Conv2D (None, 28, 28, 512) 66048
['conv3_block3_2_relu[0][0]'])

conv3_block3_3_bn (Batch Normalization (None, 28, 28, 512) 2048
['conv3_block3_3_conv[0][0]'])

conv3_block3_add (Add (None, 28, 28, 512) 0
['conv3_block2_out[0][0]',
'conv3_block3_3_bn[0][0]'])

conv3_block3_out (Activation (None, 28, 28, 512) 0
['conv3_block3_add[0][0]'])

conv3_block4_1_conv (Conv2D (None, 28, 28, 128) 65664
['conv3_block3_out[0][0]'])

conv3_block4_1_bn (Batch Normalization (None, 28, 28, 128) 512
['conv3_block4_1_conv[0][0]'])

```

```

rmalization)

conv3_block4_1_relu (Activ (None, 28, 28, 128) 0
['conv3_block4_1_bn[0][0]']
ation)

conv3_block4_2_conv (Conv2 (None, 28, 28, 128) 147584
['conv3_block4_1_relu[0][0]']
D)

conv3_block4_2_bn (BatchNo (None, 28, 28, 128) 512
['conv3_block4_2_conv[0][0]']
rmalization)

conv3_block4_2_relu (Activ (None, 28, 28, 128) 0
['conv3_block4_2_bn[0][0]']
ation)

conv3_block4_3_conv (Conv2 (None, 28, 28, 512) 66048
['conv3_block4_2_relu[0][0]']
D)

conv3_block4_3_bn (BatchNo (None, 28, 28, 512) 2048
['conv3_block4_3_conv[0][0]']
rmalization)

conv3_block4_add (Add) (None, 28, 28, 512) 0
['conv3_block3_out[0][0]',
'conv3_block4_3_bn[0][0]']

conv3_block4_out (Activati (None, 28, 28, 512) 0
['conv3_block4_add[0][0]']
on)

conv4_block1_1_conv (Conv2 (None, 14, 14, 256) 131328
['conv3_block4_out[0][0]']
D)

conv4_block1_1_bn (BatchNo (None, 14, 14, 256) 1024
['conv4_block1_1_conv[0][0]']
rmalization)

conv4_block1_1_relu (Activ (None, 14, 14, 256) 0
['conv4_block1_1_bn[0][0]']
ation)

conv4_block1_2_conv (Conv2 (None, 14, 14, 256) 590080
['conv4_block1_1_relu[0][0]']

```

D)

conv4_block1_2_bn (BatchNormal ization)	(None, 14, 14, 256)	1024
conv4_block1_2_conv [0] [0]'		
conv4_block1_2_relu (Activation)	(None, 14, 14, 256)	0
conv4_block1_2_bn [0] [0]'		
conv4_block1_0_conv (Conv2D)	(None, 14, 14, 1024)	525312
conv4_block1_3_conv (Conv2D)	(None, 14, 14, 1024)	263168
conv4_block1_0_bn (BatchNormal ization)	(None, 14, 14, 1024)	4096
conv4_block1_3_bn (BatchNormal ization)	(None, 14, 14, 1024)	4096
conv4_block1_add (Add)	(None, 14, 14, 1024)	0
conv4_block1_out (Activation)	(None, 14, 14, 1024)	0
conv4_block2_1_conv (Conv2D)	(None, 14, 14, 256)	262400
conv4_block2_1_bn (BatchNormal ization)	(None, 14, 14, 256)	1024
conv4_block2_1_relu (Activation)	(None, 14, 14, 256)	0
conv4_block2_2_conv (Conv2D)	(None, 14, 14, 256)	590080



D)

conv4_block2_2_bn (BatchNormal ization)	(None, 14, 14, 256)	1024
conv4_block2_2_conv (Conv2D)	(None, 14, 14, 256)	0
conv4_block2_2_relu (Activation)	(None, 14, 14, 256)	0
conv4_block2_3_conv (Conv2D)	(None, 14, 14, 1024)	263168
conv4_block2_3_bn (BatchNormal ization)	(None, 14, 14, 1024)	4096
conv4_block2_add (Add)	(None, 14, 14, 1024)	0
conv4_block2_out (Activation)	(None, 14, 14, 1024)	0
conv4_block3_1_conv (Conv2D)	(None, 14, 14, 256)	262400
conv4_block3_1_bn (BatchNormal ization)	(None, 14, 14, 256)	1024
conv4_block3_1_relu (Activation)	(None, 14, 14, 256)	0
conv4_block3_2_conv (Conv2D)	(None, 14, 14, 256)	590080
conv4_block3_2_bn (BatchNormal ization)	(None, 14, 14, 256)	1024
conv4_block3_2_relu (Activation)	(None, 14, 14, 256)	0

ation)	
conv4_block3_3_conv (Conv2D) (None, 14, 14, 1024)	263168
['conv4_block3_2_relu[0][0]']	
conv4_block3_3_bn (Batch Normalization) (None, 14, 14, 1024)	4096
['conv4_block3_3_conv[0][0]']	
conv4_block3_add (Add) (None, 14, 14, 1024)	0
['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	
conv4_block3_out (Activation) (None, 14, 14, 1024)	0
['conv4_block3_add[0][0]']	
conv4_block4_1_conv (Conv2D) (None, 14, 14, 256)	262400
['conv4_block3_out[0][0]']	
conv4_block4_1_bn (Batch Normalization) (None, 14, 14, 256)	1024
['conv4_block4_1_conv[0][0]']	
conv4_block4_1_relu (Activation) (None, 14, 14, 256)	0
['conv4_block4_1_bn[0][0]']	
conv4_block4_2_conv (Conv2D) (None, 14, 14, 256)	590080
['conv4_block4_1_relu[0][0]']	
conv4_block4_2_bn (Batch Normalization) (None, 14, 14, 256)	1024
['conv4_block4_2_conv[0][0]']	
conv4_block4_2_relu (Activation) (None, 14, 14, 256)	0
['conv4_block4_2_bn[0][0]']	
conv4_block4_3_conv (Conv2D) (None, 14, 14, 1024)	263168
['conv4_block4_2_relu[0][0]']	
conv4_block4_3_bn (Batch Normalization) (None, 14, 14, 1024)	4096
['conv4_block4_3_conv[0][0]']	

```

rmalization)

conv4_block4_add (Add)      (None, 14, 14, 1024)      0
['conv4_block3_out[0][0]',
'conv4_block4_3_bn[0][0]']

conv4_block4_out (Activati (None, 14, 14, 1024)      0
['conv4_block4_add[0][0]'
on)

conv4_block5_1_conv (Conv2  (None, 14, 14, 256)      262400
['conv4_block4_out[0][0]'
D)

conv4_block5_1_bn (BatchNo  (None, 14, 14, 256)      1024
['conv4_block5_1_conv[0][0]'
rmalization)

conv4_block5_1_relu (Activ  (None, 14, 14, 256)      0
['conv4_block5_1_bn[0][0]'
ation)

conv4_block5_2_conv (Conv2  (None, 14, 14, 256)      590080
['conv4_block5_1_relu[0][0]'
D)

conv4_block5_2_bn (BatchNo  (None, 14, 14, 256)      1024
['conv4_block5_2_conv[0][0]'
rmalization)

conv4_block5_2_relu (Activ  (None, 14, 14, 256)      0
['conv4_block5_2_bn[0][0]'
ation)

conv4_block5_3_conv (Conv2  (None, 14, 14, 1024)     263168
['conv4_block5_2_relu[0][0]'
D)

conv4_block5_3_bn (BatchNo  (None, 14, 14, 1024)     4096
['conv4_block5_3_conv[0][0]'
rmalization)

conv4_block5_add (Add)      (None, 14, 14, 1024)      0
['conv4_block4_out[0][0]',
'conv4_block5_3_bn[0][0]']

conv4_block5_out (Activati (None, 14, 14, 1024)      0
['conv4_block5_add[0][0]']

```

```

on)

conv4_block6_1_conv (Conv2 (None, 14, 14, 256)      262400
['conv4_block5_out[0][0]']
D)

conv4_block6_1_bn (BatchNo (None, 14, 14, 256)      1024
['conv4_block6_1_conv[0][0]']
rmalization)

conv4_block6_1_relu (Activ (None, 14, 14, 256)      0
['conv4_block6_1_bn[0][0]']
ation)

conv4_block6_2_conv (Conv2 (None, 14, 14, 256)      590080
['conv4_block6_1_relu[0][0]']
D)

conv4_block6_2_bn (BatchNo (None, 14, 14, 256)      1024
['conv4_block6_2_conv[0][0]']
rmalization)

conv4_block6_2_relu (Activ (None, 14, 14, 256)      0
['conv4_block6_2_bn[0][0]']
ation)

conv4_block6_3_conv (Conv2 (None, 14, 14, 1024)     263168
['conv4_block6_2_relu[0][0]']
D)

conv4_block6_3_bn (BatchNo (None, 14, 14, 1024)     4096
['conv4_block6_3_conv[0][0]']
rmalization)

conv4_block6_add (Add) (None, 14, 14, 1024)      0
['conv4_block5_out[0][0]',
'conv4_block6_3_bn[0][0]']

conv4_block6_out (Activati (None, 14, 14, 1024)     0
['conv4_block6_add[0][0]']
on)

conv5_block1_1_conv (Conv2 (None, 7, 7, 512)      524800
['conv4_block6_out[0][0]']
D)

conv5_block1_1_bn (BatchNo (None, 7, 7, 512)      2048
['conv5_block1_1_conv[0][0]']

```

rmalization)		
conv5_block1_1_relu (Activ ['conv5_block1_1_bn[0][0]'] ation)	(None, 7, 7, 512)	0
conv5_block1_2_conv (Conv2 ['conv5_block1_1_relu[0][0]'] D)	(None, 7, 7, 512)	2359808
conv5_block1_2_bn (BatchNo ['conv5_block1_2_conv[0][0]'] rmalization)	(None, 7, 7, 512)	2048
conv5_block1_2_relu (Activ ['conv5_block1_2_bn[0][0]'] ation)	(None, 7, 7, 512)	0
conv5_block1_0_conv (Conv2 ['conv4_block6_out[0][0]'] D)	(None, 7, 7, 2048)	2099200
conv5_block1_3_conv (Conv2 ['conv5_block1_2_relu[0][0]'] D)	(None, 7, 7, 2048)	1050624
conv5_block1_0_bn (BatchNo ['conv5_block1_0_conv[0][0]'] rmalization)	(None, 7, 7, 2048)	8192
conv5_block1_3_bn (BatchNo ['conv5_block1_3_conv[0][0]'] rmalization)	(None, 7, 7, 2048)	8192
conv5_block1_add (Add) ['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']	(None, 7, 7, 2048)	0
conv5_block1_out (Activati ['conv5_block1_add[0][0]'] on)	(None, 7, 7, 2048)	0
conv5_block2_1_conv (Conv2 ['conv5_block1_out[0][0]'] D)	(None, 7, 7, 512)	1049088
conv5_block2_1_bn (BatchNo ['conv5_block2_1_conv[0][0]']	(None, 7, 7, 512)	2048

rmalization)	
conv5_block2_1_relu (Activ (None, 7, 7, 512) ['conv5_block2_1_bn[0][0]'] ation)	0
conv5_block2_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block2_1_relu[0][0]'] D)	2359808
conv5_block2_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_2_conv[0][0]'] rmalization)	2048
conv5_block2_2_relu (Activ (None, 7, 7, 512) ['conv5_block2_2_bn[0][0]'] ation)	0
conv5_block2_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block2_2_relu[0][0]'] D)	1050624
conv5_block2_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block2_3_conv[0][0]'] rmalization)	8192
conv5_block2_add (Add) (None, 7, 7, 2048) ['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']	0
conv5_block2_out (Activati (None, 7, 7, 2048) ['conv5_block2_add[0][0]'] on)	0
conv5_block3_1_conv (Conv2 (None, 7, 7, 512) ['conv5_block2_out[0][0]'] D)	1049088
conv5_block3_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block3_1_conv[0][0]'] rmalization)	2048
conv5_block3_1_relu (Activ (None, 7, 7, 512) ['conv5_block3_1_bn[0][0]'] ation)	0
conv5_block3_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block3_1_relu[0][0]']	2359808

D)

conv5_block3_2_bn (BatchNormal- ization) ['conv5_block3_2_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block3_2_relu (Activation) ['conv5_block3_2_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block3_3_conv (Conv2D) ['conv5_block3_2_relu[0][0]']	(None, 7, 7, 2048)	1050624
conv5_block3_3_bn (BatchNormal- ization) ['conv5_block3_3_conv[0][0]']	(None, 7, 7, 2048)	8192
conv5_block3_add (Add) ['conv5_block2_out[0][0]', 'conv5_block3_3_bn[0][0]']	(None, 7, 7, 2048)	0
conv5_block3_out (Activation) ['conv5_block3_add[0][0]']	(None, 7, 7, 2048)	0
average_pooling2d (AveragePooling2D) ['conv5_block3_out[0][0]']	(None, 3, 3, 2048)	0
flatten (Flatten) ['average_pooling2d[0][0]']	(None, 18432)	0
dense (Dense) ['flatten[0][0]']	(None, 9)	165897

```
=====
=====
Total params: 23753609 (90.61 MB)
Trainable params: 23700489 (90.41 MB)
Non-trainable params: 53120 (207.50 KB)
-----
-----
```

**Last Conv Block** Plotted the last layer separately to “avoid kernel crash”

```
[ ]: for layer_name in ['conv5_block3_3_conv']:
    size = 64
    margin = 5

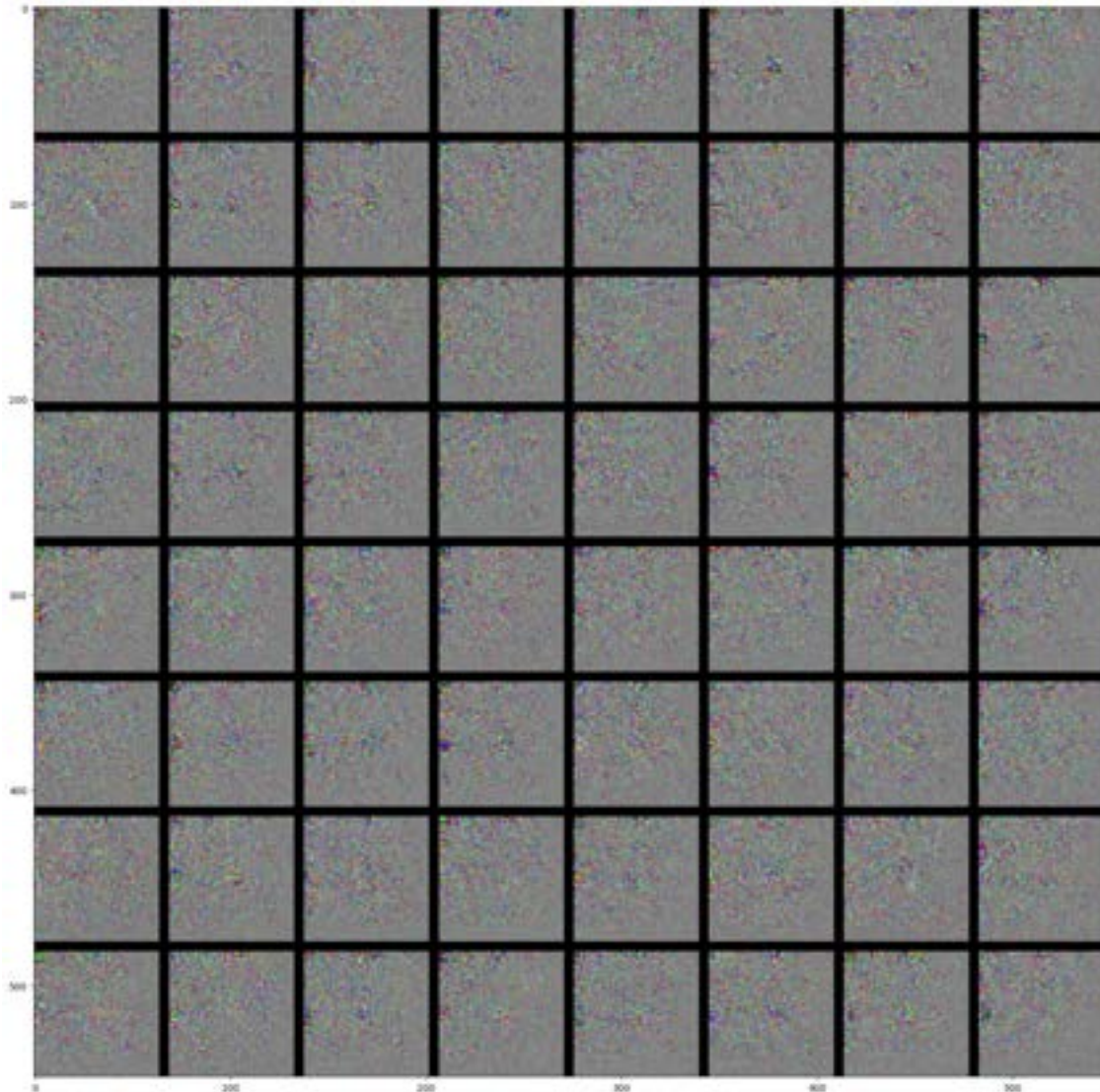
    # This a empty (black) image where we will store our results.
    results = np.zeros((8 * size + 7 * margin, 8 * size + 7 * margin, 3))

    for i in range(8): # iterate over the rows of our results grid
        for j in range(8): # iterate over the columns of our results grid
            # Generate the pattern for filter `i + (j * 8)` in `layer_name`
            filter_img = generate_pattern(layer_name, i + (j * 8), size=size)

            # Put the result in the square `(i, j)` of the results grid
            horizontal_start = i * size + i * margin
            horizontal_end = horizontal_start + size
            vertical_start = j * size + j * margin
            vertical_end = vertical_start + size
            results[horizontal_start: horizontal_end, vertical_start:
↪vertical_end, :] = filter_img

    # Display the results grid
    plt.figure(figsize=(20, 20))
    plt.imshow((results * 255).astype(np.uint8))
    plt.show()
```






### 1.1.8 PS1A. Visualizing heatmaps of class activation

```
[ ]: model = load_model('model_resnet50_aug4.h5')
```

```
2023-10-25 23:21:25.936815: W tensorflow/c/c_api.cc:305] Operation
'{name:'conv3_block4_2_bn_1/beta/Assign' id:9217 op device:{requested: '',
assigned: ''} def:{{{node conv3_block4_2_bn_1/beta/Assign}}} =
AssignVariableOp[_has_manual_control_dependencies=true, dtype=DT_FLOAT,
validate_shape=false](conv3_block4_2_bn_1/beta,
conv3_block4_2_bn_1/beta/Initializer/zeros)}}' was changed by setting attribute
after it was run by a session. This mutation will have no effect, and will
trigger an error in the future. Either don't modify nodes after running them or
create a new session.
```

WARNING:absl:At this time, the v2.11+ optimizer `tf.keras.optimizers.SGD` runs slowly on M1/M2 Macs, please use the legacy Keras optimizer instead, located at `tf.keras.optimizers.legacy.SGD`.

```
[ ]: import numpy as np

# The local path to our target image

img_path = 'refs/dog_1.jpg'

im_size = 224

# `img` is a PIL image of size 224x224
img = tf.keras.utils.load_img(img_path, target_size=(im_size, im_size))

# `x` is a float32 Numpy array of shape (224, 224, 3)
img_tensor = tf.keras.utils.img_to_array(img)

# We add a dimension to transform our array into a "batch"
# of size (1, 224, 224, 3)
img_tensor = np.expand_dims(img_tensor, axis=0)

img_tensor /= 255.

# # Its shape is (1, 224, 224, 3)
print(img_tensor.shape)
```

(1, 224, 224, 3)

```
[ ]: preds = model.predict(img_tensor)
def custom_decode_predictions(predictions, top=5, class_names=None):
    if class_names is None:
        class_names = [str(i) for i in range(predictions.shape[-1])]

    results = []
    for pred in predictions:
        top_indices = pred.argsort()[-top:][::-1]
        decoded_preds = [(class_names[i], pred[i]) for i in top_indices]
        results.append(decoded_preds)

    return results

classes = ['airplane', 'automobile', 'bird', 'cat', 'dog', 'deer', 'frog',
↳ 'horse', 'truck']

print('Predicted:', custom_decode_predictions(preds, top=5,
↳ class_names=classes))
```

```

/Users/banani/Library/Python/3.9/lib/python/site-
packages/keras/src/engine/training_v1.py:2359: UserWarning:
`Model.state_updates` will be removed in a future version. This property should
not be used in TensorFlow 2.0, as `updates` are applied automatically.
  updates=self.state_updates,
2023-10-25 23:21:32.224473: W tensorflow/c/c_api.cc:305] Operation
'{name:'dense_1/Softmax' id:11531 op device:{requested: '', assigned: ''}
def:{{{node dense_1/Softmax}} = Softmax[T=DT_FLOAT,
_has_manual_control_dependencies=true](dense_1/BiasAdd)}}' was changed by
setting attribute after it was run by a session. This mutation will have no
effect, and will trigger an error in the future. Either don't modify nodes after
running them or create a new session.
2023-10-25 23:21:32.709967: W tensorflow/c/c_api.cc:305] Operation
'{name:'total_1/Assign' id:12826 op device:{requested: '', assigned: ''}
def:{{{node total_1/Assign}} =
AssignVariableOp[_has_manual_control_dependencies=true, dtype=DT_FLOAT,
validate_shape=false](total_1, total_1/Initializer/zeros)}}' was changed by
setting attribute after it was run by a session. This mutation will have no
effect, and will trigger an error in the future. Either don't modify nodes after
running them or create a new session.

Predicted: [[('dog', 0.16175188), ('frog', 0.15776682), ('horse', 0.15381174),
('deer', 0.12696117), ('cat', 0.10170399)]]

```

### Chosen Layers or blocks

```

[ ]: layers_to_visualize = [model.get_layer('conv1_relu'),
                             model.get_layer('conv3_block2_2_relu'),
                             model.get_layer('conv5_block3_2_relu')]

for layer in layers_to_visualize:
    print(layer.name)

```

```

conv1_relu
conv3_block2_2_relu
conv5_block3_2_relu

```

```

[ ]: model.summary()

```

```

Model: "model"

```

```

-----
-----
Layer (type)                Output Shape                Param #   Connected to
=====
input_1 (InputLayer)        [(None, 224, 224, 3)]      0         []

conv1_pad (ZeroPadding2D)   (None, 230, 230, 3)        0         ['input_1[0][0]']

```

conv1_conv (Conv2D) ['conv1_pad[0][0]']	(None, 112, 112, 64)	9472
conv1_bn (BatchNormalizati on)	(None, 112, 112, 64)	256
conv1_relu (Activation) ['conv1_bn[0][0]']	(None, 112, 112, 64)	0
pool1_pad (ZeroPadding2D) ['conv1_relu[0][0]']	(None, 114, 114, 64)	0
pool1_pool (MaxPooling2D) ['pool1_pad[0][0]']	(None, 56, 56, 64)	0
conv2_block1_1_conv (Conv2 D)	(None, 56, 56, 64)	4160
conv2_block1_1_bn (BatchNo rmalization)	(None, 56, 56, 64)	256
conv2_block1_1_relu (Activ ation)	(None, 56, 56, 64)	0
conv2_block1_2_conv (Conv2 D)	(None, 56, 56, 64)	36928
conv2_block1_2_bn (BatchNo rmalization)	(None, 56, 56, 64)	256
conv2_block1_2_relu (Activ ation)	(None, 56, 56, 64)	0
conv2_block1_0_conv (Conv2 D)	(None, 56, 56, 256)	16640
conv2_block1_3_conv (Conv2 D)	(None, 56, 56, 256)	16640

conv2_block1_0_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_0_conv[0][0]'] rmalization)	1024
conv2_block1_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_3_conv[0][0]'] rmalization)	1024
conv2_block1_add (Add) (None, 56, 56, 256) ['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']	0
conv2_block1_out (Activati (None, 56, 56, 256) ['conv2_block1_add[0][0]'] on)	0
conv2_block2_1_conv (Conv2 (None, 56, 56, 64) ['conv2_block1_out[0][0]'] D)	16448
conv2_block2_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block2_1_conv[0][0]'] rmalization)	256
conv2_block2_1_relu (Activ (None, 56, 56, 64) ['conv2_block2_1_bn[0][0]'] ation)	0
conv2_block2_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block2_1_relu[0][0]'] D)	36928
conv2_block2_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block2_2_conv[0][0]'] rmalization)	256
conv2_block2_2_relu (Activ (None, 56, 56, 64) ['conv2_block2_2_bn[0][0]'] ation)	0
conv2_block2_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block2_2_relu[0][0]'] D)	16640
conv2_block2_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block2_3_conv[0][0]'] rmalization)	1024

conv2_block2_add (Add) ['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block2_out (Activation) ['conv2_block2_add[0][0]']	(None, 56, 56, 256)	0
conv2_block3_1_conv (Conv2D) ['conv2_block2_out[0][0]']	(None, 56, 56, 64)	16448
conv2_block3_1_bn (Batch Normalization) ['conv2_block3_1_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block3_1_relu (Activation) ['conv2_block3_1_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block3_2_conv (Conv2D) ['conv2_block3_1_relu[0][0]']	(None, 56, 56, 64)	36928
conv2_block3_2_bn (Batch Normalization) ['conv2_block3_2_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block3_2_relu (Activation) ['conv2_block3_2_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block3_3_conv (Conv2D) ['conv2_block3_2_relu[0][0]']	(None, 56, 56, 256)	16640
conv2_block3_3_bn (Batch Normalization) ['conv2_block3_3_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block3_add (Add) ['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block3_out (Activation) ['conv2_block3_add[0][0]']	(None, 56, 56, 256)	0

conv3_block1_1_conv (Conv2D)	(None, 28, 28, 128)	32896
conv3_block1_1_bn (Batch Normalization)	(None, 28, 28, 128)	512
conv3_block1_1_relu (Activation)	(None, 28, 28, 128)	0
conv3_block1_2_conv (Conv2D)	(None, 28, 28, 128)	147584
conv3_block1_2_bn (Batch Normalization)	(None, 28, 28, 128)	512
conv3_block1_2_relu (Activation)	(None, 28, 28, 128)	0
conv3_block1_0_conv (Conv2D)	(None, 28, 28, 512)	131584
conv3_block1_3_conv (Conv2D)	(None, 28, 28, 512)	66048
conv3_block1_0_bn (Batch Normalization)	(None, 28, 28, 512)	2048
conv3_block1_3_bn (Batch Normalization)	(None, 28, 28, 512)	2048
conv3_block1_add (Add)	(None, 28, 28, 512)	0
conv3_block1_out (Activation)	(None, 28, 28, 512)	0

conv3_block2_1_conv (Conv2D) ['conv3_block1_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block2_1_bn (Batch Normalization) ['conv3_block2_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block2_1_relu (Activation) ['conv3_block2_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block2_2_conv (Conv2D) ['conv3_block2_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block2_2_bn (Batch Normalization) ['conv3_block2_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block2_2_relu (Activation) ['conv3_block2_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block2_3_conv (Conv2D) ['conv3_block2_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block2_3_bn (Batch Normalization) ['conv3_block2_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block2_add (Add) ['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block2_out (Activation) ['conv3_block2_add[0][0]']	(None, 28, 28, 512)	0
conv3_block3_1_conv (Conv2D) ['conv3_block2_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block3_1_bn (Batch Normalization) ['conv3_block3_1_conv[0][0]']	(None, 28, 28, 128)	512



conv3_block3_1_relu (Activation) (None, 28, 28, 128) ['conv3_block3_1_bn[0][0]'])	0
conv3_block3_2_conv (Conv2D) (None, 28, 28, 128) ['conv3_block3_1_relu[0][0]'])	147584
conv3_block3_2_bn (Batch Normalization) (None, 28, 28, 128) ['conv3_block3_2_conv[0][0]'])	512
conv3_block3_2_relu (Activation) (None, 28, 28, 128) ['conv3_block3_2_bn[0][0]'])	0
conv3_block3_3_conv (Conv2D) (None, 28, 28, 512) ['conv3_block3_2_relu[0][0]'])	66048
conv3_block3_3_bn (Batch Normalization) (None, 28, 28, 512) ['conv3_block3_3_conv[0][0]'])	2048
conv3_block3_add (Add) (None, 28, 28, 512) ['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']	0
conv3_block3_out (Activation) (None, 28, 28, 512) ['conv3_block3_add[0][0]']	0
conv3_block4_1_conv (Conv2D) (None, 28, 28, 128) ['conv3_block3_out[0][0]']	65664
conv3_block4_1_bn (Batch Normalization) (None, 28, 28, 128) ['conv3_block4_1_conv[0][0]']	512
conv3_block4_1_relu (Activation) (None, 28, 28, 128) ['conv3_block4_1_bn[0][0]']	0
conv3_block4_2_conv (Conv2D) (None, 28, 28, 128) ['conv3_block4_1_relu[0][0]']	147584

conv3_block4_2_bn (BatchNormal-ization)	(None, 28, 28, 128)	512
conv3_block4_2_conv (Conv2D)	(None, 28, 28, 128)	0
conv3_block4_2_relu (Activation)	(None, 28, 28, 128)	0
conv3_block4_3_conv (Conv2D)	(None, 28, 28, 512)	66048
conv3_block4_3_bn (BatchNormal-ization)	(None, 28, 28, 512)	2048
conv3_block4_add (Add)	(None, 28, 28, 512)	0
conv3_block4_out (Activation)	(None, 28, 28, 512)	0
conv4_block1_1_conv (Conv2D)	(None, 14, 14, 256)	131328
conv4_block1_1_bn (BatchNormal-ization)	(None, 14, 14, 256)	1024
conv4_block1_1_relu (Activation)	(None, 14, 14, 256)	0
conv4_block1_2_conv (Conv2D)	(None, 14, 14, 256)	590080
conv4_block1_2_bn (BatchNormal-ization)	(None, 14, 14, 256)	1024
conv4_block1_2_relu (Activation)	(None, 14, 14, 256)	0

conv4_block1_0_conv (Conv2 (None, 14, 14, 1024) ['conv3_block4_out[0][0]'] D)	525312
conv4_block1_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block1_2_relu[0][0]'] D)	263168
conv4_block1_0_bn (BatchNo (None, 14, 14, 1024) ['conv4_block1_0_conv[0][0]'] rmalization)	4096
conv4_block1_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block1_3_conv[0][0]'] rmalization)	4096
conv4_block1_add (Add) (None, 14, 14, 1024) ['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']	0
conv4_block1_out (Activati (None, 14, 14, 1024) ['conv4_block1_add[0][0]'] on)	0
conv4_block2_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block1_out[0][0]'] D)	262400
conv4_block2_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block2_1_conv[0][0]'] rmalization)	1024
conv4_block2_1_relu (Activ (None, 14, 14, 256) ['conv4_block2_1_bn[0][0]'] ation)	0
conv4_block2_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block2_1_relu[0][0]'] D)	590080
conv4_block2_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block2_2_conv[0][0]'] rmalization)	1024
conv4_block2_2_relu (Activ (None, 14, 14, 256) ['conv4_block2_2_bn[0][0]'] ation)	0

conv4_block2_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block2_2_relu[0][0]'] D)	263168
conv4_block2_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block2_3_conv[0][0]'] rmalization)	4096
conv4_block2_add (Add) (None, 14, 14, 1024) ['conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]']	0
conv4_block2_out (Activati (None, 14, 14, 1024) ['conv4_block2_add[0][0]'] on)	0
conv4_block3_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block2_out[0][0]'] D)	262400
conv4_block3_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block3_1_conv[0][0]'] rmalization)	1024
conv4_block3_1_relu (Activ (None, 14, 14, 256) ['conv4_block3_1_bn[0][0]'] ation)	0
conv4_block3_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_1_relu[0][0]'] D)	590080
conv4_block3_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block3_2_conv[0][0]'] rmalization)	1024
conv4_block3_2_relu (Activ (None, 14, 14, 256) ['conv4_block3_2_bn[0][0]'] ation)	0
conv4_block3_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block3_2_relu[0][0]'] D)	263168
conv4_block3_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block3_3_conv[0][0]'] rmalization)	4096

conv4_block3_add (Add) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block3_out (Activation) ['conv4_block3_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block4_1_conv (Conv2D) ['conv4_block3_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block4_1_bn (Batch Normalization) ['conv4_block4_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block4_1_relu (Activation) ['conv4_block4_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block4_2_conv (Conv2D) ['conv4_block4_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block4_2_bn (Batch Normalization) ['conv4_block4_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block4_2_relu (Activation) ['conv4_block4_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block4_3_conv (Conv2D) ['conv4_block4_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block4_3_bn (Batch Normalization) ['conv4_block4_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block4_add (Add) ['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block4_out (Activation) ['conv4_block4_add[0][0]']	(None, 14, 14, 1024)	0

conv4_block5_1_conv (Conv2D) ['conv4_block4_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block5_1_bn (Batch Normalization) ['conv4_block5_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block5_1_relu (Activation) ['conv4_block5_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block5_2_conv (Conv2D) ['conv4_block5_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block5_2_bn (Batch Normalization) ['conv4_block5_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block5_2_relu (Activation) ['conv4_block5_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block5_3_conv (Conv2D) ['conv4_block5_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block5_3_bn (Batch Normalization) ['conv4_block5_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block5_add (Add) ['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block5_out (Activation) ['conv4_block5_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block6_1_conv (Conv2D) ['conv4_block5_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block6_1_bn (Batch Normalization) ['conv4_block6_1_conv[0][0]']	(None, 14, 14, 256)	1024

conv4_block6_1_relu (Activation) (None, 14, 14, 256)	0
conv4_block6_2_conv (Conv2D) (None, 14, 14, 256)	590080
conv4_block6_2_bn (Batch Normalization) (None, 14, 14, 256)	1024
conv4_block6_2_relu (Activation) (None, 14, 14, 256)	0
conv4_block6_3_conv (Conv2D) (None, 14, 14, 1024)	263168
conv4_block6_3_bn (Batch Normalization) (None, 14, 14, 1024)	4096
conv4_block6_add (Add) (None, 14, 14, 1024)	0
conv4_block6_out (Activation) (None, 14, 14, 1024)	0
conv5_block1_1_conv (Conv2D) (None, 7, 7, 512)	524800
conv5_block1_1_bn (Batch Normalization) (None, 7, 7, 512)	2048
conv5_block1_1_relu (Activation) (None, 7, 7, 512)	0
conv5_block1_2_conv (Conv2D) (None, 7, 7, 512)	2359808

conv5_block1_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block1_2_conv[0][0]'] rmalization)	2048
conv5_block1_2_relu (Activ (None, 7, 7, 512) ['conv5_block1_2_bn[0][0]'] ation)	0
conv5_block1_0_conv (Conv2 (None, 7, 7, 2048) ['conv4_block6_out[0][0]'] D)	2099200
conv5_block1_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block1_2_relu[0][0]'] D)	1050624
conv5_block1_0_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_0_conv[0][0]'] rmalization)	8192
conv5_block1_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_3_conv[0][0]'] rmalization)	8192
conv5_block1_add (Add) (None, 7, 7, 2048) ['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']	0
conv5_block1_out (Activati (None, 7, 7, 2048) ['conv5_block1_add[0][0]'] on)	0
conv5_block2_1_conv (Conv2 (None, 7, 7, 512) ['conv5_block1_out[0][0]'] D)	1049088
conv5_block2_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_1_conv[0][0]'] rmalization)	2048
conv5_block2_1_relu (Activ (None, 7, 7, 512) ['conv5_block2_1_bn[0][0]'] ation)	0
conv5_block2_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block2_1_relu[0][0]'] D)	2359808



conv5_block2_2_bn (BatchNormal ['conv5_block2_2_conv[0][0]' rmalization)	(None, 7, 7, 512)	2048
conv5_block2_2_relu (Activati ['conv5_block2_2_bn[0][0]' ation)	(None, 7, 7, 512)	0
conv5_block2_3_conv (Conv2D) ['conv5_block2_2_relu[0][0]'	(None, 7, 7, 2048)	1050624
conv5_block2_3_bn (BatchNormal ['conv5_block2_3_conv[0][0]' rmalization)	(None, 7, 7, 2048)	8192
conv5_block2_add (Add) ['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]'	(None, 7, 7, 2048)	0
conv5_block2_out (Activation) ['conv5_block2_add[0][0]'	(None, 7, 7, 2048)	0
conv5_block3_1_conv (Conv2D) ['conv5_block2_out[0][0]'	(None, 7, 7, 512)	1049088
conv5_block3_1_bn (BatchNormal ['conv5_block3_1_conv[0][0]' rmalization)	(None, 7, 7, 512)	2048
conv5_block3_1_relu (Activati ['conv5_block3_1_bn[0][0]' ation)	(None, 7, 7, 512)	0
conv5_block3_2_conv (Conv2D) ['conv5_block3_1_relu[0][0]'	(None, 7, 7, 512)	2359808
conv5_block3_2_bn (BatchNormal ['conv5_block3_2_conv[0][0]' rmalization)	(None, 7, 7, 512)	2048
conv5_block3_2_relu (Activati ['conv5_block3_2_bn[0][0]' ation)	(None, 7, 7, 512)	0

conv5_block3_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block3_2_relu[0][0]'] D)	1050624
conv5_block3_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block3_3_conv[0][0]'] rmalization)	8192
conv5_block3_add (Add) (None, 7, 7, 2048) ['conv5_block2_out[0][0]', 'conv5_block3_3_bn[0][0]']	0
conv5_block3_out (Activati (None, 7, 7, 2048) ['conv5_block3_add[0][0]'] on)	0
average_pooling2d (Average (None, 3, 3, 2048) ['conv5_block3_out[0][0]'] Pooling2D)	0
flatten (Flatten) (None, 18432) ['average_pooling2d[0][0]']	0
dense (Dense) (None, 9) ['flatten[0][0]']	165897

```
=====
=====
Total params: 23753609 (90.61 MB)
Trainable params: 23700489 (90.41 MB)
Non-trainable params: 53120 (207.50 KB)
-----
-----
```

```
[ ]: import cv2

def overlay_heatmap(heatmap, img_path, alpha=0.4):
    img = cv2.imread(img_path)

    hmap = heatmap

    # Resize the heatmap to the size of the original image
    heatmap = cv2.resize(heatmap, (img.shape[1], img.shape[0]))

    # Normalize the heatmap
    heatmap = (heatmap - np.min(heatmap)) / (np.max(heatmap) - np.min(heatmap))
```

```

# Apply a colormap to the heatmap
heatmap = cv2.applyColorMap(np.uint8(255 * heatmap), cv2.COLORMAP_JET)

# Superimpose the heatmap on the original image
superimposed_img = heatmap * alpha + img

fig, axes = plt.subplots(1, 3, figsize=(12, 8))

# Display the heatmap in the second subplot
axes[0].imshow(hmap)
axes[0].set_title('Heatmap')
axes[0].axis('off')

# Display the heatmap in the second subplot
axes[1].imshow(heatmap, cmap='jet')
axes[1].set_title('Heatmap Normalized')
axes[1].axis('off')

# Display the superimposed image in the first subplot
axes[2].imshow(cv2.cvtColor(np.uint8(superimposed_img), cv2.COLOR_BGR2RGB))
axes[2].set_title('Superimposed Image')
axes[2].axis('off')

plt.show()

# plt.imshow(cv2.cvtColor(np.uint8(superimposed_img), cv2.COLOR_BGR2RGB))
# plt.axis('off')
# plt.show()

```

## First Layer Heatmap

```

[ ]: # This is the "dog" entry in the prediction vector
dog_output = model.output[:, 4]

# This is the output feature map of the `conv3_block2_2_relu` layer,
# the first convolutional layer in ResNet50
first_conv_layer = model.get_layer('conv1_relu')

# This is the gradient of the "dog" class with regard to
# the output feature map of `conv5_block3_3_conv`
grads = K.gradients(dog_output, first_conv_layer.output)[0]

```

```

# This is a vector of shape (512,), where each entry
# is the mean intensity of the gradient over a specific feature map channel
pooled_grads = K.mean(grads, axis=(0, 1, 2))

# This function allows us to access the values of the quantities we just
↳ defined:
# `pooled_grads` and the output feature map of `block5_conv3`,
# given a sample image
iterate = K.function([model.input], [pooled_grads, first_conv_layer.output[0]])

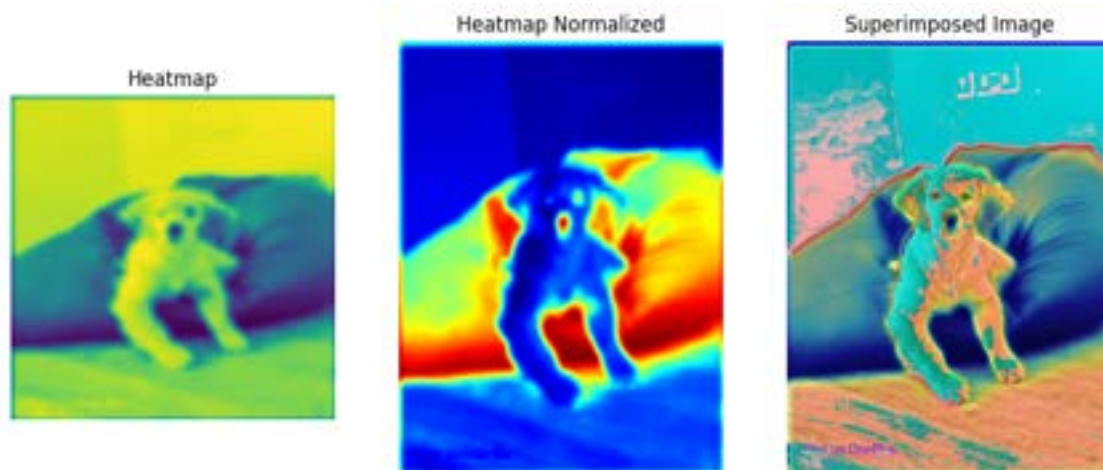
# These are the values of these two quantities, as Numpy arrays,
# given our sample image of two elephants
pooled_grads_value, conv_layer_output_value = iterate([img_tensor])

# We multiply each channel in the feature map array
# by "how important this channel is" with regard to the elephant class
for i in range(64):
    conv_layer_output_value[:, :, i] *= pooled_grads_value[i]

# The channel-wise mean of the resulting feature map
# is our heatmap of class activation
heatmap_first = np.mean(conv_layer_output_value, axis=-1)

overlay_heatmap(heatmap_first, img_path)

```



### Middle Layer Heatmap

```

[ ]: # This is the "dog" entry in the prediction vector
dog_output = model.output[:, 4]

```

```

# This is the output feature map of the `conv3_block2_2_relu` layer,
# the middle convolutional layer in ResNet50
middle_conv_layer = model.get_layer('conv3_block2_2_relu')

# This is the gradient of the "dog" class with regard to
# the output feature map of `conv5_block3_3_conv`
grads = K.gradients(dog_output, middle_conv_layer.output)[0]

# This is a vector of shape (512,), where each entry
# is the mean intensity of the gradient over a specific feature map channel
pooled_grads = K.mean(grads, axis=(0, 1, 2))

# This function allows us to access the values of the quantities we just
↳ defined:
# `pooled_grads` and the output feature map of `block5_conv3`,
# given a sample image
iterate = K.function([model.input], [pooled_grads, middle_conv_layer.output[0]])

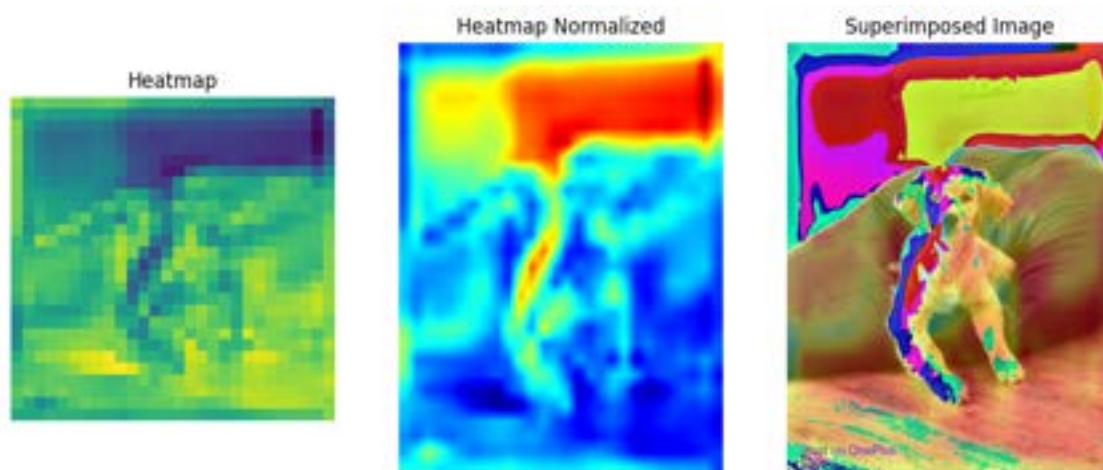
# These are the values of these two quantities, as Numpy arrays,
# given our sample image of two elephants
pooled_grads_value, conv_layer_output_value = iterate([img_tensor])

# We multiply each channel in the feature map array
# by "how important this channel is" with regard to the elephant class
for i in range(128):
    conv_layer_output_value[:, :, i] *= pooled_grads_value[i]

# The channel-wise mean of the resulting feature map
# is our heatmap of class activation
heatmap_middle = np.mean(conv_layer_output_value, axis=-1)

overlay_heatmap(heatmap_middle, img_path)

```



## Last layer heatmap

```
[ ]: # This is the "dog" entry in the prediction vector
dog_output = model.output[:, 4]

# This is the output feature map of the `conv5_block3_3_conv` layer,
# the last convolutional layer in ResNet50
last_conv_layer = model.get_layer('conv5_block3_3_conv')

# This is the gradient of the "dog" class with regard to
# the output feature map of `conv5_block3_3_conv`
grads = K.gradients(dog_output, last_conv_layer.output)[0]

# This is a vector of shape (512,), where each entry
# is the mean intensity of the gradient over a specific feature map channel
pooled_grads = K.mean(grads, axis=(0, 1, 2))

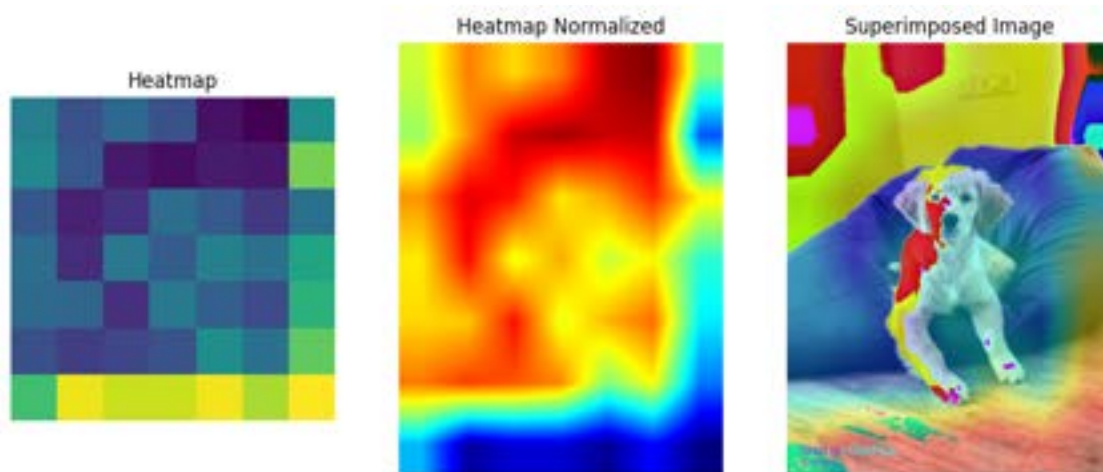
# This function allows us to access the values of the quantities we just
↳ defined:
# `pooled_grads` and the output feature map of `block5_conv3`,
# given a sample image
iterate = K.function([model.input], [pooled_grads, last_conv_layer.output[0]])

# These are the values of these two quantities, as Numpy arrays,
# given our sample image of two elephants
pooled_grads_value, conv_layer_output_value = iterate([img_tensor])

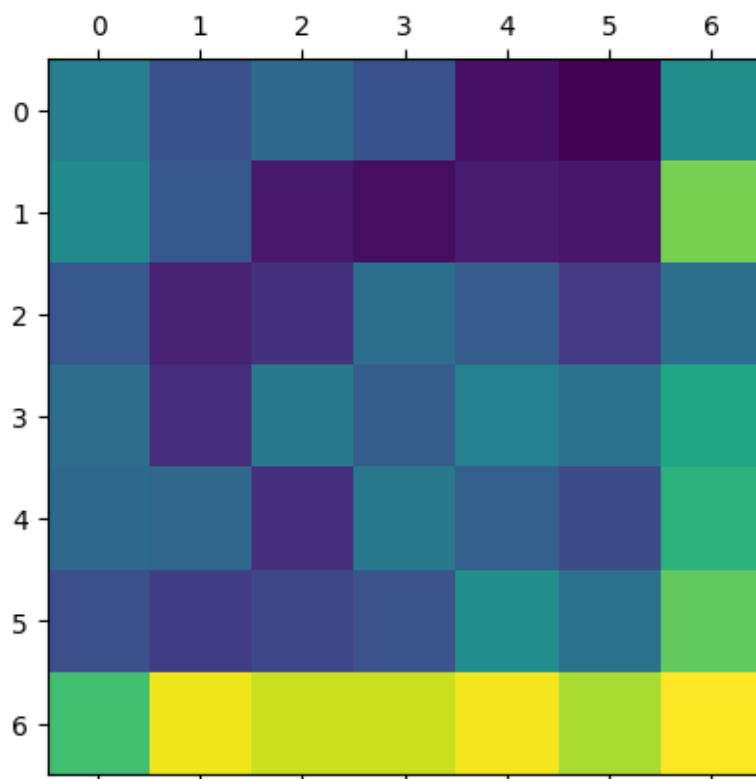
# We multiply each channel in the feature map array
# by "how important this channel is" with regard to the elephant class
for i in range(512):
    conv_layer_output_value[:, :, i] *= pooled_grads_value[i]

# The channel-wise mean of the resulting feature map
# is our heatmap of class activation
heatmap_last = np.mean(conv_layer_output_value, axis=-1)

overlay_heatmap(heatmap_last, img_path)
```



```
[ ]: ## print(heatmap)
      # heatmap = np.maximum(heatmap, 0)
      # heatmap /= np.max(heatmap)
      # plt.matshow(heatmap)
      # plt.show()
```



```
[ ]: overlay_heatmap(heatmap, img_path)
```



```
[ ]:
```

```
[ ]: # # print(heatmap)
# heatmap = np.maximum(heatmap, 0)
# heatmap /= np.max(heatmap)
# plt.matshow(heatmap)
# plt.show()
```

```
[ ]: # overlay_heatmap(heatmap, img_path)
```

## 1.2 PS 1.B - 20 points

In [this notebook](#) we showcase *transfer learning* using a pre-trained CNN model.

Perform the fine-tuning and feature extraction methods of transfer learning using the same model as in PS-1A, for the class **ship**.

Repeat the visualization of PS-1.A before and after transfer learning and write a conclusive summary as to the relative value of the two methods.

```
[ ]: # Insert your code here
```



```
#Load base_model from PS 1.A
# base_model = load_model('model_my_custom_resnet50_aug1.h5')
# base_model = tf.keras.models.load_model('model_my_custom_resnet50_1.h5')
base_model = load_model('model_resnet50_aug4.h5')
base_model.summary()
```

WARNING:absl:At this time, the v2.11+ optimizer `tf.keras.optimizers.SGD` runs slowly on M1/M2 Macs, please use the legacy Keras optimizer instead, located at `tf.keras.optimizers.legacy.SGD`.

Model: "model"

```
-----
Layer (type)                 Output Shape              Param #   Connected to
=====
input_1 (InputLayer)         [(None, 224, 224, 3)]    0         []
conv1_pad (ZeroPadding2D)    (None, 230, 230, 3)      0         ['input_1[0][0]']
conv1_conv (Conv2D)          (None, 112, 112, 64)     9472      ['conv1_pad[0][0]']
conv1_bn (BatchNormalizati  (None, 112, 112, 64)     256       ['conv1_conv[0][0]']
on)
conv1_relu (Activation)      (None, 112, 112, 64)     0         ['conv1_bn[0][0]']
pool1_pad (ZeroPadding2D)    (None, 114, 114, 64)     0         ['conv1_relu[0][0]']
pool1_pool (MaxPooling2D)    (None, 56, 56, 64)       0         ['pool1_pad[0][0]']
conv2_block1_1_conv (Conv2  (None, 56, 56, 64)       4160      ['pool1_pool[0][0]']
D)
conv2_block1_1_bn (BatchNo  (None, 56, 56, 64)       256       ['conv2_block1_1_conv[0][0]']
rmalization)
conv2_block1_1_relu (Activ  (None, 56, 56, 64)       0         ['conv2_block1_1_bn[0][0]']
ation)
```

conv2_block1_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block1_1_relu[0][0]'] D)	36928
conv2_block1_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block1_2_conv[0][0]'] rmalization)	256
conv2_block1_2_relu (Activ (None, 56, 56, 64) ['conv2_block1_2_bn[0][0]'] ation)	0
conv2_block1_0_conv (Conv2 (None, 56, 56, 256) ['pool1_pool[0][0]'] D)	16640
conv2_block1_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block1_2_relu[0][0]'] D)	16640
conv2_block1_0_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_0_conv[0][0]'] rmalization)	1024
conv2_block1_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_3_conv[0][0]'] rmalization)	1024
conv2_block1_add (Add) (None, 56, 56, 256) ['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']	0
conv2_block1_out (Activati (None, 56, 56, 256) ['conv2_block1_add[0][0]'] on)	0
conv2_block2_1_conv (Conv2 (None, 56, 56, 64) ['conv2_block1_out[0][0]'] D)	16448
conv2_block2_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block2_1_conv[0][0]'] rmalization)	256
conv2_block2_1_relu (Activ (None, 56, 56, 64) ['conv2_block2_1_bn[0][0]'] ation)	0

conv2_block2_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block2_1_relu[0][0]'] D)	36928
conv2_block2_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block2_2_conv[0][0]'] rmalization)	256
conv2_block2_2_relu (Activ (None, 56, 56, 64) ['conv2_block2_2_bn[0][0]'] ation)	0
conv2_block2_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block2_2_relu[0][0]'] D)	16640
conv2_block2_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block2_3_conv[0][0]'] rmalization)	1024
conv2_block2_add (Add) (None, 56, 56, 256) ['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']	0
conv2_block2_out (Activati (None, 56, 56, 256) ['conv2_block2_add[0][0]'] on)	0
conv2_block3_1_conv (Conv2 (None, 56, 56, 64) ['conv2_block2_out[0][0]'] D)	16448
conv2_block3_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block3_1_conv[0][0]'] rmalization)	256
conv2_block3_1_relu (Activ (None, 56, 56, 64) ['conv2_block3_1_bn[0][0]'] ation)	0
conv2_block3_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block3_1_relu[0][0]'] D)	36928
conv2_block3_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block3_2_conv[0][0]'] rmalization)	256

conv2_block3_2_relu (Activation) ['conv2_block3_2_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block3_3_conv (Conv2D) ['conv2_block3_2_relu[0][0]']	(None, 56, 56, 256)	16640
conv2_block3_3_bn (Batch Normalization) ['conv2_block3_3_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block3_add (Add) ['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block3_out (Activation) ['conv2_block3_add[0][0]']	(None, 56, 56, 256)	0
conv3_block1_1_conv (Conv2D) ['conv2_block3_out[0][0]']	(None, 28, 28, 128)	32896
conv3_block1_1_bn (Batch Normalization) ['conv3_block1_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block1_1_relu (Activation) ['conv3_block1_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block1_2_conv (Conv2D) ['conv3_block1_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block1_2_bn (Batch Normalization) ['conv3_block1_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block1_2_relu (Activation) ['conv3_block1_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block1_0_conv (Conv2D) ['conv2_block3_out[0][0]']	(None, 28, 28, 512)	131584

conv3_block1_3_conv (Conv2D) (None, 28, 28, 512)	66048
['conv3_block1_2_relu[0][0]']	
conv3_block1_0_bn (Batch Normalization) (None, 28, 28, 512)	2048
['conv3_block1_0_conv[0][0]']	
conv3_block1_3_bn (Batch Normalization) (None, 28, 28, 512)	2048
['conv3_block1_3_conv[0][0]']	
conv3_block1_add (Add) (None, 28, 28, 512)	0
['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']	
conv3_block1_out (Activation) (None, 28, 28, 512)	0
['conv3_block1_add[0][0]']	
conv3_block2_1_conv (Conv2D) (None, 28, 28, 128)	65664
['conv3_block1_out[0][0]']	
conv3_block2_1_bn (Batch Normalization) (None, 28, 28, 128)	512
['conv3_block2_1_conv[0][0]']	
conv3_block2_1_relu (Activation) (None, 28, 28, 128)	0
['conv3_block2_1_bn[0][0]']	
conv3_block2_2_conv (Conv2D) (None, 28, 28, 128)	147584
['conv3_block2_1_relu[0][0]']	
conv3_block2_2_bn (Batch Normalization) (None, 28, 28, 128)	512
['conv3_block2_2_conv[0][0]']	
conv3_block2_2_relu (Activation) (None, 28, 28, 128)	0
['conv3_block2_2_bn[0][0]']	
conv3_block2_3_conv (Conv2D) (None, 28, 28, 512)	66048
['conv3_block2_2_relu[0][0]']	

conv3_block2_3_bn (BatchNormal- ization) ['conv3_block2_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block2_add (Add) ['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block2_out (Activation) ['conv3_block2_add[0][0]']	(None, 28, 28, 512)	0
conv3_block3_1_conv (Conv2D) ['conv3_block2_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block3_1_bn (BatchNormal- ization) ['conv3_block3_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block3_1_relu (Activation) ['conv3_block3_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block3_2_conv (Conv2D) ['conv3_block3_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block3_2_bn (BatchNormal- ization) ['conv3_block3_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block3_2_relu (Activation) ['conv3_block3_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block3_3_conv (Conv2D) ['conv3_block3_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block3_3_bn (BatchNormal- ization) ['conv3_block3_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block3_add (Add) ['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']	(None, 28, 28, 512)	0

conv3_block3_out (Activation) ['conv3_block3_add[0][0]']	(None, 28, 28, 512)	0
conv3_block4_1_conv (Conv2D) ['conv3_block3_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block4_1_bn (Batch Normalization) ['conv3_block4_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_1_relu (Activation) ['conv3_block4_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_2_conv (Conv2D) ['conv3_block4_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block4_2_bn (Batch Normalization) ['conv3_block4_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_2_relu (Activation) ['conv3_block4_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_3_conv (Conv2D) ['conv3_block4_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block4_3_bn (Batch Normalization) ['conv3_block4_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block4_add (Add) ['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block4_out (Activation) ['conv3_block4_add[0][0]']	(None, 28, 28, 512)	0
conv4_block1_1_conv (Conv2D) ['conv3_block4_out[0][0]']	(None, 14, 14, 256)	131328

conv4_block1_1_bn (BatchNormal- ization) ['conv4_block1_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block1_1_relu (Activation) ['conv4_block1_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block1_2_conv (Conv2D) ['conv4_block1_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block1_2_bn (BatchNormal- ization) ['conv4_block1_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block1_2_relu (Activation) ['conv4_block1_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block1_0_conv (Conv2D) ['conv3_block4_out[0][0]']	(None, 14, 14, 1024)	525312
conv4_block1_3_conv (Conv2D) ['conv4_block1_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block1_0_bn (BatchNormal- ization) ['conv4_block1_0_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block1_3_bn (BatchNormal- ization) ['conv4_block1_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block1_add (Add) ['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block1_out (Activation) ['conv4_block1_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block2_1_conv (Conv2D) ['conv4_block1_out[0][0]']	(None, 14, 14, 256)	262400



conv4_block2_1_bn (BatchNormal ['conv4_block2_1_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block2_1_relu (Activati ['conv4_block2_1_bn[0][0]' ation)	(None, 14, 14, 256)	0
conv4_block2_2_conv (Conv2D) ['conv4_block2_1_relu[0][0]'	(None, 14, 14, 256)	590080
conv4_block2_2_bn (BatchNormal ['conv4_block2_2_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block2_2_relu (Activati ['conv4_block2_2_bn[0][0]' ation)	(None, 14, 14, 256)	0
conv4_block2_3_conv (Conv2D) ['conv4_block2_2_relu[0][0]'	(None, 14, 14, 1024)	263168
conv4_block2_3_bn (BatchNormal ['conv4_block2_3_conv[0][0]' rmalization)	(None, 14, 14, 1024)	4096
conv4_block2_add (Add) ['conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block2_out (Activation) ['conv4_block2_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block3_1_conv (Conv2D) ['conv4_block2_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block3_1_bn (BatchNormal ['conv4_block3_1_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block3_1_relu (Activati ['conv4_block3_1_bn[0][0]' ation)	(None, 14, 14, 256)	0

conv4_block3_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_1_relu[0][0]'] D)	590080
conv4_block3_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block3_2_conv[0][0]'] rmalization)	1024
conv4_block3_2_relu (Activ (None, 14, 14, 256) ['conv4_block3_2_bn[0][0]'] ation)	0
conv4_block3_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block3_2_relu[0][0]'] D)	263168
conv4_block3_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block3_3_conv[0][0]'] rmalization)	4096
conv4_block3_add (Add) (None, 14, 14, 1024) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	0
conv4_block3_out (Activati (None, 14, 14, 1024) ['conv4_block3_add[0][0]'] on)	0
conv4_block4_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_out[0][0]'] D)	262400
conv4_block4_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block4_1_conv[0][0]'] rmalization)	1024
conv4_block4_1_relu (Activ (None, 14, 14, 256) ['conv4_block4_1_bn[0][0]'] ation)	0
conv4_block4_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block4_1_relu[0][0]'] D)	590080
conv4_block4_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block4_2_conv[0][0]'] rmalization)	1024

conv4_block4_2_relu (Activation) (None, 14, 14, 256) ['conv4_block4_2_bn[0][0]'])	0
conv4_block4_3_conv (Conv2D) (None, 14, 14, 1024) ['conv4_block4_2_relu[0][0]'])	263168
conv4_block4_3_bn (Batch Normalization) (None, 14, 14, 1024) ['conv4_block4_3_conv[0][0]'])	4096
conv4_block4_add (Add) (None, 14, 14, 1024) ['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']	0
conv4_block4_out (Activation) (None, 14, 14, 1024) ['conv4_block4_add[0][0]'])	0
conv4_block5_1_conv (Conv2D) (None, 14, 14, 256) ['conv4_block4_out[0][0]']	262400
conv4_block5_1_bn (Batch Normalization) (None, 14, 14, 256) ['conv4_block5_1_conv[0][0]']	1024
conv4_block5_1_relu (Activation) (None, 14, 14, 256) ['conv4_block5_1_bn[0][0]']	0
conv4_block5_2_conv (Conv2D) (None, 14, 14, 256) ['conv4_block5_1_relu[0][0]']	590080
conv4_block5_2_bn (Batch Normalization) (None, 14, 14, 256) ['conv4_block5_2_conv[0][0]']	1024
conv4_block5_2_relu (Activation) (None, 14, 14, 256) ['conv4_block5_2_bn[0][0]']	0
conv4_block5_3_conv (Conv2D) (None, 14, 14, 1024) ['conv4_block5_2_relu[0][0]']	263168

conv4_block5_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block5_3_conv[0][0]'] rmalization)	4096
conv4_block5_add (Add) (None, 14, 14, 1024) ['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']	0
conv4_block5_out (Activati (None, 14, 14, 1024) ['conv4_block5_add[0][0]'] on)	0
conv4_block6_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block5_out[0][0]'] D)	262400
conv4_block6_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block6_1_conv[0][0]'] rmalization)	1024
conv4_block6_1_relu (Activ (None, 14, 14, 256) ['conv4_block6_1_bn[0][0]'] ation)	0
conv4_block6_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block6_1_relu[0][0]'] D)	590080
conv4_block6_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block6_2_conv[0][0]'] rmalization)	1024
conv4_block6_2_relu (Activ (None, 14, 14, 256) ['conv4_block6_2_bn[0][0]'] ation)	0
conv4_block6_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block6_2_relu[0][0]'] D)	263168
conv4_block6_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block6_3_conv[0][0]'] rmalization)	4096
conv4_block6_add (Add) (None, 14, 14, 1024) ['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']	0

conv4_block6_out (Activation)	(None, 14, 14, 1024)	0
['conv4_block6_add[0][0]']		
conv5_block1_1_conv (Conv2D)	(None, 7, 7, 512)	524800
['conv4_block6_out[0][0]']		
conv5_block1_1_bn (Batch Normalization)	(None, 7, 7, 512)	2048
['conv5_block1_1_conv[0][0]']		
conv5_block1_1_relu (Activation)	(None, 7, 7, 512)	0
['conv5_block1_1_bn[0][0]']		
conv5_block1_2_conv (Conv2D)	(None, 7, 7, 512)	2359808
['conv5_block1_1_relu[0][0]']		
conv5_block1_2_bn (Batch Normalization)	(None, 7, 7, 512)	2048
['conv5_block1_2_conv[0][0]']		
conv5_block1_2_relu (Activation)	(None, 7, 7, 512)	0
['conv5_block1_2_bn[0][0]']		
conv5_block1_0_conv (Conv2D)	(None, 7, 7, 2048)	2099200
['conv4_block6_out[0][0]']		
conv5_block1_3_conv (Conv2D)	(None, 7, 7, 2048)	1050624
['conv5_block1_2_relu[0][0]']		
conv5_block1_0_bn (Batch Normalization)	(None, 7, 7, 2048)	8192
['conv5_block1_0_conv[0][0]']		
conv5_block1_3_bn (Batch Normalization)	(None, 7, 7, 2048)	8192
['conv5_block1_3_conv[0][0]']		
conv5_block1_add (Add)	(None, 7, 7, 2048)	0
['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']		

conv5_block1_out (Activation) ['conv5_block1_add[0][0]']	(None, 7, 7, 2048)	0
conv5_block2_1_conv (Conv2D) ['conv5_block1_out[0][0]']	(None, 7, 7, 512)	1049088
conv5_block2_1_bn (Batch Normalization) ['conv5_block2_1_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block2_1_relu (Activation) ['conv5_block2_1_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block2_2_conv (Conv2D) ['conv5_block2_1_relu[0][0]']	(None, 7, 7, 512)	2359808
conv5_block2_2_bn (Batch Normalization) ['conv5_block2_2_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block2_2_relu (Activation) ['conv5_block2_2_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block2_3_conv (Conv2D) ['conv5_block2_2_relu[0][0]']	(None, 7, 7, 2048)	1050624
conv5_block2_3_bn (Batch Normalization) ['conv5_block2_3_conv[0][0]']	(None, 7, 7, 2048)	8192
conv5_block2_add (Add) ['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']	(None, 7, 7, 2048)	0
conv5_block2_out (Activation) ['conv5_block2_add[0][0]']	(None, 7, 7, 2048)	0
conv5_block3_1_conv (Conv2D) ['conv5_block2_out[0][0]']	(None, 7, 7, 512)	1049088

conv5_block3_1_bn (BatchNormal-ization)	(None, 7, 7, 512)	2048
conv5_block3_1_conv	(None, 7, 7, 512)	0
conv5_block3_1_relu	(None, 7, 7, 512)	0
conv5_block3_2_conv	(None, 7, 7, 512)	2359808
conv5_block3_2_bn	(None, 7, 7, 512)	2048
conv5_block3_2_relu	(None, 7, 7, 512)	0
conv5_block3_3_conv	(None, 7, 7, 2048)	1050624
conv5_block3_3_bn	(None, 7, 7, 2048)	8192
conv5_block3_add	(None, 7, 7, 2048)	0
conv5_block3_out	(None, 7, 7, 2048)	0
average_pooling2d	(None, 3, 3, 2048)	0
flatten	(None, 18432)	0
dense	(None, 9)	165897

=====

```
=====
Total params: 23753609 (90.61 MB)
Trainable params: 23700489 (90.41 MB)
Non-trainable params: 53120 (207.50 KB)
```

Here model 2 will get defined using `model2b = Model(input = base-model.input, output = base-model.outputs)` but update `base-model.outputs`'s last layer for sigmoid

### 1.2.1 Freeze the base layer

```
[ ]: for layer in base_model.layers:
      layer.trainable = False

base_model.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	[]
conv1_pad (ZeroPadding2D)	(None, 230, 230, 3)	0	['input_1[0][0]']
conv1_conv (Conv2D)	(None, 112, 112, 64)	9472	['conv1_pad[0][0]']
conv1_bn (BatchNormalizati on)	(None, 112, 112, 64)	256	['conv1_conv[0][0]']
conv1_relu (Activation)	(None, 112, 112, 64)	0	['conv1_bn[0][0]']
pool1_pad (ZeroPadding2D)	(None, 114, 114, 64)	0	['conv1_relu[0][0]']
pool1_pool (MaxPooling2D)	(None, 56, 56, 64)	0	['pool1_pad[0][0]']
conv2_block1_1_conv (Conv2 D)	(None, 56, 56, 64)	4160	['pool1_pool[0][0]']



conv2_block1_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block1_1_conv[0][0]'] rmalization)	256
conv2_block1_1_relu (Activ (None, 56, 56, 64) ['conv2_block1_1_bn[0][0]'] ation)	0
conv2_block1_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block1_1_relu[0][0]'] D)	36928
conv2_block1_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block1_2_conv[0][0]'] rmalization)	256
conv2_block1_2_relu (Activ (None, 56, 56, 64) ['conv2_block1_2_bn[0][0]'] ation)	0
conv2_block1_0_conv (Conv2 (None, 56, 56, 256) ['pool1_pool[0][0]'] D)	16640
conv2_block1_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block1_2_relu[0][0]'] D)	16640
conv2_block1_0_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_0_conv[0][0]'] rmalization)	1024
conv2_block1_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_3_conv[0][0]'] rmalization)	1024
conv2_block1_add (Add) (None, 56, 56, 256) ['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']	0
conv2_block1_out (Activati (None, 56, 56, 256) ['conv2_block1_add[0][0]'] on)	0
conv2_block2_1_conv (Conv2 (None, 56, 56, 64) ['conv2_block1_out[0][0]'] D)	16448

conv2_block2_1_bn (BatchNormal- ization) ['conv2_block2_1_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block2_1_relu (Activation) ['conv2_block2_1_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block2_2_conv (Conv2D) ['conv2_block2_1_relu[0][0]']	(None, 56, 56, 64)	36928
conv2_block2_2_bn (BatchNormal- ization) ['conv2_block2_2_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block2_2_relu (Activation) ['conv2_block2_2_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block2_3_conv (Conv2D) ['conv2_block2_2_relu[0][0]']	(None, 56, 56, 256)	16640
conv2_block2_3_bn (BatchNormal- ization) ['conv2_block2_3_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block2_add (Add) ['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block2_out (Activation) ['conv2_block2_add[0][0]']	(None, 56, 56, 256)	0
conv2_block3_1_conv (Conv2D) ['conv2_block2_out[0][0]']	(None, 56, 56, 64)	16448
conv2_block3_1_bn (BatchNormal- ization) ['conv2_block3_1_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block3_1_relu (Activation) ['conv2_block3_1_bn[0][0]']	(None, 56, 56, 64)	0

conv2_block3_2_conv (Conv2D) [ 'conv2_block3_1_relu[0][0]' ]	(None, 56, 56, 64)	36928
conv2_block3_2_bn (Batch Normalization) [ 'conv2_block3_2_conv[0][0]' ]	(None, 56, 56, 64)	256
conv2_block3_2_relu (Activation) [ 'conv2_block3_2_bn[0][0]' ]	(None, 56, 56, 64)	0
conv2_block3_3_conv (Conv2D) [ 'conv2_block3_2_relu[0][0]' ]	(None, 56, 56, 256)	16640
conv2_block3_3_bn (Batch Normalization) [ 'conv2_block3_3_conv[0][0]' ]	(None, 56, 56, 256)	1024
conv2_block3_add (Add) [ 'conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]' ]	(None, 56, 56, 256)	0
conv2_block3_out (Activation) [ 'conv2_block3_add[0][0]' ]	(None, 56, 56, 256)	0
conv3_block1_1_conv (Conv2D) [ 'conv2_block3_out[0][0]' ]	(None, 28, 28, 128)	32896
conv3_block1_1_bn (Batch Normalization) [ 'conv3_block1_1_conv[0][0]' ]	(None, 28, 28, 128)	512
conv3_block1_1_relu (Activation) [ 'conv3_block1_1_bn[0][0]' ]	(None, 28, 28, 128)	0
conv3_block1_2_conv (Conv2D) [ 'conv3_block1_1_relu[0][0]' ]	(None, 28, 28, 128)	147584
conv3_block1_2_bn (Batch Normalization) [ 'conv3_block1_2_conv[0][0]' ]	(None, 28, 28, 128)	512

conv3_block1_2_relu (Activation) ['conv3_block1_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block1_0_conv (Conv2D) ['conv2_block3_out[0][0]']	(None, 28, 28, 512)	131584
conv3_block1_3_conv (Conv2D) ['conv3_block1_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block1_0_bn (Batch Normalization) ['conv3_block1_0_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block1_3_bn (Batch Normalization) ['conv3_block1_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block1_add (Add) ['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block1_out (Activation) ['conv3_block1_add[0][0]']	(None, 28, 28, 512)	0
conv3_block2_1_conv (Conv2D) ['conv3_block1_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block2_1_bn (Batch Normalization) ['conv3_block2_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block2_1_relu (Activation) ['conv3_block2_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block2_2_conv (Conv2D) ['conv3_block2_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block2_2_bn (Batch Normalization) ['conv3_block2_2_conv[0][0]']	(None, 28, 28, 128)	512

conv3_block2_2_relu (Activation) ['conv3_block2_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block2_3_conv (Conv2D) ['conv3_block2_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block2_3_bn (Batch Normalization) ['conv3_block2_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block2_add (Add) ['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block2_out (Activation) ['conv3_block2_add[0][0]']	(None, 28, 28, 512)	0
conv3_block3_1_conv (Conv2D) ['conv3_block2_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block3_1_bn (Batch Normalization) ['conv3_block3_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block3_1_relu (Activation) ['conv3_block3_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block3_2_conv (Conv2D) ['conv3_block3_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block3_2_bn (Batch Normalization) ['conv3_block3_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block3_2_relu (Activation) ['conv3_block3_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block3_3_conv (Conv2D) ['conv3_block3_2_relu[0][0]']	(None, 28, 28, 512)	66048

conv3_block3_3_bn (BatchNormal- ization) ['conv3_block3_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block3_add (Add) ['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block3_out (Activation) ['conv3_block3_add[0][0]']	(None, 28, 28, 512)	0
conv3_block4_1_conv (Conv2D) ['conv3_block3_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block4_1_bn (BatchNormal- ization) ['conv3_block4_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_1_relu (Activation) ['conv3_block4_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_2_conv (Conv2D) ['conv3_block4_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block4_2_bn (BatchNormal- ization) ['conv3_block4_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_2_relu (Activation) ['conv3_block4_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_3_conv (Conv2D) ['conv3_block4_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block4_3_bn (BatchNormal- ization) ['conv3_block4_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block4_add (Add) ['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']	(None, 28, 28, 512)	0

conv3_block4_out (Activation)	(None, 28, 28, 512)	0
['conv3_block4_add[0][0]']		
conv4_block1_1_conv (Conv2D)	(None, 14, 14, 256)	131328
['conv3_block4_out[0][0]']		
conv4_block1_1_bn (Batch Normalization)	(None, 14, 14, 256)	1024
['conv4_block1_1_conv[0][0]']		
conv4_block1_1_relu (Activation)	(None, 14, 14, 256)	0
['conv4_block1_1_bn[0][0]']		
conv4_block1_2_conv (Conv2D)	(None, 14, 14, 256)	590080
['conv4_block1_1_relu[0][0]']		
conv4_block1_2_bn (Batch Normalization)	(None, 14, 14, 256)	1024
['conv4_block1_2_conv[0][0]']		
conv4_block1_2_relu (Activation)	(None, 14, 14, 256)	0
['conv4_block1_2_bn[0][0]']		
conv4_block1_0_conv (Conv2D)	(None, 14, 14, 1024)	525312
['conv3_block4_out[0][0]']		
conv4_block1_3_conv (Conv2D)	(None, 14, 14, 1024)	263168
['conv4_block1_2_relu[0][0]']		
conv4_block1_0_bn (Batch Normalization)	(None, 14, 14, 1024)	4096
['conv4_block1_0_conv[0][0]']		
conv4_block1_3_bn (Batch Normalization)	(None, 14, 14, 1024)	4096
['conv4_block1_3_conv[0][0]']		
conv4_block1_add (Add)	(None, 14, 14, 1024)	0
['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']		

conv4_block1_out (Activation) ['conv4_block1_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block2_1_conv (Conv2D) ['conv4_block1_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block2_1_bn (Batch Normalization) ['conv4_block2_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block2_1_relu (Activation) ['conv4_block2_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block2_2_conv (Conv2D) ['conv4_block2_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block2_2_bn (Batch Normalization) ['conv4_block2_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block2_2_relu (Activation) ['conv4_block2_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block2_3_conv (Conv2D) ['conv4_block2_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block2_3_bn (Batch Normalization) ['conv4_block2_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block2_add (Add) ['conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block2_out (Activation) ['conv4_block2_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block3_1_conv (Conv2D) ['conv4_block2_out[0][0]']	(None, 14, 14, 256)	262400



conv4_block3_1_bn (BatchNormal- ization) ['conv4_block3_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block3_1_relu (Activation) ['conv4_block3_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block3_2_conv (Conv2D) ['conv4_block3_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block3_2_bn (BatchNormal- ization) ['conv4_block3_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block3_2_relu (Activation) ['conv4_block3_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block3_3_conv (Conv2D) ['conv4_block3_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block3_3_bn (BatchNormal- ization) ['conv4_block3_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block3_add (Add) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block3_out (Activation) ['conv4_block3_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block4_1_conv (Conv2D) ['conv4_block3_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block4_1_bn (BatchNormal- ization) ['conv4_block4_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block4_1_relu (Activation) ['conv4_block4_1_bn[0][0]']	(None, 14, 14, 256)	0

conv4_block4_2_conv (Conv2D) [ 'conv4_block4_1_relu[0][0]' ]	(None, 14, 14, 256)	590080
conv4_block4_2_bn (Batch Normalization) [ 'conv4_block4_2_conv[0][0]' ]	(None, 14, 14, 256)	1024
conv4_block4_2_relu (Activation) [ 'conv4_block4_2_bn[0][0]' ]	(None, 14, 14, 256)	0
conv4_block4_3_conv (Conv2D) [ 'conv4_block4_2_relu[0][0]' ]	(None, 14, 14, 1024)	263168
conv4_block4_3_bn (Batch Normalization) [ 'conv4_block4_3_conv[0][0]' ]	(None, 14, 14, 1024)	4096
conv4_block4_add (Add) [ 'conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]' ]	(None, 14, 14, 1024)	0
conv4_block4_out (Activation) [ 'conv4_block4_add[0][0]' ]	(None, 14, 14, 1024)	0
conv4_block5_1_conv (Conv2D) [ 'conv4_block4_out[0][0]' ]	(None, 14, 14, 256)	262400
conv4_block5_1_bn (Batch Normalization) [ 'conv4_block5_1_conv[0][0]' ]	(None, 14, 14, 256)	1024
conv4_block5_1_relu (Activation) [ 'conv4_block5_1_bn[0][0]' ]	(None, 14, 14, 256)	0
conv4_block5_2_conv (Conv2D) [ 'conv4_block5_1_relu[0][0]' ]	(None, 14, 14, 256)	590080
conv4_block5_2_bn (Batch Normalization) [ 'conv4_block5_2_conv[0][0]' ]	(None, 14, 14, 256)	1024

conv4_block5_2_relu (Activation) ['conv4_block5_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block5_3_conv (Conv2D) ['conv4_block5_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block5_3_bn (Batch Normalization) ['conv4_block5_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block5_add (Add) ['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block5_out (Activation) ['conv4_block5_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block6_1_conv (Conv2D) ['conv4_block5_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block6_1_bn (Batch Normalization) ['conv4_block6_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block6_1_relu (Activation) ['conv4_block6_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block6_2_conv (Conv2D) ['conv4_block6_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block6_2_bn (Batch Normalization) ['conv4_block6_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block6_2_relu (Activation) ['conv4_block6_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block6_3_conv (Conv2D) ['conv4_block6_2_relu[0][0]']	(None, 14, 14, 1024)	263168

conv4_block6_3_bn (BatchNormal- ization) ['conv4_block6_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block6_add (Add) ['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block6_out (Activation) ['conv4_block6_add[0][0]']	(None, 14, 14, 1024)	0
conv5_block1_1_conv (Conv2D) ['conv4_block6_out[0][0]']	(None, 7, 7, 512)	524800
conv5_block1_1_bn (BatchNormal- ization) ['conv5_block1_1_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block1_1_relu (Activation) ['conv5_block1_1_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block1_2_conv (Conv2D) ['conv5_block1_1_relu[0][0]']	(None, 7, 7, 512)	2359808
conv5_block1_2_bn (BatchNormal- ization) ['conv5_block1_2_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block1_2_relu (Activation) ['conv5_block1_2_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block1_0_conv (Conv2D) ['conv4_block6_out[0][0]']	(None, 7, 7, 2048)	2099200
conv5_block1_3_conv (Conv2D) ['conv5_block1_2_relu[0][0]']	(None, 7, 7, 2048)	1050624
conv5_block1_0_bn (BatchNormal- ization) ['conv5_block1_0_conv[0][0]']	(None, 7, 7, 2048)	8192

conv5_block1_3_bn (BatchNormal- ization) ['conv5_block1_3_conv[0][0]']	(None, 7, 7, 2048)	8192
conv5_block1_add (Add) ['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']	(None, 7, 7, 2048)	0
conv5_block1_out (Activation) ['conv5_block1_add[0][0]']	(None, 7, 7, 2048)	0
conv5_block2_1_conv (Conv2D) ['conv5_block1_out[0][0]']	(None, 7, 7, 512)	1049088
conv5_block2_1_bn (BatchNormal- ization) ['conv5_block2_1_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block2_1_relu (Activation) ['conv5_block2_1_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block2_2_conv (Conv2D) ['conv5_block2_1_relu[0][0]']	(None, 7, 7, 512)	2359808
conv5_block2_2_bn (BatchNormal- ization) ['conv5_block2_2_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block2_2_relu (Activation) ['conv5_block2_2_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block2_3_conv (Conv2D) ['conv5_block2_2_relu[0][0]']	(None, 7, 7, 2048)	1050624
conv5_block2_3_bn (BatchNormal- ization) ['conv5_block2_3_conv[0][0]']	(None, 7, 7, 2048)	8192
conv5_block2_add (Add) ['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']	(None, 7, 7, 2048)	0

conv5_block2_out (Activation) ['conv5_block2_add[0][0]']	(None, 7, 7, 2048)	0
conv5_block3_1_conv (Conv2D) ['conv5_block2_out[0][0]']	(None, 7, 7, 512)	1049088
conv5_block3_1_bn (Batch Normalization) ['conv5_block3_1_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block3_1_relu (Activation) ['conv5_block3_1_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block3_2_conv (Conv2D) ['conv5_block3_1_relu[0][0]']	(None, 7, 7, 512)	2359808
conv5_block3_2_bn (Batch Normalization) ['conv5_block3_2_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block3_2_relu (Activation) ['conv5_block3_2_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block3_3_conv (Conv2D) ['conv5_block3_2_relu[0][0]']	(None, 7, 7, 2048)	1050624
conv5_block3_3_bn (Batch Normalization) ['conv5_block3_3_conv[0][0]']	(None, 7, 7, 2048)	8192
conv5_block3_add (Add) ['conv5_block2_out[0][0]', 'conv5_block3_3_bn[0][0]']	(None, 7, 7, 2048)	0
conv5_block3_out (Activation) ['conv5_block3_add[0][0]']	(None, 7, 7, 2048)	0
average_pooling2d (Average Pooling2D) ['conv5_block3_out[0][0]']	(None, 3, 3, 2048)	0

```

flatten (Flatten)          (None, 18432)          0
['average_pooling2d[0][0]']

dense (Dense)              (None, 9)          165897
['flatten[0][0]']

```

```

=====
Total params: 23753609 (90.61 MB)
Trainable params: 0 (0.00 Byte)
Non-trainable params: 23753609 (90.61 MB)
-----

```

### Feature Extraction

```
[ ]: # Add a new Dense layer with sigmoid activation for binary classification
```

```

num_classes = 1 # For binary classification
output_layer = Dense(num_classes, activation='sigmoid',
                      name='fc9_predictions')(base_model.layers[-2].output)

model_2b = Model(inputs=base_model.input, outputs=output_layer)

model_2b.summary()

```

Model: "model"

```

-----
Layer (type)                 Output Shape              Param #   Connected to
-----
input_1 (InputLayer)         [(None, 224, 224, 3)]    0         []
conv1_pad (ZeroPadding2D)    (None, 230, 230, 3)      0         ['input_1[0][0]']
conv1_conv (Conv2D)          (None, 112, 112, 64)     9472      ['conv1_pad[0][0]']
conv1_bn (BatchNormalizati   (None, 112, 112, 64)     256       ['conv1_conv[0][0]']
on)
conv1_relu (Activation)      (None, 112, 112, 64)     0         ['conv1_bn[0][0]']
pool1_pad (ZeroPadding2D)    (None, 114, 114, 64)     0

```

['conv1_relu[0][0]']		
pool1_pool (MaxPooling2D)	(None, 56, 56, 64)	0
['pool1_pad[0][0]']		
conv2_block1_1_conv (Conv2D)	(None, 56, 56, 64)	4160
['pool1_pool[0][0]']		
conv2_block1_1_bn (Batch Normalization)	(None, 56, 56, 64)	256
['conv2_block1_1_conv[0][0]']		
conv2_block1_1_relu (Activation)	(None, 56, 56, 64)	0
['conv2_block1_1_bn[0][0]']		
conv2_block1_2_conv (Conv2D)	(None, 56, 56, 64)	36928
['conv2_block1_1_relu[0][0]']		
conv2_block1_2_bn (Batch Normalization)	(None, 56, 56, 64)	256
['conv2_block1_2_conv[0][0]']		
conv2_block1_2_relu (Activation)	(None, 56, 56, 64)	0
['conv2_block1_2_bn[0][0]']		
conv2_block1_0_conv (Conv2D)	(None, 56, 56, 256)	16640
['pool1_pool[0][0]']		
conv2_block1_3_conv (Conv2D)	(None, 56, 56, 256)	16640
['conv2_block1_2_relu[0][0]']		
conv2_block1_0_bn (Batch Normalization)	(None, 56, 56, 256)	1024
['conv2_block1_0_conv[0][0]']		
conv2_block1_3_bn (Batch Normalization)	(None, 56, 56, 256)	1024
['conv2_block1_3_conv[0][0]']		
conv2_block1_add (Add)	(None, 56, 56, 256)	0
['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']		



conv2_block1_out (Activation) ['conv2_block1_add[0][0]']	(None, 56, 56, 256)	0
conv2_block2_1_conv (Conv2D) ['conv2_block1_out[0][0]']	(None, 56, 56, 64)	16448
conv2_block2_1_bn (Batch Normalization) ['conv2_block2_1_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block2_1_relu (Activation) ['conv2_block2_1_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block2_2_conv (Conv2D) ['conv2_block2_1_relu[0][0]']	(None, 56, 56, 64)	36928
conv2_block2_2_bn (Batch Normalization) ['conv2_block2_2_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block2_2_relu (Activation) ['conv2_block2_2_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block2_3_conv (Conv2D) ['conv2_block2_2_relu[0][0]']	(None, 56, 56, 256)	16640
conv2_block2_3_bn (Batch Normalization) ['conv2_block2_3_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block2_add (Add) ['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block2_out (Activation) ['conv2_block2_add[0][0]']	(None, 56, 56, 256)	0
conv2_block3_1_conv (Conv2D) ['conv2_block2_out[0][0]']	(None, 56, 56, 64)	16448

conv2_block3_1_bn (BatchNormal- ization) ['conv2_block3_1_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block3_1_relu (Activation) ['conv2_block3_1_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block3_2_conv (Conv2D) ['conv2_block3_1_relu[0][0]']	(None, 56, 56, 64)	36928
conv2_block3_2_bn (BatchNormal- ization) ['conv2_block3_2_conv[0][0]']	(None, 56, 56, 64)	256
conv2_block3_2_relu (Activation) ['conv2_block3_2_bn[0][0]']	(None, 56, 56, 64)	0
conv2_block3_3_conv (Conv2D) ['conv2_block3_2_relu[0][0]']	(None, 56, 56, 256)	16640
conv2_block3_3_bn (BatchNormal- ization) ['conv2_block3_3_conv[0][0]']	(None, 56, 56, 256)	1024
conv2_block3_add (Add) ['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block3_out (Activation) ['conv2_block3_add[0][0]']	(None, 56, 56, 256)	0
conv3_block1_1_conv (Conv2D) ['conv2_block3_out[0][0]']	(None, 28, 28, 128)	32896
conv3_block1_1_bn (BatchNormal- ization) ['conv3_block1_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block1_1_relu (Activation) ['conv3_block1_1_bn[0][0]']	(None, 28, 28, 128)	0

conv3_block1_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block1_1_relu[0][0]'] D)	147584
conv3_block1_2_bn (BatchNo (None, 28, 28, 128) ['conv3_block1_2_conv[0][0]'] rmalization)	512
conv3_block1_2_relu (Activ (None, 28, 28, 128) ['conv3_block1_2_bn[0][0]'] ation)	0
conv3_block1_0_conv (Conv2 (None, 28, 28, 512) ['conv2_block3_out[0][0]'] D)	131584
conv3_block1_3_conv (Conv2 (None, 28, 28, 512) ['conv3_block1_2_relu[0][0]'] D)	66048
conv3_block1_0_bn (BatchNo (None, 28, 28, 512) ['conv3_block1_0_conv[0][0]'] rmalization)	2048
conv3_block1_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block1_3_conv[0][0]'] rmalization)	2048
conv3_block1_add (Add) (None, 28, 28, 512) ['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']	0
conv3_block1_out (Activati (None, 28, 28, 512) ['conv3_block1_add[0][0]'] on)	0
conv3_block2_1_conv (Conv2 (None, 28, 28, 128) ['conv3_block1_out[0][0]'] D)	65664
conv3_block2_1_bn (BatchNo (None, 28, 28, 128) ['conv3_block2_1_conv[0][0]'] rmalization)	512
conv3_block2_1_relu (Activ (None, 28, 28, 128) ['conv3_block2_1_bn[0][0]'] ation)	0

conv3_block2_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block2_1_relu[0][0]'] D)	147584
conv3_block2_2_bn (BatchNo (None, 28, 28, 128) ['conv3_block2_2_conv[0][0]'] rmalization)	512
conv3_block2_2_relu (Activ (None, 28, 28, 128) ['conv3_block2_2_bn[0][0]'] ation)	0
conv3_block2_3_conv (Conv2 (None, 28, 28, 512) ['conv3_block2_2_relu[0][0]'] D)	66048
conv3_block2_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block2_3_conv[0][0]'] rmalization)	2048
conv3_block2_add (Add) (None, 28, 28, 512) ['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']	0
conv3_block2_out (Activati (None, 28, 28, 512) ['conv3_block2_add[0][0]'] on)	0
conv3_block3_1_conv (Conv2 (None, 28, 28, 128) ['conv3_block2_out[0][0]'] D)	65664
conv3_block3_1_bn (BatchNo (None, 28, 28, 128) ['conv3_block3_1_conv[0][0]'] rmalization)	512
conv3_block3_1_relu (Activ (None, 28, 28, 128) ['conv3_block3_1_bn[0][0]'] ation)	0
conv3_block3_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block3_1_relu[0][0]'] D)	147584
conv3_block3_2_bn (BatchNo (None, 28, 28, 128) ['conv3_block3_2_conv[0][0]'] rmalization)	512

conv3_block3_2_relu (Activation) ['conv3_block3_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block3_3_conv (Conv2D) ['conv3_block3_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block3_3_bn (Batch Normalization) ['conv3_block3_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block3_add (Add) ['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block3_out (Activation) ['conv3_block3_add[0][0]']	(None, 28, 28, 512)	0
conv3_block4_1_conv (Conv2D) ['conv3_block3_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block4_1_bn (Batch Normalization) ['conv3_block4_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_1_relu (Activation) ['conv3_block4_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_2_conv (Conv2D) ['conv3_block4_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block4_2_bn (Batch Normalization) ['conv3_block4_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_2_relu (Activation) ['conv3_block4_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_3_conv (Conv2D) ['conv3_block4_2_relu[0][0]']	(None, 28, 28, 512)	66048

conv3_block4_3_bn (BatchNormal- ization) ['conv3_block4_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block4_add (Add) ['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block4_out (Activation) ['conv3_block4_add[0][0]']	(None, 28, 28, 512)	0
conv4_block1_1_conv (Conv2D) ['conv3_block4_out[0][0]']	(None, 14, 14, 256)	131328
conv4_block1_1_bn (BatchNormal- ization) ['conv4_block1_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block1_1_relu (Activation) ['conv4_block1_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block1_2_conv (Conv2D) ['conv4_block1_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block1_2_bn (BatchNormal- ization) ['conv4_block1_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block1_2_relu (Activation) ['conv4_block1_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block1_0_conv (Conv2D) ['conv3_block4_out[0][0]']	(None, 14, 14, 1024)	525312
conv4_block1_3_conv (Conv2D) ['conv4_block1_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block1_0_bn (BatchNormal- ization) ['conv4_block1_0_conv[0][0]']	(None, 14, 14, 1024)	4096

conv4_block1_3_bn (BatchNormal ['conv4_block1_3_conv[0][0]' rmalization)	(None, 14, 14, 1024)	4096
conv4_block1_add (Add) ['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block1_out (Activation) ['conv4_block1_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block2_1_conv (Conv2D) ['conv4_block1_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block2_1_bn (BatchNormal ['conv4_block2_1_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block2_1_relu (Activation) ['conv4_block2_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block2_2_conv (Conv2D) ['conv4_block2_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block2_2_bn (BatchNormal ['conv4_block2_2_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block2_2_relu (Activation) ['conv4_block2_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block2_3_conv (Conv2D) ['conv4_block2_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block2_3_bn (BatchNormal ['conv4_block2_3_conv[0][0]' rmalization)	(None, 14, 14, 1024)	4096
conv4_block2_add (Add) ['conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]']	(None, 14, 14, 1024)	0

conv4_block2_out (Activation) ['conv4_block2_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block3_1_conv (Conv2D) ['conv4_block2_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block3_1_bn (Batch Normalization) ['conv4_block3_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block3_1_relu (Activation) ['conv4_block3_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block3_2_conv (Conv2D) ['conv4_block3_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block3_2_bn (Batch Normalization) ['conv4_block3_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block3_2_relu (Activation) ['conv4_block3_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block3_3_conv (Conv2D) ['conv4_block3_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block3_3_bn (Batch Normalization) ['conv4_block3_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block3_add (Add) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block3_out (Activation) ['conv4_block3_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block4_1_conv (Conv2D) ['conv4_block3_out[0][0]']	(None, 14, 14, 256)	262400



conv4_block4_1_bn (BatchNormal- ization) ['conv4_block4_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block4_1_relu (Activation) ['conv4_block4_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block4_2_conv (Conv2D) ['conv4_block4_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block4_2_bn (BatchNormal- ization) ['conv4_block4_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block4_2_relu (Activation) ['conv4_block4_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block4_3_conv (Conv2D) ['conv4_block4_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block4_3_bn (BatchNormal- ization) ['conv4_block4_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block4_add (Add) ['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block4_out (Activation) ['conv4_block4_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block5_1_conv (Conv2D) ['conv4_block4_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block5_1_bn (BatchNormal- ization) ['conv4_block5_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block5_1_relu (Activation) ['conv4_block5_1_bn[0][0]']	(None, 14, 14, 256)	0

conv4_block5_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block5_1_relu[0][0]'] D)	590080
conv4_block5_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block5_2_conv[0][0]'] rmalization)	1024
conv4_block5_2_relu (Activ (None, 14, 14, 256) ['conv4_block5_2_bn[0][0]'] ation)	0
conv4_block5_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block5_2_relu[0][0]'] D)	263168
conv4_block5_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block5_3_conv[0][0]'] rmalization)	4096
conv4_block5_add (Add) (None, 14, 14, 1024) ['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']	0
conv4_block5_out (Activati (None, 14, 14, 1024) ['conv4_block5_add[0][0]'] on)	0
conv4_block6_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block5_out[0][0]'] D)	262400
conv4_block6_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block6_1_conv[0][0]'] rmalization)	1024
conv4_block6_1_relu (Activ (None, 14, 14, 256) ['conv4_block6_1_bn[0][0]'] ation)	0
conv4_block6_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block6_1_relu[0][0]'] D)	590080
conv4_block6_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block6_2_conv[0][0]'] rmalization)	1024

conv4_block6_2_relu (Activation) ['conv4_block6_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block6_3_conv (Conv2D) ['conv4_block6_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block6_3_bn (Batch Normalization) ['conv4_block6_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block6_add (Add) ['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block6_out (Activation) ['conv4_block6_add[0][0]']	(None, 14, 14, 1024)	0
conv5_block1_1_conv (Conv2D) ['conv4_block6_out[0][0]']	(None, 7, 7, 512)	524800
conv5_block1_1_bn (Batch Normalization) ['conv5_block1_1_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block1_1_relu (Activation) ['conv5_block1_1_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block1_2_conv (Conv2D) ['conv5_block1_1_relu[0][0]']	(None, 7, 7, 512)	2359808
conv5_block1_2_bn (Batch Normalization) ['conv5_block1_2_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block1_2_relu (Activation) ['conv5_block1_2_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block1_0_conv (Conv2D) ['conv4_block6_out[0][0]']	(None, 7, 7, 2048)	2099200

conv5_block1_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block1_2_relu[0][0]'] D)	1050624
conv5_block1_0_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_0_conv[0][0]'] rmalization)	8192
conv5_block1_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_3_conv[0][0]'] rmalization)	8192
conv5_block1_add (Add) (None, 7, 7, 2048) ['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']	0
conv5_block1_out (Activati (None, 7, 7, 2048) ['conv5_block1_add[0][0]'] on)	0
conv5_block2_1_conv (Conv2 (None, 7, 7, 512) ['conv5_block1_out[0][0]'] D)	1049088
conv5_block2_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_1_conv[0][0]'] rmalization)	2048
conv5_block2_1_relu (Activ (None, 7, 7, 512) ['conv5_block2_1_bn[0][0]'] ation)	0
conv5_block2_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block2_1_relu[0][0]'] D)	2359808
conv5_block2_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_2_conv[0][0]'] rmalization)	2048
conv5_block2_2_relu (Activ (None, 7, 7, 512) ['conv5_block2_2_bn[0][0]'] ation)	0
conv5_block2_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block2_2_relu[0][0]'] D)	1050624

conv5_block2_3_bn (BatchNormal ization) ['conv5_block2_3_conv[0][0]']	(None, 7, 7, 2048)	8192
conv5_block2_add (Add) ['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']	(None, 7, 7, 2048)	0
conv5_block2_out (Activation) ['conv5_block2_add[0][0]']	(None, 7, 7, 2048)	0
conv5_block3_1_conv (Conv2D) ['conv5_block2_out[0][0]']	(None, 7, 7, 512)	1049088
conv5_block3_1_bn (BatchNormal ization) ['conv5_block3_1_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block3_1_relu (Activation) ['conv5_block3_1_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block3_2_conv (Conv2D) ['conv5_block3_1_relu[0][0]']	(None, 7, 7, 512)	2359808
conv5_block3_2_bn (BatchNormal ization) ['conv5_block3_2_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block3_2_relu (Activation) ['conv5_block3_2_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block3_3_conv (Conv2D) ['conv5_block3_2_relu[0][0]']	(None, 7, 7, 2048)	1050624
conv5_block3_3_bn (BatchNormal ization) ['conv5_block3_3_conv[0][0]']	(None, 7, 7, 2048)	8192
conv5_block3_add (Add) ['conv5_block2_out[0][0]', 'conv5_block3_3_bn[0][0]']	(None, 7, 7, 2048)	0

```

conv5_block3_out (Activation) (None, 7, 7, 2048) 0
['conv5_block3_add[0][0]']
on)

average_pooling2d (Average Pooling2D) (None, 3, 3, 2048) 0
['conv5_block3_out[0][0]']
Pooling2D)

flatten (Flatten) (None, 18432) 0
['average_pooling2d[0][0]']

fc9_predictions (Dense) (None, 1) 18433
['flatten[0][0]']

```

```

=====
=====
Total params: 23606145 (90.05 MB)
Trainable params: 18433 (72.00 KB)
Non-trainable params: 23587712 (89.98 MB)
-----
-----

```

## Fine Tuning

```

[ ]: # Set hyperparameters
learning_rate = 0.001
beta1 = 0.9
beta2 = 0.999
epsilon = 1e-8
num_iterations = 1000
# optimizer = Adam(learning_rate=learning_rate, beta_1=beta1, beta_2=beta2,
#                  ↪epsilon=1e-8)
optimizer = SGD(learning_rate=learning_rate, momentum=0.9, nesterov=True)

```

WARNING:absl:At this time, the v2.11+ optimizer `tf.keras.optimizers.SGD` runs slowly on M1/M2 Macs, please use the legacy Keras optimizer instead, located at `tf.keras.optimizers.legacy.SGD`.

```

[ ]: # Create the final model
transfer_model = Model(inputs=model_2b.input, outputs=model_2b.outputs)

# Compile the model with an appropriate optimizer and loss function
# transfer_model.compile(optimizer='adam', loss='categorical_crossentropy',
#                        ↪metrics=['accuracy'])

transfer_model.compile(optimizer=optimizer, loss='binary_crossentropy',
                      ↪metrics=['accuracy'])

```

```
transfer_model.summary()
```

Model: "model\_2"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	[]
conv1_pad (ZeroPadding2D) ['input_1[0][0]']	(None, 230, 230, 3)	0	
conv1_conv (Conv2D) ['conv1_pad[0][0]']	(None, 112, 112, 64)	9472	
conv1_bn (BatchNormalizati on) ['conv1_conv[0][0]']	(None, 112, 112, 64)	256	
conv1_relu (Activation) ['conv1_bn[0][0]']	(None, 112, 112, 64)	0	
pool1_pad (ZeroPadding2D) ['conv1_relu[0][0]']	(None, 114, 114, 64)	0	
pool1_pool (MaxPooling2D) ['pool1_pad[0][0]']	(None, 56, 56, 64)	0	
conv2_block1_1_conv (Conv2 D) ['pool1_pool[0][0]']	(None, 56, 56, 64)	4160	
conv2_block1_1_bn (BatchNo rmalization) ['conv2_block1_1_conv[0][0]']	(None, 56, 56, 64)	256	
conv2_block1_1_relu (Activ ation) ['conv2_block1_1_bn[0][0]']	(None, 56, 56, 64)	0	
conv2_block1_2_conv (Conv2 D) ['conv2_block1_1_relu[0][0]']	(None, 56, 56, 64)	36928	
conv2_block1_2_bn (BatchNo	(None, 56, 56, 64)	256	

['conv2_block1_2_conv[0][0]'] rmalization)		
conv2_block1_2_relu (Activ ['conv2_block1_2_bn[0][0]'] ation)	(None, 56, 56, 64)	0
conv2_block1_0_conv (Conv2 ['pool1_pool[0][0]'] D)	(None, 56, 56, 256)	16640
conv2_block1_3_conv (Conv2 ['conv2_block1_2_relu[0][0]'] D)	(None, 56, 56, 256)	16640
conv2_block1_0_bn (BatchNo ['conv2_block1_0_conv[0][0]'] rmalization)	(None, 56, 56, 256)	1024
conv2_block1_3_bn (BatchNo ['conv2_block1_3_conv[0][0]'] rmalization)	(None, 56, 56, 256)	1024
conv2_block1_add (Add) ['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block1_out (Activati ['conv2_block1_add[0][0]'] on)	(None, 56, 56, 256)	0
conv2_block2_1_conv (Conv2 ['conv2_block1_out[0][0]'] D)	(None, 56, 56, 64)	16448
conv2_block2_1_bn (BatchNo ['conv2_block2_1_conv[0][0]'] rmalization)	(None, 56, 56, 64)	256
conv2_block2_1_relu (Activ ['conv2_block2_1_bn[0][0]'] ation)	(None, 56, 56, 64)	0
conv2_block2_2_conv (Conv2 ['conv2_block2_1_relu[0][0]'] D)	(None, 56, 56, 64)	36928
conv2_block2_2_bn (BatchNo (None, 56, 56, 64)		256



['conv2_block2_2_conv[0][0]'] rmalization)		
conv2_block2_2_relu (Activ (None, 56, 56, 64) ['conv2_block2_2_bn[0][0]'] ation)		0
conv2_block2_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block2_2_relu[0][0]'] D)		16640
conv2_block2_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block2_3_conv[0][0]'] rmalization)		1024
conv2_block2_add (Add) (None, 56, 56, 256) ['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']		0
conv2_block2_out (Activati (None, 56, 56, 256) ['conv2_block2_add[0][0]'] on)		0
conv2_block3_1_conv (Conv2 (None, 56, 56, 64) ['conv2_block2_out[0][0]'] D)		16448
conv2_block3_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block3_1_conv[0][0]'] rmalization)		256
conv2_block3_1_relu (Activ (None, 56, 56, 64) ['conv2_block3_1_bn[0][0]'] ation)		0
conv2_block3_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block3_1_relu[0][0]'] D)		36928
conv2_block3_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block3_2_conv[0][0]'] rmalization)		256
conv2_block3_2_relu (Activ (None, 56, 56, 64) ['conv2_block3_2_bn[0][0]'] ation)		0
conv2_block3_3_conv (Conv2 (None, 56, 56, 256)		16640

['conv2_block3_2_relu[0][0]'] D)		
conv2_block3_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block3_3_conv[0][0]'] rmalization)		1024
conv2_block3_add (Add) (None, 56, 56, 256) ['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']		0
conv2_block3_out (Activati (None, 56, 56, 256) ['conv2_block3_add[0][0]'] on)		0
conv3_block1_1_conv (Conv2 (None, 28, 28, 128) ['conv2_block3_out[0][0]'] D)		32896
conv3_block1_1_bn (BatchNo (None, 28, 28, 128) ['conv3_block1_1_conv[0][0]'] rmalization)		512
conv3_block1_1_relu (Activ (None, 28, 28, 128) ['conv3_block1_1_bn[0][0]'] ation)		0
conv3_block1_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block1_1_relu[0][0]'] D)		147584
conv3_block1_2_bn (BatchNo (None, 28, 28, 128) ['conv3_block1_2_conv[0][0]'] rmalization)		512
conv3_block1_2_relu (Activ (None, 28, 28, 128) ['conv3_block1_2_bn[0][0]'] ation)		0
conv3_block1_0_conv (Conv2 (None, 28, 28, 512) ['conv2_block3_out[0][0]'] D)		131584
conv3_block1_3_conv (Conv2 (None, 28, 28, 512) ['conv3_block1_2_relu[0][0]'] D)		66048
conv3_block1_0_bn (BatchNo (None, 28, 28, 512)		2048

['conv3_block1_0_conv[0][0]'] rmalization)		
conv3_block1_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block1_3_conv[0][0]'] rmalization)		2048
conv3_block1_add (Add) (None, 28, 28, 512) ['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']		0
conv3_block1_out (Activati (None, 28, 28, 512) ['conv3_block1_add[0][0]'] on)		0
conv3_block2_1_conv (Conv2 (None, 28, 28, 128) ['conv3_block1_out[0][0]'] D)		65664
conv3_block2_1_bn (BatchNo (None, 28, 28, 128) ['conv3_block2_1_conv[0][0]'] rmalization)		512
conv3_block2_1_relu (Activ (None, 28, 28, 128) ['conv3_block2_1_bn[0][0]'] ation)		0
conv3_block2_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block2_1_relu[0][0]'] D)		147584
conv3_block2_2_bn (BatchNo (None, 28, 28, 128) ['conv3_block2_2_conv[0][0]'] rmalization)		512
conv3_block2_2_relu (Activ (None, 28, 28, 128) ['conv3_block2_2_bn[0][0]'] ation)		0
conv3_block2_3_conv (Conv2 (None, 28, 28, 512) ['conv3_block2_2_relu[0][0]'] D)		66048
conv3_block2_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block2_3_conv[0][0]'] rmalization)		2048
conv3_block2_add (Add) (None, 28, 28, 512)		0

```

['conv3_block1_out[0][0]',
'conv3_block2_3_bn[0][0]']

conv3_block2_out (Activation) (None, 28, 28, 512) 0
['conv3_block2_add[0][0]']

conv3_block3_1_conv (Conv2D) (None, 28, 28, 128) 65664
['conv3_block2_out[0][0]']

conv3_block3_1_bn (BatchNormalization) (None, 28, 28, 128) 512
['conv3_block3_1_conv[0][0]']

conv3_block3_1_relu (Activation) (None, 28, 28, 128) 0
['conv3_block3_1_bn[0][0]']

conv3_block3_2_conv (Conv2D) (None, 28, 28, 128) 147584
['conv3_block3_1_relu[0][0]']

conv3_block3_2_bn (BatchNormalization) (None, 28, 28, 128) 512
['conv3_block3_2_conv[0][0]']

conv3_block3_2_relu (Activation) (None, 28, 28, 128) 0
['conv3_block3_2_bn[0][0]']

conv3_block3_3_conv (Conv2D) (None, 28, 28, 512) 66048
['conv3_block3_2_relu[0][0]']

conv3_block3_3_bn (BatchNormalization) (None, 28, 28, 512) 2048
['conv3_block3_3_conv[0][0]']

conv3_block3_add (Add) (None, 28, 28, 512) 0
['conv3_block2_out[0][0]',
'conv3_block3_3_bn[0][0]']

conv3_block3_out (Activation) (None, 28, 28, 512) 0
['conv3_block3_add[0][0]']

conv3_block4_1_conv (Conv2D) (None, 28, 28, 128) 65664

```

```

['conv3_block3_out[0][0]']
D)

conv3_block4_1_bn (BatchNo (None, 28, 28, 128) 512
['conv3_block4_1_conv[0][0]']
rmalization)

conv3_block4_1_relu (Activ (None, 28, 28, 128) 0
['conv3_block4_1_bn[0][0]']
ation)

conv3_block4_2_conv (Conv2 (None, 28, 28, 128) 147584
['conv3_block4_1_relu[0][0]']
D)

conv3_block4_2_bn (BatchNo (None, 28, 28, 128) 512
['conv3_block4_2_conv[0][0]']
rmalization)

conv3_block4_2_relu (Activ (None, 28, 28, 128) 0
['conv3_block4_2_bn[0][0]']
ation)

conv3_block4_3_conv (Conv2 (None, 28, 28, 512) 66048
['conv3_block4_2_relu[0][0]']
D)

conv3_block4_3_bn (BatchNo (None, 28, 28, 512) 2048
['conv3_block4_3_conv[0][0]']
rmalization)

conv3_block4_add (Add) (None, 28, 28, 512) 0
['conv3_block3_out[0][0]',
'conv3_block4_3_bn[0][0]']

conv3_block4_out (Activati (None, 28, 28, 512) 0
['conv3_block4_add[0][0]']
on)

conv4_block1_1_conv (Conv2 (None, 14, 14, 256) 131328
['conv3_block4_out[0][0]']
D)

conv4_block1_1_bn (BatchNo (None, 14, 14, 256) 1024
['conv4_block1_1_conv[0][0]']
rmalization)

conv4_block1_1_relu (Activ (None, 14, 14, 256) 0

```

['conv4_block1_1_bn[0][0]'] ation)	
conv4_block1_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block1_1_relu[0][0]'] D)	590080
conv4_block1_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block1_2_conv[0][0]'] rmalization)	1024
conv4_block1_2_relu (Activ (None, 14, 14, 256) ['conv4_block1_2_bn[0][0]'] ation)	0
conv4_block1_0_conv (Conv2 (None, 14, 14, 1024) ['conv3_block4_out[0][0]'] D)	525312
conv4_block1_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block1_2_relu[0][0]'] D)	263168
conv4_block1_0_bn (BatchNo (None, 14, 14, 1024) ['conv4_block1_0_conv[0][0]'] rmalization)	4096
conv4_block1_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block1_3_conv[0][0]'] rmalization)	4096
conv4_block1_add (Add) (None, 14, 14, 1024) ['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']	0
conv4_block1_out (Activati (None, 14, 14, 1024) ['conv4_block1_add[0][0]'] on)	0
conv4_block2_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block1_out[0][0]'] D)	262400
conv4_block2_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block2_1_conv[0][0]'] rmalization)	1024
conv4_block2_1_relu (Activ (None, 14, 14, 256)	0

['conv4_block2_1_bn[0][0]'] ation)	
conv4_block2_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block2_1_relu[0][0]'] D)	590080
conv4_block2_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block2_2_conv[0][0]'] rmalization)	1024
conv4_block2_2_relu (Activ (None, 14, 14, 256) ['conv4_block2_2_bn[0][0]'] ation)	0
conv4_block2_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block2_2_relu[0][0]'] D)	263168
conv4_block2_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block2_3_conv[0][0]'] rmalization)	4096
conv4_block2_add (Add) (None, 14, 14, 1024) ['conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]']	0
conv4_block2_out (Activati (None, 14, 14, 1024) ['conv4_block2_add[0][0]'] on)	0
conv4_block3_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block2_out[0][0]'] D)	262400
conv4_block3_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block3_1_conv[0][0]'] rmalization)	1024
conv4_block3_1_relu (Activ (None, 14, 14, 256) ['conv4_block3_1_bn[0][0]'] ation)	0
conv4_block3_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_1_relu[0][0]'] D)	590080
conv4_block3_2_bn (BatchNo (None, 14, 14, 256)	1024

['conv4_block3_2_conv[0][0]'] rmalization)		
conv4_block3_2_relu (Activ (None, 14, 14, 256) ['conv4_block3_2_bn[0][0]'] ation)		0
conv4_block3_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block3_2_relu[0][0]'] D)		263168
conv4_block3_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block3_3_conv[0][0]'] rmalization)		4096
conv4_block3_add (Add) (None, 14, 14, 1024) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']		0
conv4_block3_out (Activati (None, 14, 14, 1024) ['conv4_block3_add[0][0]'] on)		0
conv4_block4_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_out[0][0]'] D)		262400
conv4_block4_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block4_1_conv[0][0]'] rmalization)		1024
conv4_block4_1_relu (Activ (None, 14, 14, 256) ['conv4_block4_1_bn[0][0]'] ation)		0
conv4_block4_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block4_1_relu[0][0]'] D)		590080
conv4_block4_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block4_2_conv[0][0]'] rmalization)		1024
conv4_block4_2_relu (Activ (None, 14, 14, 256) ['conv4_block4_2_bn[0][0]'] ation)		0
conv4_block4_3_conv (Conv2 (None, 14, 14, 1024)		263168



['conv4_block4_2_relu[0][0]'] D)		
conv4_block4_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block4_3_conv[0][0]'] rmalization)		4096
conv4_block4_add (Add) (None, 14, 14, 1024) ['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']		0
conv4_block4_out (Activati (None, 14, 14, 1024) ['conv4_block4_add[0][0]'] on)		0
conv4_block5_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block4_out[0][0]'] D)		262400
conv4_block5_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block5_1_conv[0][0]'] rmalization)		1024
conv4_block5_1_relu (Activ (None, 14, 14, 256) ['conv4_block5_1_bn[0][0]'] ation)		0
conv4_block5_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block5_1_relu[0][0]'] D)		590080
conv4_block5_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block5_2_conv[0][0]'] rmalization)		1024
conv4_block5_2_relu (Activ (None, 14, 14, 256) ['conv4_block5_2_bn[0][0]'] ation)		0
conv4_block5_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block5_2_relu[0][0]'] D)		263168
conv4_block5_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block5_3_conv[0][0]'] rmalization)		4096
conv4_block5_add (Add) (None, 14, 14, 1024)		0

```

['conv4_block4_out[0][0]',
'conv4_block5_3_bn[0][0]']

conv4_block5_out (Activation) (None, 14, 14, 1024) 0
['conv4_block5_add[0][0]']

conv4_block6_1_conv (Conv2D) (None, 14, 14, 256) 262400
['conv4_block5_out[0][0]']

conv4_block6_1_bn (Batch Normalization) (None, 14, 14, 256) 1024
['conv4_block6_1_conv[0][0]']

conv4_block6_1_relu (Activation) (None, 14, 14, 256) 0
['conv4_block6_1_bn[0][0]']

conv4_block6_2_conv (Conv2D) (None, 14, 14, 256) 590080
['conv4_block6_1_relu[0][0]']

conv4_block6_2_bn (Batch Normalization) (None, 14, 14, 256) 1024
['conv4_block6_2_conv[0][0]']

conv4_block6_2_relu (Activation) (None, 14, 14, 256) 0
['conv4_block6_2_bn[0][0]']

conv4_block6_3_conv (Conv2D) (None, 14, 14, 1024) 263168
['conv4_block6_2_relu[0][0]']

conv4_block6_3_bn (Batch Normalization) (None, 14, 14, 1024) 4096
['conv4_block6_3_conv[0][0]']

conv4_block6_add (Add) (None, 14, 14, 1024) 0
['conv4_block5_out[0][0]',
'conv4_block6_3_bn[0][0]']

conv4_block6_out (Activation) (None, 14, 14, 1024) 0
['conv4_block6_add[0][0]']

conv5_block1_1_conv (Conv2D) (None, 7, 7, 512) 524800

```

['conv4_block6_out[0][0]'] D)	
conv5_block1_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block1_1_conv[0][0]'] rmalization)	2048
conv5_block1_1_relu (Activ (None, 7, 7, 512) ['conv5_block1_1_bn[0][0]'] ation)	0
conv5_block1_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block1_1_relu[0][0]'] D)	2359808
conv5_block1_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block1_2_conv[0][0]'] rmalization)	2048
conv5_block1_2_relu (Activ (None, 7, 7, 512) ['conv5_block1_2_bn[0][0]'] ation)	0
conv5_block1_0_conv (Conv2 (None, 7, 7, 2048) ['conv4_block6_out[0][0]'] D)	2099200
conv5_block1_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block1_2_relu[0][0]'] D)	1050624
conv5_block1_0_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_0_conv[0][0]'] rmalization)	8192
conv5_block1_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_3_conv[0][0]'] rmalization)	8192
conv5_block1_add (Add) (None, 7, 7, 2048) ['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']	0
conv5_block1_out (Activati (None, 7, 7, 2048) ['conv5_block1_add[0][0]'] on)	0
conv5_block2_1_conv (Conv2 (None, 7, 7, 512)	1049088

['conv5_block1_out[0][0]'] D)	
conv5_block2_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_1_conv[0][0]'] rmalization)	2048
conv5_block2_1_relu (Activ (None, 7, 7, 512) ['conv5_block2_1_bn[0][0]'] ation)	0
conv5_block2_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block2_1_relu[0][0]'] D)	2359808
conv5_block2_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_2_conv[0][0]'] rmalization)	2048
conv5_block2_2_relu (Activ (None, 7, 7, 512) ['conv5_block2_2_bn[0][0]'] ation)	0
conv5_block2_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block2_2_relu[0][0]'] D)	1050624
conv5_block2_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block2_3_conv[0][0]'] rmalization)	8192
conv5_block2_add (Add) (None, 7, 7, 2048) ['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']	0
conv5_block2_out (Activati (None, 7, 7, 2048) ['conv5_block2_add[0][0]'] on)	0
conv5_block3_1_conv (Conv2 (None, 7, 7, 512) ['conv5_block2_out[0][0]'] D)	1049088
conv5_block3_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block3_1_conv[0][0]'] rmalization)	2048
conv5_block3_1_relu (Activ (None, 7, 7, 512)	0

```

['conv5_block3_1_bn[0][0]']
ation)

conv5_block3_2_conv (Conv2 (None, 7, 7, 512)          2359808
['conv5_block3_1_relu[0][0]']
D)

conv5_block3_2_bn (BatchNo (None, 7, 7, 512)          2048
['conv5_block3_2_conv[0][0]']
rmalization)

conv5_block3_2_relu (Activ (None, 7, 7, 512)          0
['conv5_block3_2_bn[0][0]']
ation)

conv5_block3_3_conv (Conv2 (None, 7, 7, 2048)         1050624
['conv5_block3_2_relu[0][0]']
D)

conv5_block3_3_bn (BatchNo (None, 7, 7, 2048)         8192
['conv5_block3_3_conv[0][0]']
rmalization)

conv5_block3_add (Add) (None, 7, 7, 2048)             0
['conv5_block2_out[0][0]',
'conv5_block3_3_bn[0][0]']

conv5_block3_out (Activati (None, 7, 7, 2048)         0
['conv5_block3_add[0][0]']
on)

average_pooling2d (Average (None, 3, 3, 2048)         0
['conv5_block3_out[0][0]']
Pooling2D)

flatten (Flatten) (None, 18432)                      0
['average_pooling2d[0][0]']

fc9_predictions (Dense) (None, 1)                    18433
['flatten[0][0]']

```

```

=====
=====
Total params: 23606145 (90.05 MB)
Trainable params: 18433 (72.00 KB)
Non-trainable params: 23587712 (89.98 MB)
-----
-----

```

### 1.2.2 Data pre-processing for 2B

```
[ ]: (x_train_new, y_train_new), (x_test_new, y_test_new) = cifar10.load_data()

classes = len(np.unique(y_train_new))
classes

[ ]: 10

[ ]: print(x_train_new.shape, y_train_new.shape, x_test_new.shape, y_test_new.shape)

(50000, 32, 32, 3) (50000, 1) (10000, 32, 32, 3) (10000, 1)

[ ]: index = np.where(y_train_new == 8)
# Filter the dataset for 9 out of 10 classes (exclude "ship index 8" class)
ship_class = 8 # Adjust this class to exclude "ship"
x_train_new = x_train_new / 255.0
# y_train_new = y_train_new[y_train_new == ship_class].astype(int)
# Create binary labels (0 for negative, 1 for positive)
y_train_new_norm = np.where(y_train_new == ship_class, 1, 0)
x_test_new = x_test_new / 255.0
y_test_new_norm = np.where(y_test_new == ship_class, 1, 0)
# y_test_new = y_test_new[y_test_new == ship_class].astype(int)

# number of classes K=9
k = 1

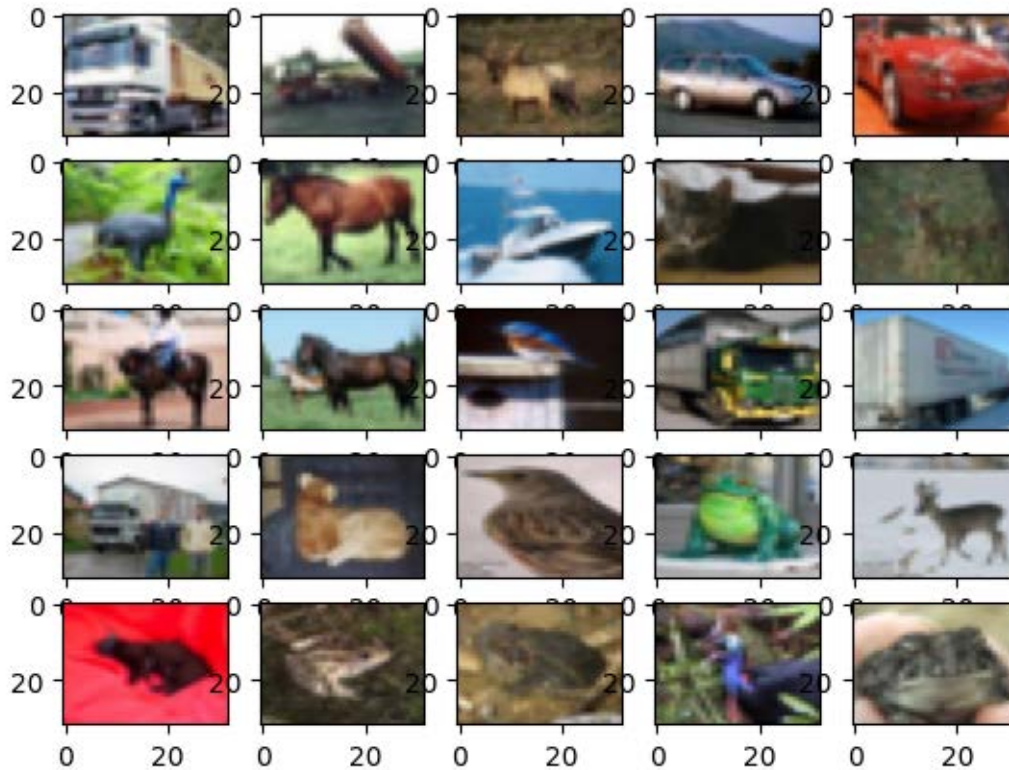
[ ]: print(x_train_new.shape, y_train_new.shape, x_test_new.shape, y_test_new.shape)

(50000, 32, 32, 3) (50000, 1) (10000, 32, 32, 3) (10000, 1)

[ ]: # visualize data by plotting images
fig, ax = plt.subplots(5, 5)
k = 1

for i in range(5):
    for j in range(5):
        ax[i][j].imshow(x_train_new[k], aspect='auto')
        k += 1

plt.show()
```



```
[ ]: # A ResNet-50 model expects 224 x 224-pixel images
im_size = 224
```

```
x_train_new_resize = Resizing(height=im_size, width=im_size,
    ↪crop_to_aspect_ratio=True)(x_train_new)
x_test_new_resize = Resizing(height=im_size, width=im_size,
    ↪crop_to_aspect_ratio=True)(x_test_new)
```

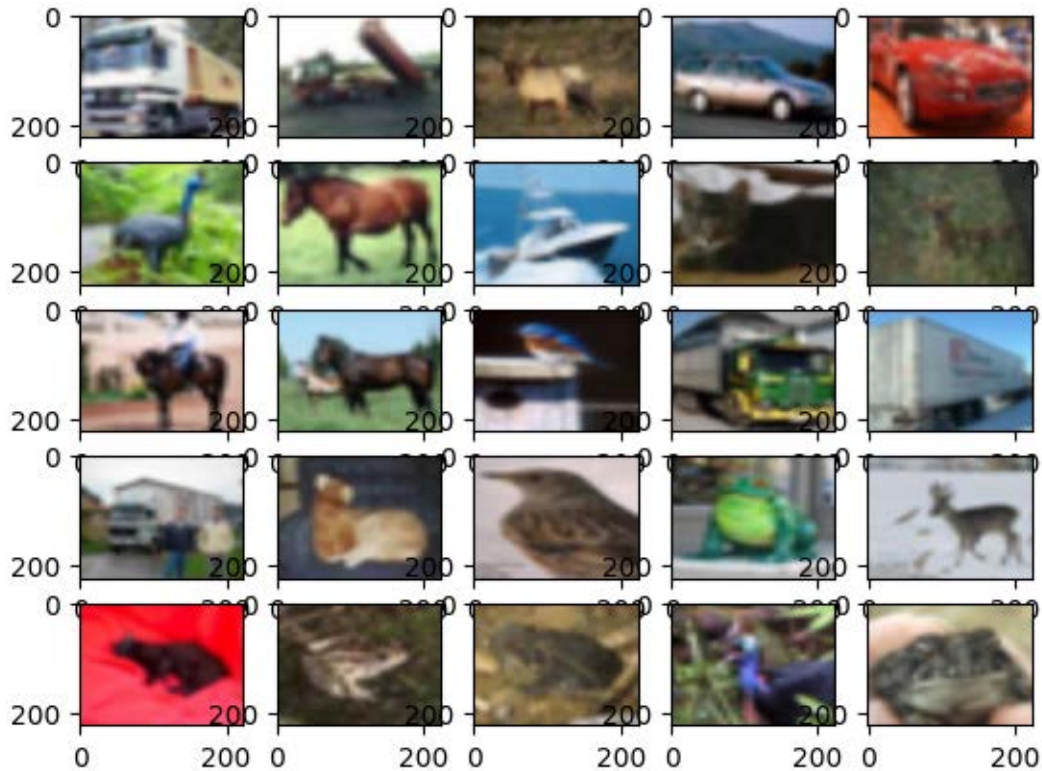
```
[ ]: print(x_train_new_resize.shape, y_train_new_norm.shape, x_test_new_resize.  
      ↪shape, y_test_new_norm.shape)
```

(50000, 224, 224, 3) (50000, 1) (10000, 224, 224, 3) (10000, 1)

```
[ ]: # visualize data by plotting images
fig, ax = plt.subplots(5, 5)
k = 1

for i in range(5):
    for j in range(5):
        ax[i][j].imshow(x_train_new_resize[k], aspect='auto')
        k += 1
```

```
plt.show()
```



```
[ ]: ## Data augmentation
batch_size = 32
data_generator = tf.keras.preprocessing.image.ImageDataGenerator(
    width_shift_range=0.1,
    height_shift_range=0.1,
    horizontal_flip=True,)

train_generator_new = data_generator.flow(x_train_new_resize,
                                          y_train_new_norm,
                                          batch_size)
steps_per_epoch = x_train_new_resize.shape[0] // batch_size
```

```
[ ]: val_loss_callback = ValidationLoss()
lr_scheduler = CustomLRScheduler()

transfer_model.fit_generator(
    train_generator_new, # Your training labels
    epochs=10,
    steps_per_epoch=10,
```



```

validation_data=(x_test_new_resize, y_test_new_norm),
callbacks=[val_loss_callback, lr_scheduler]
# Other parameters like callbacks, verbose, etc.
)

# transfer_model.fit(
#     x=x_train_new_resize, # Your training data
#     y=y_train_new_norm, # Your training labels
#     batch_size=64,
#     epochs=10,
#     steps_per_epoch=10,
#     validation_data=(x_test_new_resize, y_test_new_norm),
#     callbacks=[val_loss_callback, lr_scheduler]
#     # Other parameters like callbacks, verbose, etc.
# )

transfer_model.save('transfer_resnet50_2b_aug1.h5')

```

Epoch 1: Learning Rate = 0.001

Epoch 1/10

/var/folders/xy/1x8f1vlx569\_ch145p3563j80000gn/T/ipykernel\_17264/2315718120.py:1

6: UserWarning: `Model.fit\_generator` is deprecated and will be removed in a future version. Please use `Model.fit`, which supports generators.

transfer\_model.fit\_generator(

10/10 [=====] - 51s 5s/step - loss: 0.0952 - accuracy: 0.9831 - val\_loss: 5.3289 - val\_accuracy: 0.0000e+00

Epoch 2: Learning Rate = 0.0007071067811865475

Epoch 2/10

10/10 [=====] - 48s 5s/step - loss: 3.3305e-05 - accuracy: 1.0000 - val\_loss: 7.3087 - val\_accuracy: 0.0000e+00

Epoch 3: Learning Rate = 0.0005773502691896258

Epoch 3/10

10/10 [=====] - 50s 5s/step - loss: 2.7973e-06 - accuracy: 1.0000 - val\_loss: 8.2863 - val\_accuracy: 0.0000e+00

Epoch 4: Learning Rate = 0.0005

Epoch 4/10

10/10 [=====] - 50s 5s/step - loss: 2.1394e-06 - accuracy: 1.0000 - val\_loss: 8.8889 - val\_accuracy: 0.0000e+00

Epoch 5: Learning Rate = 0.0004472135954999579

Epoch 5/10

10/10 [=====] - 48s 5s/step - loss: 2.3662e-06 - accuracy: 1.0000 - val\_loss: 9.2786 - val\_accuracy: 0.0000e+00

Epoch 6: Learning Rate = 0.0004082482904638631

Epoch 6/10

10/10 [=====] - 47s 5s/step - loss: 1.7452e-06 -

```

accuracy: 1.0000 - val_loss: 9.6265 - val_accuracy: 0.0000e+00
Epoch 7: Learning Rate = 0.0003779644730092272
Epoch 7/10
10/10 [=====] - 48s 5s/step - loss: 3.1252e-06 -
accuracy: 1.0000 - val_loss: 9.9620 - val_accuracy: 0.0000e+00
Epoch 8: Learning Rate = 0.00035355339059327376
Epoch 8/10
10/10 [=====] - 51s 5s/step - loss: 2.1242e-06 -
accuracy: 1.0000 - val_loss: 10.3004 - val_accuracy: 0.0000e+00
Epoch 9: Learning Rate = 0.00033333333333333333
Epoch 9/10
10/10 [=====] - 49s 5s/step - loss: 3.2146e-06 -
accuracy: 1.0000 - val_loss: 10.6168 - val_accuracy: 0.0000e+00
Epoch 10: Learning Rate = 0.00031622776601683794
Epoch 10/10
10/10 [=====] - 53s 5s/step - loss: 1.6738e-06 -
accuracy: 1.0000 - val_loss: 11.0166 - val_accuracy: 0.0000e+00

/Users/banani/Library/Python/3.9/lib/python/site-
packages/keras/src/engine/training.py:3079: UserWarning: You are saving your
model as an HDF5 file via `model.save()`. This file format is considered legacy.
We recommend using instead the native Keras format, e.g.
`model.save('my_model.keras')`.
  saving_api.save_model(

```

### 1.2.3 Pre process the test image

```

[ ]: layers_to_visualize = [transfer_model.get_layer('conv1_relu'),
                           transfer_model.get_layer('conv3_block2_2_relu'),
                           transfer_model.get_layer('conv5_block3_2_relu')]

for layer in layers_to_visualize:
    print(layer.name)

```

```

conv1_relu
conv3_block2_2_relu
conv5_block3_2_relu

```

```

[ ]: # img_path = 'dogscats/subset/test/cats/cat.1700.jpg'
img_path='refs/dog_1.jpg'
# img_path='refs/dog.215.jpg'

im_size = 224

img = tf.keras.utils.load_img(img_path, target_size=(im_size, im_size))
img_tensor = tf.keras.utils.img_to_array(img)
img_tensor = np.expand_dims(img_tensor, axis=0)

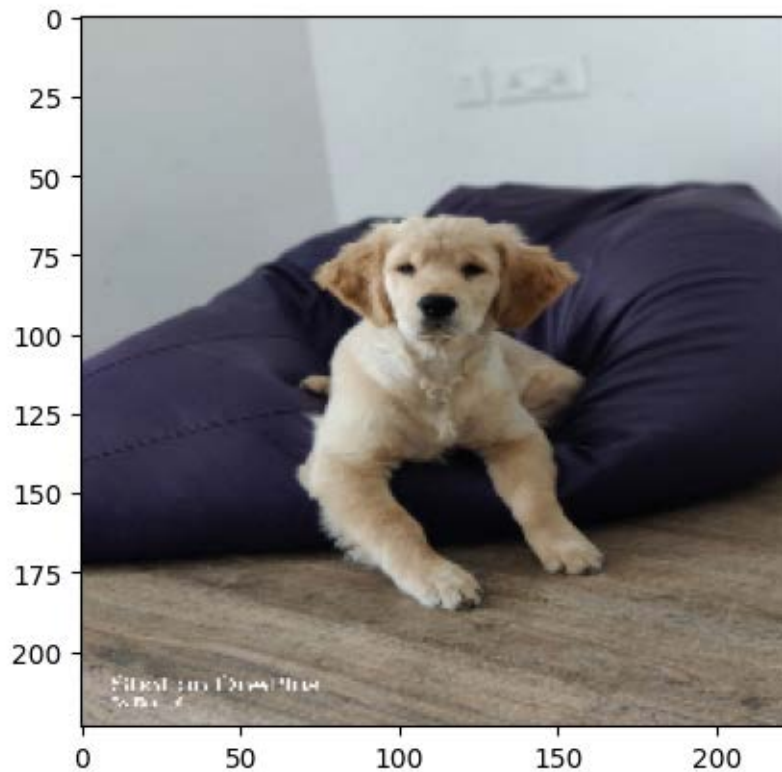
```

```
img_tensor /= 255.

# Its shape is (1, 224, 224, 3)
print(img_tensor.shape)

plt.imshow(img_tensor[0])
plt.show()
```

(1, 224, 224, 3)



#### 1.2.4 Image pre process to visualize

```
[ ]: # import tensorflow as tf
transfer_model = tf.keras.models.load_model('transfer_resnet50_2b_aug1.h5')
transfer_model.summary()
```

WARNING:absl:At this time, the v2.11+ optimizer `tf.keras.optimizers.SGD` runs slowly on M1/M2 Macs, please use the legacy Keras optimizer instead, located at `tf.keras.optimizers.legacy.SGD`.

Model: "model\_1"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	[]
conv1_pad (ZeroPadding2D) ['input_1[0][0]']	(None, 230, 230, 3)	0	
conv1_conv (Conv2D) ['conv1_pad[0][0]']	(None, 112, 112, 64)	9472	
conv1_bn (BatchNormalizati on) ['conv1_conv[0][0]']	(None, 112, 112, 64)	256	
conv1_relu (Activation) ['conv1_bn[0][0]']	(None, 112, 112, 64)	0	
pool1_pad (ZeroPadding2D) ['conv1_relu[0][0]']	(None, 114, 114, 64)	0	
pool1_pool (MaxPooling2D) ['pool1_pad[0][0]']	(None, 56, 56, 64)	0	
conv2_block1_1_conv (Conv2 D) ['pool1_pool[0][0]']	(None, 56, 56, 64)	4160	
conv2_block1_1_bn (BatchNo rmalization) ['conv2_block1_1_conv[0][0]']	(None, 56, 56, 64)	256	
conv2_block1_1_relu (Activ ation) ['conv2_block1_1_bn[0][0]']	(None, 56, 56, 64)	0	
conv2_block1_2_conv (Conv2 D) ['conv2_block1_1_relu[0][0]']	(None, 56, 56, 64)	36928	
conv2_block1_2_bn (BatchNo rmalization) ['conv2_block1_2_conv[0][0]']	(None, 56, 56, 64)	256	
conv2_block1_2_relu (Activ ation) ['conv2_block1_2_bn[0][0]']	(None, 56, 56, 64)	0	

conv2_block1_0_conv (Conv2 (None, 56, 56, 256) ['pool1_pool[0][0]'] D)	16640
conv2_block1_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block1_2_relu[0][0]'] D)	16640
conv2_block1_0_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_0_conv[0][0]'] rmalization)	1024
conv2_block1_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_3_conv[0][0]'] rmalization)	1024
conv2_block1_add (Add) (None, 56, 56, 256) ['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']	0
conv2_block1_out (Activati (None, 56, 56, 256) ['conv2_block1_add[0][0]'] on)	0
conv2_block2_1_conv (Conv2 (None, 56, 56, 64) ['conv2_block1_out[0][0]'] D)	16448
conv2_block2_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block2_1_conv[0][0]'] rmalization)	256
conv2_block2_1_relu (Activ (None, 56, 56, 64) ['conv2_block2_1_bn[0][0]'] ation)	0
conv2_block2_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block2_1_relu[0][0]'] D)	36928
conv2_block2_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block2_2_conv[0][0]'] rmalization)	256
conv2_block2_2_relu (Activ (None, 56, 56, 64) ['conv2_block2_2_bn[0][0]'] ation)	0

conv2_block2_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block2_2_relu[0][0]'] D)	16640
conv2_block2_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block2_3_conv[0][0]'] rmalization)	1024
conv2_block2_add (Add) (None, 56, 56, 256) ['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']	0
conv2_block2_out (Activati (None, 56, 56, 256) ['conv2_block2_add[0][0]'] on)	0
conv2_block3_1_conv (Conv2 (None, 56, 56, 64) ['conv2_block2_out[0][0]'] D)	16448
conv2_block3_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block3_1_conv[0][0]'] rmalization)	256
conv2_block3_1_relu (Activ (None, 56, 56, 64) ['conv2_block3_1_bn[0][0]'] ation)	0
conv2_block3_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block3_1_relu[0][0]'] D)	36928
conv2_block3_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block3_2_conv[0][0]'] rmalization)	256
conv2_block3_2_relu (Activ (None, 56, 56, 64) ['conv2_block3_2_bn[0][0]'] ation)	0
conv2_block3_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block3_2_relu[0][0]'] D)	16640
conv2_block3_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block3_3_conv[0][0]'] rmalization)	1024

conv2_block3_add (Add) ['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']	(None, 56, 56, 256)	0
conv2_block3_out (Activation) ['conv2_block3_add[0][0]']	(None, 56, 56, 256)	0
conv3_block1_1_conv (Conv2D) ['conv2_block3_out[0][0]']	(None, 28, 28, 128)	32896
conv3_block1_1_bn (Batch Normalization) ['conv3_block1_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block1_1_relu (Activation) ['conv3_block1_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block1_2_conv (Conv2D) ['conv3_block1_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block1_2_bn (Batch Normalization) ['conv3_block1_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block1_2_relu (Activation) ['conv3_block1_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block1_0_conv (Conv2D) ['conv2_block3_out[0][0]']	(None, 28, 28, 512)	131584
conv3_block1_3_conv (Conv2D) ['conv3_block1_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block1_0_bn (Batch Normalization) ['conv3_block1_0_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block1_3_bn (Batch Normalization) ['conv3_block1_3_conv[0][0]']	(None, 28, 28, 512)	2048

conv3_block1_add (Add) ['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block1_out (Activation) ['conv3_block1_add[0][0]']	(None, 28, 28, 512)	0
conv3_block2_1_conv (Conv2D) ['conv3_block1_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block2_1_bn (Batch Normalization) ['conv3_block2_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block2_1_relu (Activation) ['conv3_block2_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block2_2_conv (Conv2D) ['conv3_block2_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block2_2_bn (Batch Normalization) ['conv3_block2_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block2_2_relu (Activation) ['conv3_block2_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block2_3_conv (Conv2D) ['conv3_block2_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block2_3_bn (Batch Normalization) ['conv3_block2_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block2_add (Add) ['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block2_out (Activation) ['conv3_block2_add[0][0]']	(None, 28, 28, 512)	0



conv3_block3_1_conv (Conv2D) ['conv3_block2_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block3_1_bn (Batch Normalization) ['conv3_block3_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block3_1_relu (Activation) ['conv3_block3_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block3_2_conv (Conv2D) ['conv3_block3_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block3_2_bn (Batch Normalization) ['conv3_block3_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block3_2_relu (Activation) ['conv3_block3_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block3_3_conv (Conv2D) ['conv3_block3_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block3_3_bn (Batch Normalization) ['conv3_block3_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block3_add (Add) ['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block3_out (Activation) ['conv3_block3_add[0][0]']	(None, 28, 28, 512)	0
conv3_block4_1_conv (Conv2D) ['conv3_block3_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block4_1_bn (Batch Normalization) ['conv3_block4_1_conv[0][0]']	(None, 28, 28, 128)	512

conv3_block4_1_relu (Activation) (None, 28, 28, 128) ['conv3_block4_1_bn[0][0]']	0
conv3_block4_2_conv (Conv2D) (None, 28, 28, 128) ['conv3_block4_1_relu[0][0]']	147584
conv3_block4_2_bn (Batch Normalization) (None, 28, 28, 128) ['conv3_block4_2_conv[0][0]']	512
conv3_block4_2_relu (Activation) (None, 28, 28, 128) ['conv3_block4_2_bn[0][0]']	0
conv3_block4_3_conv (Conv2D) (None, 28, 28, 512) ['conv3_block4_2_relu[0][0]']	66048
conv3_block4_3_bn (Batch Normalization) (None, 28, 28, 512) ['conv3_block4_3_conv[0][0]']	2048
conv3_block4_add (Add) (None, 28, 28, 512) ['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']	0
conv3_block4_out (Activation) (None, 28, 28, 512) ['conv3_block4_add[0][0]']	0
conv4_block1_1_conv (Conv2D) (None, 14, 14, 256) ['conv3_block4_out[0][0]']	131328
conv4_block1_1_bn (Batch Normalization) (None, 14, 14, 256) ['conv4_block1_1_conv[0][0]']	1024
conv4_block1_1_relu (Activation) (None, 14, 14, 256) ['conv4_block1_1_bn[0][0]']	0
conv4_block1_2_conv (Conv2D) (None, 14, 14, 256) ['conv4_block1_1_relu[0][0]']	590080

conv4_block1_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block1_2_conv[0][0]'] rmalization)	1024
conv4_block1_2_relu (Activ (None, 14, 14, 256) ['conv4_block1_2_bn[0][0]'] ation)	0
conv4_block1_0_conv (Conv2 (None, 14, 14, 1024) ['conv3_block4_out[0][0]'] D)	525312
conv4_block1_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block1_2_relu[0][0]'] D)	263168
conv4_block1_0_bn (BatchNo (None, 14, 14, 1024) ['conv4_block1_0_conv[0][0]'] rmalization)	4096
conv4_block1_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block1_3_conv[0][0]'] rmalization)	4096
conv4_block1_add (Add) (None, 14, 14, 1024) ['conv4_block1_0_bn[0][0]', 'conv4_block1_3_bn[0][0]']	0
conv4_block1_out (Activati (None, 14, 14, 1024) ['conv4_block1_add[0][0]'] on)	0
conv4_block2_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block1_out[0][0]'] D)	262400
conv4_block2_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block2_1_conv[0][0]'] rmalization)	1024
conv4_block2_1_relu (Activ (None, 14, 14, 256) ['conv4_block2_1_bn[0][0]'] ation)	0
conv4_block2_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block2_1_relu[0][0]'] D)	590080

conv4_block2_2_bn (BatchNormal [ 'conv4_block2_2_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block2_2_relu (Activation [ 'conv4_block2_2_bn[0][0]' ation)	(None, 14, 14, 256)	0
conv4_block2_3_conv (Conv2D [ 'conv4_block2_2_relu[0][0]' D)	(None, 14, 14, 1024)	263168
conv4_block2_3_bn (BatchNormal [ 'conv4_block2_3_conv[0][0]' rmalization)	(None, 14, 14, 1024)	4096
conv4_block2_add (Add [ 'conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]'	(None, 14, 14, 1024)	0
conv4_block2_out (Activation [ 'conv4_block2_add[0][0]' on)	(None, 14, 14, 1024)	0
conv4_block3_1_conv (Conv2D [ 'conv4_block2_out[0][0]' D)	(None, 14, 14, 256)	262400
conv4_block3_1_bn (BatchNormal [ 'conv4_block3_1_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block3_1_relu (Activation [ 'conv4_block3_1_bn[0][0]' ation)	(None, 14, 14, 256)	0
conv4_block3_2_conv (Conv2D [ 'conv4_block3_1_relu[0][0]' D)	(None, 14, 14, 256)	590080
conv4_block3_2_bn (BatchNormal [ 'conv4_block3_2_conv[0][0]' rmalization)	(None, 14, 14, 256)	1024
conv4_block3_2_relu (Activation [ 'conv4_block3_2_bn[0][0]' ation)	(None, 14, 14, 256)	0

conv4_block3_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block3_2_relu[0][0]'] D)	263168
conv4_block3_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block3_3_conv[0][0]'] rmalization)	4096
conv4_block3_add (Add) (None, 14, 14, 1024) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	0
conv4_block3_out (Activati (None, 14, 14, 1024) ['conv4_block3_add[0][0]'] on)	0
conv4_block4_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_out[0][0]'] D)	262400
conv4_block4_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block4_1_conv[0][0]'] rmalization)	1024
conv4_block4_1_relu (Activ (None, 14, 14, 256) ['conv4_block4_1_bn[0][0]'] ation)	0
conv4_block4_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block4_1_relu[0][0]'] D)	590080
conv4_block4_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block4_2_conv[0][0]'] rmalization)	1024
conv4_block4_2_relu (Activ (None, 14, 14, 256) ['conv4_block4_2_bn[0][0]'] ation)	0
conv4_block4_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block4_2_relu[0][0]'] D)	263168
conv4_block4_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block4_3_conv[0][0]'] rmalization)	4096

conv4_block4_add (Add) ['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block4_out (Activation) ['conv4_block4_add[0][0]']	(None, 14, 14, 1024)	0
conv4_block5_1_conv (Conv2D) ['conv4_block4_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block5_1_bn (Batch Normalization) ['conv4_block5_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block5_1_relu (Activation) ['conv4_block5_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block5_2_conv (Conv2D) ['conv4_block5_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block5_2_bn (Batch Normalization) ['conv4_block5_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block5_2_relu (Activation) ['conv4_block5_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block5_3_conv (Conv2D) ['conv4_block5_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block5_3_bn (Batch Normalization) ['conv4_block5_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block5_add (Add) ['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block5_out (Activation) ['conv4_block5_add[0][0]']	(None, 14, 14, 1024)	0

conv4_block6_1_conv (Conv2D) ['conv4_block5_out[0][0]']	(None, 14, 14, 256)	262400
conv4_block6_1_bn (Batch Normalization) ['conv4_block6_1_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block6_1_relu (Activation) ['conv4_block6_1_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block6_2_conv (Conv2D) ['conv4_block6_1_relu[0][0]']	(None, 14, 14, 256)	590080
conv4_block6_2_bn (Batch Normalization) ['conv4_block6_2_conv[0][0]']	(None, 14, 14, 256)	1024
conv4_block6_2_relu (Activation) ['conv4_block6_2_bn[0][0]']	(None, 14, 14, 256)	0
conv4_block6_3_conv (Conv2D) ['conv4_block6_2_relu[0][0]']	(None, 14, 14, 1024)	263168
conv4_block6_3_bn (Batch Normalization) ['conv4_block6_3_conv[0][0]']	(None, 14, 14, 1024)	4096
conv4_block6_add (Add) ['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']	(None, 14, 14, 1024)	0
conv4_block6_out (Activation) ['conv4_block6_add[0][0]']	(None, 14, 14, 1024)	0
conv5_block1_1_conv (Conv2D) ['conv4_block6_out[0][0]']	(None, 7, 7, 512)	524800
conv5_block1_1_bn (Batch Normalization) ['conv5_block1_1_conv[0][0]']	(None, 7, 7, 512)	2048

conv5_block1_1_relu (Activation) (None, 7, 7, 512) ['conv5_block1_1_bn[0][0]']	0
conv5_block1_2_conv (Conv2D) (None, 7, 7, 512) ['conv5_block1_1_relu[0][0]']	2359808
conv5_block1_2_bn (Batch Normalization) (None, 7, 7, 512) ['conv5_block1_2_conv[0][0]']	2048
conv5_block1_2_relu (Activation) (None, 7, 7, 512) ['conv5_block1_2_bn[0][0]']	0
conv5_block1_0_conv (Conv2D) (None, 7, 7, 2048) ['conv4_block6_out[0][0]']	2099200
conv5_block1_3_conv (Conv2D) (None, 7, 7, 2048) ['conv5_block1_2_relu[0][0]']	1050624
conv5_block1_0_bn (Batch Normalization) (None, 7, 7, 2048) ['conv5_block1_0_conv[0][0]']	8192
conv5_block1_3_bn (Batch Normalization) (None, 7, 7, 2048) ['conv5_block1_3_conv[0][0]']	8192
conv5_block1_add (Add) (None, 7, 7, 2048) ['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']	0
conv5_block1_out (Activation) (None, 7, 7, 2048) ['conv5_block1_add[0][0]']	0
conv5_block2_1_conv (Conv2D) (None, 7, 7, 512) ['conv5_block1_out[0][0]']	1049088
conv5_block2_1_bn (Batch Normalization) (None, 7, 7, 512) ['conv5_block2_1_conv[0][0]']	2048



conv5_block2_1_relu (Activation) (None, 7, 7, 512) ['conv5_block2_1_bn[0][0]']	0
conv5_block2_2_conv (Conv2D) (None, 7, 7, 512) ['conv5_block2_1_relu[0][0]']	2359808
conv5_block2_2_bn (Batch Normalization) (None, 7, 7, 512) ['conv5_block2_2_conv[0][0]']	2048
conv5_block2_2_relu (Activation) (None, 7, 7, 512) ['conv5_block2_2_bn[0][0]']	0
conv5_block2_3_conv (Conv2D) (None, 7, 7, 2048) ['conv5_block2_2_relu[0][0]']	1050624
conv5_block2_3_bn (Batch Normalization) (None, 7, 7, 2048) ['conv5_block2_3_conv[0][0]']	8192
conv5_block2_add (Add) (None, 7, 7, 2048) ['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']	0
conv5_block2_out (Activation) (None, 7, 7, 2048) ['conv5_block2_add[0][0]']	0
conv5_block3_1_conv (Conv2D) (None, 7, 7, 512) ['conv5_block2_out[0][0]']	1049088
conv5_block3_1_bn (Batch Normalization) (None, 7, 7, 512) ['conv5_block3_1_conv[0][0]']	2048
conv5_block3_1_relu (Activation) (None, 7, 7, 512) ['conv5_block3_1_bn[0][0]']	0
conv5_block3_2_conv (Conv2D) (None, 7, 7, 512) ['conv5_block3_1_relu[0][0]']	2359808

conv5_block3_2_bn (Batch Normalization)	(None, 7, 7, 512)	2048
conv5_block3_2_relu (Activation)	(None, 7, 7, 512)	0
conv5_block3_3_conv (Conv2D)	(None, 7, 7, 2048)	1050624
conv5_block3_3_bn (Batch Normalization)	(None, 7, 7, 2048)	8192
conv5_block3_add (Add)	(None, 7, 7, 2048)	0
conv5_block3_out (Activation)	(None, 7, 7, 2048)	0
average_pooling2d (Average Pooling2D)	(None, 3, 3, 2048)	0
flatten (Flatten)	(None, 18432)	0
fc9_predictions (Dense)	(None, 1)	18433

=====

=====

Total params: 23606145 (90.05 MB)  
 Trainable params: 23553025 (89.85 MB)  
 Non-trainable params: 53120 (207.50 KB)

-----

-----

```
[ ]: # Visualize predictions on test data
      predictions = transfer_model.predict(x_test_new_resize)
      predicted_labels = (predictions > 0.5).astype(int)
      sample_images = x_test_new_resize[:10]
```

```

sample_labels = y_test_new_norm[:10]

for i in range(10):
    plt.subplot(2, 5, i + 1)
    plt.imshow(sample_images[i])
    plt.title(f'Predicted: {predicted_labels[i][0]}\nActual: ↪{sample_labels[i][0]}')
    plt.axis('off')

plt.show()

```

32/32 [=====] - 21s 637ms/step

Predicted: 1 Predicted: 1 Predicted: 1 Predicted: 1 Predicted: 1  
 Actual: 0.0 Actual: 0.0 Actual: 0.0 Actual: 0.0 Actual: 0.0



Predicted: 1 Predicted: 1 Predicted: 1 Predicted: 1 Predicted: 1  
 Actual: 0.0 Actual: 0.0 Actual: 0.0 Actual: 0.0 Actual: 0.0



### 1.2.5 PS1.B Visualize Activation

```

[ ]: # img_path = 'dogscats/subset/test/cats/cat.1700.jpg'
img_path='refs/dog_1.jpg'
# img_path='refs/dog.215.jpg'

# We preprocess the image into a 4D tensor
from keras.preprocessing import image
import numpy as np

im_size = 224

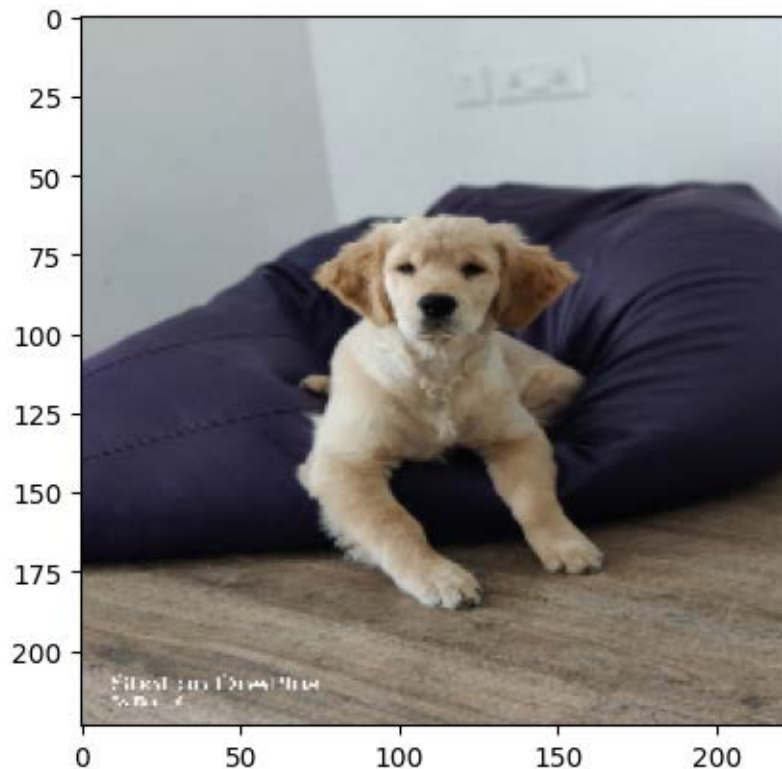
```

```
img = tf.keras.utils.load_img(img_path, target_size=(im_size, im_size))
img_tensor = tf.keras.utils.img_to_array(img)
img_tensor = np.expand_dims(img_tensor, axis=0)
# Remember that the model was trained on inputs
# that were preprocessed in the following way:
img_tensor /= 255.

# Its shape is (1, 150, 150, 3)
print(img_tensor.shape)
```

(1, 224, 224, 3)

```
[ ]: plt.imshow(img_tensor[0])
plt.show()
```



### Chosen layer

```
[ ]: layers_to_visualize = [transfer_model.get_layer('conv1_relu'),
                           transfer_model.get_layer('conv3_block2_2_relu'),
                           transfer_model.get_layer('conv5_block3_2_relu')]

for layer in layers_to_visualize:
```

```
print(layer.name)
```

```
conv1_relu
conv3_block2_2_relu
conv5_block3_2_relu
```

```
[ ]: act_input1 = transfer_model.input
# Extracts the outputs of the top 50 layers:
# layer_outputs = [layer.output for layer in model.layers[1:51]]
layer_outputs1 = [layer.output for layer in layers_to_visualize]

# First block : conv1
# Middle block: conv2c_branch2b
# Last block: conv5c_branch2c

# layer_outputs = [layer.output for layer in layers_to_visualize]
# Creates a model that will return these outputs, given the model input:
activation_model1 = Model(inputs=act_input1, outputs=layer_outputs1)
activation_model1.summary()
```

Model: "model\_2"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	[]
conv1_pad (ZeroPadding2D)	(None, 230, 230, 3)	0	['input_1[0][0]']
conv1_conv (Conv2D)	(None, 112, 112, 64)	9472	['conv1_pad[0][0]']
conv1_bn (BatchNormalizati on)	(None, 112, 112, 64)	256	['conv1_conv[0][0]']
conv1_relu (Activation)	(None, 112, 112, 64)	0	['conv1_bn[0][0]']
pool1_pad (ZeroPadding2D)	(None, 114, 114, 64)	0	['conv1_relu[0][0]']
pool1_pool (MaxPooling2D)	(None, 56, 56, 64)	0	['pool1_pad[0][0]']
conv2_block1_1_conv (Conv2	(None, 56, 56, 64)	4160	

```

['pool1_pool[0][0]']
D)

conv2_block1_1_bn (BatchNo (None, 56, 56, 64)          256
['conv2_block1_1_conv[0][0]']
rmalization)

conv2_block1_1_relu (Activ (None, 56, 56, 64)          0
['conv2_block1_1_bn[0][0]']
ation)

conv2_block1_2_conv (Conv2 (None, 56, 56, 64)          36928
['conv2_block1_1_relu[0][0]']
D)

conv2_block1_2_bn (BatchNo (None, 56, 56, 64)          256
['conv2_block1_2_conv[0][0]']
rmalization)

conv2_block1_2_relu (Activ (None, 56, 56, 64)          0
['conv2_block1_2_bn[0][0]']
ation)

conv2_block1_0_conv (Conv2 (None, 56, 56, 256)         16640
['pool1_pool[0][0]']
D)

conv2_block1_3_conv (Conv2 (None, 56, 56, 256)         16640
['conv2_block1_2_relu[0][0]']
D)

conv2_block1_0_bn (BatchNo (None, 56, 56, 256)         1024
['conv2_block1_0_conv[0][0]']
rmalization)

conv2_block1_3_bn (BatchNo (None, 56, 56, 256)         1024
['conv2_block1_3_conv[0][0]']
rmalization)

conv2_block1_add (Add) (None, 56, 56, 256)          0
['conv2_block1_0_bn[0][0]',
'conv2_block1_3_bn[0][0]']

conv2_block1_out (Activati (None, 56, 56, 256)          0
['conv2_block1_add[0][0]']
on)

conv2_block2_1_conv (Conv2 (None, 56, 56, 64)          16448

```

```

['conv2_block1_out[0][0]']
D)

conv2_block2_1_bn (BatchNo (None, 56, 56, 64)          256
['conv2_block2_1_conv[0][0]']
rmalization)

conv2_block2_1_relu (Activ (None, 56, 56, 64)          0
['conv2_block2_1_bn[0][0]']
ation)

conv2_block2_2_conv (Conv2 (None, 56, 56, 64)          36928
['conv2_block2_1_relu[0][0]']
D)

conv2_block2_2_bn (BatchNo (None, 56, 56, 64)          256
['conv2_block2_2_conv[0][0]']
rmalization)

conv2_block2_2_relu (Activ (None, 56, 56, 64)          0
['conv2_block2_2_bn[0][0]']
ation)

conv2_block2_3_conv (Conv2 (None, 56, 56, 256)         16640
['conv2_block2_2_relu[0][0]']
D)

conv2_block2_3_bn (BatchNo (None, 56, 56, 256)         1024
['conv2_block2_3_conv[0][0]']
rmalization)

conv2_block2_add (Add) (None, 56, 56, 256)             0
['conv2_block1_out[0][0]',
'conv2_block2_3_bn[0][0]']

conv2_block2_out (Activati (None, 56, 56, 256)         0
['conv2_block2_add[0][0]']
on)

conv2_block3_1_conv (Conv2 (None, 56, 56, 64)          16448
['conv2_block2_out[0][0]']
D)

conv2_block3_1_bn (BatchNo (None, 56, 56, 64)          256
['conv2_block3_1_conv[0][0]']
rmalization)

conv2_block3_1_relu (Activ (None, 56, 56, 64)          0

```

['conv2_block3_1_bn[0][0]'] ation)	
conv2_block3_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block3_1_relu[0][0]'] D)	36928
conv2_block3_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block3_2_conv[0][0]'] rmalization)	256
conv2_block3_2_relu (Activ (None, 56, 56, 64) ['conv2_block3_2_bn[0][0]'] ation)	0
conv2_block3_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block3_2_relu[0][0]'] D)	16640
conv2_block3_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block3_3_conv[0][0]'] rmalization)	1024
conv2_block3_add (Add) (None, 56, 56, 256) ['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']	0
conv2_block3_out (Activati (None, 56, 56, 256) ['conv2_block3_add[0][0]'] on)	0
conv3_block1_1_conv (Conv2 (None, 28, 28, 128) ['conv2_block3_out[0][0]'] D)	32896
conv3_block1_1_bn (BatchNo (None, 28, 28, 128) ['conv3_block1_1_conv[0][0]'] rmalization)	512
conv3_block1_1_relu (Activ (None, 28, 28, 128) ['conv3_block1_1_bn[0][0]'] ation)	0
conv3_block1_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block1_1_relu[0][0]'] D)	147584
conv3_block1_2_bn (BatchNo (None, 28, 28, 128)	512



['conv3_block1_2_conv[0][0]'] rmalization)		
conv3_block1_2_relu (Activ ['conv3_block1_2_bn[0][0]'] ation)	(None, 28, 28, 128)	0
conv3_block1_0_conv (Conv2 ['conv2_block3_out[0][0]'] D)	(None, 28, 28, 512)	131584
conv3_block1_3_conv (Conv2 ['conv3_block1_2_relu[0][0]'] D)	(None, 28, 28, 512)	66048
conv3_block1_0_bn (BatchNo ['conv3_block1_0_conv[0][0]'] rmalization)	(None, 28, 28, 512)	2048
conv3_block1_3_bn (BatchNo ['conv3_block1_3_conv[0][0]'] rmalization)	(None, 28, 28, 512)	2048
conv3_block1_add (Add) ['conv3_block1_0_bn[0][0]', 'conv3_block1_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block1_out (Activati ['conv3_block1_add[0][0]'] on)	(None, 28, 28, 512)	0
conv3_block2_1_conv (Conv2 ['conv3_block1_out[0][0]'] D)	(None, 28, 28, 128)	65664
conv3_block2_1_bn (BatchNo ['conv3_block2_1_conv[0][0]'] rmalization)	(None, 28, 28, 128)	512
conv3_block2_1_relu (Activ ['conv3_block2_1_bn[0][0]'] ation)	(None, 28, 28, 128)	0
conv3_block2_2_conv (Conv2 ['conv3_block2_1_relu[0][0]'] D)	(None, 28, 28, 128)	147584
conv3_block2_2_bn (BatchNo (None, 28, 28, 128)		512

['conv3_block2_2_conv[0][0]'] rmalization)		
conv3_block2_2_relu (Activ ['conv3_block2_2_bn[0][0]'] ation)	(None, 28, 28, 128)	0
conv3_block2_3_conv (Conv2 ['conv3_block2_2_relu[0][0]'] D)	(None, 28, 28, 512)	66048
conv3_block2_3_bn (BatchNo ['conv3_block2_3_conv[0][0]'] rmalization)	(None, 28, 28, 512)	2048
conv3_block2_add (Add) ['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block2_out (Activati ['conv3_block2_add[0][0]'] on)	(None, 28, 28, 512)	0
conv3_block3_1_conv (Conv2 ['conv3_block2_out[0][0]'] D)	(None, 28, 28, 128)	65664
conv3_block3_1_bn (BatchNo ['conv3_block3_1_conv[0][0]'] rmalization)	(None, 28, 28, 128)	512
conv3_block3_1_relu (Activ ['conv3_block3_1_bn[0][0]'] ation)	(None, 28, 28, 128)	0
conv3_block3_2_conv (Conv2 ['conv3_block3_1_relu[0][0]'] D)	(None, 28, 28, 128)	147584
conv3_block3_2_bn (BatchNo ['conv3_block3_2_conv[0][0]'] rmalization)	(None, 28, 28, 128)	512
conv3_block3_2_relu (Activ ['conv3_block3_2_bn[0][0]'] ation)	(None, 28, 28, 128)	0
conv3_block3_3_conv (Conv2	(None, 28, 28, 512)	66048

['conv3_block3_2_relu[0][0]'] D)		
conv3_block3_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block3_3_conv[0][0]'] rmalization)		2048
conv3_block3_add (Add) (None, 28, 28, 512) ['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']		0
conv3_block3_out (Activati (None, 28, 28, 512) ['conv3_block3_add[0][0]'] on)		0
conv3_block4_1_conv (Conv2 (None, 28, 28, 128) ['conv3_block3_out[0][0]'] D)		65664
conv3_block4_1_bn (BatchNo (None, 28, 28, 128) ['conv3_block4_1_conv[0][0]'] rmalization)		512
conv3_block4_1_relu (Activ (None, 28, 28, 128) ['conv3_block4_1_bn[0][0]'] ation)		0
conv3_block4_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block4_1_relu[0][0]'] D)		147584
conv3_block4_2_bn (BatchNo (None, 28, 28, 128) ['conv3_block4_2_conv[0][0]'] rmalization)		512
conv3_block4_2_relu (Activ (None, 28, 28, 128) ['conv3_block4_2_bn[0][0]'] ation)		0
conv3_block4_3_conv (Conv2 (None, 28, 28, 512) ['conv3_block4_2_relu[0][0]'] D)		66048
conv3_block4_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block4_3_conv[0][0]'] rmalization)		2048
conv3_block4_add (Add) (None, 28, 28, 512)		0

```

['conv3_block3_out[0][0]',
'conv3_block4_3_bn[0][0]']

conv3_block4_out (Activation) (None, 28, 28, 512) 0
['conv3_block4_add[0][0]']

conv4_block1_1_conv (Conv2D) (None, 14, 14, 256) 131328
['conv3_block4_out[0][0]']

conv4_block1_1_bn (BatchNormalization) (None, 14, 14, 256) 1024
['conv4_block1_1_conv[0][0]']

conv4_block1_1_relu (Activation) (None, 14, 14, 256) 0
['conv4_block1_1_bn[0][0]']

conv4_block1_2_conv (Conv2D) (None, 14, 14, 256) 590080
['conv4_block1_1_relu[0][0]']

conv4_block1_2_bn (BatchNormalization) (None, 14, 14, 256) 1024
['conv4_block1_2_conv[0][0]']

conv4_block1_2_relu (Activation) (None, 14, 14, 256) 0
['conv4_block1_2_bn[0][0]']

conv4_block1_0_conv (Conv2D) (None, 14, 14, 1024) 525312
['conv3_block4_out[0][0]']

conv4_block1_3_conv (Conv2D) (None, 14, 14, 1024) 263168
['conv4_block1_2_relu[0][0]']

conv4_block1_0_bn (BatchNormalization) (None, 14, 14, 1024) 4096
['conv4_block1_0_conv[0][0]']

conv4_block1_3_bn (BatchNormalization) (None, 14, 14, 1024) 4096
['conv4_block1_3_conv[0][0]']

conv4_block1_add (Add) (None, 14, 14, 1024) 0

```

```

['conv4_block1_0_bn[0][0]',
'conv4_block1_3_bn[0][0]']

conv4_block1_out (Activation) (None, 14, 14, 1024) 0
['conv4_block1_add[0][0]']

conv4_block2_1_conv (Conv2D) (None, 14, 14, 256) 262400
['conv4_block1_out[0][0]']

conv4_block2_1_bn (BatchNormalization) (None, 14, 14, 256) 1024
['conv4_block2_1_conv[0][0]']

conv4_block2_1_relu (Activation) (None, 14, 14, 256) 0
['conv4_block2_1_bn[0][0]']

conv4_block2_2_conv (Conv2D) (None, 14, 14, 256) 590080
['conv4_block2_1_relu[0][0]']

conv4_block2_2_bn (BatchNormalization) (None, 14, 14, 256) 1024
['conv4_block2_2_conv[0][0]']

conv4_block2_2_relu (Activation) (None, 14, 14, 256) 0
['conv4_block2_2_bn[0][0]']

conv4_block2_3_conv (Conv2D) (None, 14, 14, 1024) 263168
['conv4_block2_2_relu[0][0]']

conv4_block2_3_bn (BatchNormalization) (None, 14, 14, 1024) 4096
['conv4_block2_3_conv[0][0]']

conv4_block2_add (Add) (None, 14, 14, 1024) 0
['conv4_block1_out[0][0]',
'conv4_block2_3_bn[0][0]']

conv4_block2_out (Activation) (None, 14, 14, 1024) 0
['conv4_block2_add[0][0]']

conv4_block3_1_conv (Conv2D) (None, 14, 14, 256) 262400

```

['conv4_block2_out[0][0]'] D)	
conv4_block3_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block3_1_conv[0][0]'] rmalization)	1024
conv4_block3_1_relu (Activ (None, 14, 14, 256) ['conv4_block3_1_bn[0][0]'] ation)	0
conv4_block3_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_1_relu[0][0]'] D)	590080
conv4_block3_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block3_2_conv[0][0]'] rmalization)	1024
conv4_block3_2_relu (Activ (None, 14, 14, 256) ['conv4_block3_2_bn[0][0]'] ation)	0
conv4_block3_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block3_2_relu[0][0]'] D)	263168
conv4_block3_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block3_3_conv[0][0]'] rmalization)	4096
conv4_block3_add (Add) (None, 14, 14, 1024) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	0
conv4_block3_out (Activati (None, 14, 14, 1024) ['conv4_block3_add[0][0]'] on)	0
conv4_block4_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_out[0][0]'] D)	262400
conv4_block4_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block4_1_conv[0][0]'] rmalization)	1024
conv4_block4_1_relu (Activ (None, 14, 14, 256)	0

['conv4_block4_1_bn[0][0]'] ation)	
conv4_block4_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block4_1_relu[0][0]'] D)	590080
conv4_block4_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block4_2_conv[0][0]'] rmalization)	1024
conv4_block4_2_relu (Activ (None, 14, 14, 256) ['conv4_block4_2_bn[0][0]'] ation)	0
conv4_block4_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block4_2_relu[0][0]'] D)	263168
conv4_block4_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block4_3_conv[0][0]'] rmalization)	4096
conv4_block4_add (Add) (None, 14, 14, 1024) ['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']	0
conv4_block4_out (Activati (None, 14, 14, 1024) ['conv4_block4_add[0][0]'] on)	0
conv4_block5_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block4_out[0][0]'] D)	262400
conv4_block5_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block5_1_conv[0][0]'] rmalization)	1024
conv4_block5_1_relu (Activ (None, 14, 14, 256) ['conv4_block5_1_bn[0][0]'] ation)	0
conv4_block5_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block5_1_relu[0][0]'] D)	590080
conv4_block5_2_bn (BatchNo (None, 14, 14, 256)	1024

['conv4_block5_2_conv[0][0]'] rmalization)		
conv4_block5_2_relu (Activ (None, 14, 14, 256) ['conv4_block5_2_bn[0][0]'] ation)		0
conv4_block5_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block5_2_relu[0][0]'] D)		263168
conv4_block5_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block5_3_conv[0][0]'] rmalization)		4096
conv4_block5_add (Add) (None, 14, 14, 1024) ['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']		0
conv4_block5_out (Activati (None, 14, 14, 1024) ['conv4_block5_add[0][0]'] on)		0
conv4_block6_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block5_out[0][0]'] D)		262400
conv4_block6_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block6_1_conv[0][0]'] rmalization)		1024
conv4_block6_1_relu (Activ (None, 14, 14, 256) ['conv4_block6_1_bn[0][0]'] ation)		0
conv4_block6_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block6_1_relu[0][0]'] D)		590080
conv4_block6_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block6_2_conv[0][0]'] rmalization)		1024
conv4_block6_2_relu (Activ (None, 14, 14, 256) ['conv4_block6_2_bn[0][0]'] ation)		0
conv4_block6_3_conv (Conv2 (None, 14, 14, 1024)		263168



['conv4_block6_2_relu[0][0]'] D)		
conv4_block6_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block6_3_conv[0][0]'] rmalization)		4096
conv4_block6_add (Add) (None, 14, 14, 1024) ['conv4_block5_out[0][0]', 'conv4_block6_3_bn[0][0]']		0
conv4_block6_out (Activati (None, 14, 14, 1024) ['conv4_block6_add[0][0]'] on)		0
conv5_block1_1_conv (Conv2 (None, 7, 7, 512) ['conv4_block6_out[0][0]'] D)		524800
conv5_block1_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block1_1_conv[0][0]'] rmalization)		2048
conv5_block1_1_relu (Activ (None, 7, 7, 512) ['conv5_block1_1_bn[0][0]'] ation)		0
conv5_block1_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block1_1_relu[0][0]'] D)		2359808
conv5_block1_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block1_2_conv[0][0]'] rmalization)		2048
conv5_block1_2_relu (Activ (None, 7, 7, 512) ['conv5_block1_2_bn[0][0]'] ation)		0
conv5_block1_0_conv (Conv2 (None, 7, 7, 2048) ['conv4_block6_out[0][0]'] D)		2099200
conv5_block1_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block1_2_relu[0][0]'] D)		1050624
conv5_block1_0_bn (BatchNo (None, 7, 7, 2048)		8192

['conv5_block1_0_conv[0][0]'] rmalization)		
conv5_block1_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block1_3_conv[0][0]'] rmalization)		8192
conv5_block1_add (Add) (None, 7, 7, 2048) ['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']		0
conv5_block1_out (Activati (None, 7, 7, 2048) ['conv5_block1_add[0][0]'] on)		0
conv5_block2_1_conv (Conv2 (None, 7, 7, 512) ['conv5_block1_out[0][0]'] D)		1049088
conv5_block2_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_1_conv[0][0]'] rmalization)		2048
conv5_block2_1_relu (Activ (None, 7, 7, 512) ['conv5_block2_1_bn[0][0]'] ation)		0
conv5_block2_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block2_1_relu[0][0]'] D)		2359808
conv5_block2_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block2_2_conv[0][0]'] rmalization)		2048
conv5_block2_2_relu (Activ (None, 7, 7, 512) ['conv5_block2_2_bn[0][0]'] ation)		0
conv5_block2_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block2_2_relu[0][0]'] D)		1050624
conv5_block2_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block2_3_conv[0][0]'] rmalization)		8192
conv5_block2_add (Add) (None, 7, 7, 2048)		0

```
['conv5_block1_out[0][0]',
'conv5_block2_3_bn[0][0]']

conv5_block2_out (Activation) (None, 7, 7, 2048) 0
['conv5_block2_add[0][0]']

conv5_block3_1_conv (Conv2D) (None, 7, 7, 512) 1049088
['conv5_block2_out[0][0]']

conv5_block3_1_bn (Batch Normalization) (None, 7, 7, 512) 2048
['conv5_block3_1_conv[0][0]']

conv5_block3_1_relu (Activation) (None, 7, 7, 512) 0
['conv5_block3_1_bn[0][0]']

conv5_block3_2_conv (Conv2D) (None, 7, 7, 512) 2359808
['conv5_block3_1_relu[0][0]']

conv5_block3_2_bn (Batch Normalization) (None, 7, 7, 512) 2048
['conv5_block3_2_conv[0][0]']

conv5_block3_2_relu (Activation) (None, 7, 7, 512) 0
['conv5_block3_2_bn[0][0]']
```

```
=====
=====
```

```
Total params: 22528896 (85.94 MB)
Trainable params: 22479872 (85.75 MB)
Non-trainable params: 49024 (191.50 KB)
```

```
-----
-----
```

### First Block Activation

```
[ ]: activations = activation_model1.predict([img_tensor])

first_layer_activation = activations[0]
print(first_layer_activation.shape)

for activation in activations:
    print(activation.shape)
```

```

fig, axes = plt.subplots(1, 3,  figsize=(20, 8))

# Display the first activation
axes[0].matshow(first_layer_activation[0, :, :, 6], cmap='viridis')
axes[0].set_title('Layer 1')

# Display the second activation
axes[1].matshow(first_layer_activation[0, :, :, 10], cmap='viridis')
axes[1].set_title('Layer 10')

# Display the second activation
axes[2].matshow(first_layer_activation[0, :, :, 15], cmap='viridis')
axes[2].set_title('Layer 26')

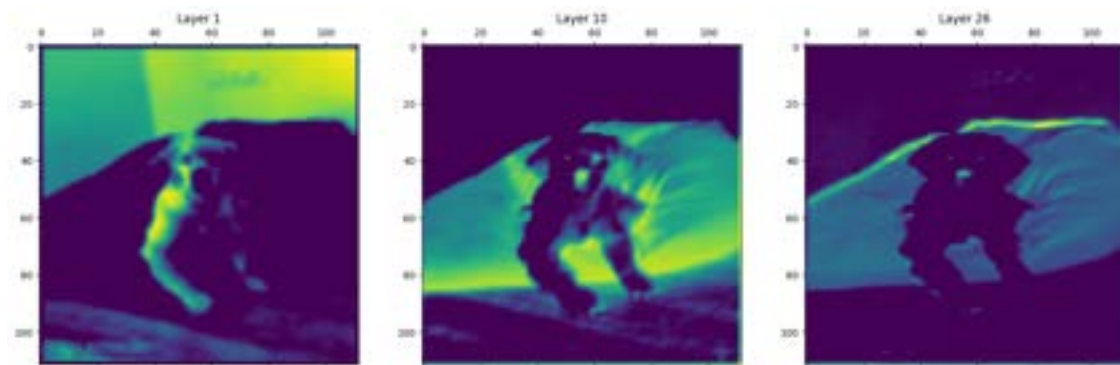
# Show the subplots
plt.show()

```

```

1/1 [=====] - 0s 55ms/step
(1, 112, 112, 64)
(1, 112, 112, 64)
(1, 28, 28, 128)
(1, 7, 7, 512)

```



This nose and eyes looks more bright green and is further clear or better to encode dog eyes and nose in the learnt model

At this point, let's go and plot a complete visualization of all the activations in the network.

### Middle Block Activation

```

[ ]: activations = activation_model1.predict([img_tensor])

middle_layer_activation = activations[1]
print(middle_layer_activation.shape)

```

```

for activation in activations:
    print(activation.shape)

fig, axes = plt.subplots(1, 3,  figsize=(20, 8))

# Display the first activation
axes[0].matshow(middle_layer_activation[0, :, :, 6], cmap='viridis')
axes[0].set_title('Layer 1')

# Display the second activation
axes[1].matshow(middle_layer_activation[0, :, :, 10], cmap='viridis')
axes[1].set_title('Layer 10')

# Display the second activation
axes[2].matshow(middle_layer_activation[0, :, :, 15], cmap='viridis')
axes[2].set_title('Layer 26')

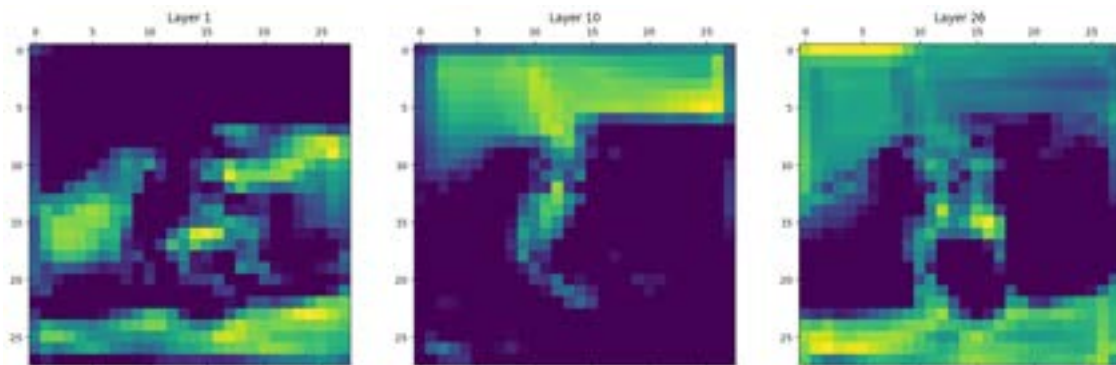
# Show the subplots
plt.show()

```

```

1/1 [=====] - 0s 62ms/step
(1, 28, 28, 128)
(1, 112, 112, 64)
(1, 28, 28, 128)
(1, 7, 7, 512)

```



### Last Block Activation

```

[ ]: activations = activation_model1.predict([img_tensor])

last_layer_activation = activations[2]
print(first_layer_activation.shape)

```

```

for activation in activations:
    print(activation.shape)

fig, axes = plt.subplots(1, 3, figsize=(20, 8))

# Display the first activation
axes[0].matshow(last_layer_activation[0, :, :, 1], cmap='viridis')
axes[0].set_title('Layer 1')

# Display the second activation
axes[1].matshow(last_layer_activation[0, :, :, 11], cmap='viridis')
axes[1].set_title('Layer 10')

# Display the second activation
axes[2].matshow(last_layer_activation[0, :, :, 15], cmap='viridis')
axes[2].set_title('Layer 26')

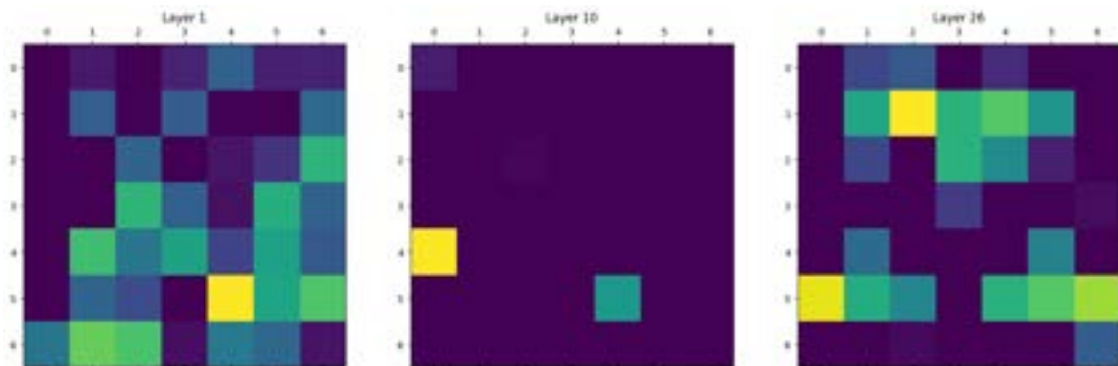
# Show the subplots
plt.show()

```

```

1/1 [=====] - 0s 53ms/step
(1, 112, 112, 64)
(1, 112, 112, 64)
(1, 28, 28, 128)
(1, 7, 7, 512)

```



### All Block channels

```

[ ]: # These are the names of the layers, so can have them as part of our plot
layer_names = []
# layer_names = layers_to_visualize
# for layer in model.layers[:50]:
#     layer_names.append(layer.name)
for layer in layers_to_visualize:

```

```

print(layer.name)
layer_names.append(layer.name)

# layer_names = layers_to_visualize

images_per_row = 16

# Now let's display our feature maps
for layer_name, layer_activation in zip(layer_names, activations):
    # This is the number of features in the feature map
    n_features = layer_activation.shape[-1]

    # The feature map has shape (1, size, size, n_features)
    size = layer_activation.shape[1]

    # We will tile the activation channels in this matrix
    n_cols = n_features // images_per_row
    display_grid = np.zeros((size * n_cols, images_per_row * size))

    # We'll tile each filter into this big horizontal grid
    for col in range(n_cols):
        for row in range(images_per_row):
            channel_image = layer_activation[0,
                                             :, :,
                                             col * images_per_row + row]

            # Post-process the feature to make it visually palatable
            channel_image -= channel_image.mean()
            channel_image /= channel_image.std()
            channel_image *= 64
            channel_image += 128
            channel_image = np.clip(channel_image, 0, 255).astype('uint8')
            display_grid[col * size : (col + 1) * size,
                          row * size : (row + 1) * size] = channel_image

    # Display the grid
    scale = 1. / size
    plt.figure(figsize=(scale * display_grid.shape[1],
                        scale * display_grid.shape[0]))
    plt.title(layer_name)
    plt.grid(False)
    plt.imshow(display_grid, aspect='auto', cmap='viridis')

plt.show()

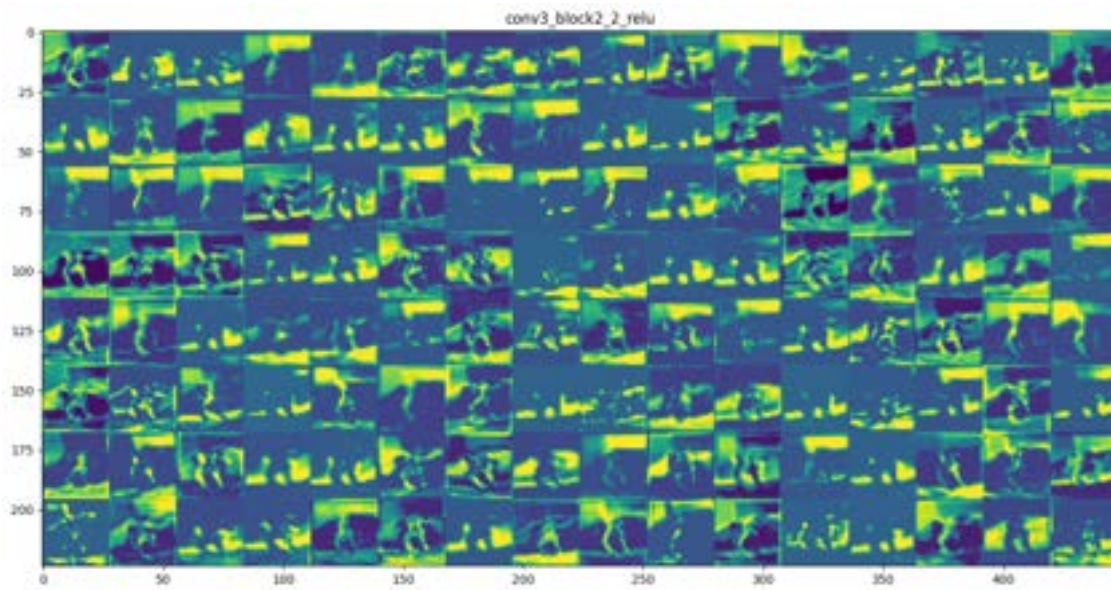
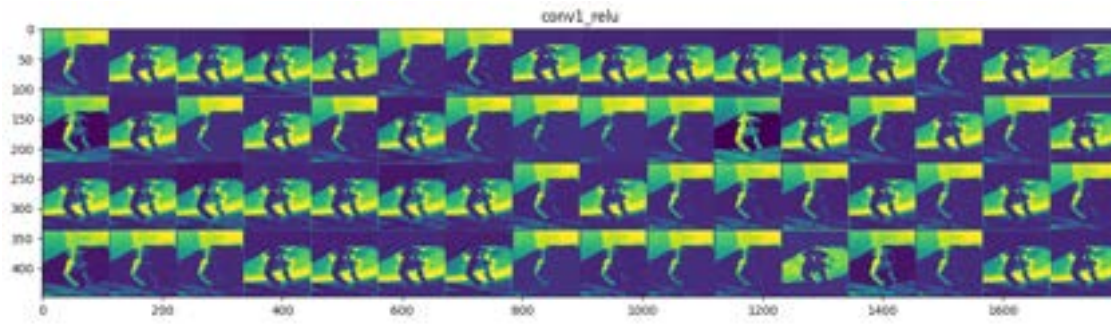
```

```

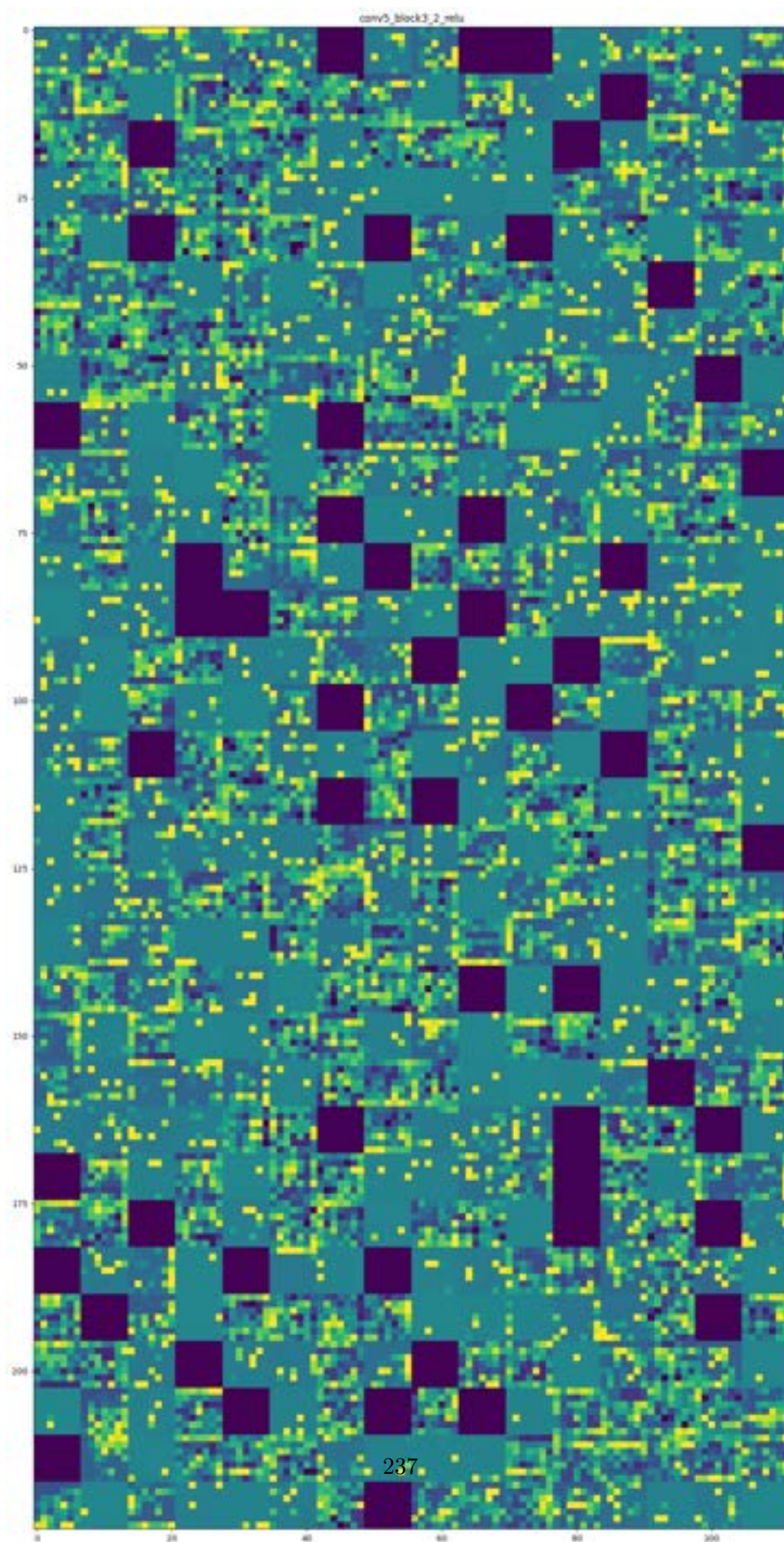
conv1_relu
conv3_block2_2_relu
conv5_block3_2_relu

```

```
/var/folders/xy/1x8f1v1x569_ch145p3563j80000gn/T/ipykernel_17264/1518767542.py:3  
7: RuntimeWarning: invalid value encountered in cast  
channel_image = np.clip(channel_image, 0, 255).astype('uint8')
```







Please scroll for all the plots

### 1.2.6 PS1.B Visualize convnet Filter

```
[ ]: layers_to_visualize = [transfer_model.get_layer('conv1_conv'),
                           transfer_model.get_layer('conv3_block2_2_conv'),
                           transfer_model.get_layer('conv5_block3_3_conv')]

for layer in layers_to_visualize:
    print(layer.name)
```

```
conv1_conv
conv3_block2_2_conv
conv5_block3_3_conv
```

```
[ ]: def deprocess_image(x):
    # normalize tensor: center on 0., ensure std is 0.1
    x -= x.mean()
    x /= (x.std() + 1e-5)
    x *= 0.1

    # clip to [0, 1]
    x += 0.5
    x = np.clip(x, 0, 1)

    # convert to RGB array
    x *= 255
    x = np.clip(x, 0, 255).astype('uint8')
    return x
```

```
[ ]: # Disable eager execution
tf.compat.v1.disable_eager_execution()
```

```
[ ]: def generate_pattern_transfer(layer_name, filter_index, size=224):

    layer = transfer_model.get_layer(layer_name)
    input_img = transfer_model.input

    loss = K.mean(layer.output[:, :, :, filter_index])
    grads = K.gradients(loss, input_img)[0]
    grads /= (K.sqrt(K.mean(K.square(grads))) + 1e-5)

    iterate = K.function([input_img], [loss, grads])

    input_img_data = np.random.random((1, size, size, 3)) * 20 + 128.
```

```

step = 1.
for i in range(40):
    loss_value, grads_value = iterate([input_img_data])
    input_img_data += grads_value * step

img = input_img_data[0]
return deprocess_image(img)

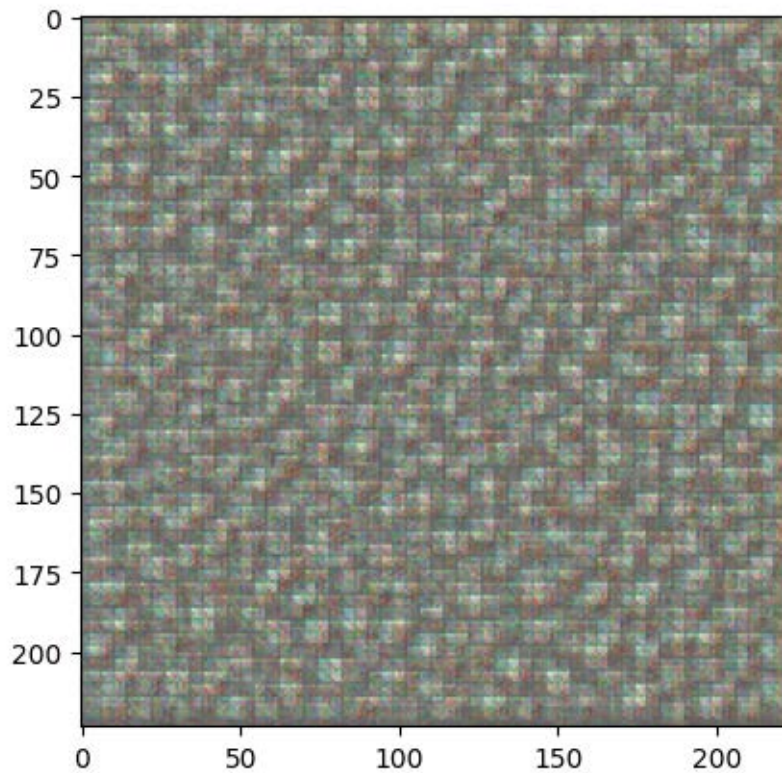
```

```

[ ]: # Now you can use generate_pattern function
layer_name = 'conv1_block1_1_conv'
filter_index = 0

pattern = generate_pattern_transfer(layer_name, filter_index)
plt.imshow(pattern)
plt.show()

```



### First and Middle Block

```

[ ]: # Now you can use generate_pattern function
layer_name = 'conv2_block1_1_conv'
filter_index = 0

```

```
pattern = generate_pattern(layer_name, filter_index)
plt.imshow(pattern)
plt.show()
```

WARNING:tensorflow:From /Users/banani/Library/Python/3.9/lib/python/site-packages/keras/src/layers/normalization/batch\_normalization.py:883: \_colocate\_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version.

Instructions for updating:

Colocations handled automatically by placer.

WARNING:tensorflow:From /Users/banani/Library/Python/3.9/lib/python/site-packages/keras/src/layers/normalization/batch\_normalization.py:883: \_colocate\_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version.

Instructions for updating:

Colocations handled automatically by placer.

2023-10-26 03:59:55.775561: I

tensorflow/compiler/mlir/mlir\_graph\_optimization\_pass.cc:382] MLIR V1 optimization pass is not enabled

2023-10-26 03:59:55.973099: W tensorflow/c/c\_api.cc:305] Operation

'{name:'conv5\_block1\_3\_bn/moving\_mean/Assign' id:3409 op device:{requested: '', assigned: ''} def:{{{node conv5\_block1\_3\_bn/moving\_mean/Assign}}} =

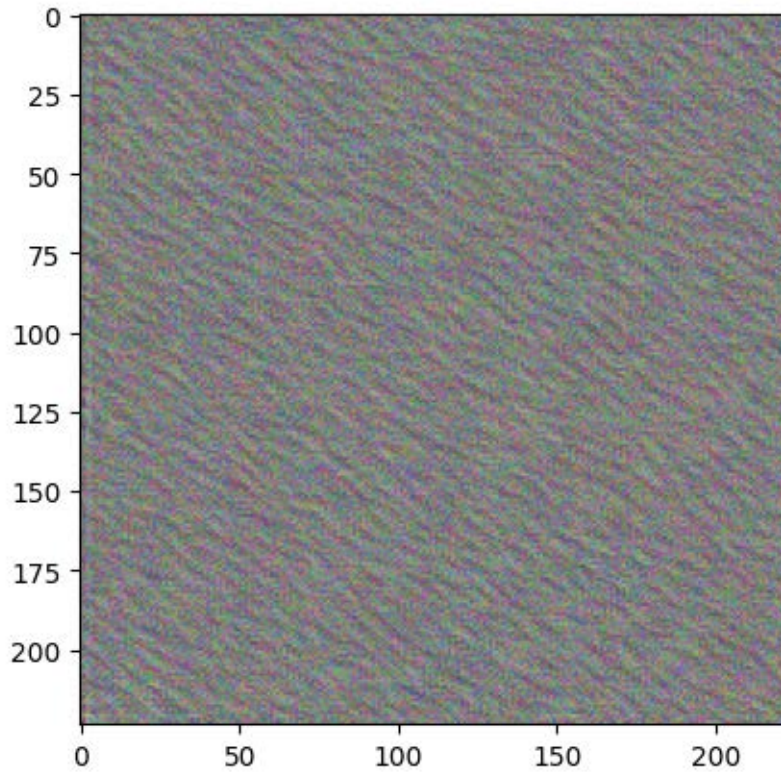
AssignVariableOp[\_has\_manual\_control\_dependencies=true, dtype=DT\_FLOAT,

validate\_shape=false](conv5\_block1\_3\_bn/moving\_mean,

conv5\_block1\_3\_bn/moving\_mean/Initializer/zeros)}}' was changed by setting

attribute after it was run by a session. This mutation will have no effect, and will trigger an error in the future. Either don't modify nodes after running them or create a new session.





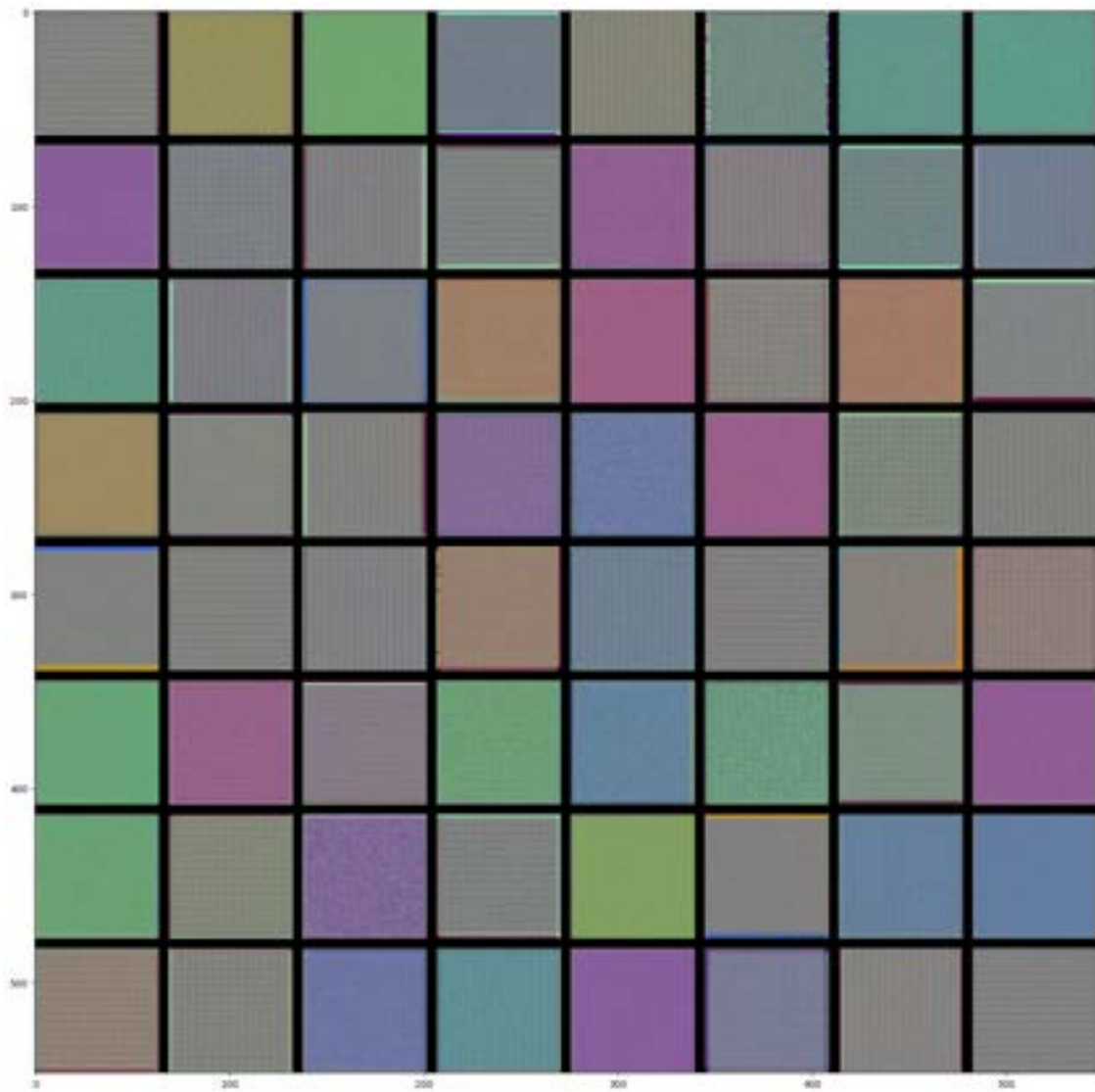
```
[ ]: for layer_name in ['conv1_conv', 'conv3_block3_2_conv']:
    size = 64
    margin = 5

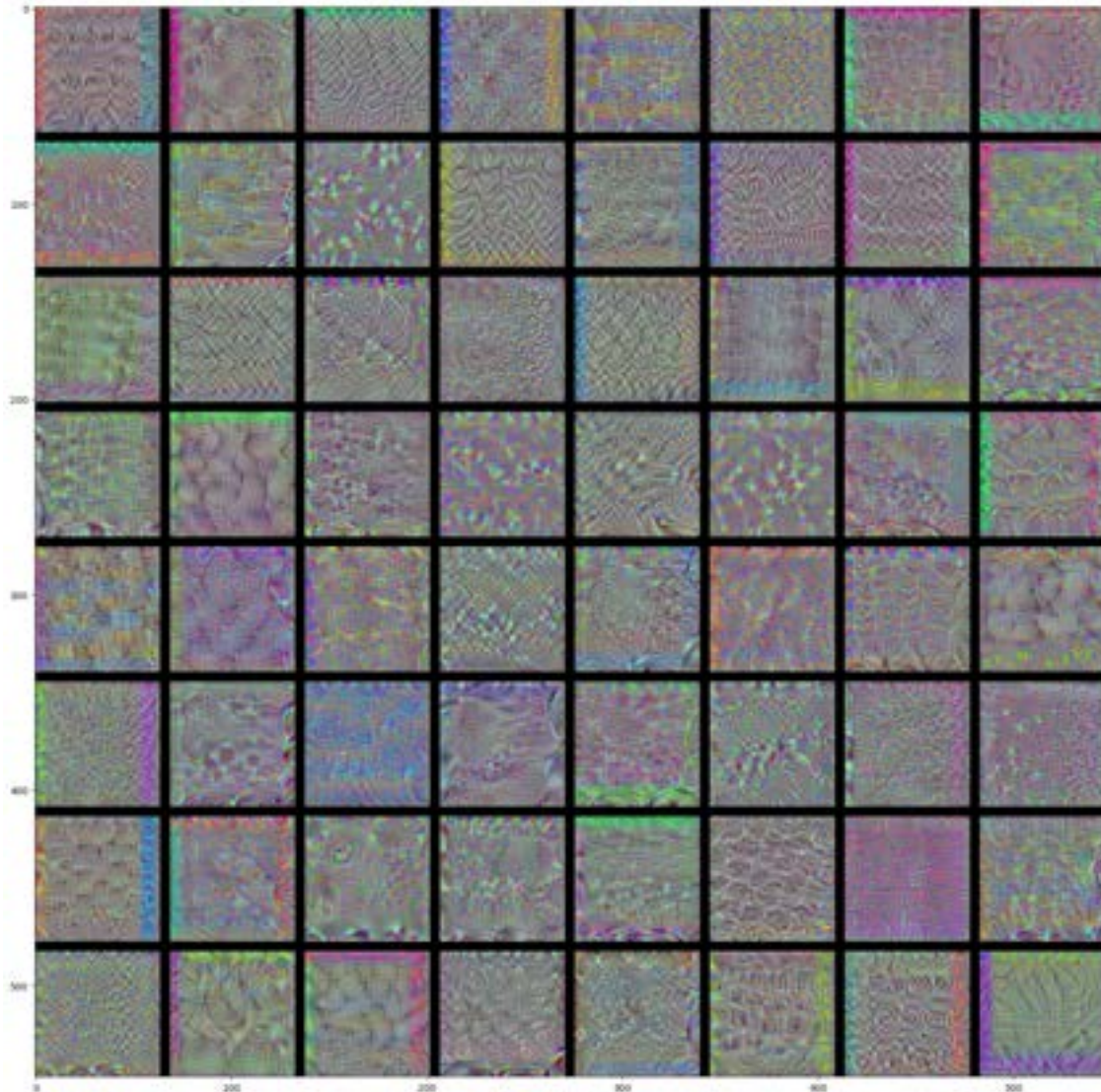
    results = np.zeros((8 * size + 7 * margin, 8 * size + 7 * margin, 3))

    for i in range(8):
        for j in range(8):
            filter_img = generate_pattern(layer_name, i + (j * 8), size=size)

            horizontal_start = i * size + i * margin
            horizontal_end = horizontal_start + size
            vertical_start = j * size + j * margin
            vertical_end = vertical_start + size
            results[horizontal_start: horizontal_end, vertical_start:
↪vertical_end, :] = filter_img

    plt.figure(figsize=(20, 20))
    plt.imshow((results * 255).astype(np.uint8))
    plt.show()
```





**Last Conv Block** To avoid kernel crash

```
[ ]: for layer_name in ['conv5_block3_3_conv']:
    size = 64
    margin = 5

    # This a empty (black) image where we will store our results.
    results = np.zeros((8 * size + 7 * margin, 8 * size + 7 * margin, 3))

    for i in range(8): # iterate over the rows of our results grid
        for j in range(8): # iterate over the columns of our results grid
            # Generate the pattern for filter `i + (j * 8)` in `layer_name`
            filter_img = generate_pattern(layer_name, i + (j * 8), size=size)
```

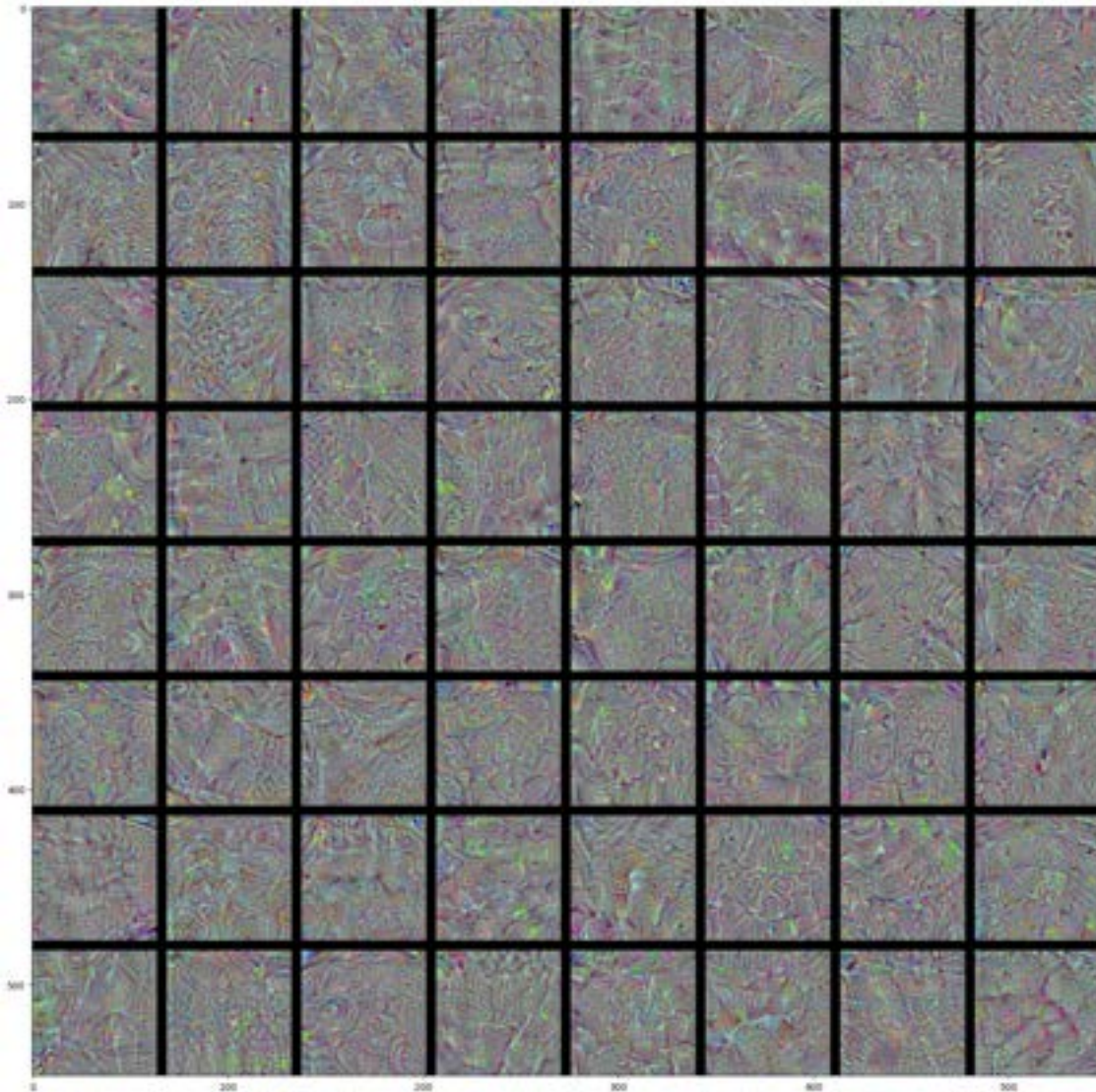


```

    # Put the result in the square `(i, j)` of the results grid
    horizontal_start = i * size + i * margin
    horizontal_end = horizontal_start + size
    vertical_start = j * size + j * margin
    vertical_end = vertical_start + size
    results[horizontal_start: horizontal_end, vertical_start:
↪vertical_end, :] = filter_img

# Display the results grid
plt.figure(figsize=(20, 20))
plt.imshow((results * 255).astype(np.uint8))
plt.show()

```





```
[ ]:
```

### 1.2.7 PS1.B Visualize Heatmaps

```
[ ]: # import tensorflow as tf
transfer_model = tf.keras.models.load_model('transfer_resnet50_2b_aug1.h5')
transfer_model.summary()
```

WARNING:absl:At this time, the v2.11+ optimizer `tf.keras.optimizers.SGD` runs slowly on M1/M2 Macs, please use the legacy Keras optimizer instead, located at `tf.keras.optimizers.legacy.SGD`.

Model: "model\_1"

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 224, 224, 3)]	0	[]
conv1_pad (ZeroPadding2D) ['input_1[0][0]']	(None, 230, 230, 3)	0	
conv1_conv (Conv2D) ['conv1_pad[0][0]']	(None, 112, 112, 64)	9472	
conv1_bn (BatchNormalizati on) ['conv1_conv[0][0]']	(None, 112, 112, 64)	256	
conv1_relu (Activation) ['conv1_bn[0][0]']	(None, 112, 112, 64)	0	
pool1_pad (ZeroPadding2D) ['conv1_relu[0][0]']	(None, 114, 114, 64)	0	
pool1_pool (MaxPooling2D) ['pool1_pad[0][0]']	(None, 56, 56, 64)	0	
conv2_block1_1_conv (Conv2 D) ['pool1_pool[0][0]']	(None, 56, 56, 64)	4160	
conv2_block1_1_bn (BatchNo rmalization) ['conv2_block1_1_conv[0][0]']	(None, 56, 56, 64)	256	
conv2_block1_1_relu (Activ ation) ['conv2_block1_1_bn[0][0]']	(None, 56, 56, 64)	0	

['conv2_block1_1_bn[0][0]'] ation)		
conv2_block1_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block1_1_relu[0][0]'] D)		36928
conv2_block1_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block1_2_conv[0][0]'] rmalization)		256
conv2_block1_2_relu (Activ (None, 56, 56, 64) ['conv2_block1_2_bn[0][0]'] ation)		0
conv2_block1_0_conv (Conv2 (None, 56, 56, 256) ['pool1_pool[0][0]'] D)		16640
conv2_block1_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block1_2_relu[0][0]'] D)		16640
conv2_block1_0_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_0_conv[0][0]'] rmalization)		1024
conv2_block1_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block1_3_conv[0][0]'] rmalization)		1024
conv2_block1_add (Add) (None, 56, 56, 256) ['conv2_block1_0_bn[0][0]', 'conv2_block1_3_bn[0][0]']		0
conv2_block1_out (Activati (None, 56, 56, 256) ['conv2_block1_add[0][0]'] on)		0
conv2_block2_1_conv (Conv2 (None, 56, 56, 64) ['conv2_block1_out[0][0]'] D)		16448
conv2_block2_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block2_1_conv[0][0]'] rmalization)		256
conv2_block2_1_relu (Activ (None, 56, 56, 64)		0

['conv2_block2_1_bn[0][0]'] ation)	
conv2_block2_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block2_1_relu[0][0]'] D)	36928
conv2_block2_2_bn (BatchNo (None, 56, 56, 64) ['conv2_block2_2_conv[0][0]'] rmalization)	256
conv2_block2_2_relu (Activ (None, 56, 56, 64) ['conv2_block2_2_bn[0][0]'] ation)	0
conv2_block2_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block2_2_relu[0][0]'] D)	16640
conv2_block2_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block2_3_conv[0][0]'] rmalization)	1024
conv2_block2_add (Add) (None, 56, 56, 256) ['conv2_block1_out[0][0]', 'conv2_block2_3_bn[0][0]']	0
conv2_block2_out (Activati (None, 56, 56, 256) ['conv2_block2_add[0][0]'] on)	0
conv2_block3_1_conv (Conv2 (None, 56, 56, 64) ['conv2_block2_out[0][0]'] D)	16448
conv2_block3_1_bn (BatchNo (None, 56, 56, 64) ['conv2_block3_1_conv[0][0]'] rmalization)	256
conv2_block3_1_relu (Activ (None, 56, 56, 64) ['conv2_block3_1_bn[0][0]'] ation)	0
conv2_block3_2_conv (Conv2 (None, 56, 56, 64) ['conv2_block3_1_relu[0][0]'] D)	36928
conv2_block3_2_bn (BatchNo (None, 56, 56, 64)	256

['conv2_block3_2_conv[0][0]'] rmalization)		
conv2_block3_2_relu (Activ (None, 56, 56, 64) ['conv2_block3_2_bn[0][0]'] ation)		0
conv2_block3_3_conv (Conv2 (None, 56, 56, 256) ['conv2_block3_2_relu[0][0]'] D)		16640
conv2_block3_3_bn (BatchNo (None, 56, 56, 256) ['conv2_block3_3_conv[0][0]'] rmalization)		1024
conv2_block3_add (Add) (None, 56, 56, 256) ['conv2_block2_out[0][0]', 'conv2_block3_3_bn[0][0]']		0
conv2_block3_out (Activati (None, 56, 56, 256) ['conv2_block3_add[0][0]'] on)		0
conv3_block1_1_conv (Conv2 (None, 28, 28, 128) ['conv2_block3_out[0][0]'] D)		32896
conv3_block1_1_bn (BatchNo (None, 28, 28, 128) ['conv3_block1_1_conv[0][0]'] rmalization)		512
conv3_block1_1_relu (Activ (None, 28, 28, 128) ['conv3_block1_1_bn[0][0]'] ation)		0
conv3_block1_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block1_1_relu[0][0]'] D)		147584
conv3_block1_2_bn (BatchNo (None, 28, 28, 128) ['conv3_block1_2_conv[0][0]'] rmalization)		512
conv3_block1_2_relu (Activ (None, 28, 28, 128) ['conv3_block1_2_bn[0][0]'] ation)		0
conv3_block1_0_conv (Conv2 (None, 28, 28, 512)		131584

```

['conv2_block3_out[0][0]']
D)

conv3_block1_3_conv (Conv2 (None, 28, 28, 512) 66048
['conv3_block1_2_relu[0][0]']
D)

conv3_block1_0_bn (BatchNo (None, 28, 28, 512) 2048
['conv3_block1_0_conv[0][0]']
rmalization)

conv3_block1_3_bn (BatchNo (None, 28, 28, 512) 2048
['conv3_block1_3_conv[0][0]']
rmalization)

conv3_block1_add (Add) (None, 28, 28, 512) 0
['conv3_block1_0_bn[0][0]',
'conv3_block1_3_bn[0][0]']

conv3_block1_out (Activati (None, 28, 28, 512) 0
['conv3_block1_add[0][0]']
on)

conv3_block2_1_conv (Conv2 (None, 28, 28, 128) 65664
['conv3_block1_out[0][0]']
D)

conv3_block2_1_bn (BatchNo (None, 28, 28, 128) 512
['conv3_block2_1_conv[0][0]']
rmalization)

conv3_block2_1_relu (Activ (None, 28, 28, 128) 0
['conv3_block2_1_bn[0][0]']
ation)

conv3_block2_2_conv (Conv2 (None, 28, 28, 128) 147584
['conv3_block2_1_relu[0][0]']
D)

conv3_block2_2_bn (BatchNo (None, 28, 28, 128) 512
['conv3_block2_2_conv[0][0]']
rmalization)

conv3_block2_2_relu (Activ (None, 28, 28, 128) 0
['conv3_block2_2_bn[0][0]']
ation)

conv3_block2_3_conv (Conv2 (None, 28, 28, 512) 66048

```

['conv3_block2_2_relu[0][0]'] D)		
conv3_block2_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block2_3_conv[0][0]'] rmalization)		2048
conv3_block2_add (Add) (None, 28, 28, 512) ['conv3_block1_out[0][0]', 'conv3_block2_3_bn[0][0]']		0
conv3_block2_out (Activati (None, 28, 28, 512) ['conv3_block2_add[0][0]'] on)		0
conv3_block3_1_conv (Conv2 (None, 28, 28, 128) ['conv3_block2_out[0][0]'] D)		65664
conv3_block3_1_bn (BatchNo (None, 28, 28, 128) ['conv3_block3_1_conv[0][0]'] rmalization)		512
conv3_block3_1_relu (Activ (None, 28, 28, 128) ['conv3_block3_1_bn[0][0]'] ation)		0
conv3_block3_2_conv (Conv2 (None, 28, 28, 128) ['conv3_block3_1_relu[0][0]'] D)		147584
conv3_block3_2_bn (BatchNo (None, 28, 28, 128) ['conv3_block3_2_conv[0][0]'] rmalization)		512
conv3_block3_2_relu (Activ (None, 28, 28, 128) ['conv3_block3_2_bn[0][0]'] ation)		0
conv3_block3_3_conv (Conv2 (None, 28, 28, 512) ['conv3_block3_2_relu[0][0]'] D)		66048
conv3_block3_3_bn (BatchNo (None, 28, 28, 512) ['conv3_block3_3_conv[0][0]'] rmalization)		2048
conv3_block3_add (Add) (None, 28, 28, 512)		0

['conv3_block2_out[0][0]', 'conv3_block3_3_bn[0][0]']		
conv3_block3_out (Activation) ['conv3_block3_add[0][0]']	(None, 28, 28, 512)	0
conv3_block4_1_conv (Conv2D) ['conv3_block3_out[0][0]']	(None, 28, 28, 128)	65664
conv3_block4_1_bn (Batch Normalization) ['conv3_block4_1_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_1_relu (Activation) ['conv3_block4_1_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_2_conv (Conv2D) ['conv3_block4_1_relu[0][0]']	(None, 28, 28, 128)	147584
conv3_block4_2_bn (Batch Normalization) ['conv3_block4_2_conv[0][0]']	(None, 28, 28, 128)	512
conv3_block4_2_relu (Activation) ['conv3_block4_2_bn[0][0]']	(None, 28, 28, 128)	0
conv3_block4_3_conv (Conv2D) ['conv3_block4_2_relu[0][0]']	(None, 28, 28, 512)	66048
conv3_block4_3_bn (Batch Normalization) ['conv3_block4_3_conv[0][0]']	(None, 28, 28, 512)	2048
conv3_block4_add (Add) ['conv3_block3_out[0][0]', 'conv3_block4_3_bn[0][0]']	(None, 28, 28, 512)	0
conv3_block4_out (Activation) ['conv3_block4_add[0][0]']	(None, 28, 28, 512)	0
conv4_block1_1_conv (Conv2D)	(None, 14, 14, 256)	131328

```

['conv3_block4_out[0][0]']
D)

conv4_block1_1_bn (BatchNo (None, 14, 14, 256) 1024
['conv4_block1_1_conv[0][0]']
rmalization)

conv4_block1_1_relu (Activ (None, 14, 14, 256) 0
['conv4_block1_1_bn[0][0]']
ation)

conv4_block1_2_conv (Conv2 (None, 14, 14, 256) 590080
['conv4_block1_1_relu[0][0]']
D)

conv4_block1_2_bn (BatchNo (None, 14, 14, 256) 1024
['conv4_block1_2_conv[0][0]']
rmalization)

conv4_block1_2_relu (Activ (None, 14, 14, 256) 0
['conv4_block1_2_bn[0][0]']
ation)

conv4_block1_0_conv (Conv2 (None, 14, 14, 1024) 525312
['conv3_block4_out[0][0]']
D)

conv4_block1_3_conv (Conv2 (None, 14, 14, 1024) 263168
['conv4_block1_2_relu[0][0]']
D)

conv4_block1_0_bn (BatchNo (None, 14, 14, 1024) 4096
['conv4_block1_0_conv[0][0]']
rmalization)

conv4_block1_3_bn (BatchNo (None, 14, 14, 1024) 4096
['conv4_block1_3_conv[0][0]']
rmalization)

conv4_block1_add (Add) (None, 14, 14, 1024) 0
['conv4_block1_0_bn[0][0]',
'conv4_block1_3_bn[0][0]']

conv4_block1_out (Activati (None, 14, 14, 1024) 0
['conv4_block1_add[0][0]']
on)

conv4_block2_1_conv (Conv2 (None, 14, 14, 256) 262400

```



['conv4_block1_out[0][0]'] D)	
conv4_block2_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block2_1_conv[0][0]'] rmalization)	1024
conv4_block2_1_relu (Activ (None, 14, 14, 256) ['conv4_block2_1_bn[0][0]'] ation)	0
conv4_block2_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block2_1_relu[0][0]'] D)	590080
conv4_block2_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block2_2_conv[0][0]'] rmalization)	1024
conv4_block2_2_relu (Activ (None, 14, 14, 256) ['conv4_block2_2_bn[0][0]'] ation)	0
conv4_block2_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block2_2_relu[0][0]'] D)	263168
conv4_block2_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block2_3_conv[0][0]'] rmalization)	4096
conv4_block2_add (Add) (None, 14, 14, 1024) ['conv4_block1_out[0][0]', 'conv4_block2_3_bn[0][0]']	0
conv4_block2_out (Activati (None, 14, 14, 1024) ['conv4_block2_add[0][0]'] on)	0
conv4_block3_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block2_out[0][0]'] D)	262400
conv4_block3_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block3_1_conv[0][0]'] rmalization)	1024
conv4_block3_1_relu (Activ (None, 14, 14, 256)	0

['conv4_block3_1_bn[0][0]'] ation)	
conv4_block3_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_1_relu[0][0]'] D)	590080
conv4_block3_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block3_2_conv[0][0]'] rmalization)	1024
conv4_block3_2_relu (Activ (None, 14, 14, 256) ['conv4_block3_2_bn[0][0]'] ation)	0
conv4_block3_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block3_2_relu[0][0]'] D)	263168
conv4_block3_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block3_3_conv[0][0]'] rmalization)	4096
conv4_block3_add (Add) (None, 14, 14, 1024) ['conv4_block2_out[0][0]', 'conv4_block3_3_bn[0][0]']	0
conv4_block3_out (Activati (None, 14, 14, 1024) ['conv4_block3_add[0][0]'] on)	0
conv4_block4_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block3_out[0][0]'] D)	262400
conv4_block4_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block4_1_conv[0][0]'] rmalization)	1024
conv4_block4_1_relu (Activ (None, 14, 14, 256) ['conv4_block4_1_bn[0][0]'] ation)	0
conv4_block4_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block4_1_relu[0][0]'] D)	590080
conv4_block4_2_bn (BatchNo (None, 14, 14, 256)	1024

['conv4_block4_2_conv[0][0]'] rmalization)		
conv4_block4_2_relu (Activ (None, 14, 14, 256) ['conv4_block4_2_bn[0][0]'] ation)		0
conv4_block4_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block4_2_relu[0][0]'] D)		263168
conv4_block4_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block4_3_conv[0][0]'] rmalization)		4096
conv4_block4_add (Add) (None, 14, 14, 1024) ['conv4_block3_out[0][0]', 'conv4_block4_3_bn[0][0]']		0
conv4_block4_out (Activati (None, 14, 14, 1024) ['conv4_block4_add[0][0]'] on)		0
conv4_block5_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block4_out[0][0]'] D)		262400
conv4_block5_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block5_1_conv[0][0]'] rmalization)		1024
conv4_block5_1_relu (Activ (None, 14, 14, 256) ['conv4_block5_1_bn[0][0]'] ation)		0
conv4_block5_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block5_1_relu[0][0]'] D)		590080
conv4_block5_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block5_2_conv[0][0]'] rmalization)		1024
conv4_block5_2_relu (Activ (None, 14, 14, 256) ['conv4_block5_2_bn[0][0]'] ation)		0
conv4_block5_3_conv (Conv2 (None, 14, 14, 1024)		263168

['conv4_block5_2_relu[0][0]'] D)		
conv4_block5_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block5_3_conv[0][0]'] rmalization)		4096
conv4_block5_add (Add) (None, 14, 14, 1024) ['conv4_block4_out[0][0]', 'conv4_block5_3_bn[0][0]']		0
conv4_block5_out (Activati (None, 14, 14, 1024) ['conv4_block5_add[0][0]'] on)		0
conv4_block6_1_conv (Conv2 (None, 14, 14, 256) ['conv4_block5_out[0][0]'] D)		262400
conv4_block6_1_bn (BatchNo (None, 14, 14, 256) ['conv4_block6_1_conv[0][0]'] rmalization)		1024
conv4_block6_1_relu (Activ (None, 14, 14, 256) ['conv4_block6_1_bn[0][0]'] ation)		0
conv4_block6_2_conv (Conv2 (None, 14, 14, 256) ['conv4_block6_1_relu[0][0]'] D)		590080
conv4_block6_2_bn (BatchNo (None, 14, 14, 256) ['conv4_block6_2_conv[0][0]'] rmalization)		1024
conv4_block6_2_relu (Activ (None, 14, 14, 256) ['conv4_block6_2_bn[0][0]'] ation)		0
conv4_block6_3_conv (Conv2 (None, 14, 14, 1024) ['conv4_block6_2_relu[0][0]'] D)		263168
conv4_block6_3_bn (BatchNo (None, 14, 14, 1024) ['conv4_block6_3_conv[0][0]'] rmalization)		4096
conv4_block6_add (Add) (None, 14, 14, 1024)		0

```

['conv4_block5_out[0][0]',
'conv4_block6_3_bn[0][0]']

conv4_block6_out (Activation) (None, 14, 14, 1024) 0
['conv4_block6_add[0][0]']

conv5_block1_1_conv (Conv2D) (None, 7, 7, 512) 524800
['conv4_block6_out[0][0]']

conv5_block1_1_bn (Batch Normalization) (None, 7, 7, 512) 2048
['conv5_block1_1_conv[0][0]']

conv5_block1_1_relu (Activation) (None, 7, 7, 512) 0
['conv5_block1_1_bn[0][0]']

conv5_block1_2_conv (Conv2D) (None, 7, 7, 512) 2359808
['conv5_block1_1_relu[0][0]']

conv5_block1_2_bn (Batch Normalization) (None, 7, 7, 512) 2048
['conv5_block1_2_conv[0][0]']

conv5_block1_2_relu (Activation) (None, 7, 7, 512) 0
['conv5_block1_2_bn[0][0]']

conv5_block1_0_conv (Conv2D) (None, 7, 7, 2048) 2099200
['conv4_block6_out[0][0]']

conv5_block1_3_conv (Conv2D) (None, 7, 7, 2048) 1050624
['conv5_block1_2_relu[0][0]']

conv5_block1_0_bn (Batch Normalization) (None, 7, 7, 2048) 8192
['conv5_block1_0_conv[0][0]']

conv5_block1_3_bn (Batch Normalization) (None, 7, 7, 2048) 8192
['conv5_block1_3_conv[0][0]']

conv5_block1_add (Add) (None, 7, 7, 2048) 0

```

['conv5_block1_0_bn[0][0]', 'conv5_block1_3_bn[0][0]']		
conv5_block1_out (Activation) ['conv5_block1_add[0][0]']	(None, 7, 7, 2048)	0
conv5_block2_1_conv (Conv2D) ['conv5_block1_out[0][0]']	(None, 7, 7, 512)	1049088
conv5_block2_1_bn (Batch Normalization) ['conv5_block2_1_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block2_1_relu (Activation) ['conv5_block2_1_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block2_2_conv (Conv2D) ['conv5_block2_1_relu[0][0]']	(None, 7, 7, 512)	2359808
conv5_block2_2_bn (Batch Normalization) ['conv5_block2_2_conv[0][0]']	(None, 7, 7, 512)	2048
conv5_block2_2_relu (Activation) ['conv5_block2_2_bn[0][0]']	(None, 7, 7, 512)	0
conv5_block2_3_conv (Conv2D) ['conv5_block2_2_relu[0][0]']	(None, 7, 7, 2048)	1050624
conv5_block2_3_bn (Batch Normalization) ['conv5_block2_3_conv[0][0]']	(None, 7, 7, 2048)	8192
conv5_block2_add (Add) ['conv5_block1_out[0][0]', 'conv5_block2_3_bn[0][0]']	(None, 7, 7, 2048)	0
conv5_block2_out (Activation) ['conv5_block2_add[0][0]']	(None, 7, 7, 2048)	0
conv5_block3_1_conv (Conv2D)	(None, 7, 7, 512)	1049088

['conv5_block2_out[0][0]'] D)		
conv5_block3_1_bn (BatchNo (None, 7, 7, 512) ['conv5_block3_1_conv[0][0]'] rmalization)		2048
conv5_block3_1_relu (Activ (None, 7, 7, 512) ['conv5_block3_1_bn[0][0]'] ation)		0
conv5_block3_2_conv (Conv2 (None, 7, 7, 512) ['conv5_block3_1_relu[0][0]'] D)		2359808
conv5_block3_2_bn (BatchNo (None, 7, 7, 512) ['conv5_block3_2_conv[0][0]'] rmalization)		2048
conv5_block3_2_relu (Activ (None, 7, 7, 512) ['conv5_block3_2_bn[0][0]'] ation)		0
conv5_block3_3_conv (Conv2 (None, 7, 7, 2048) ['conv5_block3_2_relu[0][0]'] D)		1050624
conv5_block3_3_bn (BatchNo (None, 7, 7, 2048) ['conv5_block3_3_conv[0][0]'] rmalization)		8192
conv5_block3_add (Add) (None, 7, 7, 2048) ['conv5_block2_out[0][0]', 'conv5_block3_3_bn[0][0]']		0
conv5_block3_out (Activati (None, 7, 7, 2048) ['conv5_block3_add[0][0]'] on)		0
average_pooling2d (Average (None, 3, 3, 2048) ['conv5_block3_out[0][0]'] Pooling2D)		0
flatten (Flatten) (None, 18432) ['average_pooling2d[0][0]']		0
fc9_predictions (Dense) (None, 1) ['flatten[0][0]']		18433

```

=====
Total params: 23606145 (90.05 MB)
Trainable params: 23553025 (89.85 MB)
Non-trainable params: 53120 (207.50 KB)
-----

```

```

[ ]: import numpy as np

# The local path to our target image
#img_path = 'creative_commons_elephant.jpg'
# img_path = 'refs/dog_1.jpg'
img_path = 'refs/ship_1.jpg'

im_size = 224

# `img` is a PIL image of size 224x224
img = tf.keras.utils.load_img(img_path, target_size=(im_size, im_size))

# `x` is a float32 Numpy array of shape (224, 224, 3)
img_tensor = tf.keras.utils.img_to_array(img)

# We add a dimension to transform our array into a "batch"
# of size (1, 224, 224, 3)
img_tensor = np.expand_dims(img_tensor, axis=0)

img_tensor /= 255.

# # Its shape is (1, 224, 224, 3)
print(img_tensor.shape)

```

```
(1, 224, 224, 3)
```

```

[ ]: preds = transfer_model.predict(img_tensor)
def custom_decode_predictions(predictions, top=5, class_names=None):
    if class_names is None:
        class_names = [str(i) for i in range(predictions.shape[-1])]

    results = []
    for pred in predictions:
        top_indices = pred.argsort()[-top:][::-1]
        decoded_preds = [(class_names[i], pred[i]) for i in top_indices]
        results.append(decoded_preds)

    return results

```



```
# classes = ['ship']
classes = ['airplane', 'automobile', 'bird', 'cat', 'dog', 'deer', 'frog',
↳ 'horse', 'ship', 'truck']

print('Predicted:', custom_decode_predictions(preds, top=5,
↳ class_names=classes))
```

Predicted: [['airplane', 0.9999543]]

### Chosen Layers

```
[ ]: img_path
layers_to_visualize = [transfer_model.get_layer('conv1_relu'),
                        transfer_model.get_layer('conv3_block2_2_relu'),
                        transfer_model.get_layer('conv5_block3_2_relu')]

for layer in layers_to_visualize:
    print(layer.name)
```

```
conv1_relu
conv3_block2_2_relu
conv5_block3_2_relu
```

```
[ ]: import cv2

def overlay_heatmap(heatmap, img_path, alpha=0.4):
    img = cv2.imread(img_path)

    hmap = heatmap

    # Resize the heatmap to the size of the original image
    heatmap = cv2.resize(heatmap, (img.shape[1], img.shape[0]))

    # Normalize the heatmap
    heatmap = (heatmap - np.min(heatmap)) / (np.max(heatmap) - np.min(heatmap))

    # Apply a colormap to the heatmap
    heatmap = cv2.applyColorMap(np.uint8(255 * heatmap), cv2.COLORMAP_JET)

    # Superimpose the heatmap on the original image
    superimposed_img = heatmap * alpha + img

    fig, axes = plt.subplots(1, 3, figsize=(12, 8))

    # Display the heatmap in the second subplot
```

```

axes[0].imshow(hmap)
axes[0].set_title('Heatmap')
axes[0].axis('off')

# Display the heatmap in the second subplot
axes[1].imshow(heatmap, cmap='jet')
axes[1].set_title('Heatmap Normalized')
axes[1].axis('off')

# Display the superimposed image in the first subplot
axes[2].imshow(cv2.cvtColor(np.uint8(superimposed_img), cv2.COLOR_BGR2RGB))
axes[2].set_title('Superimposed Image')
axes[2].axis('off')

plt.show()

# plt.imshow(cv2.cvtColor(np.uint8(superimposed_img), cv2.COLOR_BGR2RGB))
# plt.axis('off')
# plt.show()

```

## First Layer Heatmap

```

[ ]: # This is the "ship" entry in the prediction vector
ship_output = transfer_model.output

# This is the output feature map of the `conv3_block2_2_relu` layer,
# the first convolutional layer in ResNet50
first_conv_layer = transfer_model.get_layer('conv1_relu')

# This is the gradient of the "dog" class with regard to
# the output feature map of `conv5_block3_3_conv`
grads = K.gradients(ship_output, first_conv_layer.output)[0]

# This is a vector of shape (512,), where each entry
# is the mean intensity of the gradient over a specific feature map channel
pooled_grads = K.mean(grads, axis=(0, 1, 2))

# This function allows us to access the values of the quantities we just
↳ defined:
# `pooled_grads` and the output feature map of `block5_conv3`,
# given a sample image
iterate = K.function([transfer_model.input], [pooled_grads, first_conv_layer.
↳ output[0]])

```

```

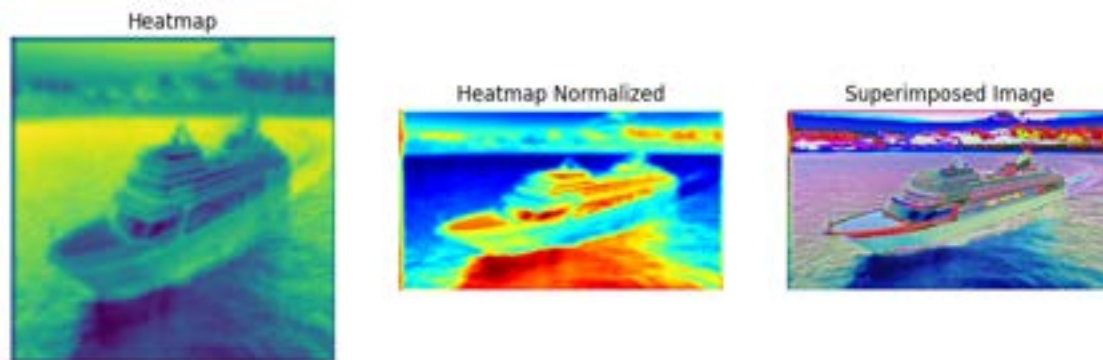
# These are the values of these two quantities, as Numpy arrays,
# given our sample image of two elephants
pooled_grads_value, conv_layer_output_value = iterate([img_tensor])

# We multiply each channel in the feature map array
# by "how important this channel is" with regard to the elephant class
for i in range(64):
    conv_layer_output_value[:, :, i] *= pooled_grads_value[i]

# The channel-wise mean of the resulting feature map
# is our heatmap of class activation
heatmap_first = np.mean(conv_layer_output_value, axis=-1)

overlay_heatmap(heatmap_first, img_path)

```



## Middle Layer Heatmap

```

[ ]: # This is the "ship" entry in the prediction vector
ship_output = transfer_model.output

# This is the output feature map of the `conv3_block2_2_relu` layer,
# the first convolutional layer in ResNet50
middle_conv_layer = transfer_model.get_layer('conv3_block2_2_relu')

# This is the gradient of the "dog" class with regard to
# the output feature map of `conv5_block3_3_conv`
grads = K.gradients(ship_output, middle_conv_layer.output)[0]

# This is a vector of shape (512,), where each entry
# is the mean intensity of the gradient over a specific feature map channel
pooled_grads = K.mean(grads, axis=(0, 1, 2))

# This function allows us to access the values of the quantities we just
↳ defined:

```

```

# `pooled_grads` and the output feature map of `block5_conv3`,
# given a sample image
iterate = K.function([transfer_model.input], [pooled_grads, middle_conv_layer.
    ↳output[0]])

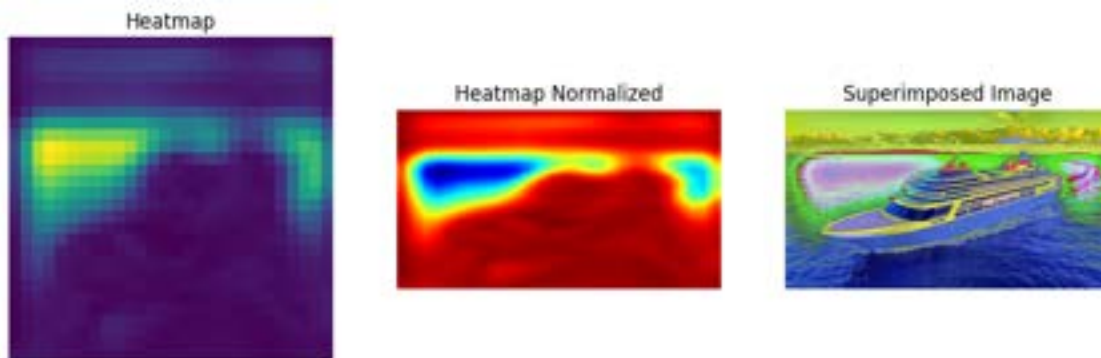
# These are the values of these two quantities, as Numpy arrays,
# given our sample image of two elephants
pooled_grads_value, conv_layer_output_value = iterate([img_tensor])

# We multiply each channel in the feature map array
# by "how important this channel is" with regard to the elephant class
for i in range(64):
    conv_layer_output_value[:, :, i] *= pooled_grads_value[i]

# The channel-wise mean of the resulting feature map
# is our heatmap of class activation
heatmap_middle = np.mean(conv_layer_output_value, axis=-1)

overlay_heatmap(heatmap_middle, img_path)

```



### Last Layer Heatmap

```

[ ]: # This is the "ship" entry in the prediction vector
ship_output = transfer_model.output

# This is the output feature map of the `conv3_block2_2_relu` layer,
# the first convolutional layer in ResNet50
last_conv_layer = transfer_model.get_layer('conv5_block3_2_relu')

# This is the gradient of the "dog" class with regard to
# the output feature map of `conv5_block3_3_conv`
grads = K.gradients(ship_output, last_conv_layer.output)[0]

# This is a vector of shape (512,), where each entry

```

```

# is the mean intensity of the gradient over a specific feature map channel
pooled_grads = K.mean(grads, axis=(0, 1, 2))

# This function allows us to access the values of the quantities we just
↳ defined:
# `pooled_grads` and the output feature map of `block5_conv3`,
# given a sample image
iterate = K.function([transfer_model.input], [pooled_grads, last_conv_layer.
↳ output[0]])

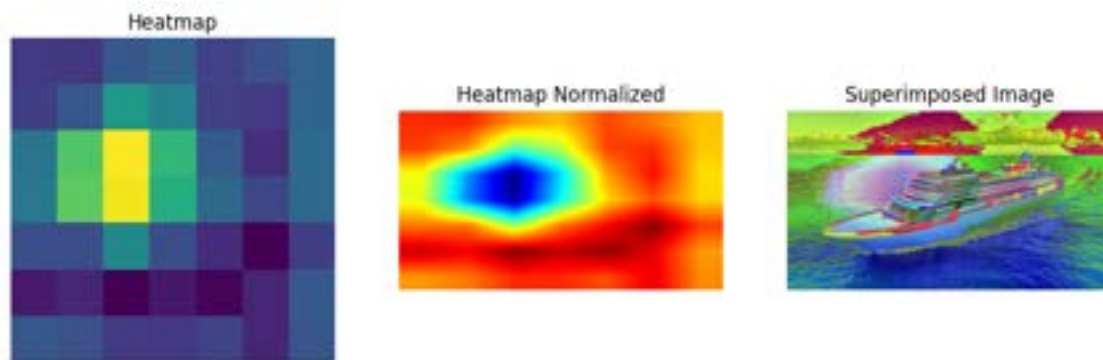
# These are the values of these two quantities, as Numpy arrays,
# given our sample image of two elephants
pooled_grads_value, conv_layer_output_value = iterate([img_tensor])

# We multiply each channel in the feature map array
# by "how important this channel is" with regard to the elephant class
for i in range(64):
    conv_layer_output_value[:, :, i] *= pooled_grads_value[i]

# The channel-wise mean of the resulting feature map
# is our heatmap of class activation
heatmap_last = np.mean(conv_layer_output_value, axis=-1)

overlay_heatmap(heatmap_last, img_path)

```



### 1.3 END