1.go to https://github.com/miaow1988/SqueezeNet_v1.2 and download the 'symbol.json' and '.params' files (there is not a 'synset.txt' file! so don't use these lines, Hint: just comment these lines).

Install MXNet v1.5 (hint: create a new conda environmet with python 3, pip install mxnet==1.5.1) and follow the same steps of the lecture (part: Using pre-trained models as feature extractors). Find the flatten output layer and create a feature extractor (hint: It should be a numpy array of 1000 elements).

Download the dogs versus cats training folder from https://www.kaggle.com/c/dogs-vs-cats-redux-kernels-edition/data (Remember the number of images is 12500 for each class).

Extract the array of features for different number of images (N: 10, 100, 500, 1000, also 5000 and 12500) and for each value train your favorite binary classifier (only one!!!) using GridSearch to optimize some hyperparameters. Consider to use https://notebooks.csc.fi if you have computational limitations.

Report the accuracy for each value of N and the computational time during the training step.

```
In [8]:
         !pip install mxnet==1.5.1
        Collecting mxnet==1.5.1
          Downloading mxnet-1.5.1-py2.py3-none-manylinux1 x86 64.whl (23.1 MB)
                                              23.1 MB 42.9 MB/s eta 0:00:01
                                   8.7 MB 1.9 MB/s eta 0:00:081.9 MB/s eta 0:00:07/s
        eta 0:00:06
                                                    13.7 MB 1.9 MB/s eta 0:00:060:04.
        5 MB 1.9 MB/s eta 0:00:03
                                                 20.4 MB 1.9 MB/s eta 0:00:02
        2.8 MB 1.9 MB/s eta 0:00:01
        Requirement already satisfied: numpy<2.0.0,>1.16.0 in /opt/conda/lib/python3.7
        /site-packages (from mxnet==1.5.1) (1.18.2)
        Requirement already satisfied: requests<3,>=2.20.0 in /opt/conda/lib/python3.7
        /site-packages (from mxnet==1.5.1) (2.23.0)
        Collecting graphviz<0.9.0,>=0.8.1
          Downloading graphviz-0.8.4-py2.py3-none-any.whl (16 kB)
        Requirement already satisfied: idna<3,>=2.5 in /opt/conda/lib/python3.7/site-p
        ackages (from requests<3,>=2.20.0->mxnet==1.5.1) (2.9)
        Requirement already satisfied: chardet<4,>=3.0.2 in /opt/conda/lib/python3.7/s
        ite-packages (from requests<3,>=2.20.0->mxnet==1.5.1) (3.0.4)
        Requirement already satisfied: certifi>=2017.4.17 in /opt/conda/lib/python3.7/
        site-packages (from requests<3,>=2.20.0->mxnet==1.5.1) (2019.11.28)
        Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /opt
        /conda/lib/python3.7/site-packages (from requests<3,>=2.20.0->mxnet==1.5.1) (1
        .25.7)
        Installing collected packages: graphviz, mxnet
          Attempting uninstall: graphviz
            Found existing installation: graphviz 0.13.2
            Uninstalling graphviz-0.13.2:
              Successfully uninstalled graphviz-0.13.2
        Successfully installed graphviz-0.8.4 mxnet-1.5.1
```

In [10]:

```
--2020-11-07 10:23:47-- https://github.com/miaow1988/SqueezeNet v1.2/raw/mast
er/model-symbol.json
Resolving github.com (github.com)... 140.82.121.3
Connecting to github.com (github.com) | 140.82.121.3 | :443... connected.
HTTP request sent, awaiting response... 302 Found
Location: https://raw.githubusercontent.com/miaow1988/SqueezeNet v1.2/master/m
odel-symbol.json [following]
--2020-11-07 10:23:47-- https://raw.githubusercontent.com/miaow1988/SqueezeNe
t v1.2/master/model-symbol.json
Resolving raw.githubusercontent.com (raw.githubusercontent.com)... 151.101.84.
Connecting to raw.githubusercontent.com (raw.githubusercontent.com) | 151.101.84
.133 :443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 24187 (24K) [text/plain]
Saving to: 'model-symbol.json'
model-symbol.json
                 2020-11-07 10:23:47 (1.57 MB/s) - 'model-symbol.json' saved [24187/24187]
--2020-11-07 10:23:48-- https://github.com/miaow1988/SqueezeNet_v1.2/raw/mast
er/model-0000.params
Resolving github.com (github.com)... 140.82.121.3
Connecting to github.com (github.com) | 140.82.121.3 | :443... connected.
HTTP request sent, awaiting response... 302 Found
Location: https://raw.githubusercontent.com/miaow1988/SqueezeNet v1.2/master/m
odel-0000.params [following]
--2020-11-07 10:23:49-- https://raw.githubusercontent.com/miaow1988/SqueezeNe
t v1.2/master/model-0000.params
Resolving raw.githubusercontent.com (raw.githubusercontent.com)... 151.101.84.
Connecting to raw.githubusercontent.com (raw.githubusercontent.com) 151.101.84
.133 :443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 4945034 (4.7M) [application/octet-stream]
Saving to: 'model-0000.params'
model-0000.params
                  4.72M 12.8MB/s
                                                                 in 0.4s
2020-11-07 10:23:50 (12.8 MB/s) - 'model-0000.params' saved [4945034/4945034]
! pip install opency-python-headless
! pip install --user kaggle --upgrade
Collecting opency-python-headless
 Downloading opency python headless-4.4.0.46-cp37-cp37m-manylinux2014 x86 64.
whl (36.7 MB)
                                    | 36.7 MB 42.8 MB/s eta 0:00:01MB 2.1 M
B/s eta 0:00:13
                                 13.6 MB 2.1 MB/s eta 0:00:12
17.5 MB 2.1 MB/s eta 0:00:10
                                                            18.9 MB 2.1 MB
```

```
21.1 MB 2.1 MB/s eta 0:00:08
/s eta 0:00:09
                                                  28.9 MB 42.8 MB/s eta 0:
23.7 MB 42.8 MB/s eta 0:00:01101
00:01 | 30.8 MB 42.8 MB/s eta 0:00:01
                                                                       33.1
MB 42.8 MB/s eta 0:00:01
Requirement already satisfied: numpy>=1.14.5 in /opt/conda/lib/python3.7/site-
packages (from opency-python-headless) (1.18.2)
Installing collected packages: opency-python-headless
Successfully installed opency-python-headless-4.4.0.46
Collecting kaggle
  Downloading kaggle-1.5.9.tar.gz (58 kB)
                 58 kB 1.5 MB/s eta 0:00:011
Requirement already satisfied, skipping upgrade: six>=1.10 in /opt/conda/lib/p
ython3.7/site-packages (from kaggle) (1.14.0)
Requirement already satisfied, skipping upgrade: certifi in /opt/conda/lib/pyt
hon3.7/site-packages (from kaggle) (2019.11.28)
Requirement already satisfied, skipping upgrade: python-dateutil in /opt/conda
/lib/python3.7/site-packages (from kaggle) (2.8.1)
Requirement already satisfied, skipping upgrade: requests in /opt/conda/lib/py
thon3.7/site-packages (from kaggle) (2.23.0)
Requirement already satisfied, skipping upgrade: tqdm in /opt/conda/lib/python
3.7/site-packages (from kaggle) (4.43.0)
Collecting python-slugify
  Downloading python-slugify-4.0.1.tar.gz (11 kB)
Collecting slugify
  Downloading slugify-0.0.1.tar.gz (1.2 kB)
Requirement already satisfied, skipping upgrade: urllib3 in /opt/conda/lib/pyt
hon3.7/site-packages (from kaggle) (1.25.7)
Requirement already satisfied, skipping upgrade: idna<3,>=2.5 in /opt/conda/li
b/python3.7/site-packages (from requests->kaggle) (2.9)
Requirement already satisfied, skipping upgrade: chardet<4,>=3.0.2 in /opt/con
da/lib/python3.7/site-packages (from requests->kaggle) (3.0.4)
Collecting text-unidecode>=1.3
  Downloading text unidecode-1.3-py2.py3-none-any.whl (78 kB)
                        78 kB 3.4 MB/s eta 0:00:011
Building wheels for collected packages: kaggle, python-slugify, slugify
  Building wheel for kaggle (setup.py) ... done
  Created wheel for kaggle: filename=kaggle-1.5.9-py3-none-any.whl size=73265
sha256=d2a32e42c1c9ad13d7feeaae1eac66f910874e3ef556195531f1285a0431c762
  Stored in directory: /home/jovyan/.cache/pip/wheels/09/25/76/1bbe8ad0c423e86
55942b6d3c781f58e0ea2791bf8ee8985b3
  Building wheel for python-slugify (setup.py) ... done
  Created wheel for python-slugify: filename=python slugify-4.0.1-py2.py3-none
-any.whl size=6767 sha256=28a60d02dcab1924182c050daff80e8dba8e396d89a7d5361aec
9dd8f76d7fa5
  Stored in directory: /home/jovyan/.cache/pip/wheels/48/1b/6f/5c1cfab22eacbe0
095fc619786da6571b55253653c71324b5c
  Building wheel for slugify (setup.py) ... done
  Created wheel for slugify: filename=slugify-0.0.1-py3-none-any.whl size=1908
\verb|sha| 256 = 33bd827d718d4f5f03d9bf292f89e4e5a74fbaba47dc0df4fa6efb32f1503135| \\
  Stored in directory: /home/jovyan/.cache/pip/wheels/d4/7b/0d/bd65011a1b44284
3bb4043e396f727ab0f1e76050355b9156a
Successfully built kaggle python-slugify slugify
Installing collected packages: text-unidecode, python-slugify, slugify, kaggle
```

```
WARNING: The script slugify is installed in '/home/jovyan/.local/bin' which is not on PATH.
```

Consider adding this directory to PATH or, if you prefer to suppress this wa rning, use --no-warn-script-location.

WARNING: The script kaggle is installed in '/home/jovyan/.local/bin' which is not on PATH.

Consider adding this directory to PATH or, if you prefer to suppress this warning, use --no-warn-script-location.

Successfully installed kaggle-1.5.9 python-slugify-4.0.1 slugify-0.0.1 text-un idecode-1.3

```
In [15]:
# if "opencv-python" not in pkgs:
# !pip install opencv-python
# if "opencv-contrib-python" not in pkgs:
# !pip install opencv-contrib-python
```

Upload your https://github.com/Kaggle/kaggle-api#api-credentials

```
In [11]:
          mkdir -/.kaggle
         mkdir: cannot create directory '/home/jovyan/.kaggle': File exists
In [12]:
          ! mv kaggle.json .kaggle/kaggle.json
In [13]:
          ! chmod 600 ~/.kaggle/kaggle.json
In [17]:
          ! ls -l ~/.kaggle
         total 4
         -rw----. 1 jovyan root 63 Nov 7 10:19 kaggle.json
In [20]:
          ! ls -l ~/.local/bin
         total 8
         -rwxr-xr-x. 1 jovyan root 217 Nov 7 10:24 kaggle
         -rwxr-xr-x. 1 jovyan root 333 Nov 7 10:24 slugify
In [19]:
          !pip show kaggle
```

```
Name: kaggle
         Version: 1.5.9
         Summary: Kaggle API
         Home-page: https://github.com/Kaggle/kaggle-api
         Author: Kaggle
         Author-email: support@kaggle.com
         License: Apache 2.0
         Location: /home/jovyan/.local/lib/python3.7/site-packages
         Requires: tqdm, certifi, requests, urllib3, six, slugify, python-slugify, pyth
         on-dateutil
         Required-by:
In [21]:
          ! find / -name kaggle 2>/dev/null
         /home/jovyan/.local/lib/python3.7/site-packages/kaggle
         /home/jovyan/.local/bin/kaggle
In [24]:
          #!kaggle competitions download -c dogs-vs-cats-redux-kernels-edition
 In [1]:
          ! unzip dogs-vs-cats-redux-kernels-edition.zip
         Archive: dogs-vs-cats-redux-kernels-edition.zip
            creating: dogs-vs-cats-redux-kernels-edition/
           inflating: dogs-vs-cats-redux-kernels-edition/test.zip
            creating: __MACOSX/
            creating: __MACOSX/dogs-vs-cats-redux-kernels-edition/
           inflating: __MACOSX/dogs-vs-cats-redux-kernels-edition/._test.zip
           inflating: dogs-vs-cats-redux-kernels-edition/train.zip
           inflating: __MACOSX/dogs-vs-cats-redux-kernels-edition/._train.zip
           inflating: dogs-vs-cats-redux-kernels-edition/sample submission.csv
           inflating: MACOSX/dogs-vs-cats-redux-kernels-edition/. sample submission.c
           inflating: MACOSX/. dogs-vs-cats-redux-kernels-edition
 In [5]:
          ! unzip ~/dogs-vs-cats-redux-kernels-edition/train.zip
         Archive:
                   /home/jovyan/dogs-vs-cats-redux-kernels-edition/train.zip
            creating: train/
           inflating: train/cat.0.jpg
           inflating: train/cat.1.jpg
           inflating: train/cat.10.jpg
           inflating: train/cat.100.jpg
           inflating: train/cat.1000.jpg
           inflating: train/cat.10000.jpg
           inflating: train/cat.10001.jpg
           inflating: train/cat.10002.jpg
           inflating: train/cat.10003.jpg
           inflating: train/cat.10004.jpg
           inflating: train/cat.10005.jpg
           inflating: train/cat.10006.jpg
```

```
inflating: train/dog.9977.jpg
           inflating: train/dog.9978.jpg
           inflating: train/dog.9979.jpg
           inflating: train/dog.998.jpg
           inflating: train/dog.9980.jpg
           inflating: train/dog.9981.jpg
           inflating: train/dog.9982.jpg
           inflating: train/dog.9983.jpg
           inflating: train/dog.9984.jpg
           inflating: train/dog.9985.jpg
           inflating: train/dog.9986.jpg
           inflating: train/dog.9987.jpg
           inflating: train/dog.9988.jpg
           inflating: train/dog.9989.jpg
           inflating: train/dog.999.jpg
           inflating: train/dog.9990.jpg
           inflating: train/dog.9991.jpg
           inflating: train/dog.9992.jpg
           inflating: train/dog.9993.jpg
           inflating: train/dog.9994.jpg
           inflating: train/dog.9995.jpg
           inflating: train/dog.9996.jpg
           inflating: train/dog.9997.jpg
           inflating: train/dog.9998.jpg
           inflating: train/dog.9999.jpg
 In [9]:
          import mxnet as mx
          import numpy as np
          from mxnet import nd, autograd
          from mxnet import gluon
          %matplotlib inline
          import matplotlib.pyplot as plt
          import cv2
          # define a simple data batch
          from collections import namedtuple
          Batch = namedtuple('Batch', ['data'])
In [11]:
          context=mx.cpu()
In [13]:
          mod = mx.mod.Module(symbol=sym, context=mx.cpu(), label_names=None)
          mod.bind(for_training=False, data_shapes=[('data', (1,3,224,224))],
                   label_shapes=mod._label_shapes)
          mod.set_params(arg_params, aux_params, allow_missing=True)
          # with open('pretrained/synset.txt', 'r') as f:
                labels = [l.rstrip() for l in f]
```

```
In [100...
          sym, arg_params, aux_params = mx.model.load_checkpoint('model', 0)
In [17]:
          def get image(url, show=False):
              if url.startswith('http'):
                  # download and show the image
                  fname = mx.test utils.download(url)
              else:
                  fname = url
              img = cv2.cvtColor(cv2.imread(fname), cv2.COLOR BGR2RGB)
              if img is None:
                   return None
              if show:
                   plt.imshow(img)
                   plt.axis('off')
              # convert into format (batch, RGB, width, height)
              img = cv2.resize(img, (224, 224))
              img = np.swapaxes(img, 0, 2)
              img = np.swapaxes(img, 1, 2)
              img = img[np.newaxis, :]
              return imq
          def predict(url):
              img = get image(url, show=True)
              # compute the predict probabilities
              mod.forward(Batch([mx.nd.array(img)]))
              prob = mod.get outputs()[0].asnumpy()
              # print the top-5
              prob = np.squeeze(prob)
              a = np.argsort(prob)[::-1]
              for i in a[0:5]:
                  print('probability=%f, class=%s' %(prob[i], labels[i]))
In [48]:
          # list the last 10 layers
          all layers = sym.get internals()
          all layers.list outputs()[-3:]
         ['flatten0 output', 'softmax label', 'softmax output']
Out[48]:
In [14]:
          fe_sym = all_layers['flatten0_output']
          fe mod = mx.mod.Module(symbol=fe_sym, context=mx.cpu(), label_names=None)
          fe_mod.bind(for_training=False, data_shapes=[('data', (1,3,224,224))])
          fe_mod.set_params(arg_params, aux_params)
```

```
In [15]:
           def get features(img):
               fe_mod.forward(Batch([mx.nd.array(img)]))
               features = fe_mod.get_outputs()[0].asnumpy()
               return features
In [18]:
           img = get image('https://icatcare.org/app/uploads/2018/07/Thinking-of-getting
           features = get_features(img)
           print("{}\n shape: {}".format(features, features.shape))
          [[ 2.731811
                          3.9008982
                                       4.8866773
                                                    6.902369
                                                                 5.587853
                                                                              5.5976596
             3.6299744
                          5.6956005
                                       5.783274
                                                    5.9859447
                                                                10.328437
                                                                              7.249119
                                       5.579315
            10.089799
                         11.63578
                                                   10.29896
                                                                 9.659985
                                                                              9.824672
             5.997461
                          9.134588
                                      10.0658655
                                                   10.166415
                                                                 9.8080845
                                                                              6.289982
                          1.9723082
            13.240795
                                       7.6495514
                                                    3.712172
                                                                 2.7841299
                                                                              6.2934337
             6.9453053
                          8.631689
                                       7.29645
                                                    6.801901
                                                                 6.52334
                                                                              5.698734
             8.282388
                          3.830843
                                      10.181621
                                                    8.7683935
                                                                12.962755
                                                                              9.04188
                                                    5.7258263
             9.2452
                         10.77944
                                      11.305762
                                                                11.200751
                                                                             10.964636
             3.4149942
                          5.8307867
                                       7.5116644
                                                    9.0251
                                                                 3.9738111
                                                                              3.9767187
             8.057208
                          4.70411
                                       8.180446
                                                    4.212362
                                                                 7.947
                                                                              5.900941
             7.274863
                          9.651376
                                       6.8402615
                                                    3.0084639
                                                                 4.359813
                                                                              4.1929684
             8.628105
                          3.9853938
                                       5.851279
                                                    7.3413982
                                                                 7.4339814
                                                                              6.6312613
             8.691614
                          8.439708
                                       8.843549
                                                    9.059312
                                                                 9.449075
                                                                             12.678203
            11.697756
                          7.8836474
                                       7.3444357
                                                    6.451958
                                                                 6.8108234
                                                                             11.757459
            10.983695
                          7.6246867
                                       5.50387
                                                    7.03473
                                                                12.915622
                                                                              8.201405
             3.994502
                          5.4100966
                                       6.856489
                                                    6.924605
                                                                 9.808682
                                                                              6.971078
             4.193388
                          3.5059521
                                       1.6024159
                                                    4.7404647
                                                                 3.3018126
                                                                              2.9084258
             2.415486
                          5.3541465
                                       8.748269
                                                    5.6931176
                                                                 5.7850704
                                                                              6.1290493
             5.1263065
                          5.997225
                                       7.454259
                                                    6.640497
                                                                11.303702
                                                                              7.120902
             9.347335
                          8.703756
                                       4.8624525
                                                   12.27395
                                                                 6.1802354
                                                                              4.480561
             5.6459785
                          1.0452362
                                       6.122116
                                                    5.4823976
                                                                11.299996
                                                                              4.1871934
             7.719489
                          2.937037
                                       4.467442
                                                    4.9877334
                                                                 5.112029
                                                                              6.401343
                          8.80435
             9.139223
                                       6.144146
                                                    2.6501632
                                                                 4.6860394
                                                                              3.7774022
            10.005826
                          8.510313
                                       7.547855
                                                    7.3255305
                                                                 7.330723
                                                                              3.8226004
             4.816145
                          6.8668685
                                       4.516519
                                                    7.42678
                                                                 6.5116243
                                                                              3.608323
             7.267048
                         10.171126
                                       7.746412
                                                    5.281473
                                                                 8.630499
                                                                              5.626668
             6.502434
                         13.557568
                                       9.849358
                                                    4.837753
                                                                 4.3339353
                                                                              3.7696319
             5.360929
                          2.2633295
                                       2.9870687
                                                    2.431563
                                                                 3.9928882
                                                                              2.2981818
             3.342405
                          6.635318
                                       2.1925945
                                                    8.712073
                                                                              8.373492
                                                                 4.962669
             4.807753
                          2.6049435
                                       4.978068
                                                    3.2130506
                                                                 7.113287
                                                                              1.5388217
             2.2567575
                          3.4583302
                                       2.7490807
                                                    4.909944
                                                                 4.181
                                                                              3.6504633
             5.1258836
                          7.6492605
                                       5.345396
                                                    4.6102114
                                                                 3.7957761
                                                                              2.6940234
             6.086538
                          5.387833
                                       1.9049265
                                                    8.116254
                                                                 7.423186
                                                                              3.3709123
             5.39928
                          6.287825
                                       3.9034204
                                                    6.0577965
                                                                 4.336672
                                                                              7.3916006
             5.458247
                          3.5410912
                                       0.81163764
                                                    4.913134
                                                                 3.077363
                                                                              2.524287
             3.4502792
                          5.504821
                                       5.5890636
                                                    3.1113174
                                                                 1.8214495
                                                                              4.9293275
             4.6584773
                          5.1635823
                                       3.283465
                                                    3.5414634
                                                                 2.3487635
                                                                              1.1612074
             4.398783
                          7.2059193
                                       7.16017
                                                    5.134449
                                                                 3.4099345
                                                                              7.284354
             1.0771716
                          3.487802
                                       7.6779857
                                                    6.426784
                                                                 6.5381465
                                                                              2.5813906
             4.531569
                          5.8846455
                                       6.2358975
                                                    7.270987
                                                                 3.3144867
                                                                              5.6216626
             4.9576645
                          2.6886454
                                       7.497139
                                                    3.8266113
                                                                 2.013793
                                                                              6.8338447
```

```
6.7536345 17.710531
                                    8.317595
                                               11.241352
                                                            7.7849245 11.963486
            3.5570066
                      8.755916
                                    5.8202972
                                                6.244552
                                                            9.4146185
                                                                       8.909975
           11.538079
                        6.4253035 17.255009
                                               13.751663
                                                           11.622289
                                                                       15.469001
                                                           18.205471
            4.237758
                       7.736162
                                   10.365318
                                               13.24044
                                                                       14.377344
            9.415742
                       10.363938
                                    3.6586409
                                                9.332913
                                                            9.287965
                                                                       9.225867
           10.726653
                       2.3643985
                                    3.2395904
                                                2.773296
                                                            7.177313
                                                                       10.607796
           13.280321
                        7.7800007
                                    3.320212
                                               10.017647
                                                            6.189191
                                                                       6.967745
            5.764682
                       8.264951
                                    2.065886
                                                9.877997
                                                            9.344261
                                                                       11.44254
                                                            4.4888806
            5.910216
                       4.108808
                                    5.1819353
                                                2.406356
                                                                        4.482684
            3.3455486 12.267346
                                    6.304007
                                                5.7975836
                                                            5.489004
                                                                        2.9057488
            8.029712
                       7.311815
                                    5.331028
                                                6.780764
                                                            7.2698026
                                                                        7.03161
            6.9609914
                        8.205074
                                    6.2120485
                                                9.704169
                                                            4.3887997
                                                                        9.308927
            5.334109
                        6.4561872
                                    5.7876287
                                                4.3589535
                                                            7.1413755
                                                                        4.1641693
            8.003898
                        8.619182
                                    2.5742774
                                                2.4626925
                                                            3.8087509
                                                                        5.987408
           10.418253
                        9.594042
                                                                        9.242509
                                   18.342413
                                               11.671421
                                                            2.637681
            2.5423734
                                              5.139831
                                                            1.7740538
                        6.47289
                                    5.5148525
                                                                        3.9243221
            5.780046
                        5.5538406
                                    2.6732502
                                                7.5682645
                                                            8.417677
                                                                        2.763147
           10.0679455
                        9.356203
                                    5.7668076 13.894629
                                                            7.128599
                                                                        8.015945
                        7.279767
                                    3.9210467
                                               4.57002
                                                            6.2065
                                                                        5.7564244
            4.268324
            7.1734467
                        3.3807797 11.519496
                                               16.3085
                                                          11
          shape: (1, 1000)
In [19]:
          # from https://www.kaggle.com/c/dogs-vs-cats-redux-kernels-edition/data
          from os import listdir
          from os.path import isfile, join
          import os
          mypath = join(os.getcwd(),'train')
          cats_imgs = [join(mypath,f) for f in listdir(mypath) if f.startswith('cat')]
          dogs imgs = [join(mypath,f) for f in listdir(mypath) if f.startswith('dog')]
In [20]:
          print("cats: {} and dogs: {}".format(len(cats imgs),len(dogs imgs)))
         cats: 12500 and dogs: 12500
In [22]:
          Nmax = 100 \# (N: 10, 100, 500, 1000, also 5000 and 12500)
          cats_features = [get_features(get_image(img)).ravel() for img in cats_imgs[:N
          dogs_features = [get_features(get_image(img)).ravel() for img in dogs_imgs[:N
In [24]:
          Y cats = np.array(Nmax * [1])
          Y dogs = np.array(Nmax * [0])
In [25]:
          X cvd = np.vstack([cats features,dogs features])
          Y cvd = np.vstack([Y cats,Y dogs]).ravel()
```

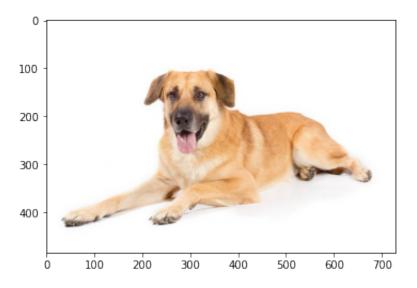
```
In [26]:
          from sklearn.model selection import train test split
          X train, X test, y train, y test = train test split(X cvd, Y cvd, random stat
In [27]:
          from sklearn.linear model import LogisticRegression
          lg = LogisticRegression().fit(X_train, y_train)
          print("Test set score: {:.2f}".format(lg.score(X_test, y_test)))
         Test set score: 0.88
         /opt/conda/lib/python3.7/site-packages/sklearn/linear model/ logistic.py:940:
         ConvergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear model.html#logistic-regress
         ion
           extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG)
In [30]:
          img = get image('https://s3.amazonaws.com/cdn-origin-etr.akc.org/wp-content/u
          features = get features(img)
          print("{}\n shape: {}".format(features, features.shape))
         [[ 3.8680282
                         3.0770736
                                     3.8007517
                                                  3.9388447
                                                              4.4794497
                                                                          2.513034
                                                  1.1125107
            1.0257548
                         3.3912466
                                     4.1260457
                                                              3.0114415
                                                                          2.2824717
            3.308097
                         3.6617973
                                     2.2812922
                                                 1.8932449
                                                              4.174476
                                                                          3.710502
            0.45999247 4.1961174
                                     1.2289764
                                                 9.014702
                                                              6.0741653
                                                                          5.186479
            4.293989
                         3.2768373
                                     4.330217
                                                  3.7503705
                                                              6.209451
                                                                          4.7506633
            3.809687
                         8.039387
                                     5.815378
                                                  4.479296
                                                              4.644457
                                                                          2.287041
            4.105688
                         3.3356824
                                     8.155916
                                                 4.587722
                                                              4.465425
                                                                          3.855896
            4.043045
                         5.1172795
                                     4.4702134
                                                 3.342398
                                                              5.4953494
                                                                          6.7198224
            3.84865
                         1.182495
                                     2.0083091
                                                  9.924515
                                                              2.7754364
                                                                          2.005786
            3.0824454
                         1.4339454
                                     2.6161735
                                                  1.7753203
                                                              1.2789447
                                                                          2.4843435
            1.5804738
                         6.9447136
                                     4.049789
                                                  2.9182622
                                                              1.8067827
                                                                          1.929847
            2.0441859
                         2.8852274
                                     2.3605494
                                                  5.1194777
                                                              2.5484276
                                                                          5.3415647
            3.779574
                         3.719067
                                     4.224366
                                                  4.0597167
                                                              2.472064
                                                                          2.177067
            6.892461
                         4.2934427
                                     4.110542
                                                  5.4240932
                                                              6.0283465
                                                                          4.938808
            1.0311867
                         1.9734994
                                     4.873224
                                                10.826043
                                                              6.586989
                                                                         10.874734
                         2.369898
                                                  1.7927071
                                                              7.2485766
            3.7613354
                                     2.6786683
                                                                          1.7895155
            2.0912905
                                                              1.7799224
                                                                          8.096338
                         4.3573813
                                     1.0156956
                                                  4.274913
            3.7537284
                         7.285345
                                     7.295822
                                                  5.0679016
                                                              2.7826424
                                                                          6.1655016
            2.4368806
                         1.3257196
                                     4.5575867
                                                  6.473884
                                                             11.895296
                                                                          6.8320866
            5.361
                         2.6531858
                                                  5.8689027
                                                                          2.6203759
                                     4.564212
                                                              1.9365345
            5.183484
                         1.7165029
                                     3.0182095
                                                  2.274215
                                                              4.6858997
                                                                          5.2946324
            7.652743
                         1.6650032
                                     1.8852196
                                                  4.6898293
                                                              2.1779253
                                                                          1.8760023
            1.9162517
                         3.1303358
                                     2.3637066
                                                  0.35587963
                                                              2.2771838
                                                                          1.5375512
```

```
7.3680015
                           9.480289
                                        5.6531177
                                                                  4.543409
  5.629182
                                                     7.636259
  2.8130362
              2.5628386
                          10.775148
                                        8.83947
                                                     9.697046
                                                                  8.9554405
  5.122448
              2.8479238
                           6.6367116
                                      10.020451
                                                     5.7062306
                                                                  5.554301
             11.314349
 10.600308
                           5.623864
                                        9.765764
                                                     9.736605
                                                                  8.770255
  6.3605294
               4.037713
                           4.383122
                                        8.11133
                                                     3.4699535
                                                                  5.010916
  9.313533
             12.367925
                           6.7642603
                                        3.3070233
                                                     7.72455
                                                                  3.3671126
  2.723685
             13.013277
                           4.3527923
                                        8.93426
                                                     3.0037897
                                                                  2.9002788
                                        5.4077883
  9.682216
               2.4089184
                         10.428298
                                                     3.6294236
                                                                  0.82561
                           8.439979
  8.775564
                                        2.38039
              8.402793
                                                     5.6820807
                                                                  4.6136727
  8.871776
               3.9851723
                           9.375548
                                        9.182731
                                                     3.058218
                                                                  3.459215
  5.4962754
              4.018234
                           8.839286
                                        8.8314085
                                                     9.505224
                                                                  2.3476098
  8.823815
             12.072752
                           8.601356
                                        7.9471993
                                                    11.292472
                                                                 11.457238
  6.027743
              7.071573
                           7.0377216
                                        2.9373167
                                                     7.307343
                                                                  8.48312
               2.9167256
 19.111702
                           5.7694364
                                        9.05401
                                                     7.0445848
                                                                  5.576129
  2.3362818
               8.641402
                           5.7351575
                                        6.7036204
                                                     5.4995666
                                                                  6.012248
  2.507907
               6.9093833
                           5.001413
                                        4.078641
                                                     4.557381
                                                                  9.426562
  4.4181395
              2.9529035
                           8.198942
                                        2.3040478
                                                     2.800392
                                                                  4.689731
 13.018832
               2.6840668
                           3.003842
                                        6.659847
                                                     4.55896
                                                                  5.778887
 10.350336
               5.926081
                           4.4054236
                                       14.795793
                                                     1.2430875
                                                                  7.5658617
  0.87765217
              7.2287545
                           6.850049
                                        4.951525
                                                     8.017592
                                                                  7.024554
  5.953328
               7.7224193
                           6.606814
                                        6.7617393
                                                    11.722911
                                                                  6.332674
  3.7297676
              3.657504
                          12.947633
                                       12.445496
                                                     3.9683583
                                                                  4.0251756
  5.04758
              7.196997
                           4.534546
                                        5.3412795
                                                     9.387548
                                                                 11.5132
  4.3527474
              3.6554115
                           3.1876783
                                        3.529498
                                                     8.873071
                                                                  4.952435
  6.8690434
              2.9971051
                           2.8173728
                                        6.308332
                                                     1.3162148
                                                                  7.098621
  5.2898364
               3.9596841
                           1.9113549
                                        1.9215572
                                                     7.324026
                                                                  9.437474
  7.8335223
               5.1925454
                           4.6072373
                                        4.2747693
                                                     6.805999
                                                                  3.2848585
  8.721183
              7.721843
                           5.8097773
                                        4.0527506
                                                     4.101082
                                                                  5.9977326
  4.32023
               3.37436
                           6.048842
                                        5.917717
                                                     2.8842163
                                                                  7.8604107
                           3.9107037
  3.3678865
              4.4946995
                                        4.710452
                                                     3.8326082
                                                                  3.741211
  6.5966315
              4.096897
                           7.036597
                                        6.277786
                                                     7.1407843
                                                                  5.955763
               7.523023
                           5.278676
                                                     3.9909554
                                                                  8.753346
  7.9515915
                                        2.6221385
  6.462921
               1.9766583
                           5.333998
                                        3.7846956
                                                     2.9319718
                                                                  6.691585
  6.369771
               2.7630775
                           2.607686
                                        3.20876
                                                     3.6895769
                                                                  5.622818
               3.9689782
                                                     7.9343553
                                                                  2.7381692
  5.5627694
                           2.0877512
                                        5.8020372
  2.6250763
              3.7928612
                           1.1395389
                                        5.831917
                                                     5.012505
                                                                  4.5884542
  4.925548
                                        6.383615
                                                     3.1053205
                                                                  6.64485
               5.024918
                           3.991863
  9.6873665
               5.221062
                           5.8723288
                                       13.390061
                                                   ]]
shape: (1, 1000)
```

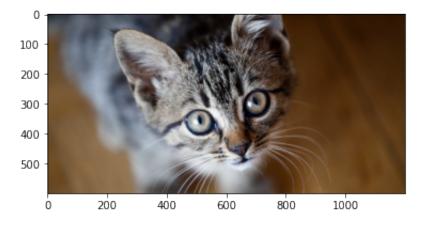
```
import matplotlib.image as mpimg

dog_test_path = join(os.getcwd(),'Chinook-On-White-03.jpg')
    cat_test_path = join(os.getcwd(),'Thinking-of-getting-a-cat.png')

img = mpimg.imread(dog_test_path)
    imgplot = plt.imshow(img)
    plt.show()
```



```
img = mpimg.imread(cat_test_path)
imgplot = plt.imshow(img)
plt.show()
```



```
In [35]:
    features_out = get_features(get_image(cat_test_path))
    prob = lg.predict_proba(features_out)
    pred = lg.predict(features_out)

if pred[0] == 1:
        fpred = 'cat'
else:
        fpred = 'dog'

print("prob: {} and prediction: {}".format(prob,fpred))
```

prob: [[7.82527617e-08 9.99999922e-01]] and prediction: cat

Repeat all previous steps using MobileNet V2 (https://github.com/KeyKy/mobilenet-mxnet). How the two networks compare?

In [36]:

```
!wget https://github.com/KeyKy/mobilenet-mxnet/raw/master/mobilenet_v2-symbol
!wget https://github.com/KeyKy/mobilenet-mxnet/raw/master/mobilenet_v2-0000.p
```

```
--2020-11-07 12:04:12--
                        https://github.com/KeyKy/mobilenet-mxnet/raw/master/m
obilenet v2-symbol.json
Resolving github.com (github.com)... 140.82.121.4
Connecting to github.com (github.com) | 140.82.121.4 | :443... connected.
HTTP request sent, awaiting response... 302 Found
Location: https://raw.githubusercontent.com/KeyKy/mobilenet-mxnet/master/mobil
enet_v2-symbol.json [following]
--2020-11-07 12:04:13-- https://raw.githubusercontent.com/KeyKy/mobilenet-mxn
et/master/mobilenet v2-symbol.json
Resolving raw.githubusercontent.com (raw.githubusercontent.com)... 151.101.84.
133
Connecting to raw.githubusercontent.com (raw.githubusercontent.com) | 151.101.84
.133 :443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 108557 (106K) [text/plain]
Saving to: 'mobilenet v2-symbol.json'
mobilenet v2-symbol 100%[============] 106.01K --.-KB/s
                                                                  in 0.05s
2020-11-07 12:04:13 (1.99 MB/s) - 'mobilenet_v2-symbol.json' saved [108557/108
5571
--2020-11-07 12:04:14-- https://github.com/KeyKy/mobilenet-mxnet/raw/master/m
obilenet v2-0000.params
Resolving github.com (github.com)... 140.82.121.4
Connecting to github.com (github.com) | 140.82.121.4 | :443... connected.
HTTP request sent, awaiting response... 302 Found
Location: https://raw.githubusercontent.com/KeyKy/mobilenet-mxnet/master/mobil
enet v2-0000.params [following]
--2020-11-07 12:04:14-- https://raw.githubusercontent.com/KeyKy/mobilenet-mxn
et/master/mobilenet v2-0000.params
Resolving raw.githubusercontent.com (raw.githubusercontent.com)... 151.101.84.
Connecting to raw.githubusercontent.com (raw.githubusercontent.com) | 151.101.84
.133 :443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 14178924 (14M) [application/octet-stream]
Saving to: 'mobilenet_v2-0000.params'
in 0.4s
2020-11-07 12:04:16 (34.5 MB/s) - 'mobilenet v2-0000.params' saved [14178924/1
41789241
```

```
In [102...
          # define a simple data batch
          from collections import namedtuple
          Batch = namedtuple('Batch', ['data'])
In [103...
          context=mx.cpu()
In [104...
          def get_image(url, show=False):
              if url.startswith('http'):
                  # download and show the image
                  fname = mx.test utils.download(url)
              else:
                  fname = url
              img = cv2.cvtColor(cv2.imread(fname), cv2.COLOR BGR2RGB)
              if img is None:
                   return None
              if show:
                   plt.imshow(img)
                   plt.axis('off')
              # convert into format (batch, RGB, width, height)
              img = cv2.resize(img, (224, 224))
              img = np.swapaxes(img, 0, 2)
              img = np.swapaxes(img, 1, 2)
              img = img[np.newaxis, :]
              return imq
          def predict(url):
              img = get image(url, show=True)
              # compute the predict probabilities
              mod.forward(Batch([mx.nd.array(img)]))
              prob = mod.get_outputs()[0].asnumpy()
              # print the top-5
              prob = np.squeeze(prob)
              a = np.argsort(prob)[::-1]
              for i in a[0:5]:
                  print('probability=%f, class=%s' %(prob[i], labels[i]))
In [105...
          sym, arg params, aux params = mx.model.load_checkpoint('mobilenet_v2', 0)
In [106...
          # list the last 10 layers
          all layers = sym.get internals()
          all layers.list outputs()[-5:]
         ['fc7_bias', 'fc7_output', 'fc7_flatten_output', 'prob_label', 'prob_output']
Out [106...
```

```
In [107...
          fe_sym = all_layers['fc7_flatten_output']
          fe mod = mx.mod.Module(symbol=fe_sym, context=mx.cpu(), label_names=None)
          fe_mod.bind(for_training=False, data_shapes=[('data', (1,3,224,224))])
          fe_mod.set_params(arg_params, aux_params)
In [108...
          def get features(img):
              fe_mod.forward(Batch([mx.nd.array(img)]))
              features = fe_mod.get_outputs()[0].asnumpy()
              return features
In [109...
          img = get image('https://icatcare.org/app/uploads/2018/07/Thinking-of-getting
          features = get features(img)
          print("{}\n shape: {}".format(features, features.shape))
         [-8.73116531e+01 \quad 2.83228577e+02 \quad -3.59607025e+02 \quad -4.03708984e+02]
           -6.46255798e+02 1.48531494e+02 -3.47858490e+02 6.66299515e+01
            2.78768982e+02 -1.49736664e+02 8.38041992e+01 -1.37439499e+02
           -4.69580017e+02 -9.14800415e+01 -2.76609131e+02 1.35187820e+02
            1.80837677e+02 -3.41249298e+02 1.15875607e+01 -5.60398926e+02
           -2.54541214e+02 3.32528320e+02 2.52942486e+01 4.03285645e+02
            2.14639755e+02 -4.70253792e+01 -4.01036255e+02 -2.38460938e+02
           -5.36336288e+01 -6.14872131e+02 -3.40083069e+02 -2.73898651e+02
           -3.69430878e+02 -6.40303528e+02 -2.56124176e+02 -4.16391968e+02
           -1.71422821e+02 -2.47153595e+02 -1.99477921e+02 -1.43629562e+02
           -3.21645264e+02 -4.46027039e+02 -2.44809418e+02 1.03301361e+02
           -1.46011917e+02 -4.25061462e+02 -3.44877625e+02 -5.04795135e+02
           -1.92245743e+02 -4.46315369e+02 -7.29045776e+02 6.43142151e+02
           -5.43736694e+02 -3.97714417e+02 -5.26605774e+02 -7.36024475e+02
           -1.60982224e+02 -2.95890625e+02 -7.81803040e+02 -3.25579132e+02
           -8.40681839e+01 -6.43729919e+02 4.14718666e+01 -3.34889435e+02
           -5.07559692e+02 -1.62592194e+02 -1.88286270e+02 -6.47604553e+02
           -1.27652222e+02 -1.61137512e+02 1.22373108e+02 -4.67065430e+02
           -2.27845261e+02 4.93836746e+01 -8.06280289e+01 5.41381775e+02
           -2.29811203e+02 -4.32604919e+02 1.78805542e+02 -2.30924088e+02
            7.84939880e+02 -3.04035858e+02 2.51537018e+02 5.64096863e+02
           -1.44117874e+02 -2.62628815e+02 2.57276184e+02 -1.06053314e+02
            3.47322205e+02 -1.54493103e+02 5.06014801e+02 4.19897217e+02
            3.65872421e+01 3.79745605e+02 -3.07246521e+02 -9.59804459e+01
            3.53265350e+02 -1.74011169e+02 2.35754410e+02 1.05712608e+02
           -6.54890060e+01 -3.04422974e+02 -1.38272079e+02 2.41708221e+02
            1.80800858e+02 -2.11239624e+02 -2.68215027e+01 -2.94282776e+02
           -2.98940186e+02 2.11309464e+02 -7.89026062e+02 -1.89356308e+02
           -1.41223419e+02 -2.24167953e+02 -2.93801178e+02 -3.43777405e+02
           -4.11104340e+02 -1.95378922e+02 -2.99134857e+02 -3.35203796e+02
           -3.44139679e+02 -5.81374756e+02 -2.53917221e+02 -3.83287689e+02
           -2.50095322e+02 -3.55800873e+02 2.17763214e+02 -5.51281616e+02
            1.30699783e+02 2.94620914e+01 -1.50848785e+02 -4.85231018e+02
           -1.26667007e+02 -2.28896347e+02 1.63994827e+02 2.31013756e+01
           -1.94970383e+02 1.25436058e+02 -4.93182648e+02 -8.36208313e+02
```

```
-3.89129456e+02 3.36298065e+01 -2.52807846e+02 -3.09167664e+02
           -1.47562958e+02 -4.39925537e+02 -1.87984955e+02 -3.16677673e+02
           -9.73731308e+01 -4.17394531e+02 8.80572357e+01 -2.53235886e+02]]
          shape: (1, 1000)
In [110...
          from os import listdir
          from os.path import isfile, join
          import os
          mypath = join(os.getcwd(), 'train')
          cats_imgs = [join(mypath,f) for f in listdir(mypath) if f.startswith('cat')]
          dogs_imgs = [join(mypath,f) for f in listdir(mypath) if f.startswith('dog')]
In [111...
          print("cats: {} and dogs: {}".format(len(cats imgs),len(dogs imgs)))
         cats: 12500 and dogs: 12500
In [112...
          Nmax = 100
          cats_features = [get_features(get_image(img)).ravel() for img in cats_imgs[:N
          dogs_features = [get_features(get_image(img)).ravel() for img in dogs_imgs[:N
In [113...
          Y_cats = np.array(Nmax * [1])
          Y dogs = np.array(Nmax * [0])
In [114...
          X_cvd = np.vstack([cats_features,dogs_features])
          Y cvd = np.vstack([Y cats,Y dogs]).ravel()
In [115...
          from sklearn.model selection import train test split
          X train, X test, y train, y test = train test split(X cvd, Y cvd, random stat
In [116...
          from sklearn.linear model import LogisticRegression
          lg = LogisticRegression().fit(X train, y train)
          print("Test set score: {:.2f}".format(lg.score(X_test, y_test)))
```

```
Test set score: 0.54
```

/opt/conda/lib/python3.7/site-packages/sklearn/linear_model/_logistic.py:940: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:
 https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
 https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

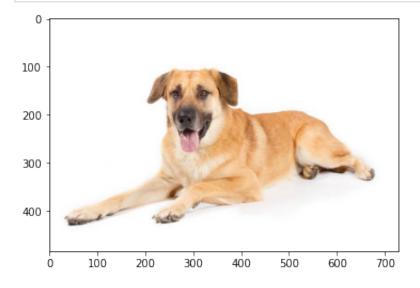
extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG)

```
In [117...
```

```
import matplotlib.image as mpimg

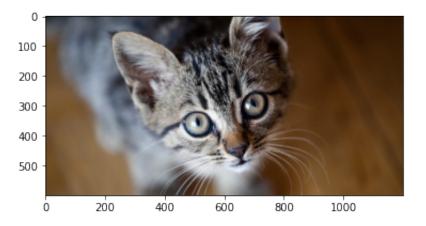
dog_test_path = join(os.getcwd(), 'Chinook-On-White-03.jpg')
cat_test_path = join(os.getcwd(), 'Thinking-of-getting-a-cat.png')

img = mpimg.imread(dog_test_path)
imgplot = plt.imshow(img)
plt.show()
```



```
In [118...
```

```
img = mpimg.imread(cat_test_path)
imgplot = plt.imshow(img)
plt.show()
```



```
In [119...
    features_out = get_features(get_image(cat_test_path))
    prob = lg.predict_proba(features_out)
    pred = lg.predict(features_out)

if pred[0] == 1:
        fpred = 'cat'
else:
        fpred = 'dog'

    print("prob: {} and prediction: {}".format(prob,fpred))
```

prob: [[0.74842734 0.25157266]] and prediction: dog

Different networks apparently produced different results

In []: