Detecting Asteroids with Neural Networks using PyBrain

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The goal

Build and train a neural network to correctly identify asteroids in astrophotography data, using **PyBrain**, a modular machine learning library for Python.

Disclaimer

- I am not an expert;
- ► This is not (quite) my field;
- Some things might be wrong!

The data

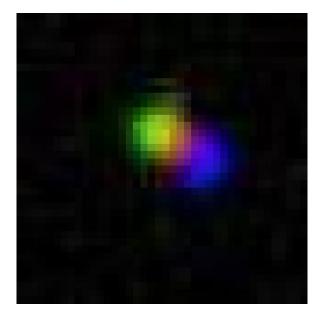
The Sloan Digital Sky Survey:

- "One of the most ambitious and influential surveys in the history of astronomy."
- ► Approx 35% of sky;
- Largest uniform survey of the sky yet accomplished;
- Data is freely available online;
- ► Each image is 922x680 pixels.





An example asteroid



Why use a Neural Network?

This type of classification is well suited for a neural network:

- We have a clear set of training data;
- There is a small amount of input features which can accurately define an item:
 - Ratio valid hues to non-valid hues
 - ▶ Best possible cluster collinearity
 - Best possible average cluster distance
- ▶ Each of the input features can be resolved to a $0 \rightarrow 1$ metric;
- ► The output is either affirmative (1) or negative (0);
- Neural network activation will be fast!

Getting started

Getting the initial training data:

- Small tool to extract potential candidates from full-scale images;
- Extremely naïve, approx 100:5 false positives to actual positives;
- Very low false negatives (approx 1:1000);
- Incredibly slow (complex scan of 100Ks of potentials);
- Manual classification, somewhat slow;
- Yields approx 250 valid items, 500 invalid items;
- ► Form is a set of 20x20px images.

Making the data set

from pybrain.datasets import SupervisedDataSet

```
def make_dataset(source):
15
        data = SupervisedDataSet(3, 1)
16
17
       print("Adding valid training data")
18
       for i in glob(source + "valid/*.jpg"):
19
            data.addSample(functions.values(i), [1])
20
21
       print("Adding invalid training data")
22
        for i in glob(source + "invalid/*.jpg"):
23
            data.addSample(functions.values(i), [0])
24
25
        return data
26
```

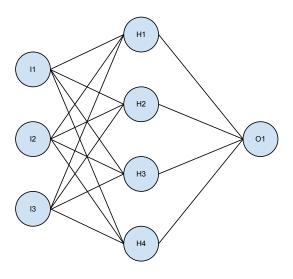
Building and training the network

```
from pybrain.tools.shortcuts import buildNetwork
from pybrain.supervised import BackpropTrainer
```

```
def train_network(d, iterations):
29
        print("Training")
30
        n = buildNetwork(d.indim, 4, d.outdim, bias=True)
31
        t = BackpropTrainer(
32
            n,
33
            d.
34
            learningrate=0.01,
35
            momentum=0.99,
36
            verbose=False)
37
        for epoch in range(iterations):
38
            t.train()
39
        return n
40
```

Building the neural network

The resulting neural network:



Training the network

- Approx 250 valid items;
- Approx 500 invalid items;
- Trained for 5,000 iterations;
- ► Took approx. 3 hours;
- Probably could have gotten by with less iterations.

Testing the network

```
import shutil
9
   import os
10
   def test(path, source, net, cutoff):
43
       val = net.activate(functions.values(path))
44
        base = os.path.basename(path)
45
        if val > cutoff:
46
            print path, val, "(Valid)"
47
            shutil.copy(path, source + 'valid/' + base)
48
        else:
49
            print path, val, "(Invalid)"
50
            shutil.copy(path, source + 'invalid/' + base)
51
```

Putting it all together

```
data = make_dataset('./training_data')
net = train_network(data, iterations=5000)

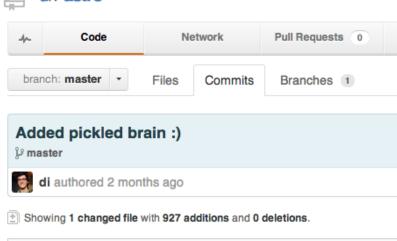
for path in glob('./' + sys.argv[1] + "*.jpg"):
test(path, './' + sys.argv[1], net, cutoff=0.9)
```

Storing your neural network

```
8 import pickle
```

```
if __name__ == "__main__":
53
        try:
54
            f = open('_learned', 'r')
55
            net = pickle.load(f)
56
            f.close()
57
        except:
58
            data = make_dataset('./training_data')
59
            net = train_network(data, iterations=5000)
60
            f = open('_learned', 'w')
61
            pickle.dump(net, f)
62
            f.close()
63
```







Thanks!

- Contact me: dustin@drexel.edu
- Source for this talk: https://github.com/di/astro
- ► The Sloan Digital Sky Survey: http://www.sdss.org/
- PyBrain: http://pybrain.org/