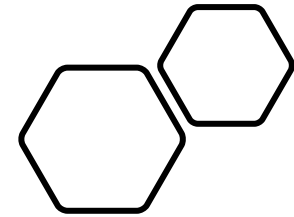
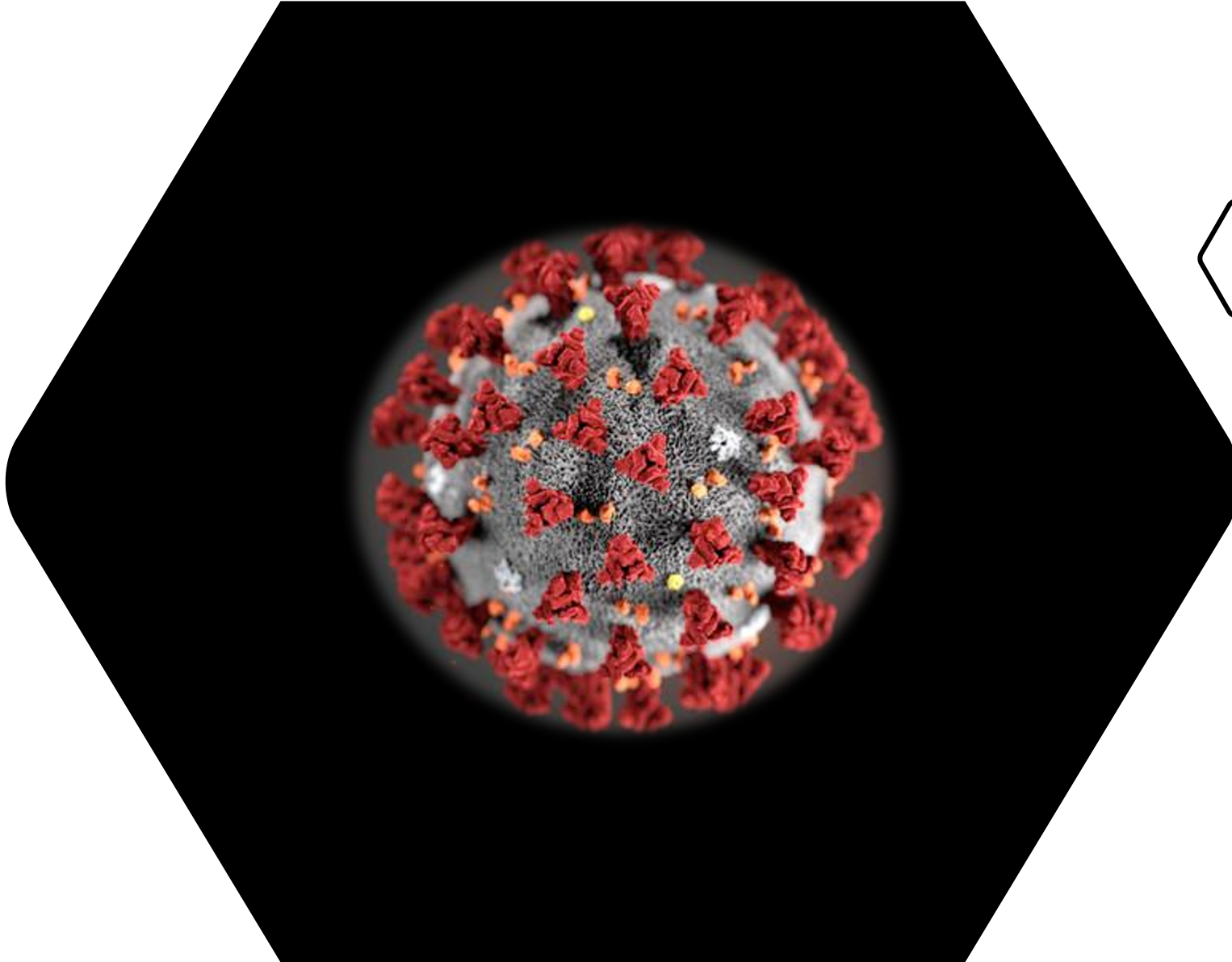




Beyond CS106A

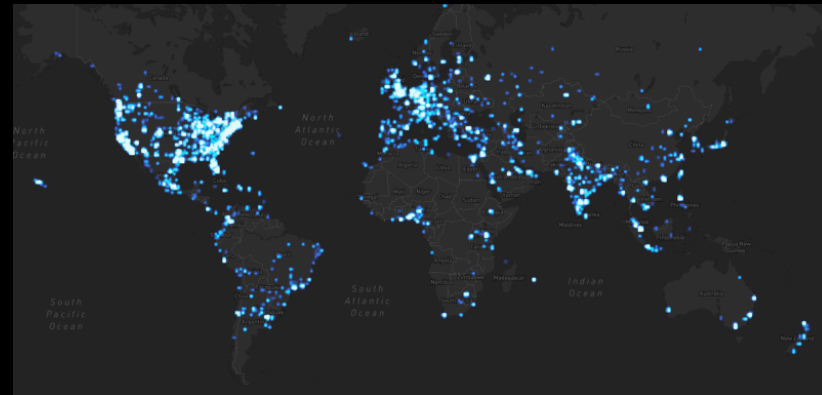
Chris Piech
CS106A, Stanford University

Did you know?

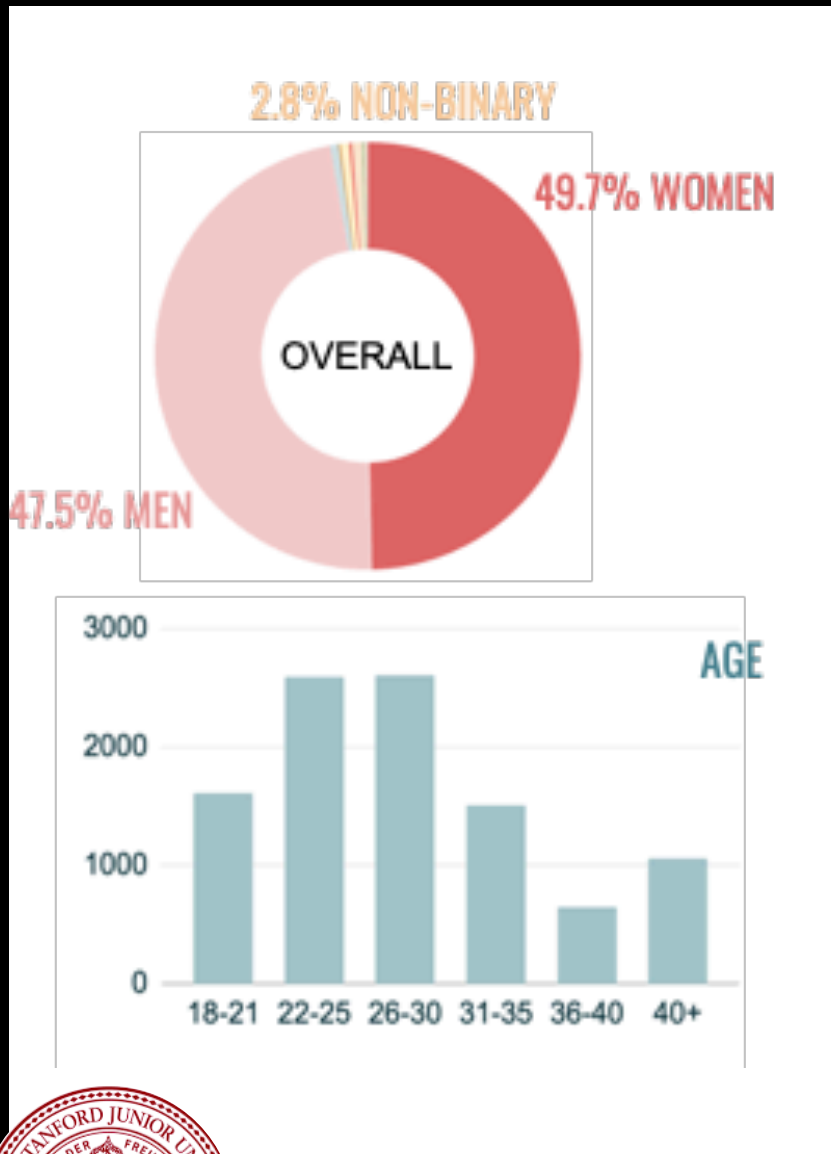


Code in Place : Course with the most section leaders?

905 section leaders teach
10,000 students
First half of Stanford CS106A



20% experienced job loss or home loss
10x retention vs baseline MOOC
99% wanted more (after first section)
6k hours of live teaching
60k hours of lecture watched
Better CS106A for **Stanford** students

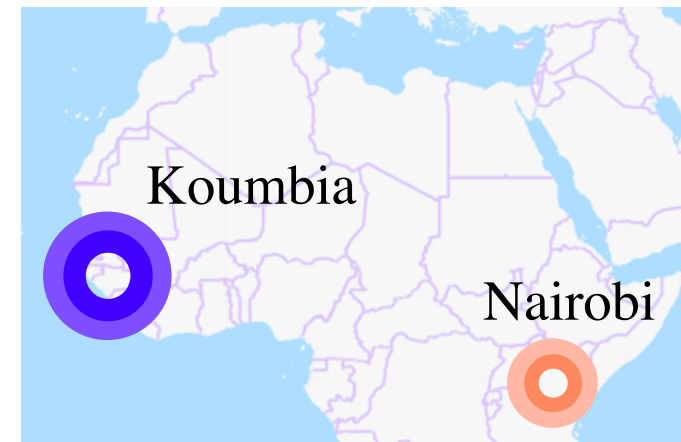
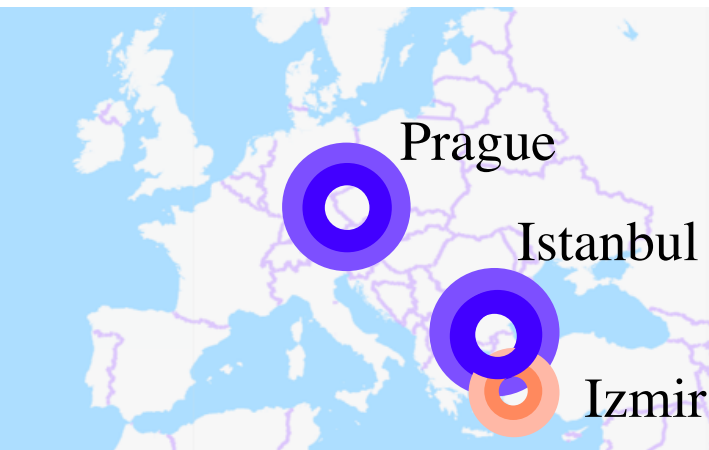


Have you thought about teaching?

Great way to learn

Great way to give back

In-person teaching around the world.



1. How to make your own project
2. What other languages look like
3. Deep Learning in Python

Life after CS106A

Programming Languages



Python

```
evens = []  
for i in range(100):  
    if i % 2 == 0:  
        evens.append(i)  
print evens
```

prints [2, 4, 6, 8, 10, 12, ...]



C++

```
Vector<double> evens;  
for(int i = 0; i < 100; i++) {  
    if(i % 2 == 0) {  
        evens.add(i);  
    }  
}  
cout << evens << endl;
```

prints [2, 4, 6, 8, 10, 12, ...]



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        evens.add(i);  
    }  
}  
cout << evens << endl;
```

prints [2, 4, 6, 8, 10, 12, ...]



Java

```
ArrayList<Double> evens = new ArrayList<Double>();  
for(int i = 0; i < 100; i++) {  
    if(i % 2 == 0) {  
        evens.add(i);  
    }  
}  
println(evens);
```

prints [2, 4, 6, 8, 10, 12, ...]



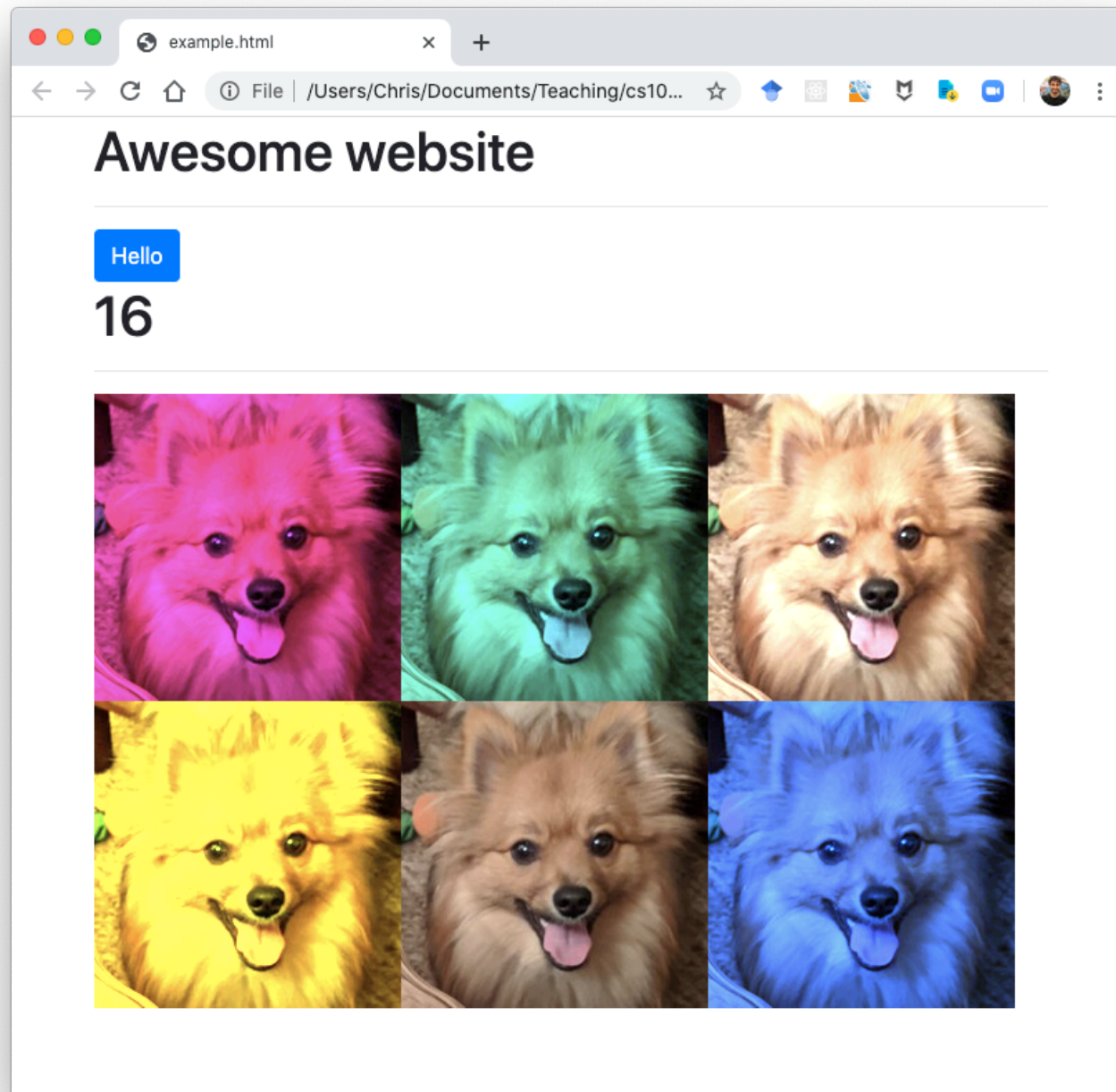
Javascript

```
var evens = []  
for(var i = 0; i < 100; i++) {  
    if(i % 2 == 0) {  
        evens.push(i)  
    }  
}  
console.log(evens)
```

prints [2, 4, 6, 8, 10, 12, ...]



Lets Play



There is something going on
in the world of AI

[suspense]

Self Driving Cars



Computers Making Art



The Last Remaining Board Game

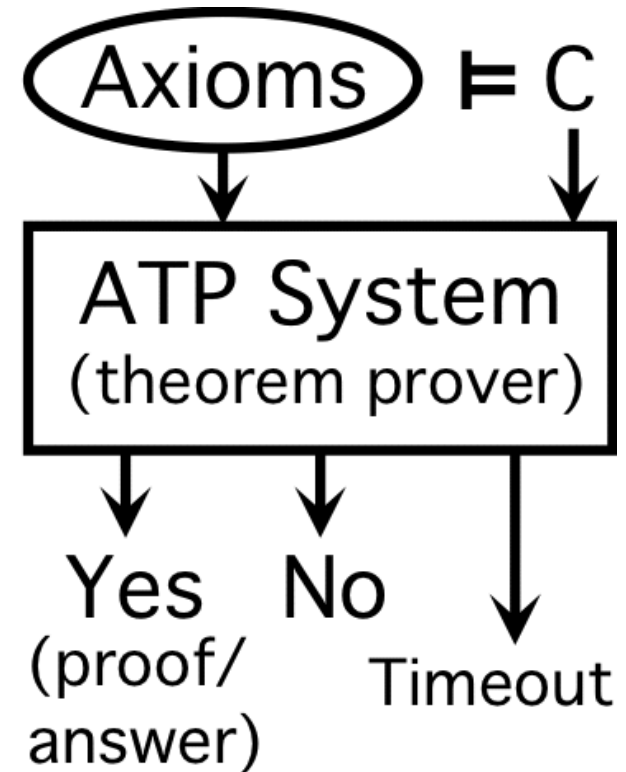


Early Optimism 1950

1952



1955



Computer Vision



Piech, CS106A, Stanford University



Classification



That is a picture
of a **one**



Classification



That is a picture
of a **zero**



Classification



That is a picture
of an **zero**



* It doesn't have to
be correct all of the
time



Identifying Cats

Here's one way you might code this...

```
def is_cat(image):  
    if contains_two_eyes(image):  
        if has_whiskers(image):  
            if has_pointy_ears(image):  
                return True  
    return False
```

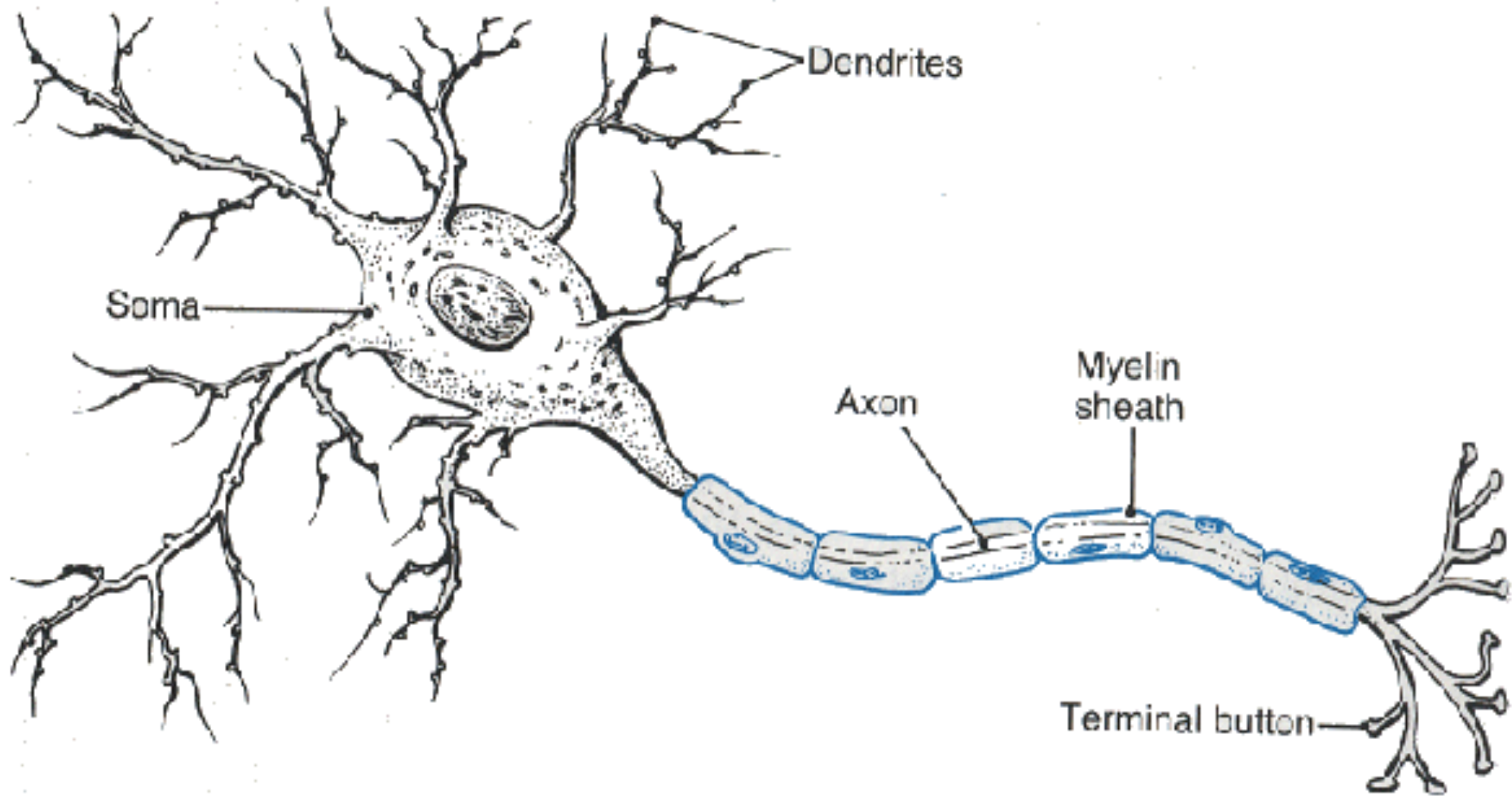


Some Tricky Cases

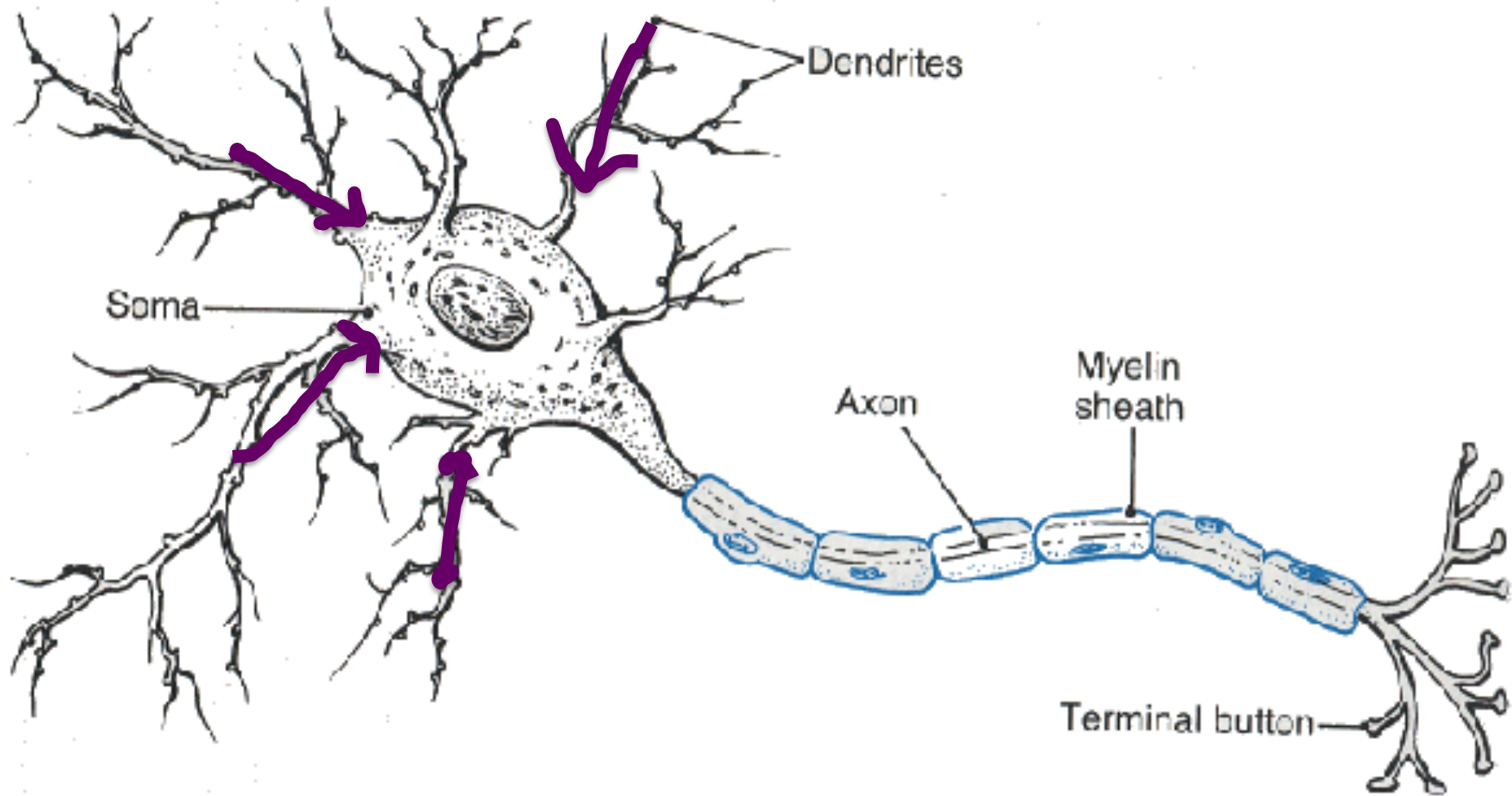


Great idea inspired by biology

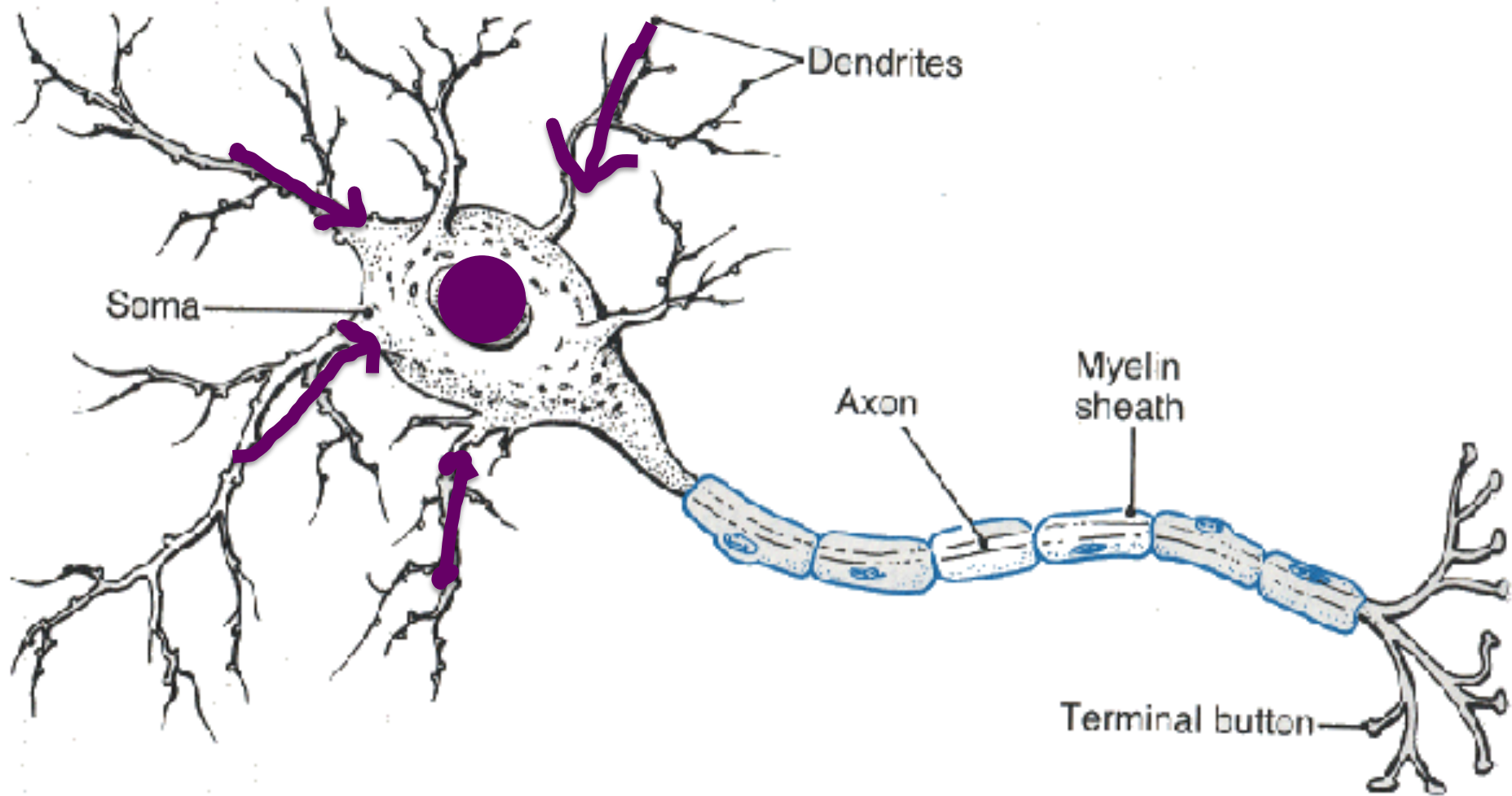
Neuron



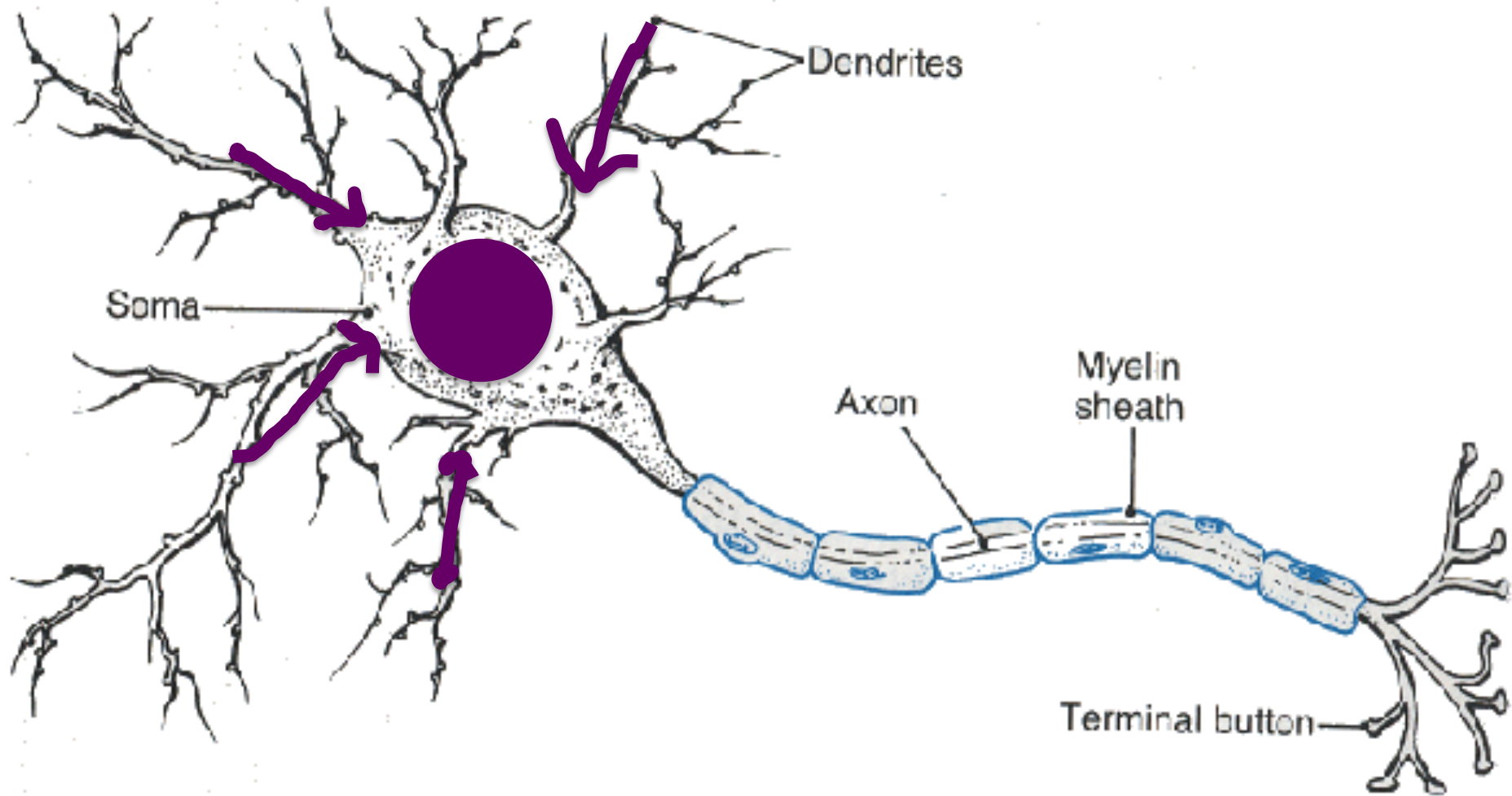
Neuron



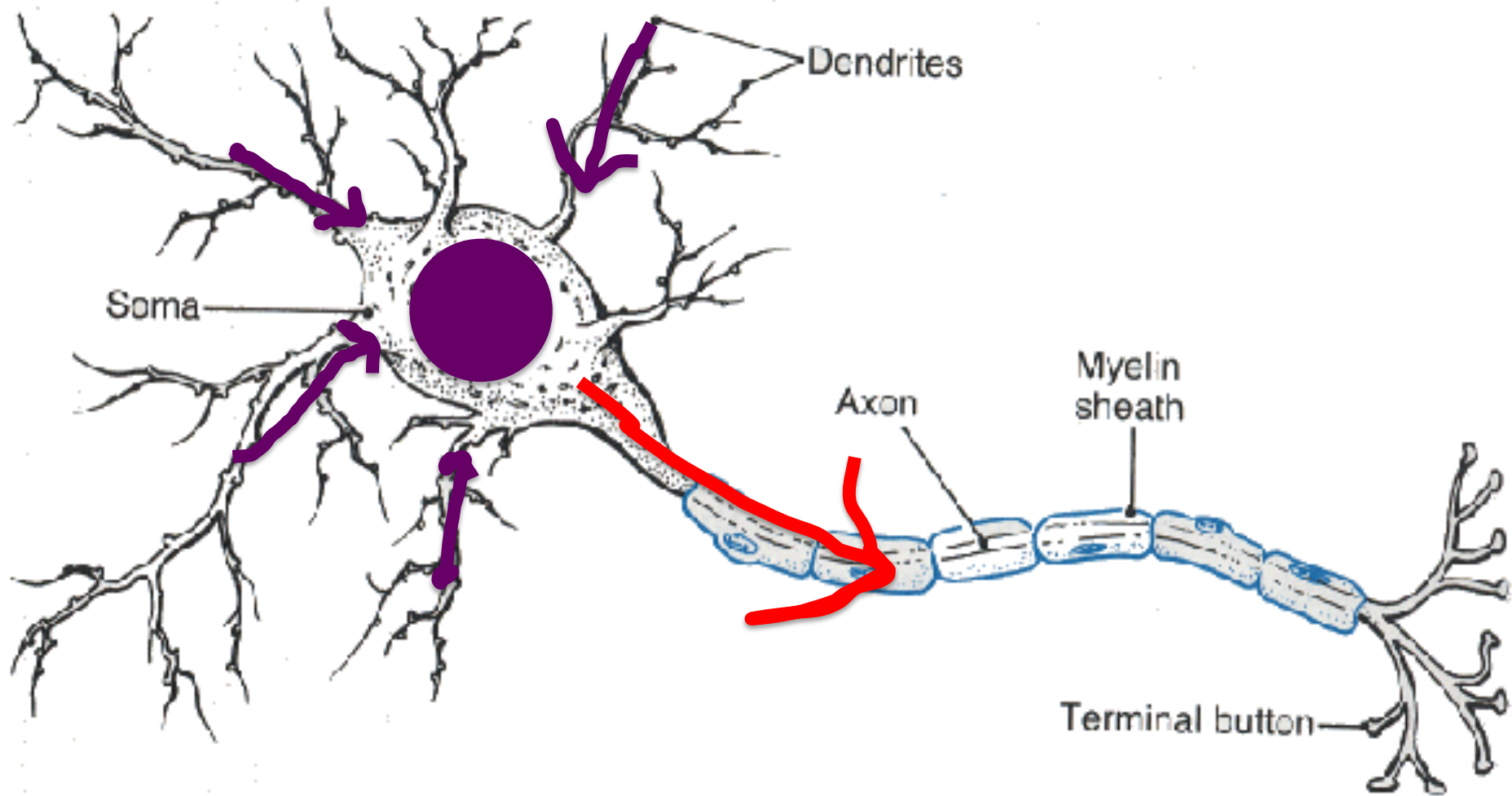
Neuron



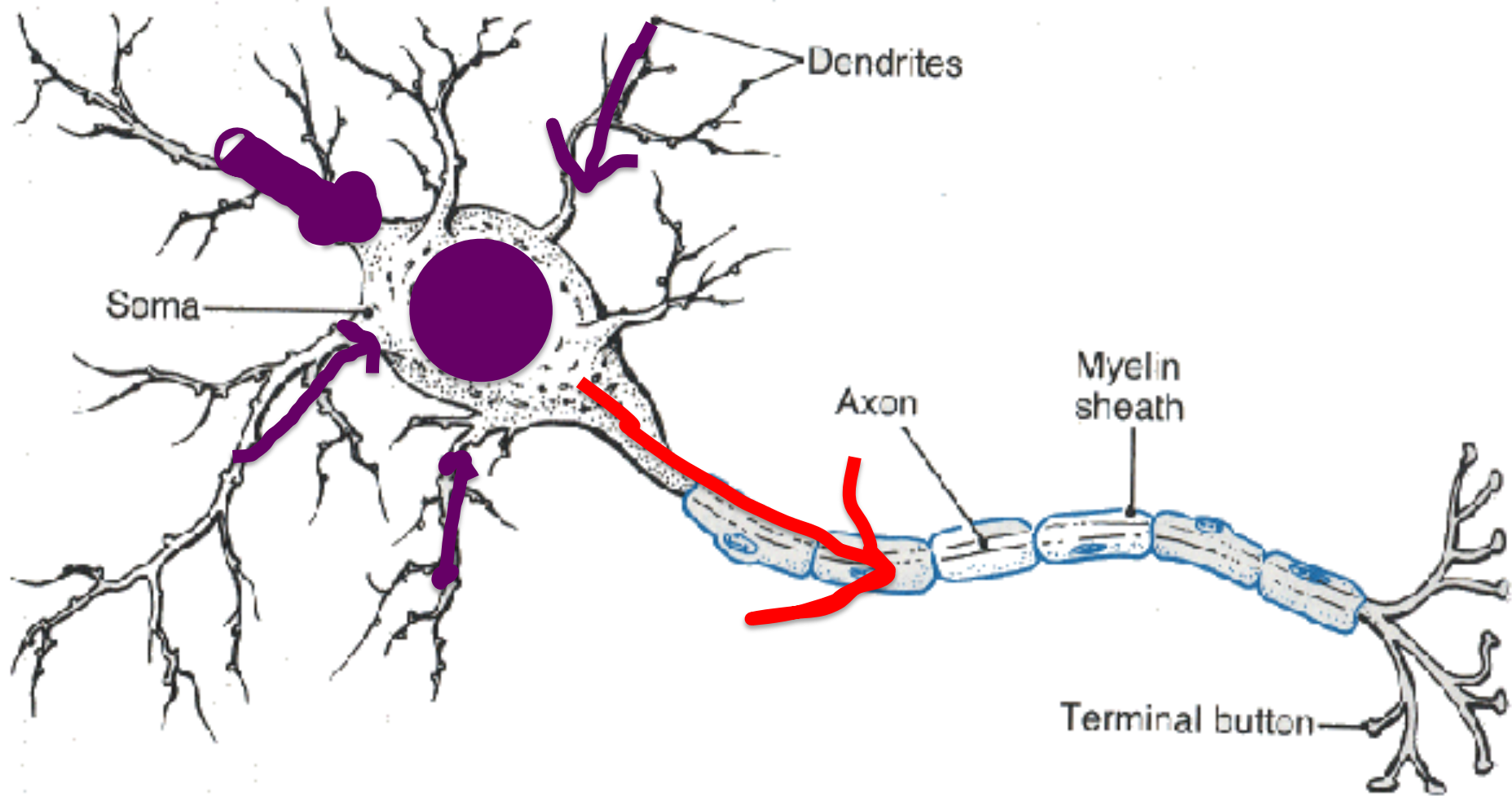
Neuron



Neuron



Some Inputs are More Important



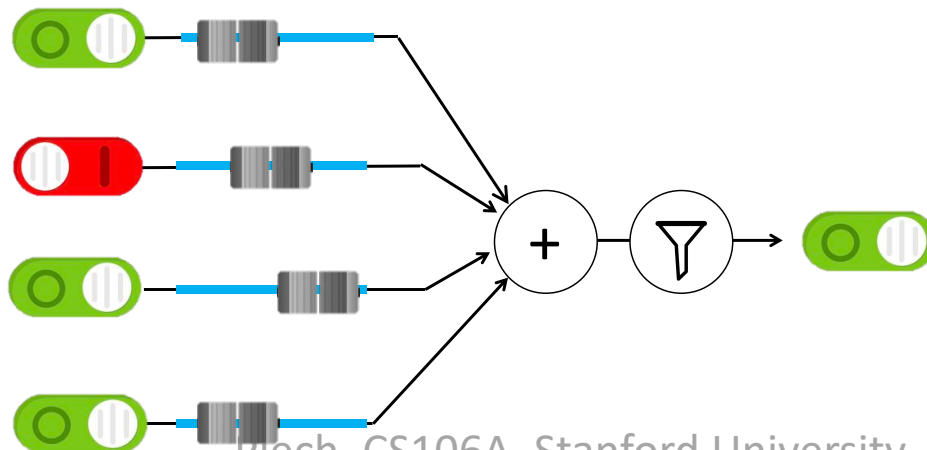
Artificial Neuron

calculate the activation of a neuron

```
def activate(weights_list, inputs_list):  
    weighted_sum = 0;  
    for i in range(len(inputs_list)):  
        weighted_sum += weights_list[i] * inputs_list[i]  
  
    return squash(weighted_sum);
```

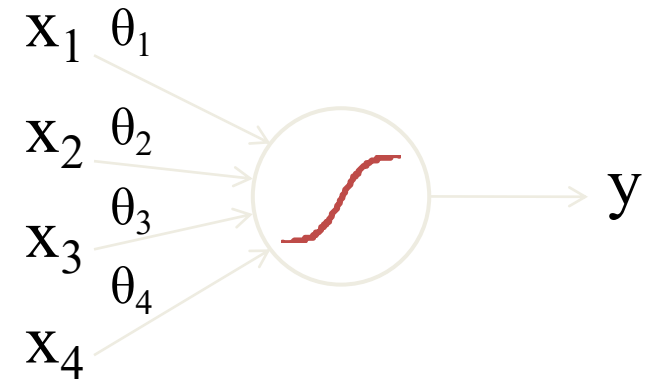
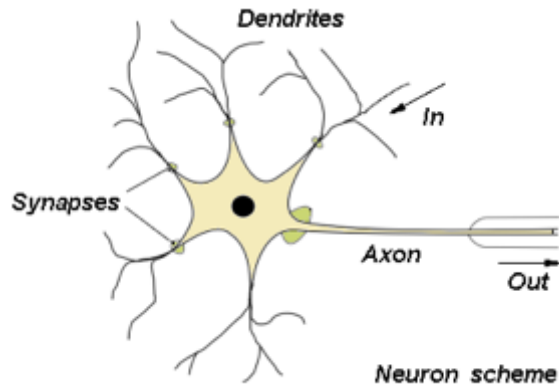
the sigmoid function forces a value to be between 0 and 1

```
def squash(value):  
    return 1 / (1 + math.exp(-value));
```

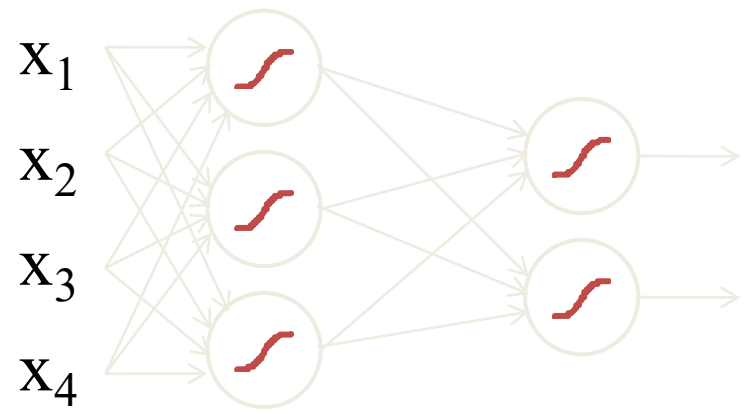
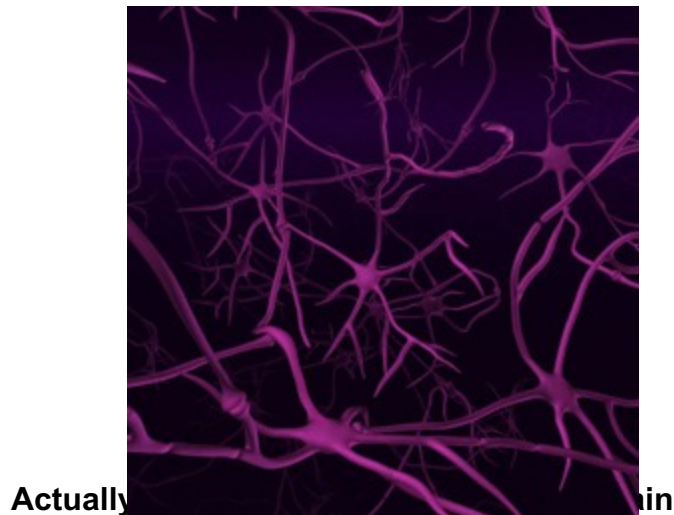


Biological Basis for Neural Networks

- A neuron



- Your brain



Demonstration

Draw your number here



Downsampled drawing:

First guess:

Second guess:

Layer visibility

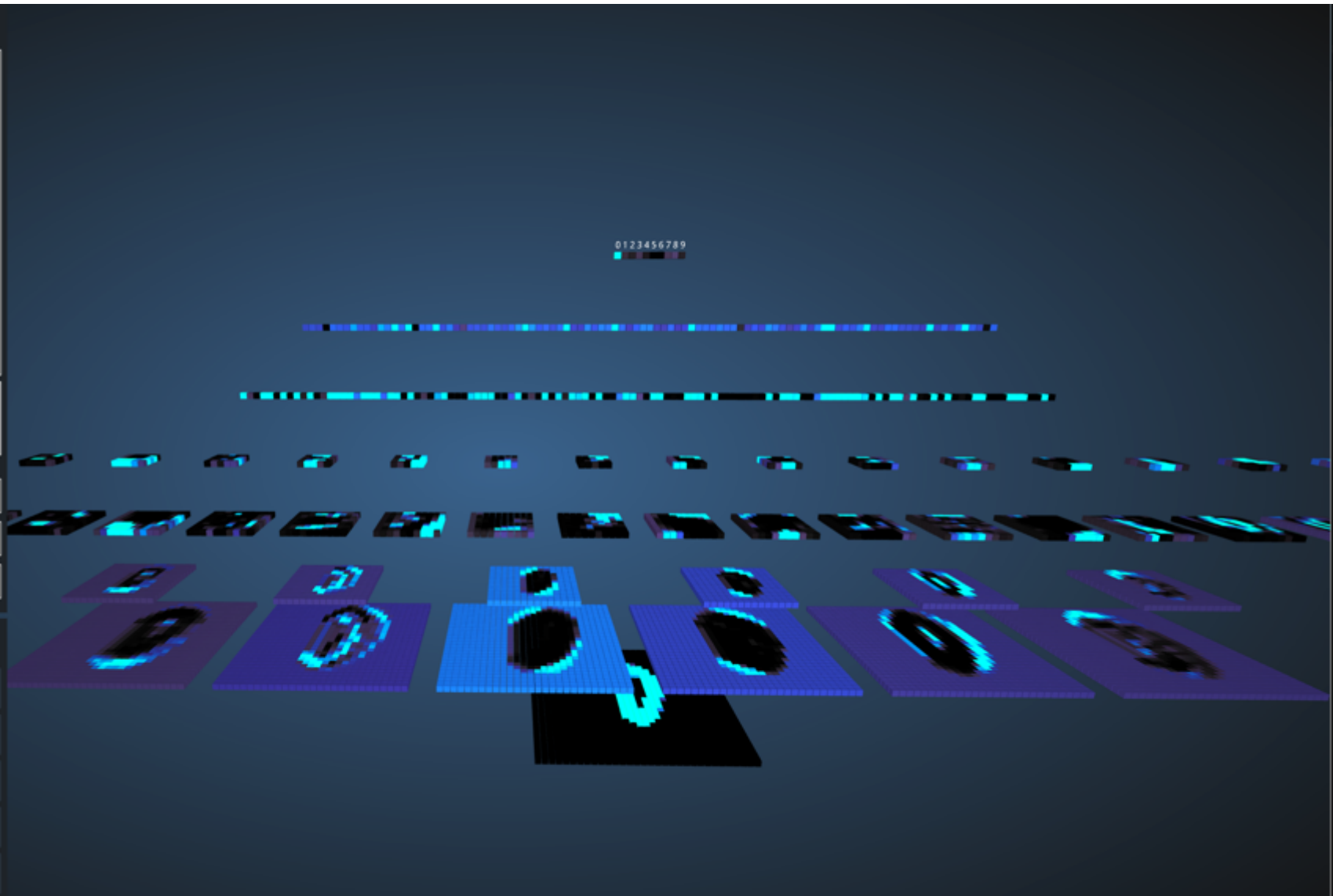
Input layer

Convolution layer 1

Downsampling layer 1

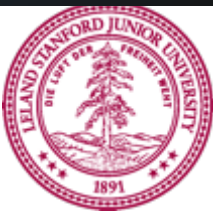
Convolution layer 2

Downsampling layer 2



<http://scs.ryerson.ca/~aharley/vis/conv/>

Piech, CS106A, Stanford University



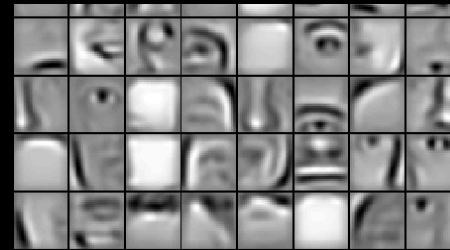
Visualize the Weights



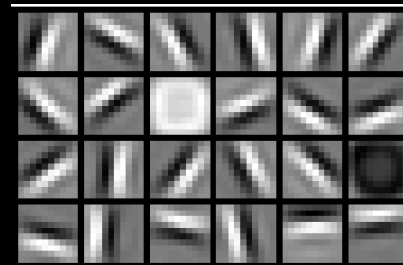
Training set: Aligned images of faces.



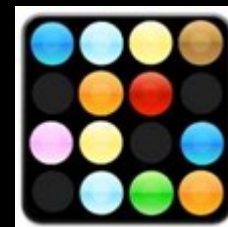
object models



object parts
(combination
of edges)

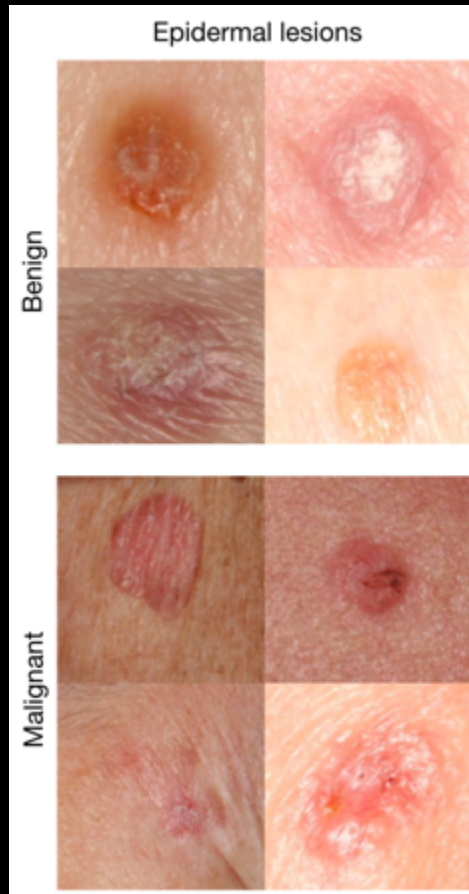


edges



pixels

Where is this useful?

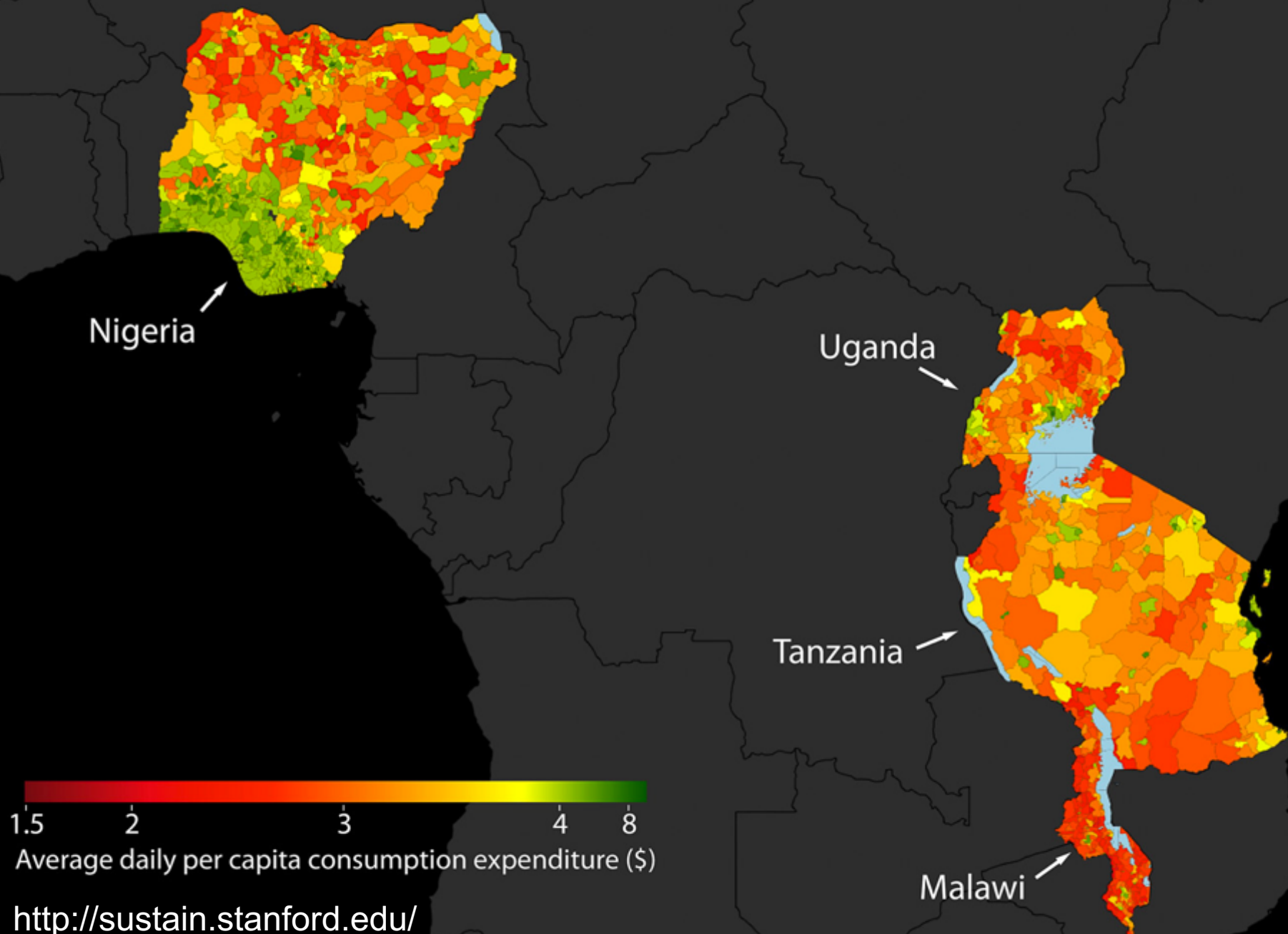


A machine learning algorithm performs **better than** the best dermatologists.

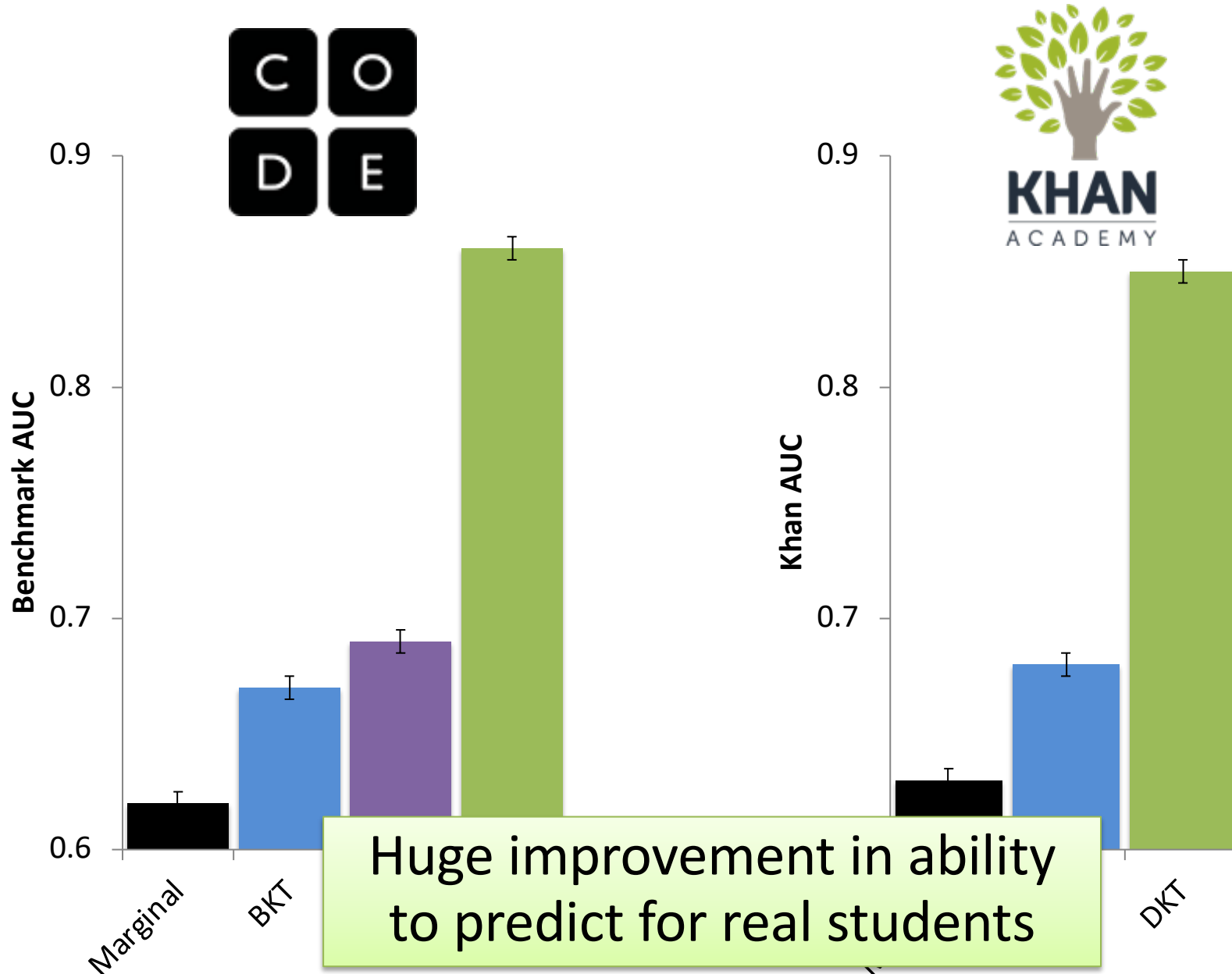
Developed this year, at Stanford.

Esteva, Andre, et al. "Dermatologist-level classification of skin cancer with deep neural networks." *Nature* 542.7639 (2017): 115-118.

Estimated daily per capita expenditure, 2012-2015



Understanding Students



What does it look like?

```
deep_learning.py x
12
13 import torch.nn as nn
14 from torch.optim import Adam
15 from torch.optim import SGD
16 import torch
17 import torch.nn.functional as F
18 import math
19 import pickle
20 import numpy as np
21 import csv
22
23 N_COUNTRIES = 329
24
25 def main():
26     # 1. get some data (either simulate or load)
27     data = load_NWEA_data()
28     # 2. build a pytorch model
29     model = DeepLearningModel()
30     # 3. run optimization
31     optimize(model, data)
32
33 # sometimes its more fun to load data...
34 def loadData():
35     return pickle.load(open('data.pkl', 'rb'))
36
37 # Define the model you are optimizing over...
38 # Written as a pytorch model so we can optimize
39 class DeepLearningModel(nn.Module):
40     # the initialize method
41     def __init__(self):
42         super().__init__() # necessary
43
44         # learning rates for each of the countries
45         self.theta = nn.Parameter(torch.ones(N_COUNTRIES))
46
47         # the global parameters for the function. In the writeup
48         # I call these phis (but in code I give them names so I
49         # dont get confused).
50         self.offset_param = nn.Parameter(torch.ones(1) * 150.0)
51         self.scale_param = nn.Parameter(torch.ones(1) * 1500.0)
52         self.amplitude_param = nn.Parameter(torch.ones(1) * 9.0)
53         self.floor_param = nn.Parameter(torch.ones(1) * -1)
54
55         # This is the parametric-family written in pytorch code.
56         # See the writeup for details. Note how we use one-hot
57         # vectors to select the theta from a student's country.
58         # The input to this function is all your data as matrices
59     def forward(self, alpha, country_one_hot):
60         # n x 1 vector which has the country theta selected for
61         # each student in the batch. (basically chose a theta from
62         # the country thetas for each student)
63         ability = torch.matmul(country_one_hot, self.theta).unsqueeze(1)
64
65         # The normal distribution exponential
```

