



Beyond CS106A

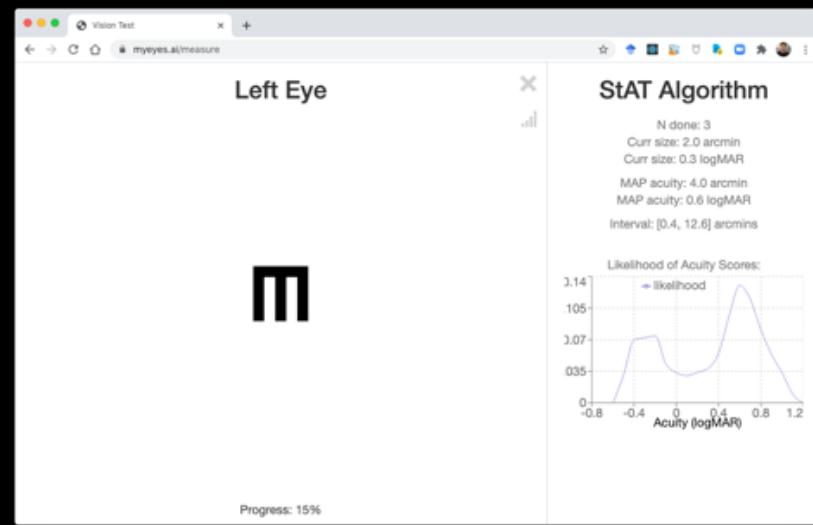
Chris Piech and Mehran Sahami
CS106A, Stanford University

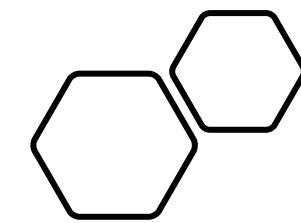
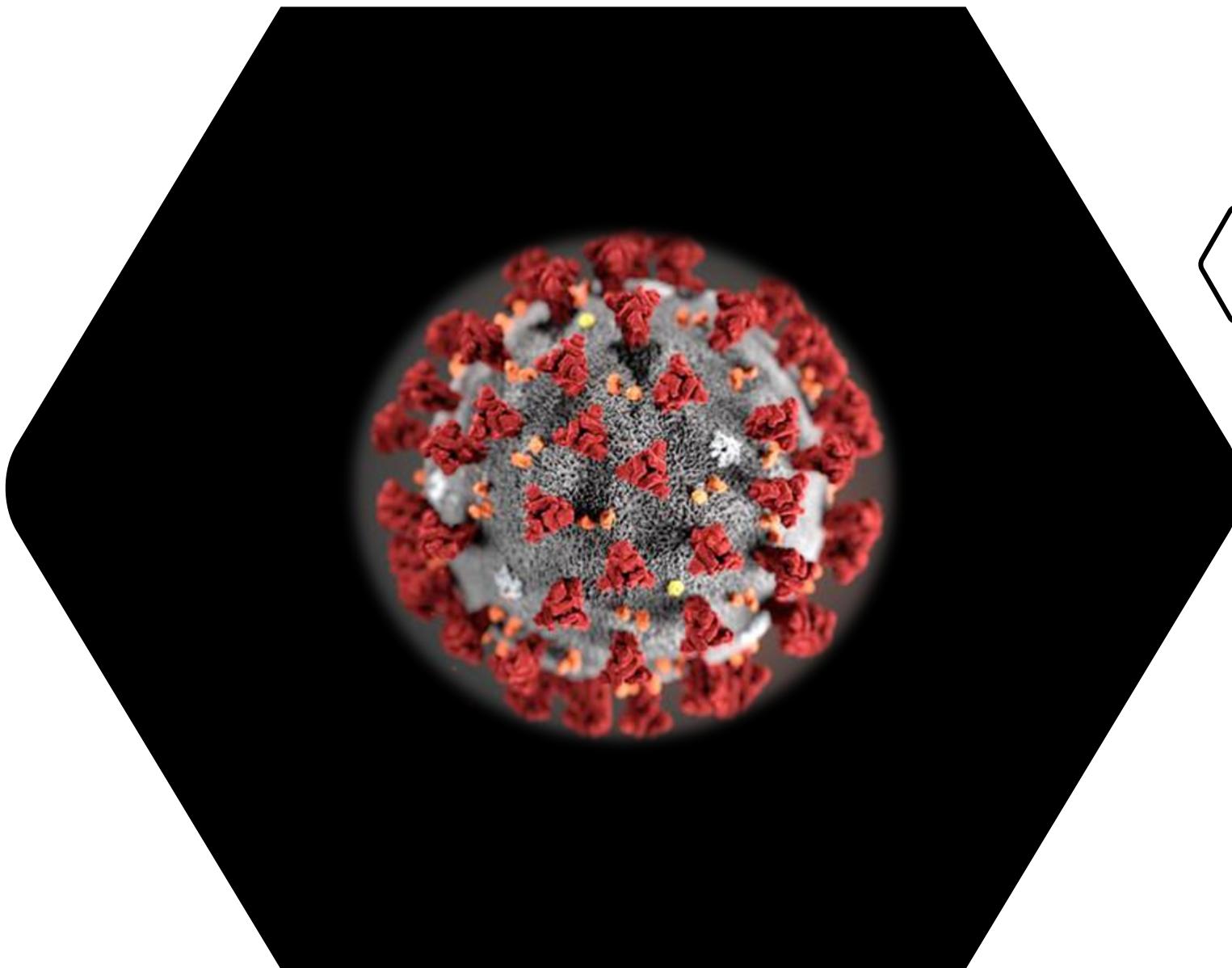
You all have your own **identity**
and goals.

CS is a versatile tool.

Life beyond CS106A?
Learn more so you can
express yourself.

I should be blind!



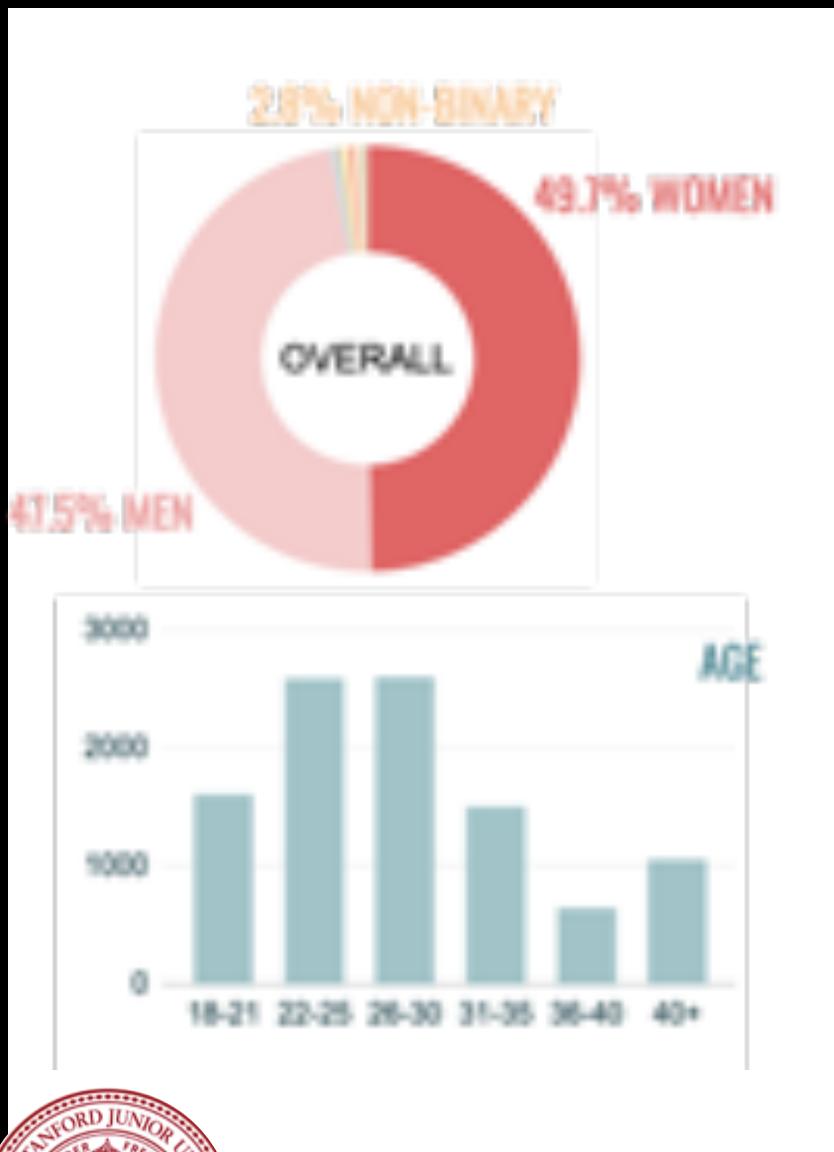


Piech and Sahami, Code in Place



Code in Place : Course with the most section leaders?

**1,000 volunteer section leaders teach
10,000 students**
First half of Stanford CS106A



20% experienced job loss or home loss
10x retention vs baseline MOOC
6k hours of live teaching
60k hours of lecture watched
Better CS106A for Stanford students





The magnitude of people
who **want to teach** is
roughly proportional to the
magnitude of people who
want to learn.

Teaching is **learning**



Teaching is **joyful**

The education community has barely scratched
the surface of the **potential** in this claim.



Some of our best teachers
were students who had
only taken CS106A

Have you thought about teaching?

Great way to learn
Great way to give back

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

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2. What other languages look like
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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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    if(i % 2 == 0) {
        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, ...]



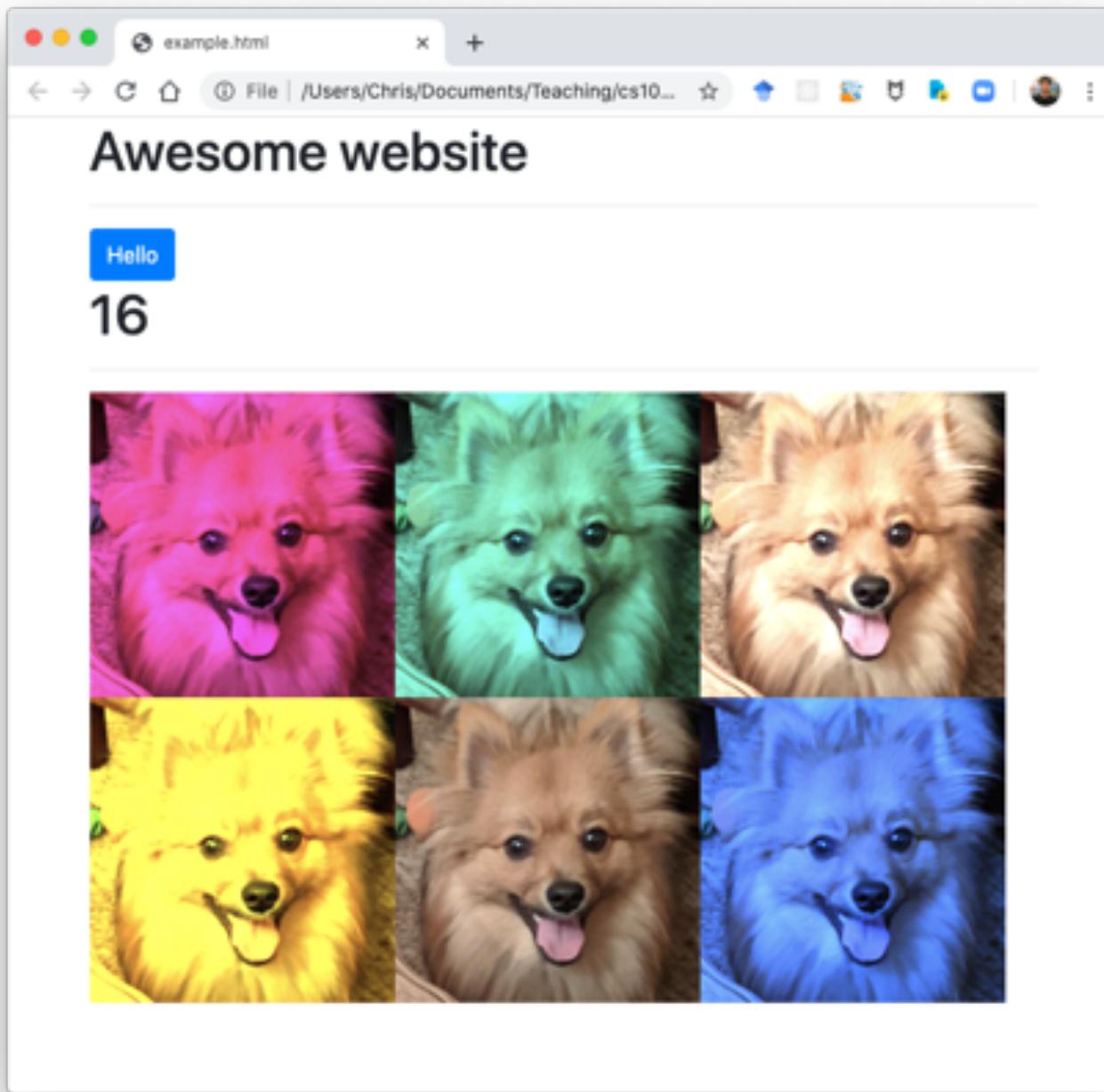
Javascript

```
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

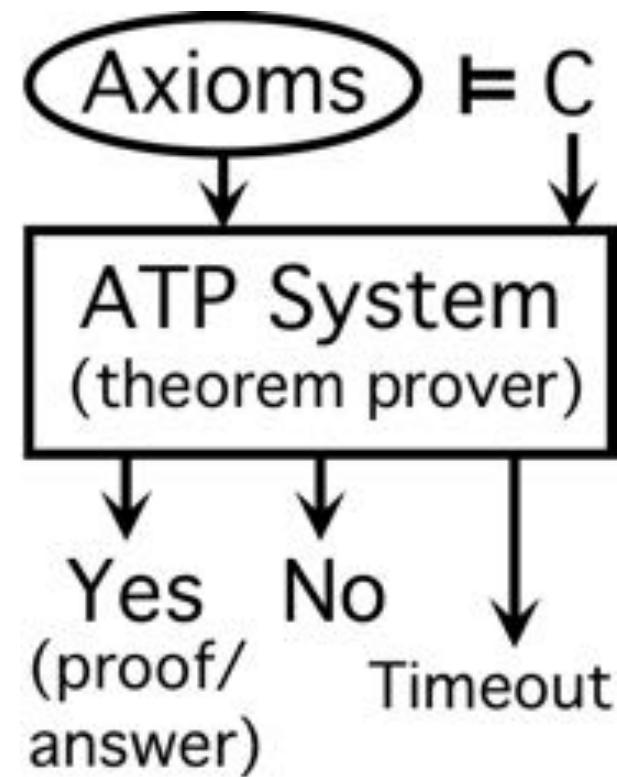


Early Optimism 1950

1952



1955



Computer Vision



Piech + Sahami, 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
```



Identifying Cats

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

```
def is_cat(image):
    if not contains_two_eyes(image):
        return False
    if not has_whiskers(image):
        return False
    if not has_pointy_ears(image):
        return False
    return True
```



Some Tricky Cases

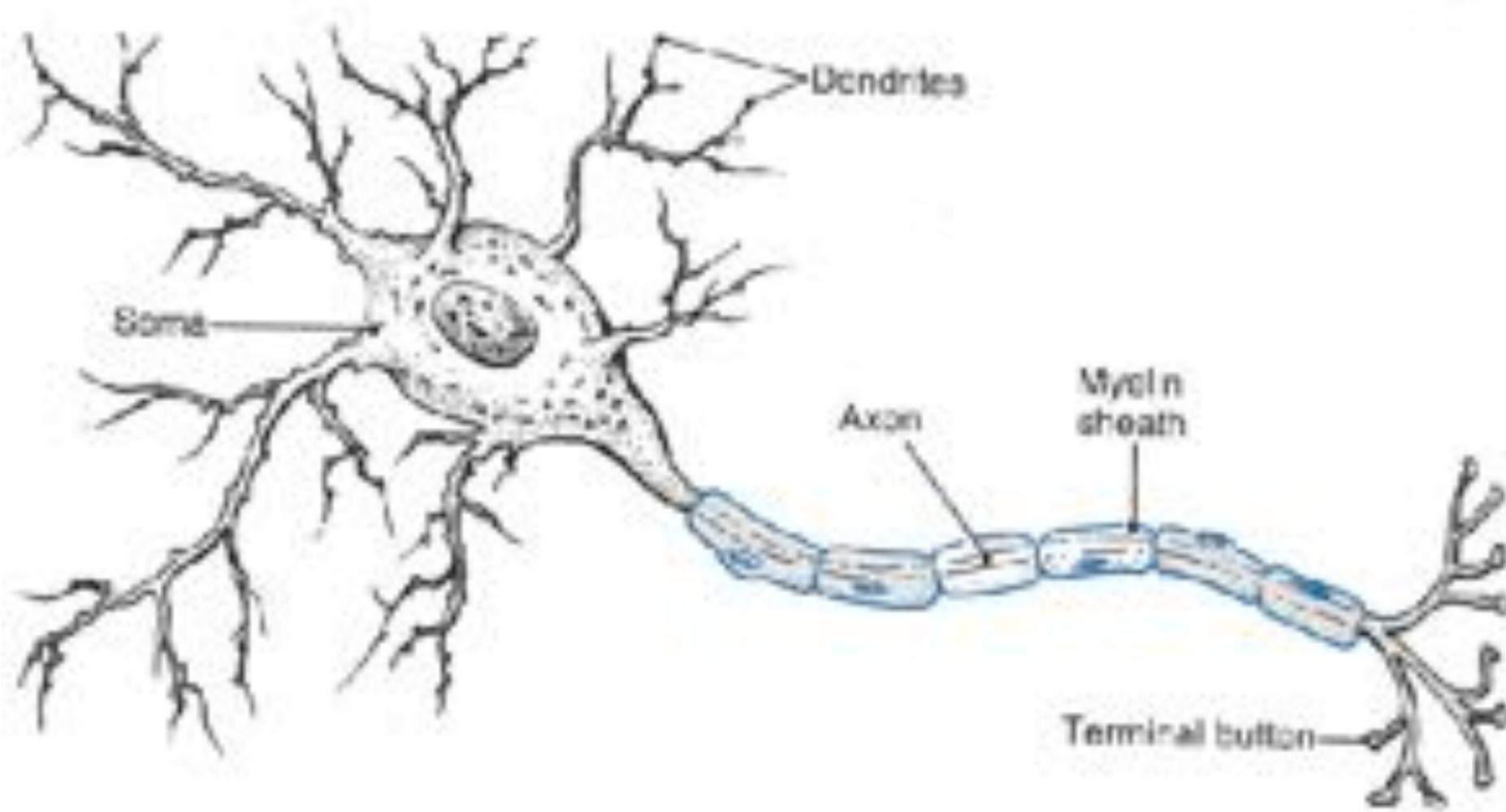


Megan Sarami, CS183A, Stanford University

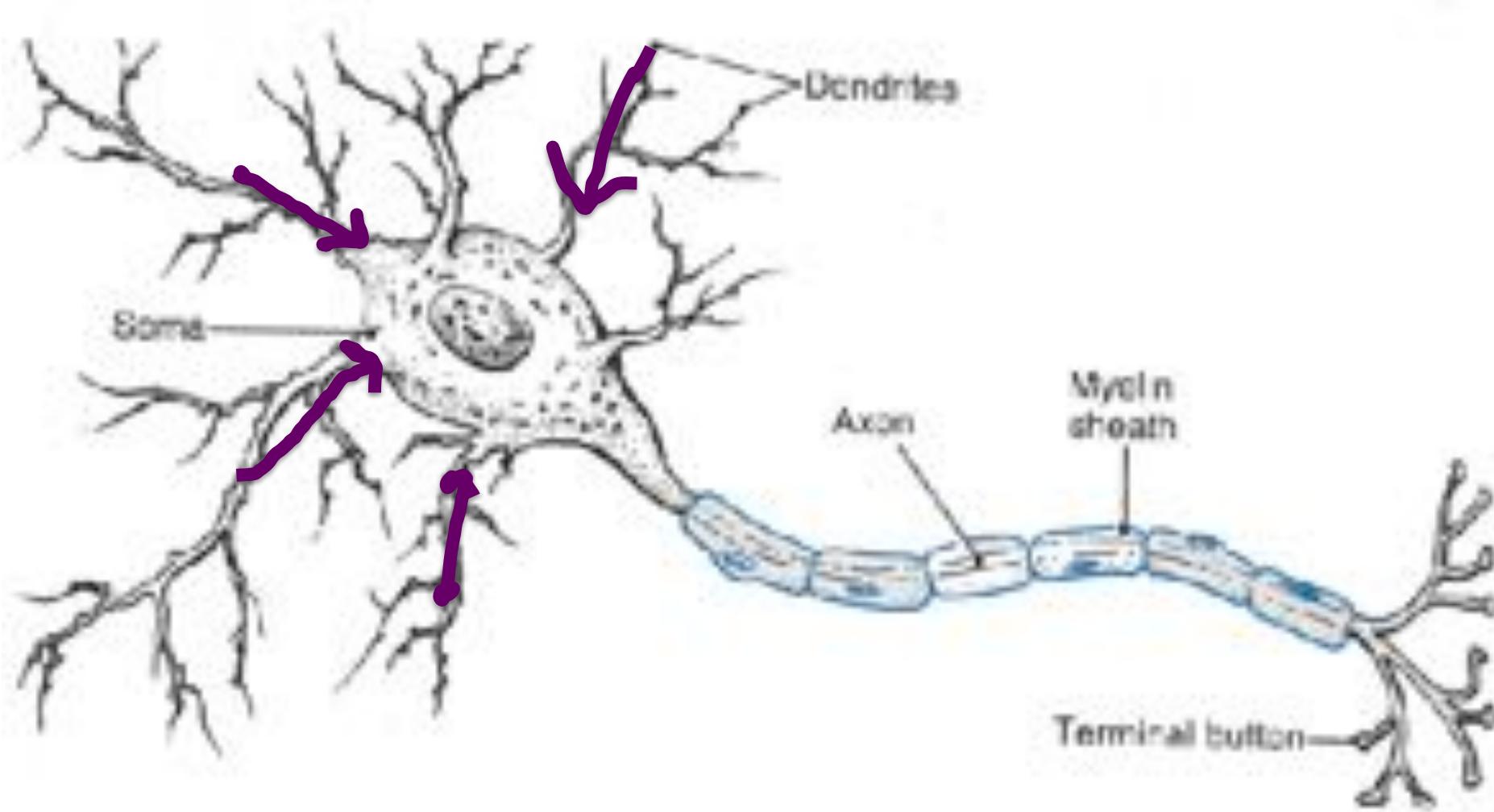


Great idea inspired by biology

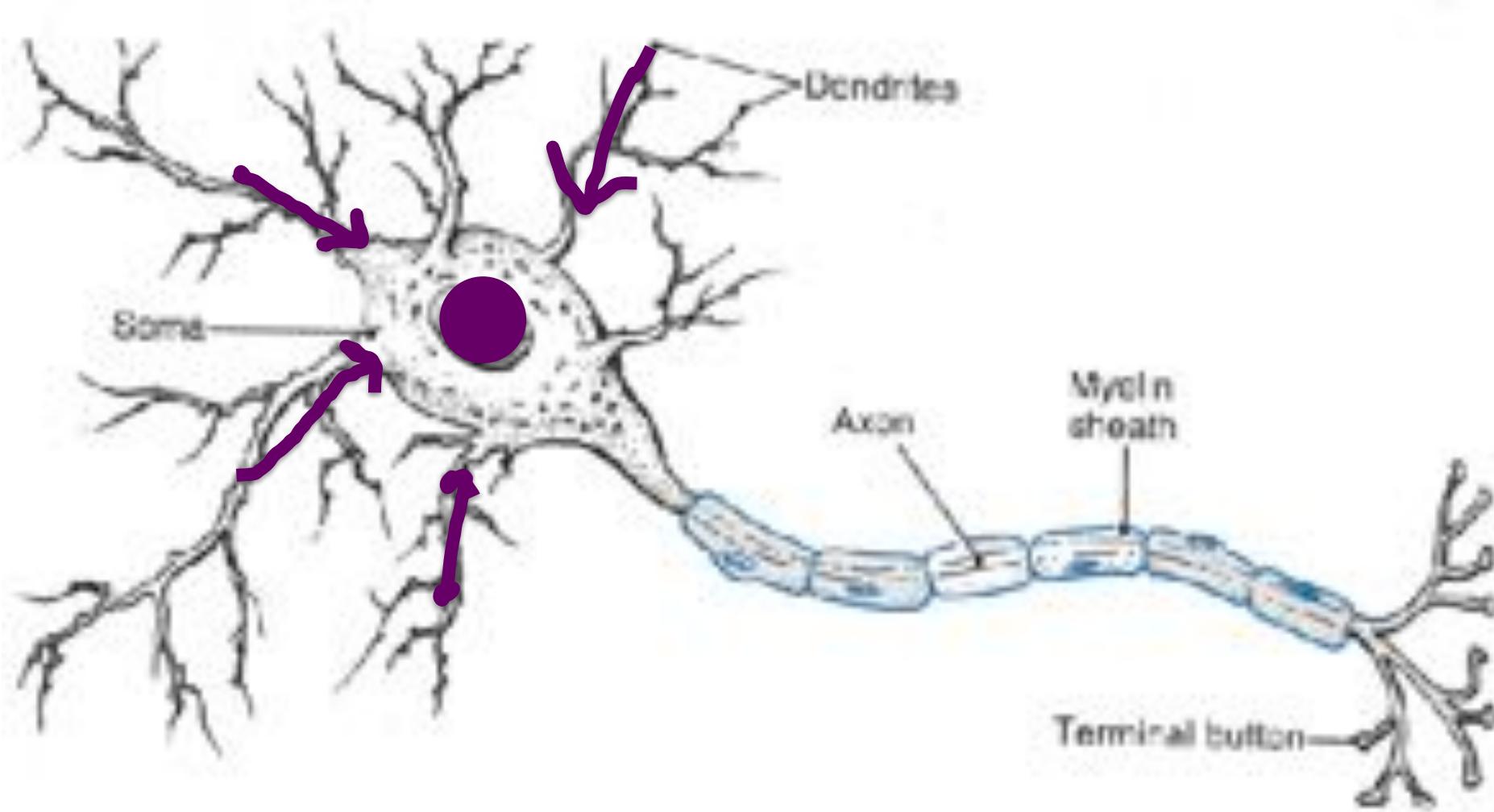
Neuron



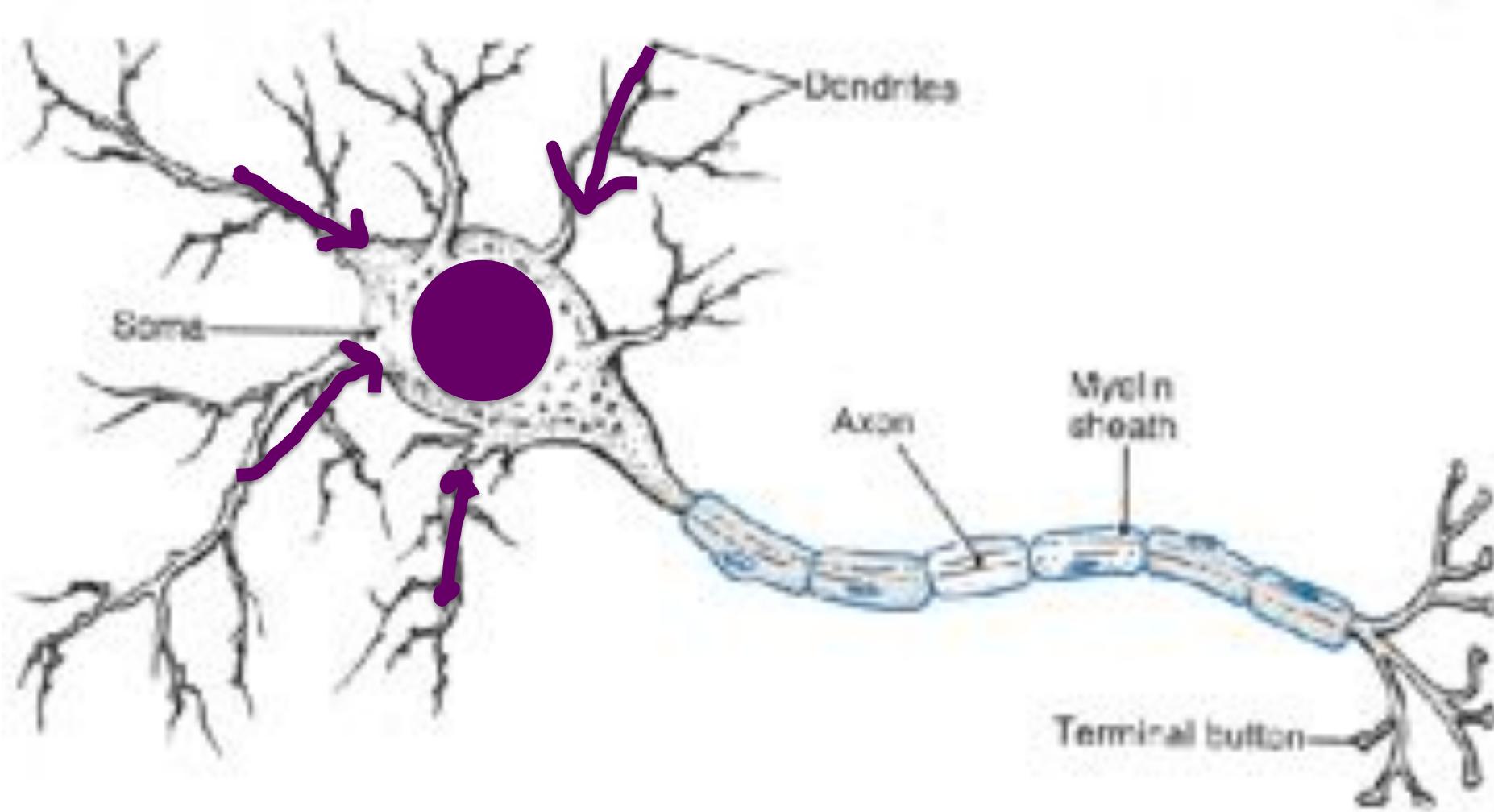
Neuron



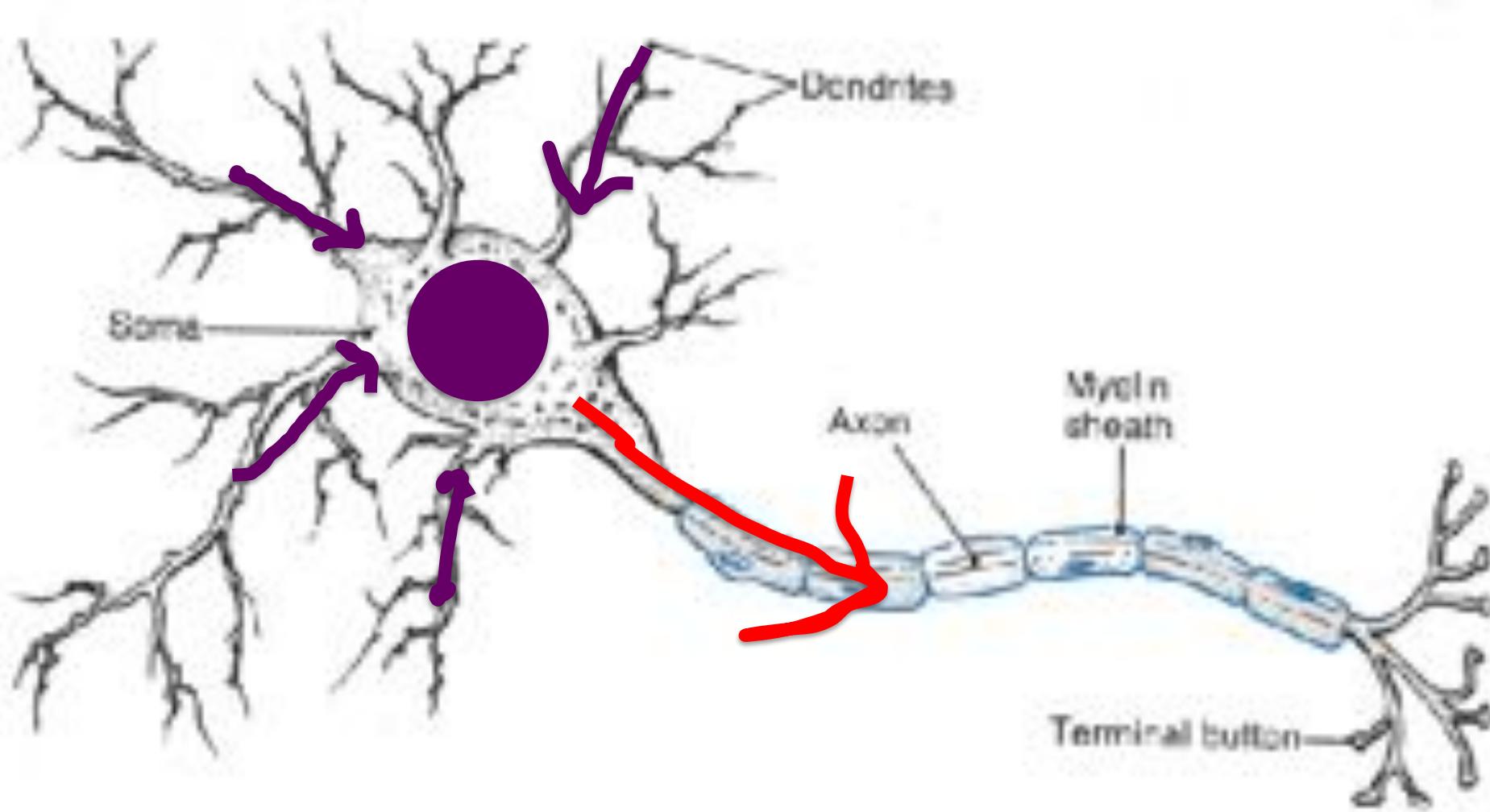
Neuron



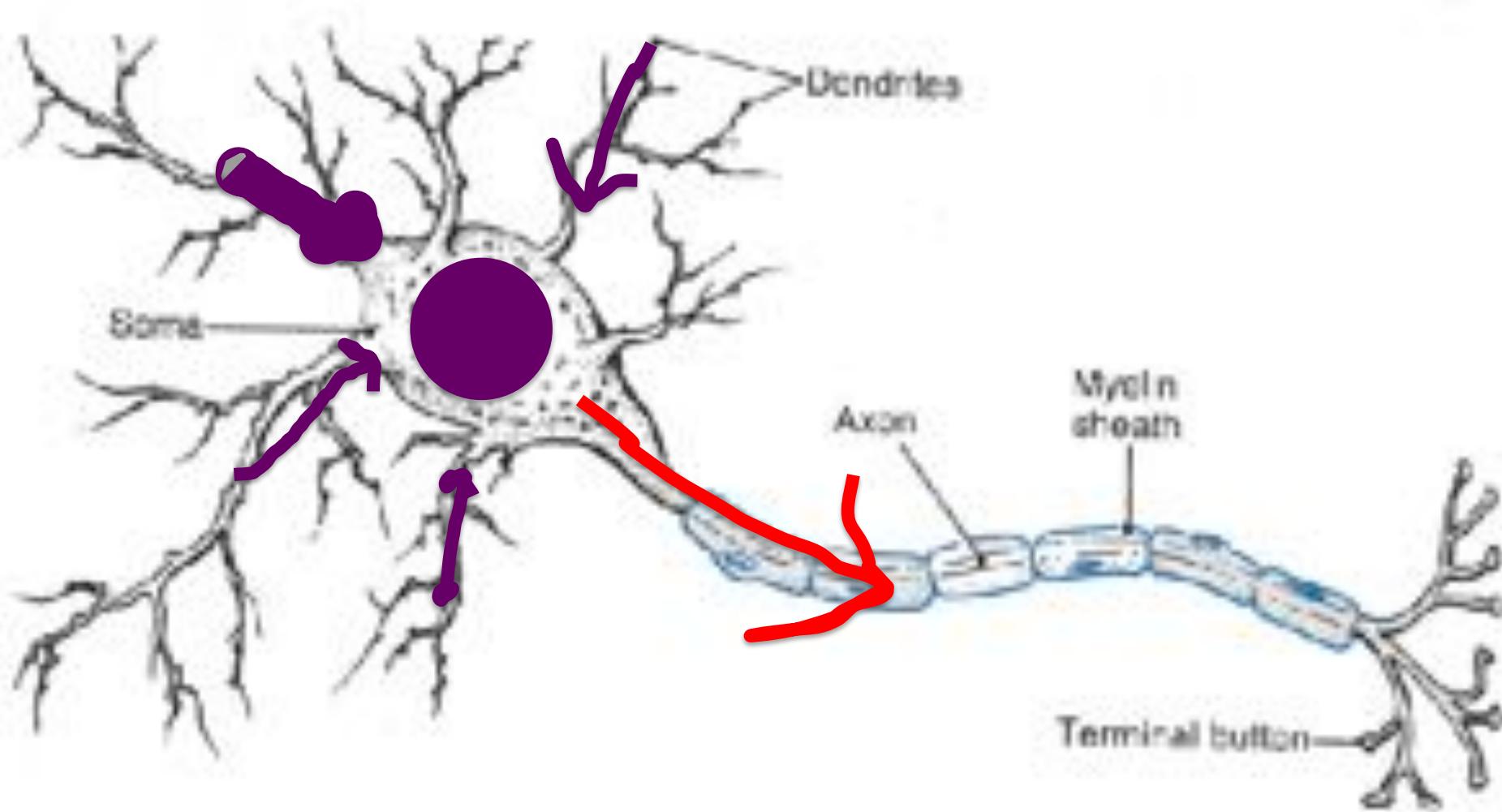
Neuron



Neuron



Some Inputs are More Important

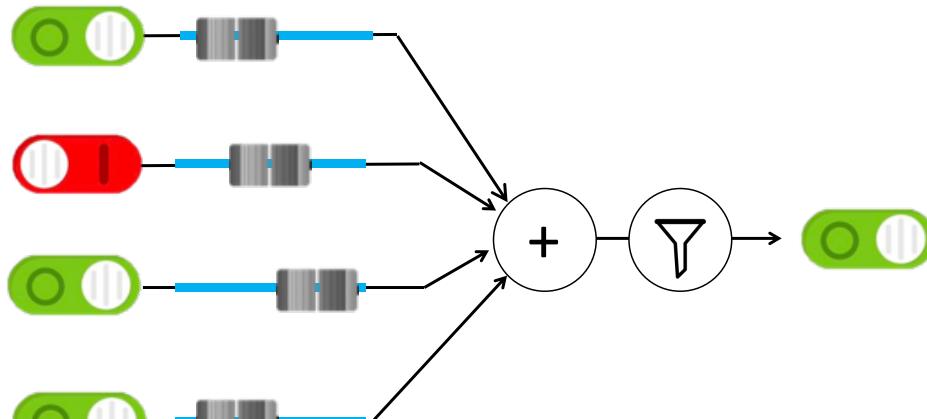


Artificial Neuron

```
# calculate the activation of a neuron
def activate(weights_list, inputs_list):
    n = len(inputs_list)
    weighted_sum = 0
    for i in range(n):
        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));
```



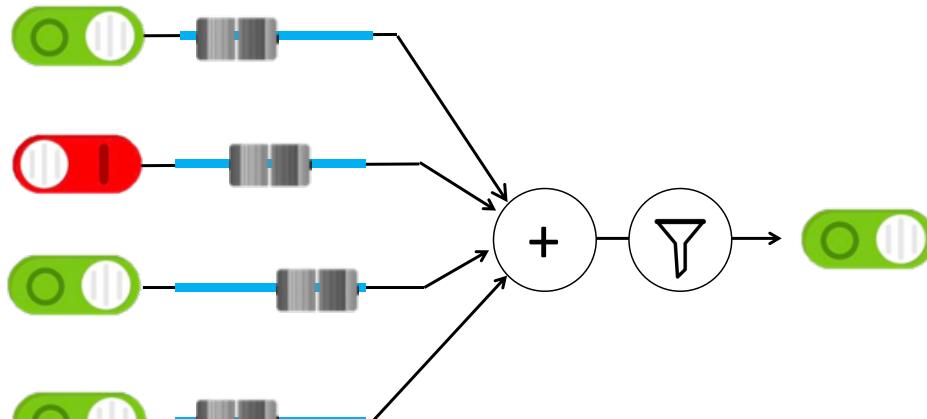
Piech + Sahami, CS106A, Stanford University



Artificial Neuron

```
# calculate the activation of a neuron
def activate(weights_list, inputs_list):
    n = len(inputs_list)
    # using list comprehensions like Juliette showed us
    weighted = [weights_list[i] * inputs_list[i] for i in range(n)]
    weighted_sum = sum(weighted)
    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));
```

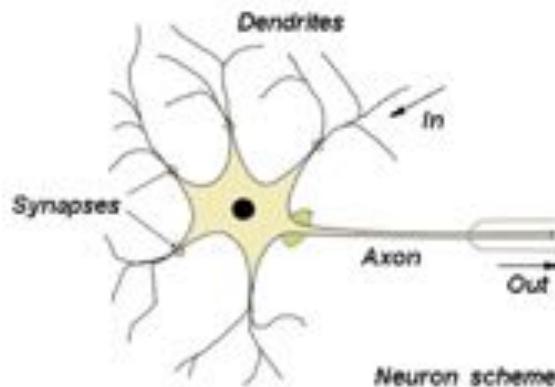


Piech + Sahami, CS106A, Stanford University

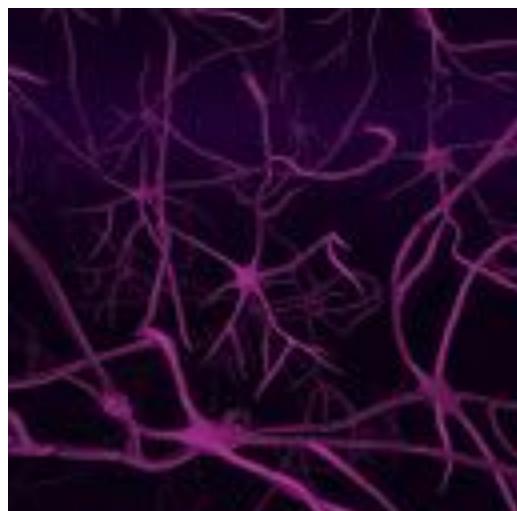


Biological Basis for Neural Networks

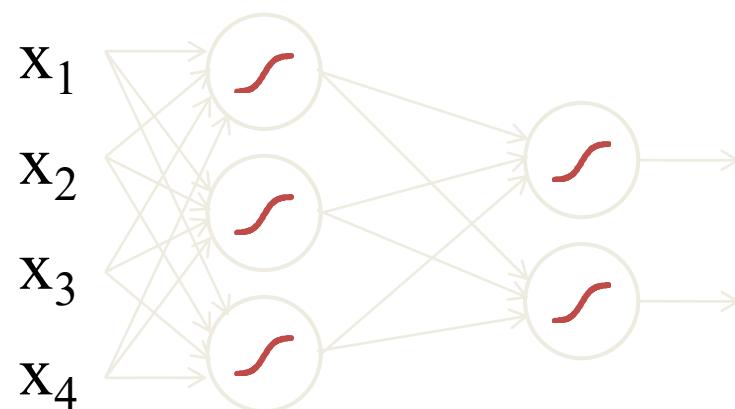
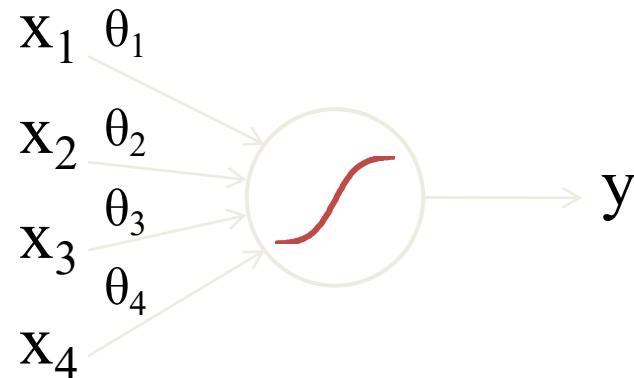
- A neuron



- Your brain



Actually, it's probably someone else's brain



Demonstration

Draw your number here



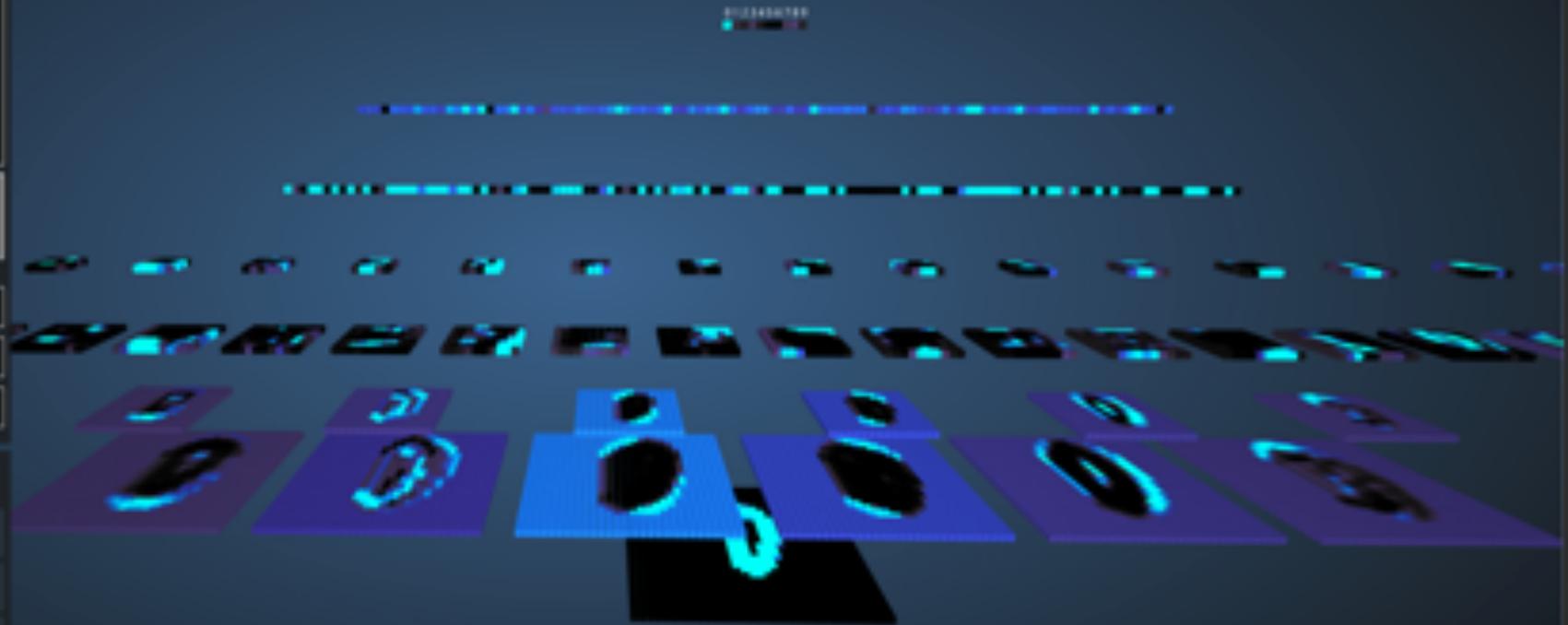
Downsampled drawing:

First guess:

Second guess:

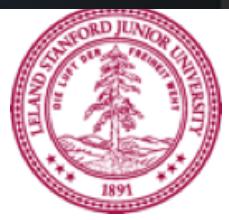
Layer visibility

Input layer	Show
Convolution layer 1	Show
Downsampling layer 1	Show
Convolution layer 2	Show
Downsampling layer 2	Show



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

Piech + Sahami, CS106A, Stanford University



Visualize the Weights



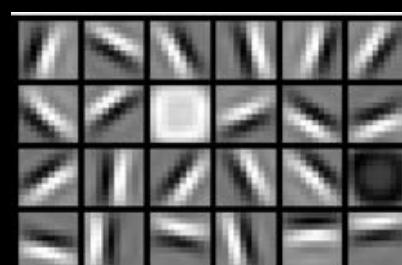
Training set: Aligned images of faces.



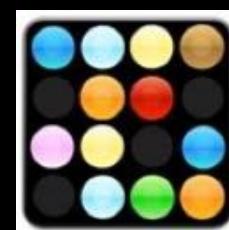
object models



object parts
(combination
of edges)

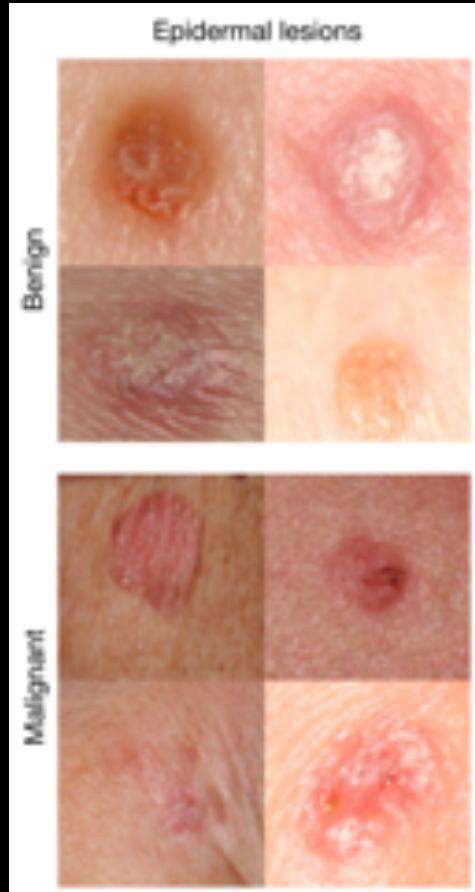


edges



pixels

Where is this useful?

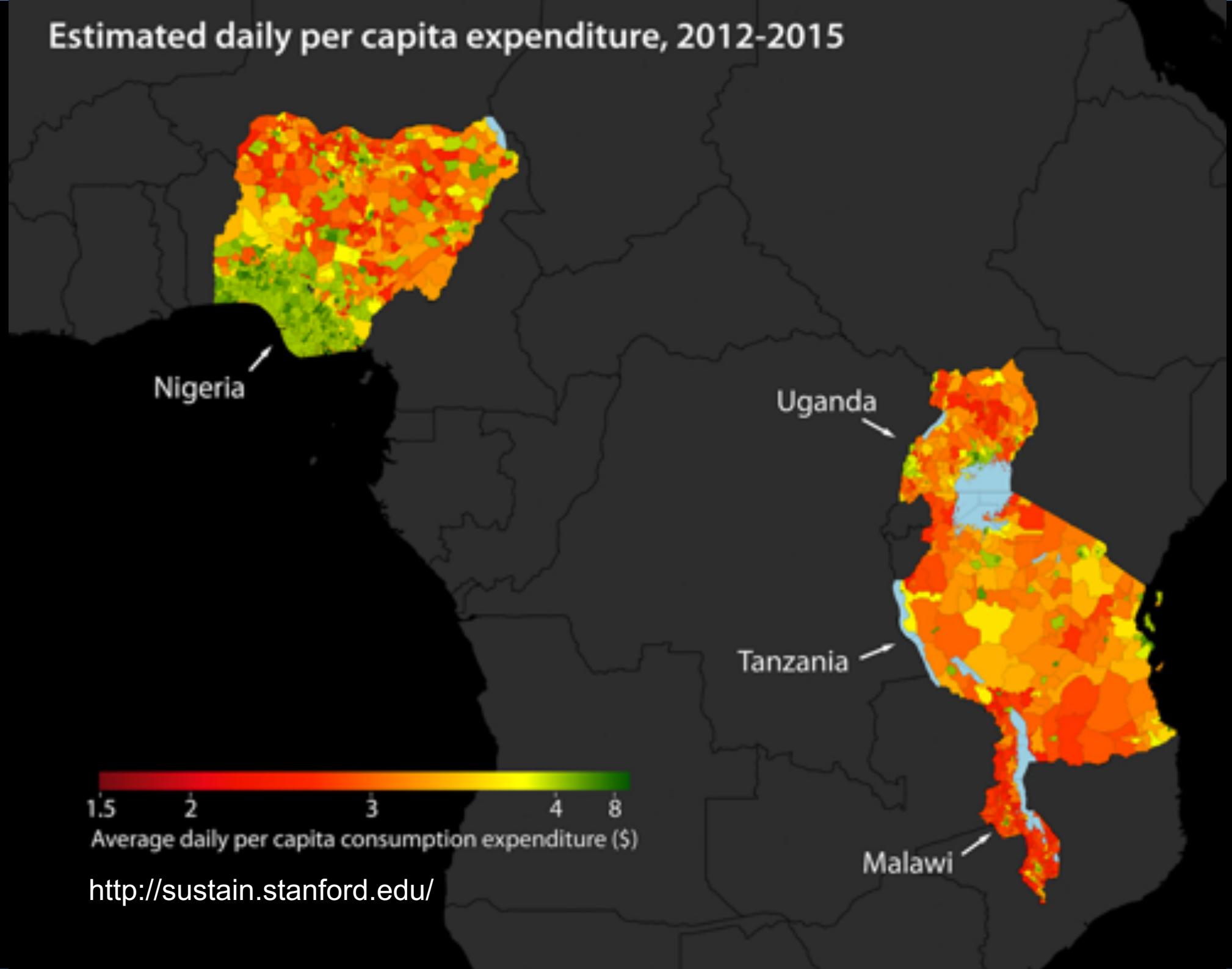


A machine learning algorithm performs **better than** the best dermatologists.

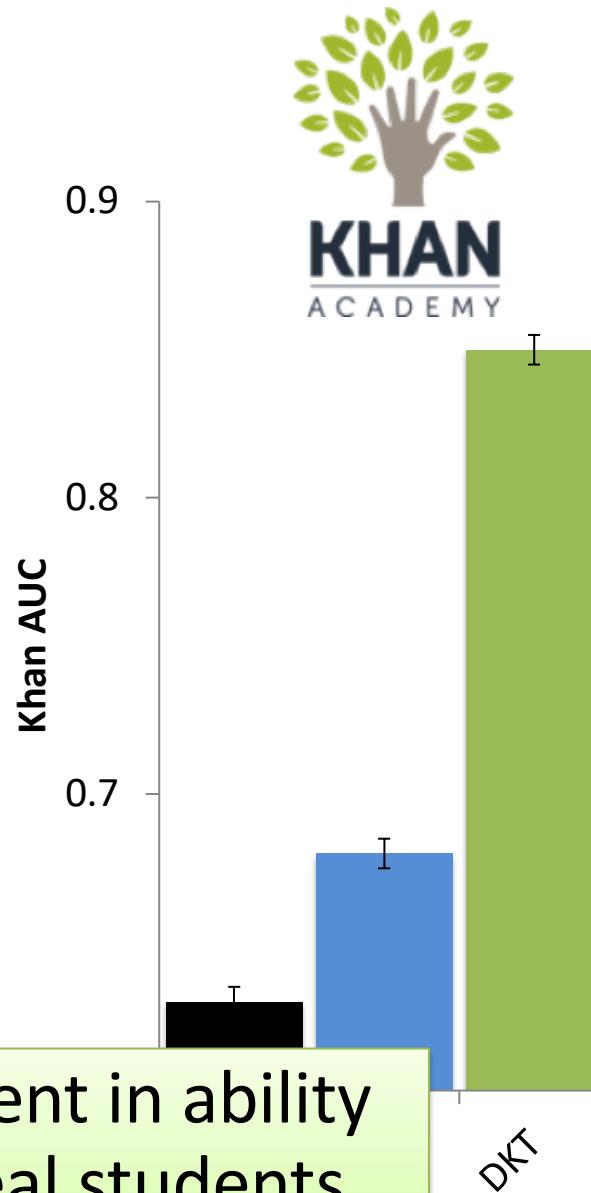
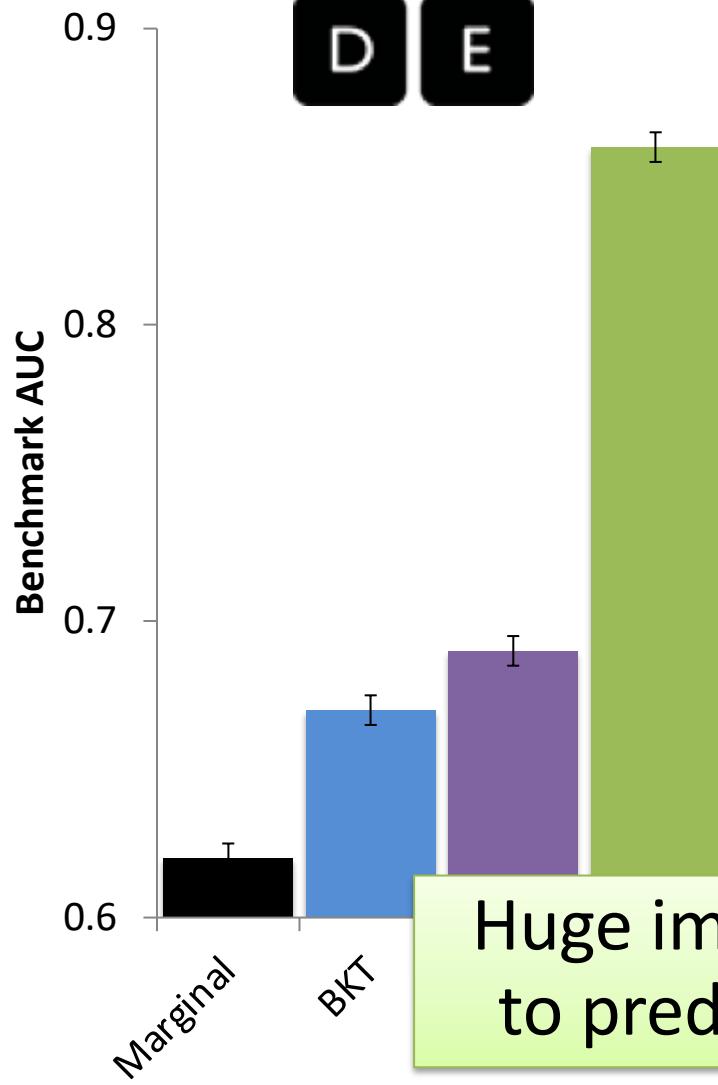
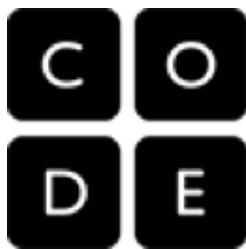
Developed this year, at Stanford.

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



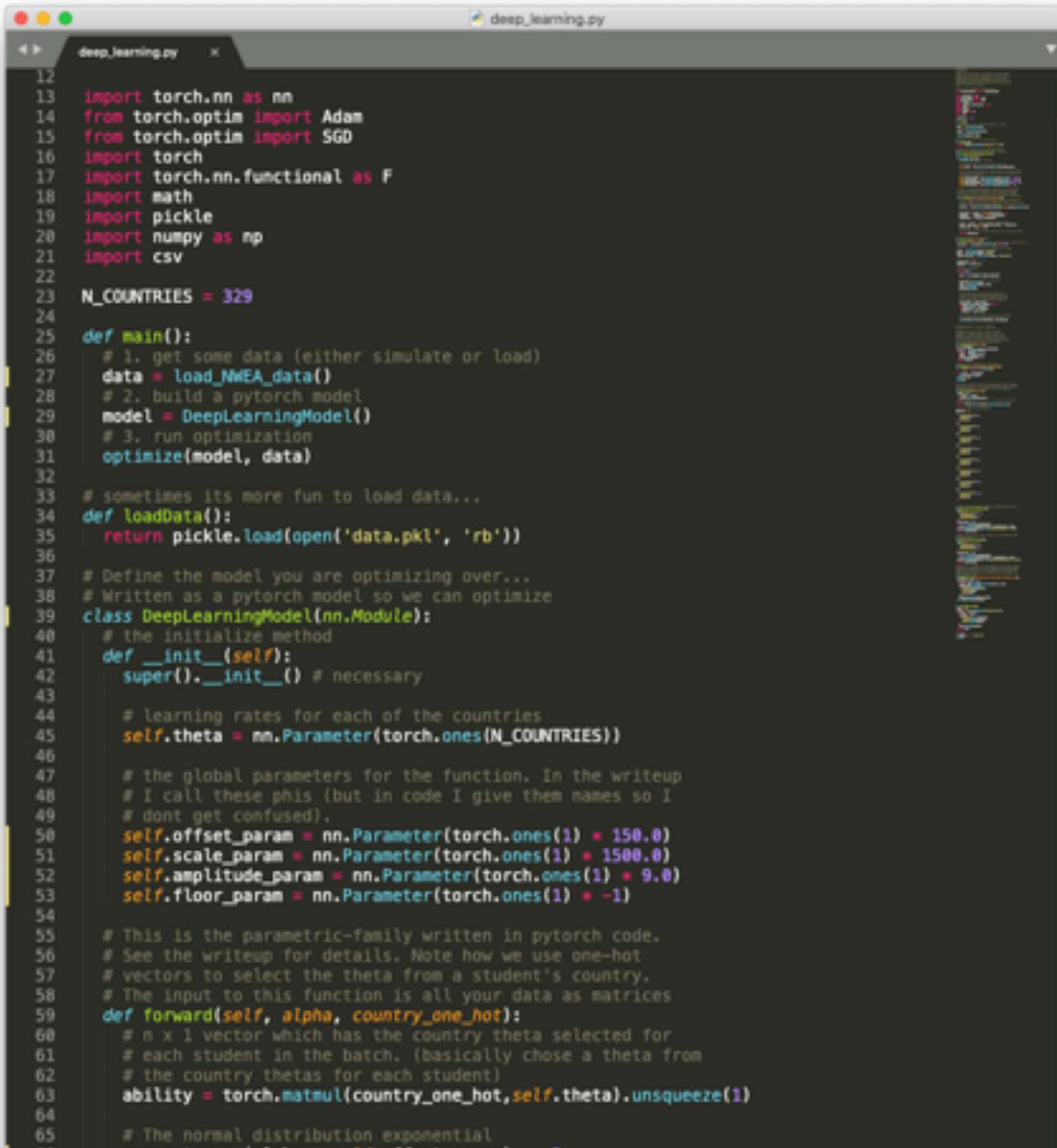
Understanding Students



Huge improvement in ability
to predict for real students

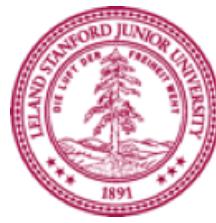


What does it look like?



A screenshot of a Jupyter Notebook cell titled "deep_learning.py". The code is written in Python and uses PyTorch libraries. It defines a main function that loads data, builds a PyTorch model, and runs optimization. The model is a DeepLearningModel that initializes parameters and defines a forward method. The code includes comments explaining the purpose of various variables and methods.

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



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

Chris:
CS109

Mehran:
CS182

What's life after CS106A going to
be like for you and Mehran?

