



Beyond CS106A

Chris Piech
CS106A, Stanford University



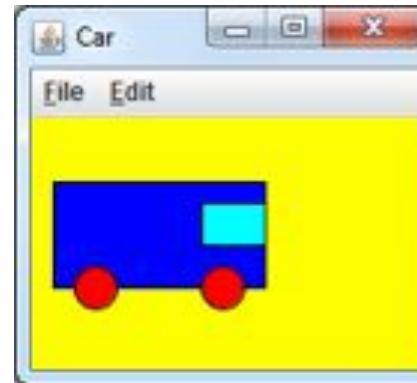
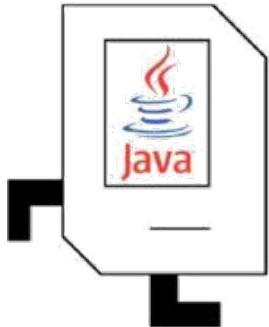
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Life after ACM

Life After The ACM Libraries

- All quarter we have relied on the **ACM Java libraries**.
 - Karel, ConsoleProgram, RandomGenerator
 - GraphicsProgram, GOval, GRect, GOval, GLine, GImage, ...



- Today we will see how **standard Java** programs are made.



Using the ACM Libraries

```
import acm.program.*;

public class MyProgram extends ConsoleProgram {
    public void run() {
        println("Hello, world!");
    }
}
```

- This is a console program written using the ACM libraries.
 - It uses the **ConsoleProgram** class to represent a console.
 - The **run** method contains the program code.
 - The **println** method prints output to the graphical console.



A Barebones Java Program

```
public class Hello {  
    public static void main(String[] args) {  
        System.out.println("Hello, world!");  
    }  
}
```

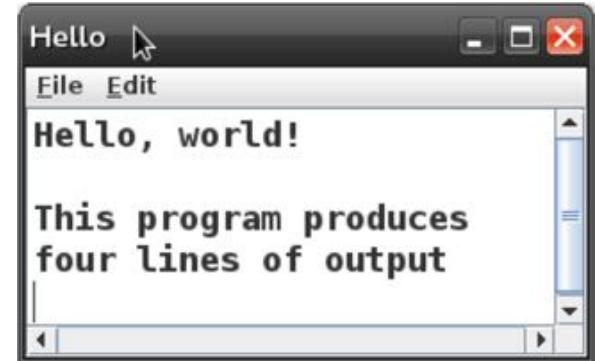
- The method **main** is the true entry point for a Java program.
 - It must have the exact heading shown above.
 - The **String[] args** are "command line arguments" (ignored).
 - The **println** command's true name is **System.out.println**.
 - Standard Java methods are **static** unless part of a class of objects.



Lets make Hello World!

Console Programs

- What does the **ConsoleProgram** library class do?
 - Creates a new graphical **window**
 - Puts a scrollable **text area** into it
 - Provides **print** and **println** commands to send text **output** to that window
 - contains a **main method** that calls your program class's **run** method
 - **ConsoleProgram**'s **run** is empty, but you extend and override it



```
public class Hello extends ConsoleProgram {  
    public void run() {  
        println("Hello, world!");  
    }  
}
```



ACM console input

```
public class Age extends ConsoleProgram {  
    public void run() {  
        String name = readLine("What's your name? ");  
        int age = readInt("How old are you? ");  
        int years = 65 - age;  
        println(name + " has " + years  
                + " years until retirement!");  
    }  
}
```

- The ACM library has simple console input commands like `readLine`, `readInt`, `readDouble`, and so on.
- These methods display a 'prompt' message, wait for input, re-prompt if the user types a bad value, and return the input.



Java console input

```
public class Age {  
    public static void main(String[] args) {  
        Scanner console = new Scanner(System.in);  
        System.out.print("What's your name? ");  
        String name = console.nextLine();  
        System.out.print("How old are you? ");  
        int age = console.nextInt();  
        int years = 65 - age;  
        System.out.println(name + " has " + years  
                           + " years until retirement!");  
    }  
}
```

- In standard Java, you must create a Scanner or similar object to read input from the console, which is also called `System.in`.
 - It does not automatically re-prompt and can crash on bad input.



Graphics Programs

The ACM library does several things to make graphics easier:

- Automatically creates and displays a **window** on the screen.
 - In standard Java, we must do this ourselves; it is called a JFrame.
- Sets up a **drawing canvas** in the center of the window
 - In standard Java, we must create our own drawing canvas.
- Provides convenient methods to listen for mouse events.
 - In standard Java, event handling takes a bit more code to set up.



ACM GUI example

```
public class ColorFun extends Program {  
    public void init() {  
        JButton button1 = new JButton("Red!");  
        JButton button2 = new JButton("Blue!");  
        add(button1, SOUTH);  
        add(button2, SOUTH);  
        addActionListeners();  
    }  
    public void actionPerformed(ActionEvent event) {  
        if (event.getActionCommand().equals("Red!")) {  
            setBackground(Color.BLUE);  
        } else {  
            setBackground(Color.RED);  
        }  
    }  
}
```



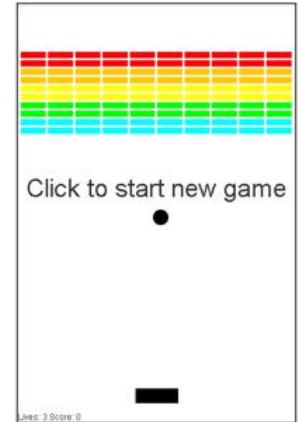
Java GUI example

```
public class ColorFun implements ActionListener {
    public static void main(String[] args) {
        new ColorFun().init();
    }
    private JFrame frame;
    public void init() {
        frame = new JFrame("ColorFun");
        frame.setSize(500, 300);
        JButton button1 = new JButton("Red!");
        JButton button2 = new JButton("Blue!");
        button1.addActionListener(this);
        button2.addActionListener(this);
        frame.add(button1, "South");
        frame.add(button2, "South");
        frame.setVisible(true);
    }
    public void actionPerformed(ActionEvent event) {
        if (event.getActionCommand().equals("Red!")) {
            frame.setBackground(Color.BLUE);
        } else {
            frame.setBackground(Color.RED);
        }
    }
}
```

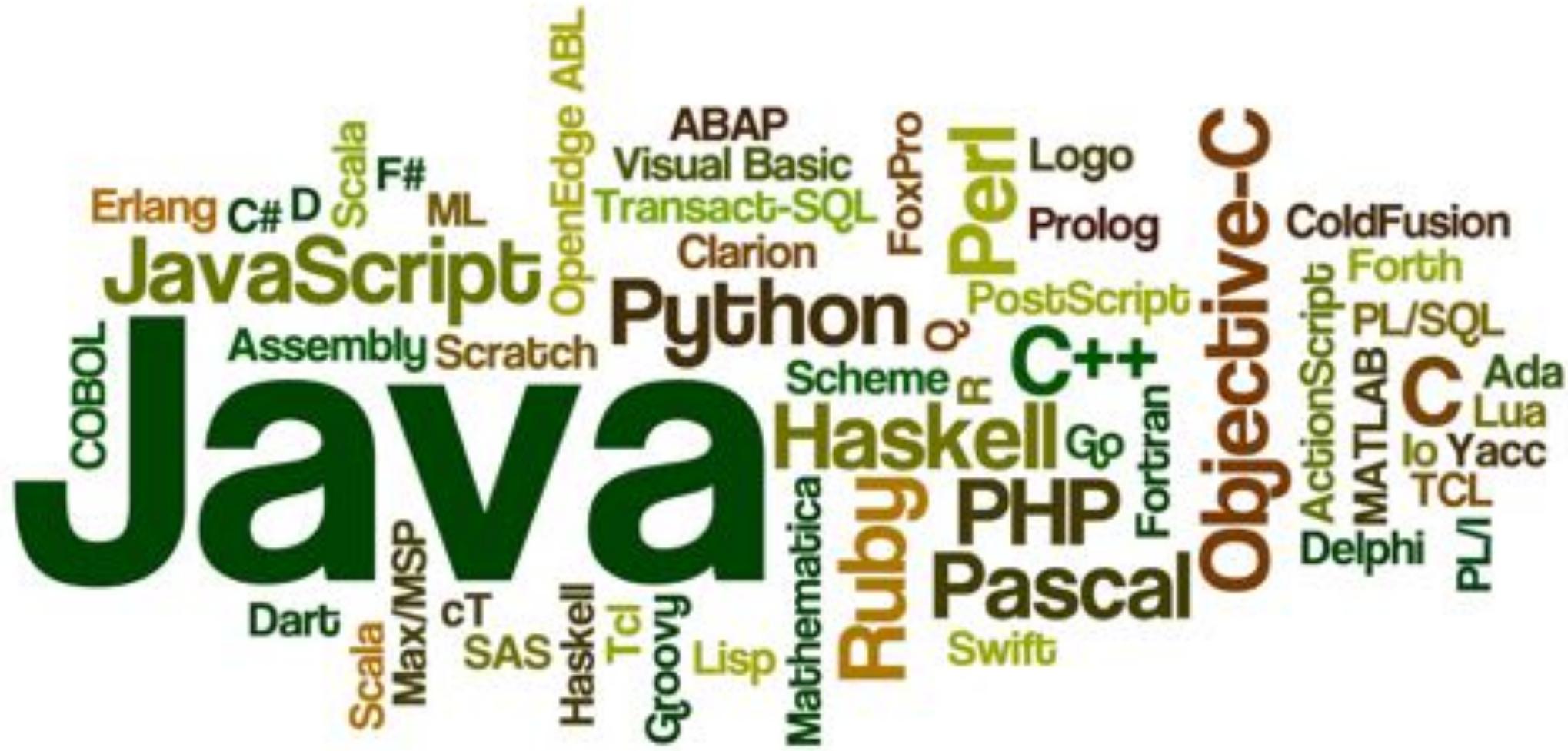


Summary

- **Benefits of libraries:**
 - simplify syntax/rough edges of language/API
 - avoid re-writing the same code over and over
 - possible to make advanced programs quickly
 - leverage work of others
- **Drawbacks of libraries:**
 - limitations on usage; e.g. ACM library cannot be re-distributed for commercial purposes



Programming Languages



Java

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



Python

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

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



Javascript

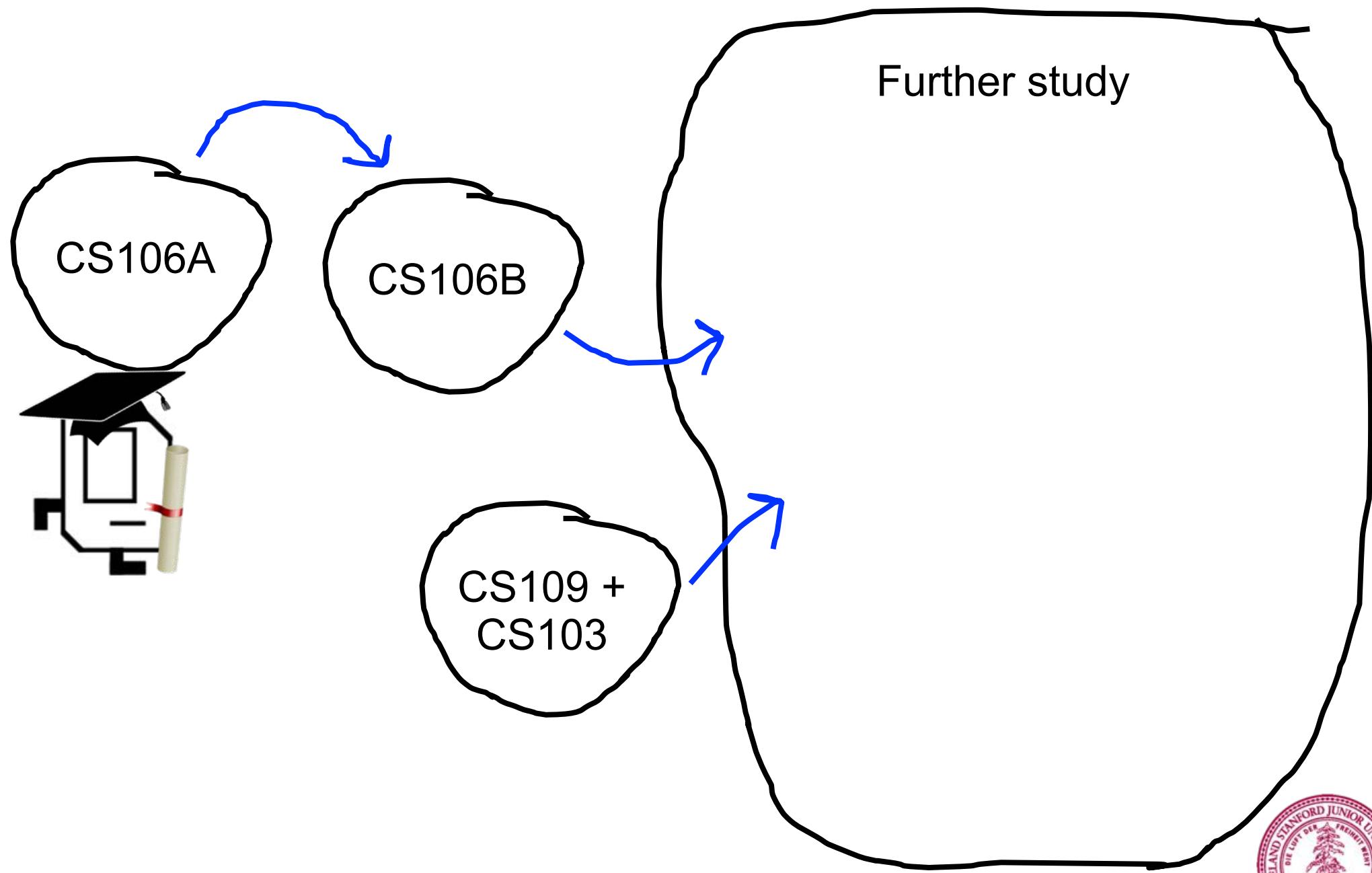
```
var evens = []
for(var i = 0; i < 100; i++) {
    if(i % 2 == 0) {
        evens.push(i)
    }
}
console.log(evens)
```

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

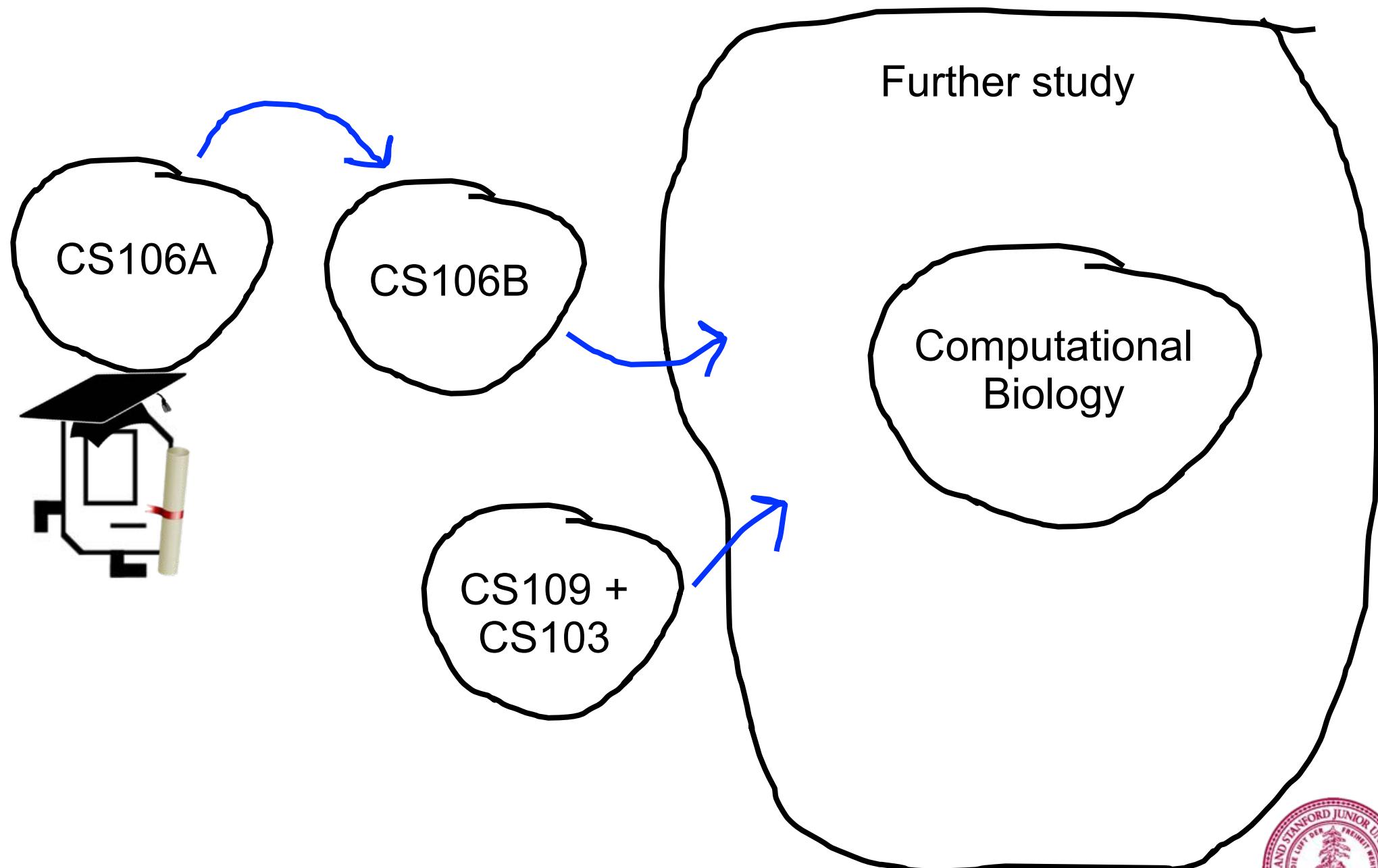


Future in the CS curriculum

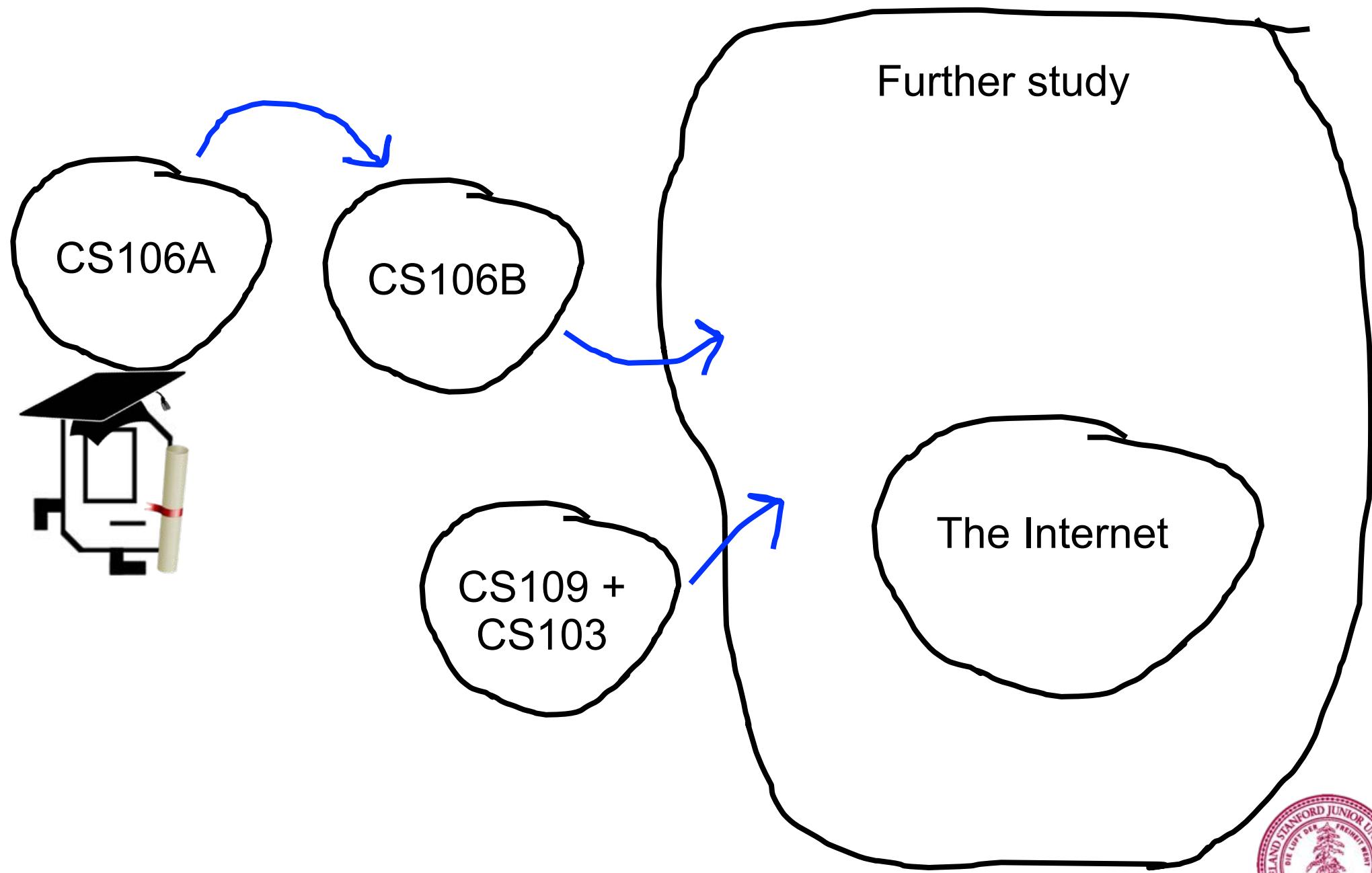
The CS Curriculum



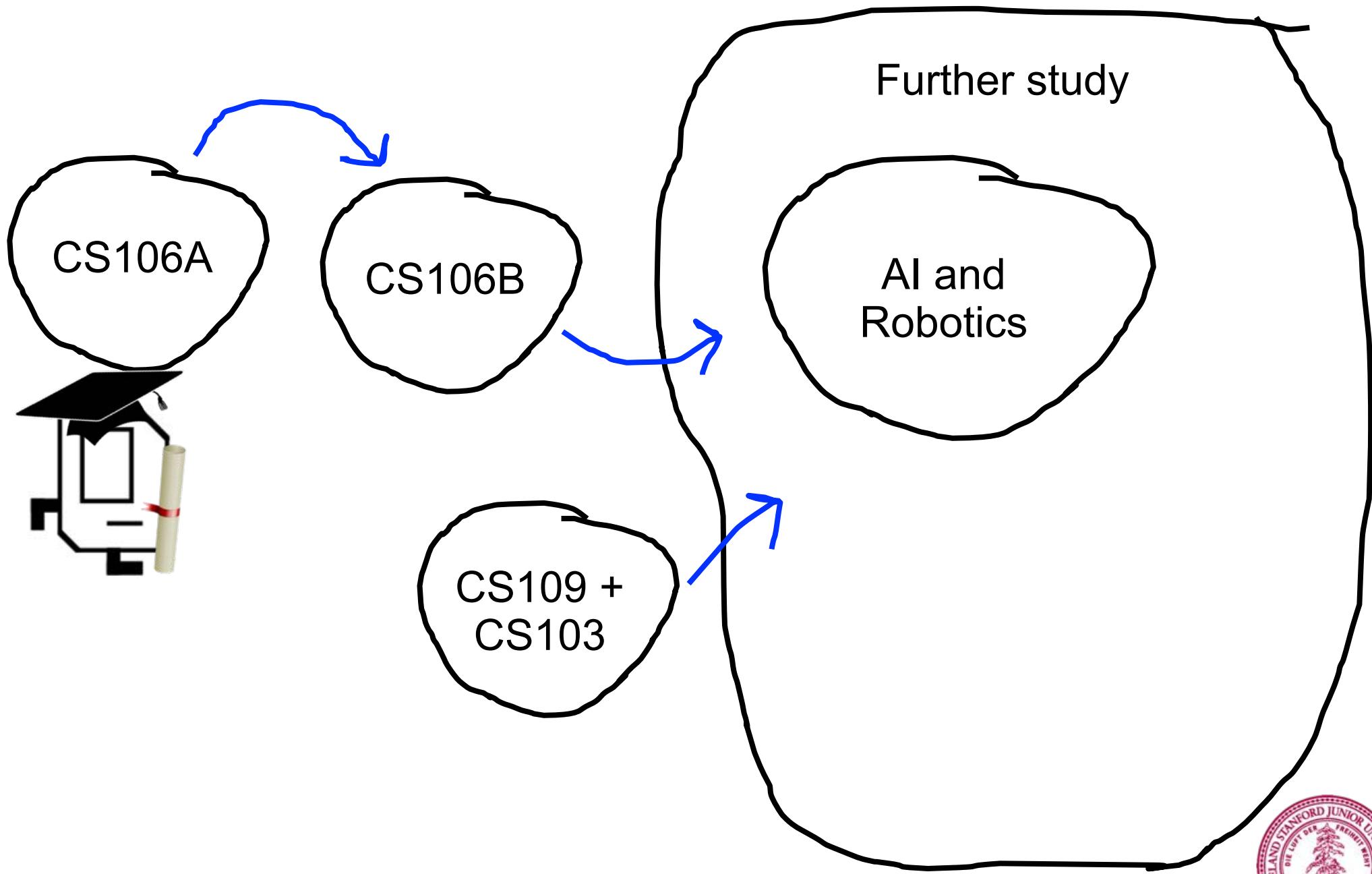
The CS Curriculum



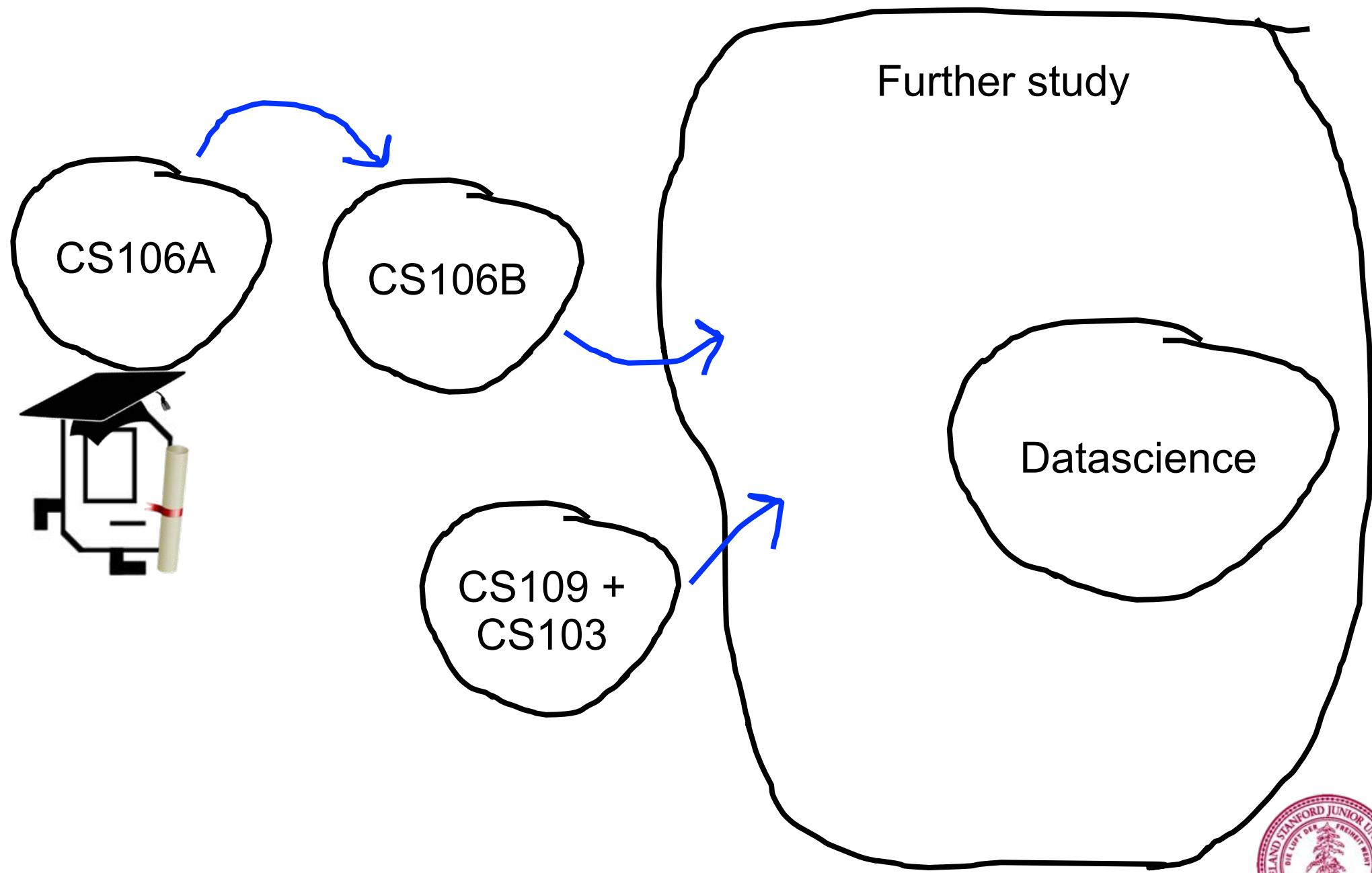
The CS Curriculum



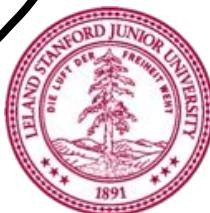
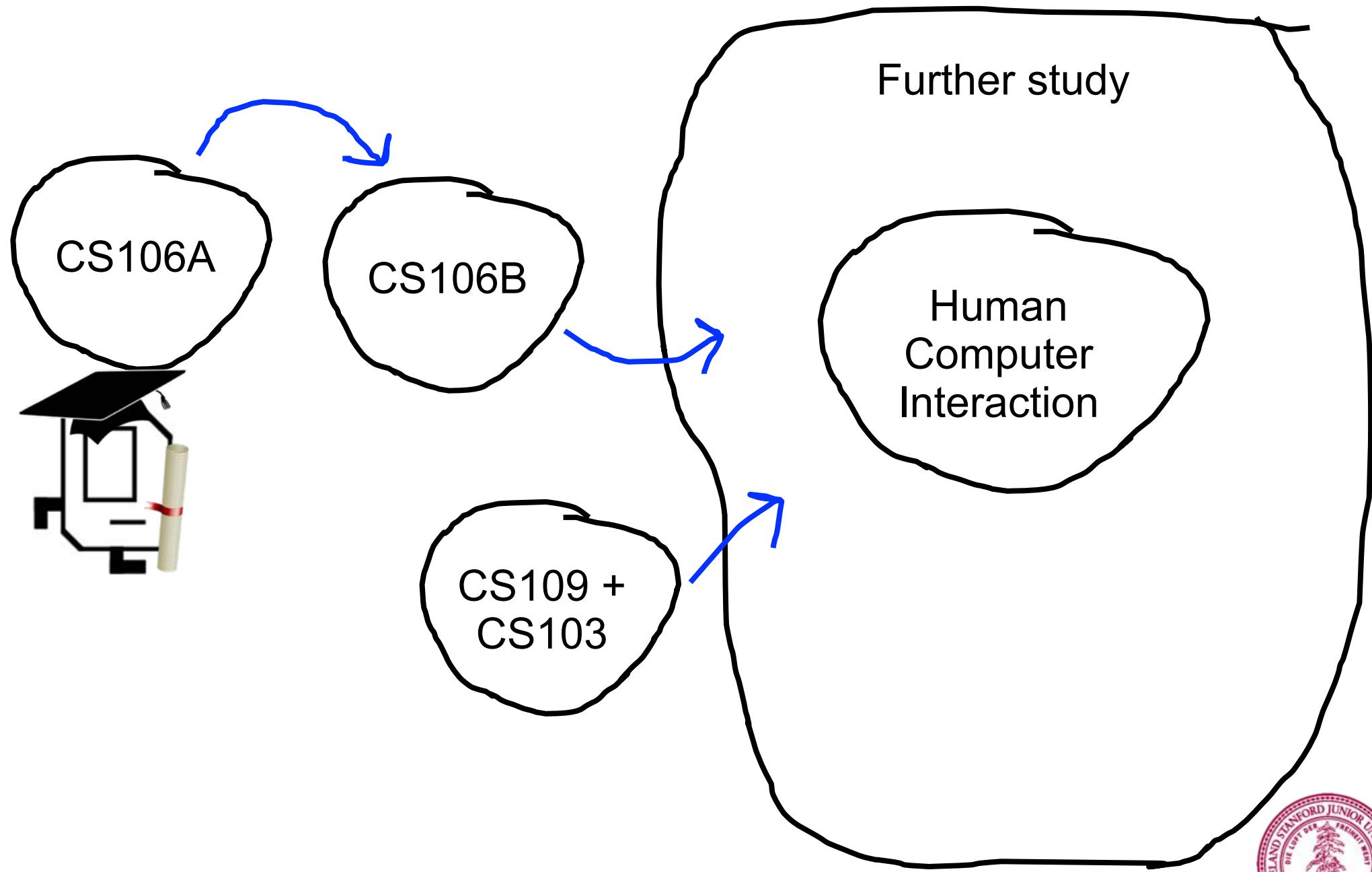
The CS Curriculum



The CS Curriculum



The CS Curriculum



Machine Learning

Machine Learning
or, How we learned to decompose

There is something going on
in the world of AI

Something big (for us)...

[suspense]

How can we develop intelligent agents?



Volunteer



Computer
programs

How can we develop intelligent agents?

Better than
chance

As well as
humans



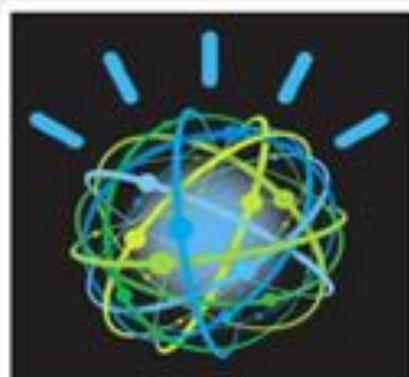
Big Milestones Starting in 1997



1997 Deep Blue



2005 Stanley



2011 Watson



Self Driving Cars



Computers Making Art



The Last Remaining Board Game

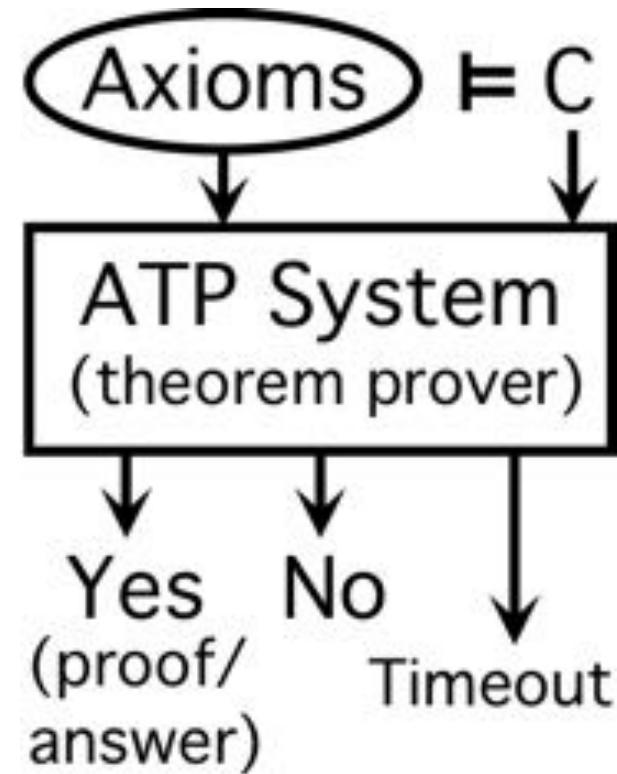


Early Optimism 1950

1952



1955



Early Optimism 1950

“Machines will be capable,
within twenty years, of
doing any work a man can
do.”

—Herbert Simon, 1952



Underwhelming Results 1950s to 1980s

The spirit is willing but the flesh is weak.



(Russian)



The vodka is good but the meat is rotten.

The world is too complex



BRACE YOURSELVES



WINTER IS COMING

Simple Example: Identifying Cats

- We have a picture and we want to know if it's a cat or not.



→ true



→ false



→ true



→ false



WILHELM HÜHN
KIRCH-ON-DER-WILLEBECK
DRESDEN-TEUCHEN

25

Identifying Cats

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

```
private void isCat(GImage animal) {  
    int[][] pixels = animal.getPixelArray();  
    if (containsTwoEyes(pixels)){  
        if (hasWhiskers(pixels)){  
            if (hasPointyEars(pixels)){  
                return true;  
            }  
        }  
    }  
    return false;  
}
```



Some Tricky Cases



Picn., CS100A, Stanford University



Pros / Cons

- Pros
 - Matches our human intuition about what a cat is
 - Easy to understand the code
- Cons
 - Requires us to explicitly enumerate every feature that's important, and know how important it is
 - Need to write code to detect eyes, and whiskers, and the pointiness of ears
 - Will never improve... cannot learn from its mistakes



Hard problems seemed impossible.

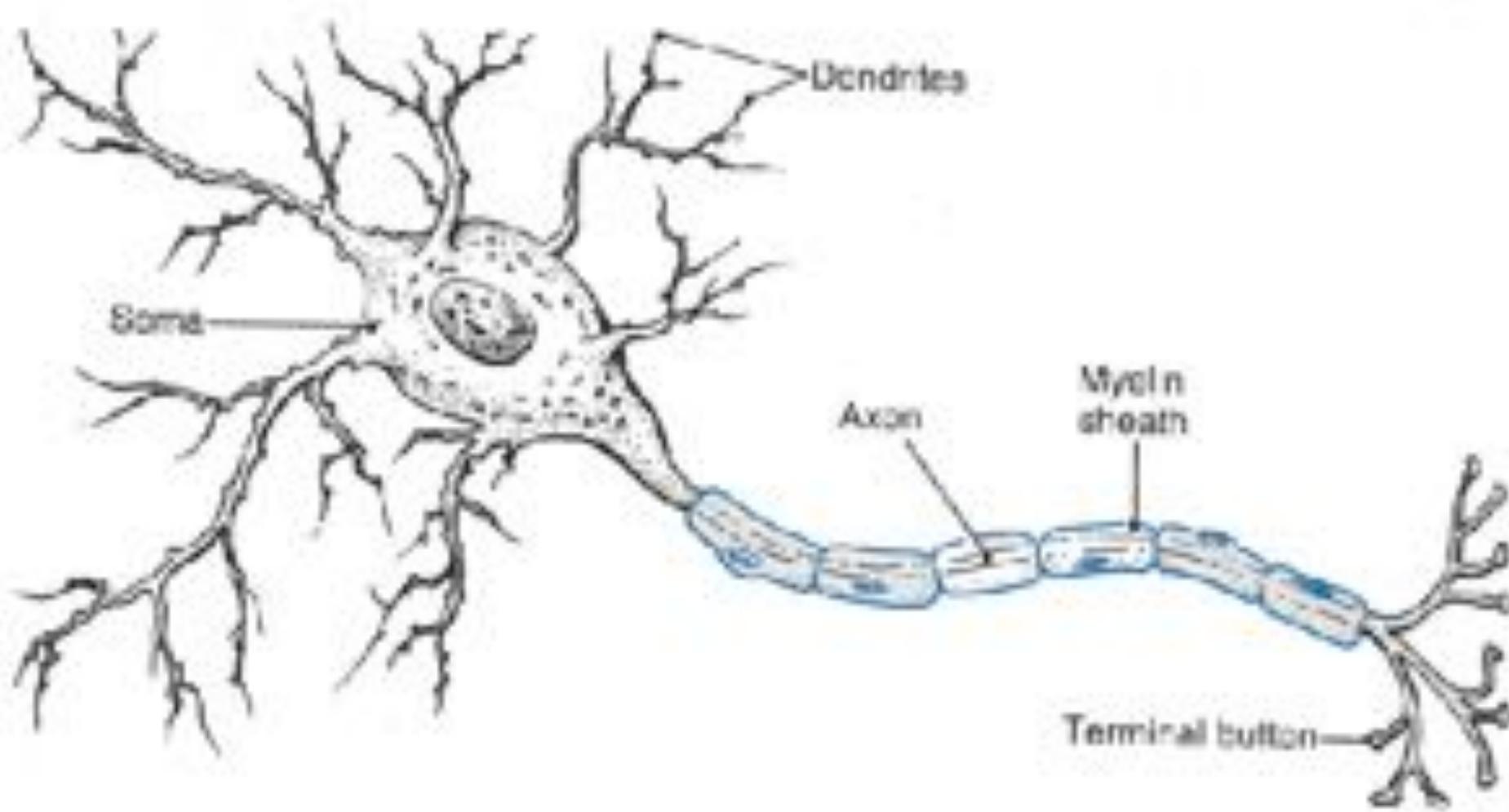
Great idea #1 learn from experience

Machine Learning: Learn From Experience

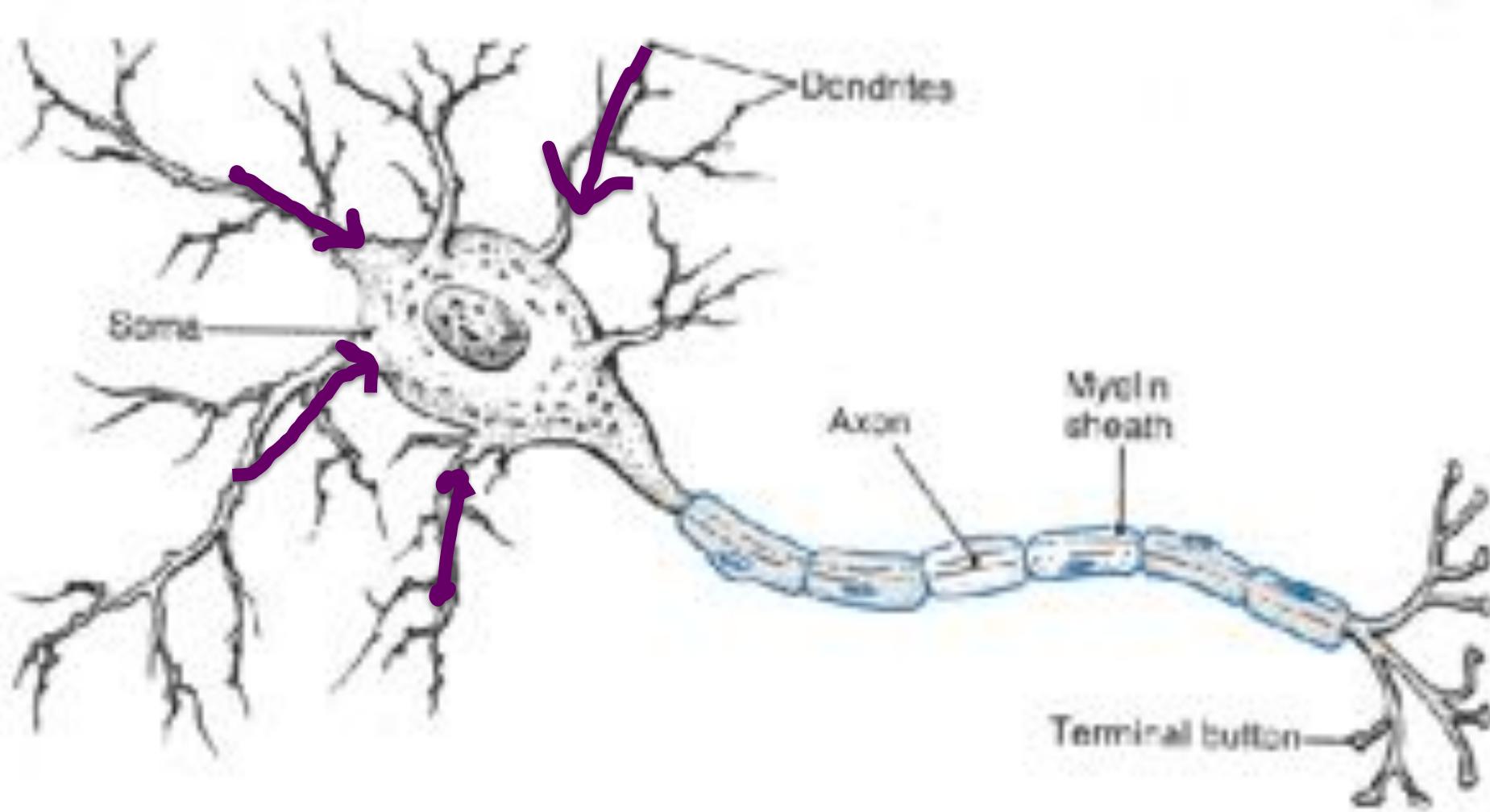


Great idea #2 inspired by biology

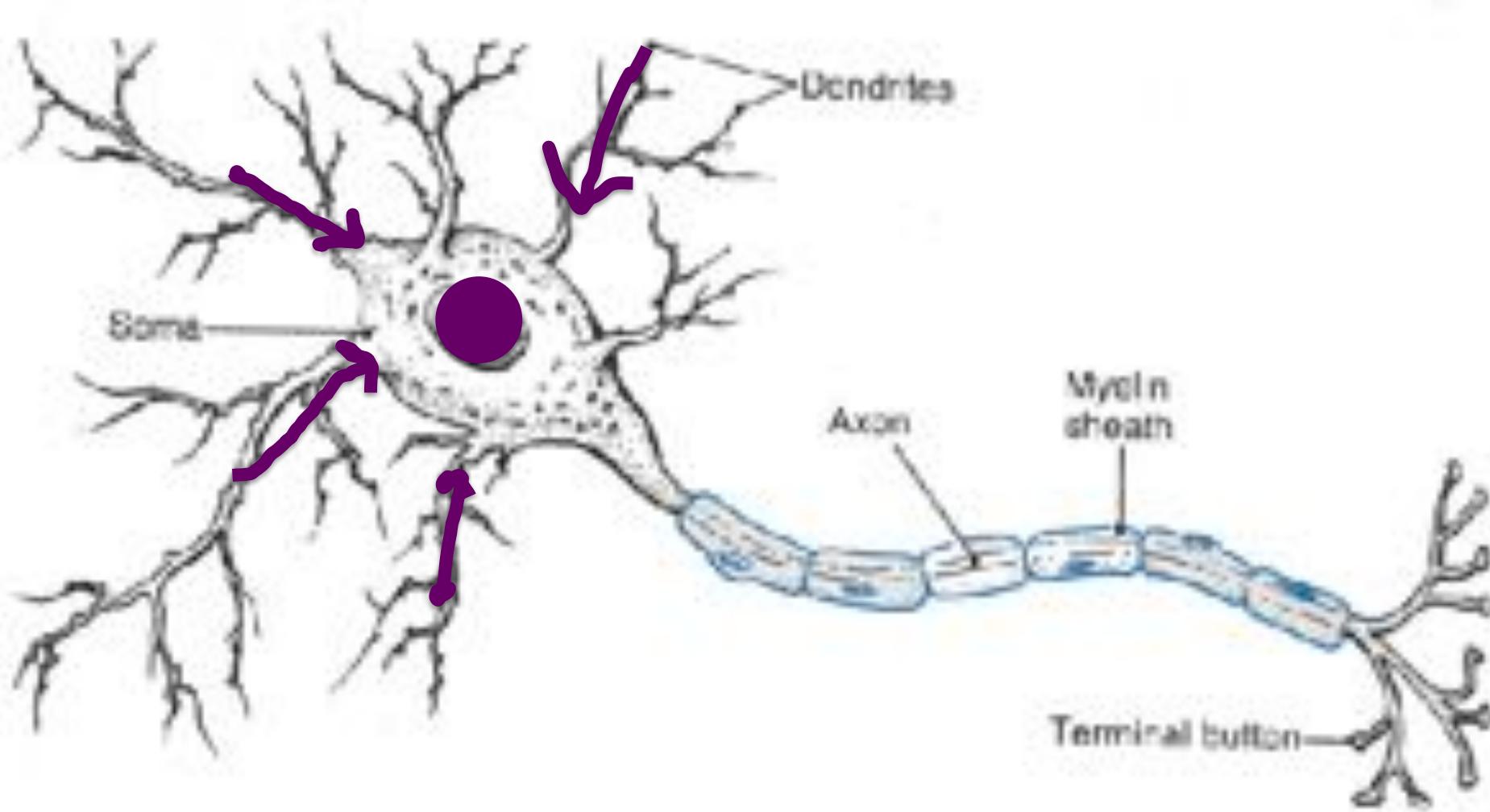
Neuron



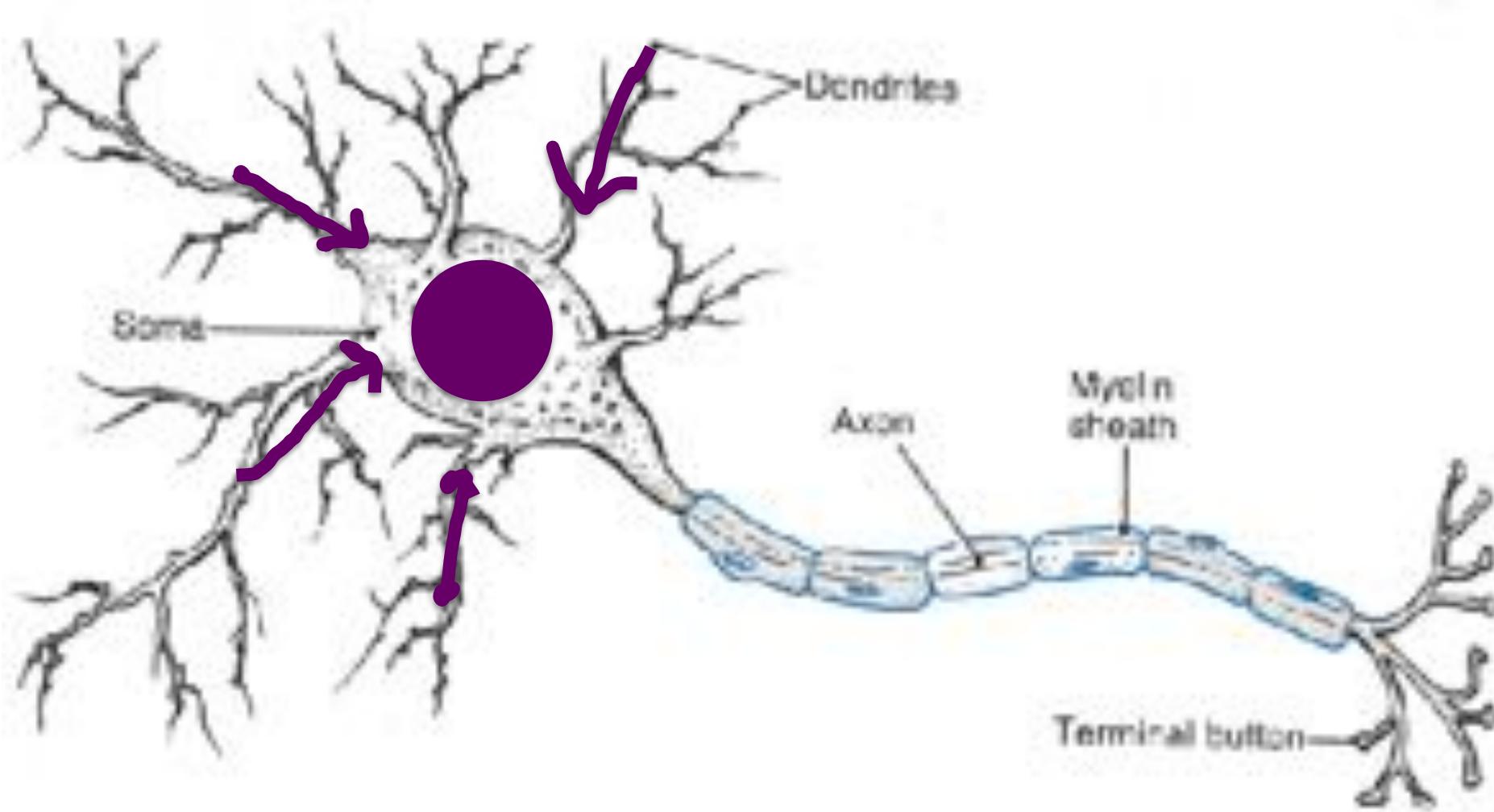
Neuron



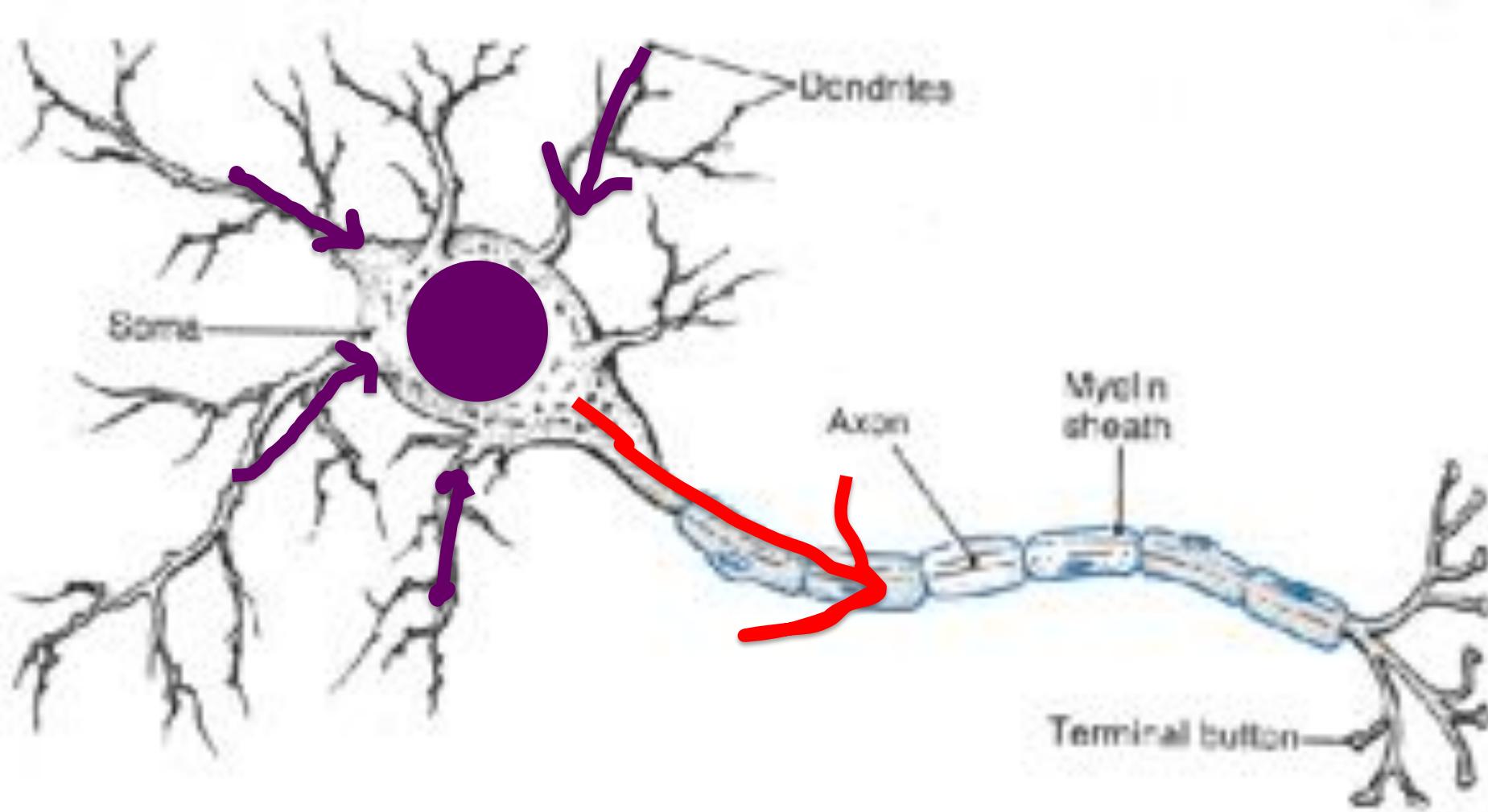
Neuron



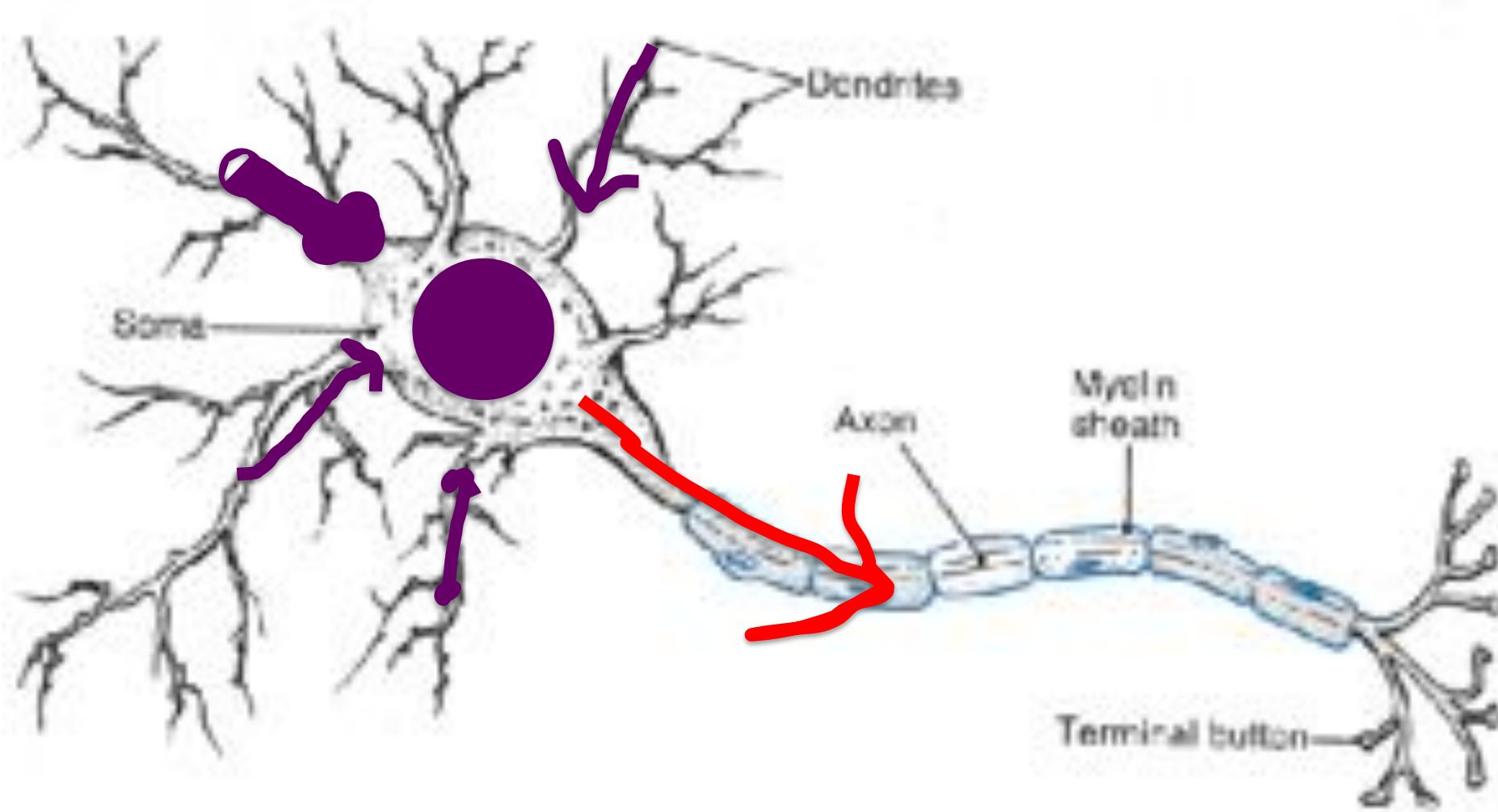
Neuron



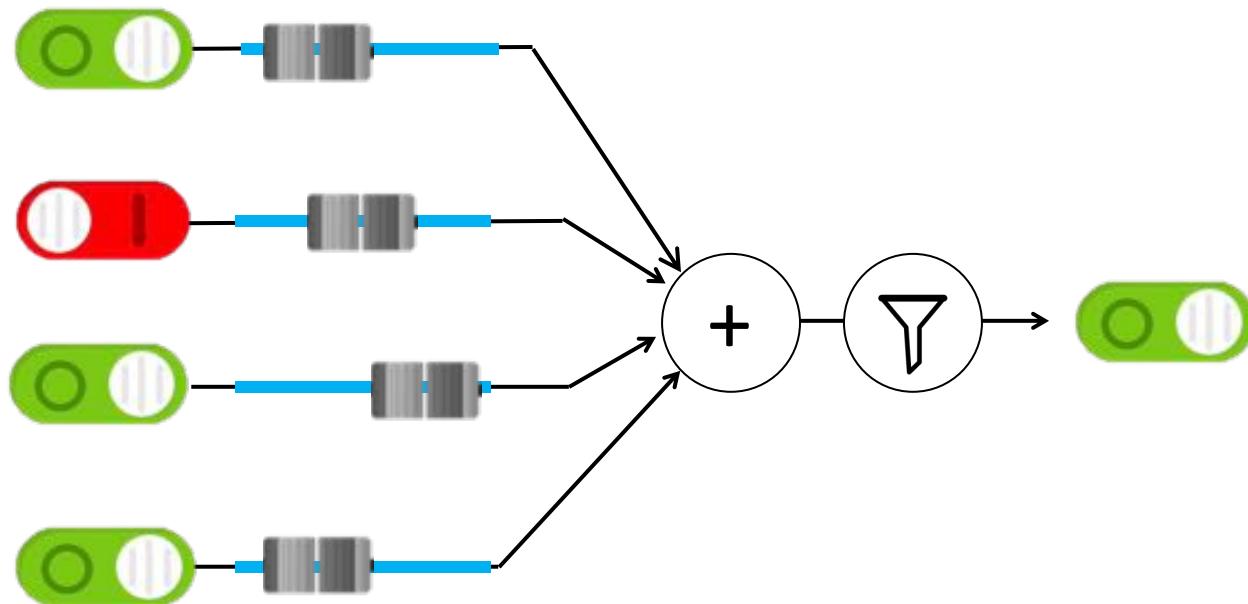
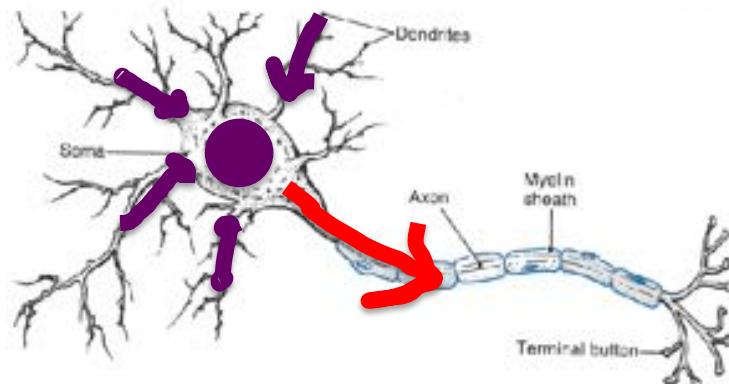
Neuron



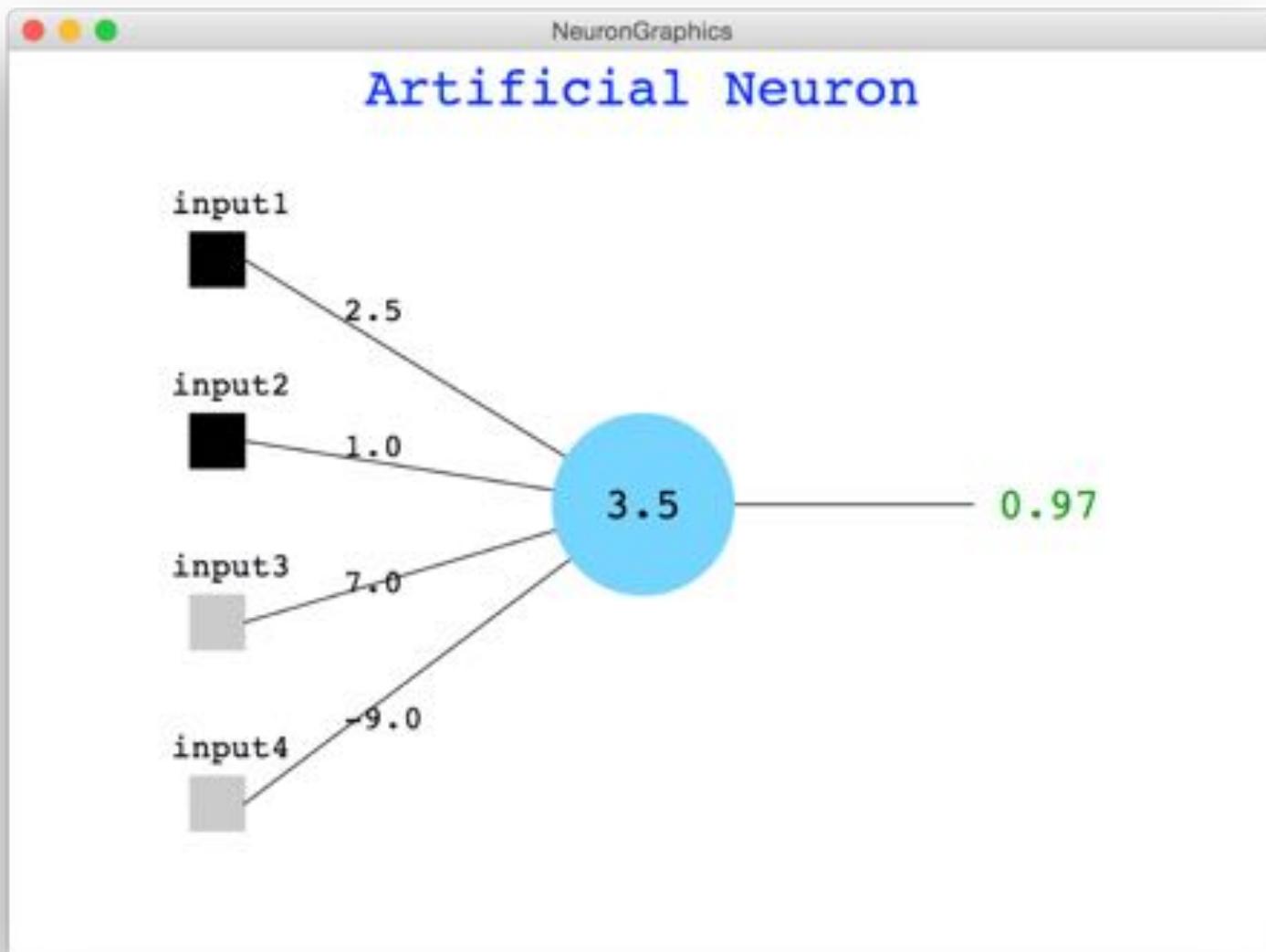
Some Inputs are More Important



Artificial Neurons



Java Demo

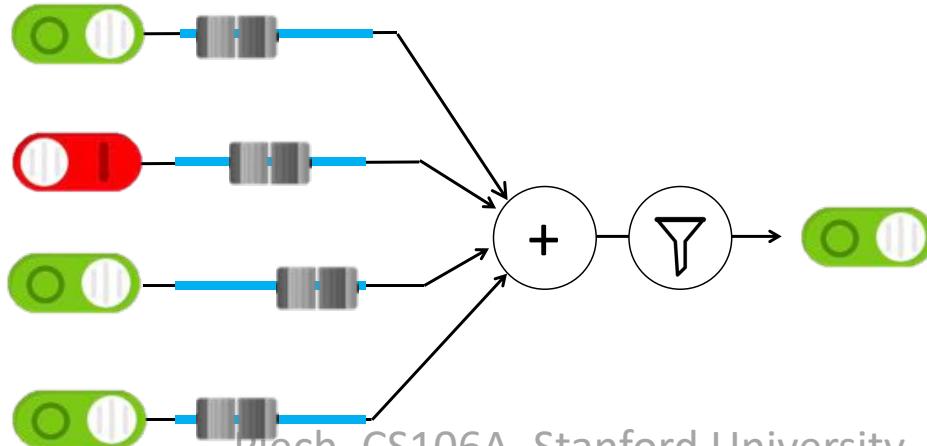


Artificial Neuron

```
// calculate the activation of a neuron
private double activate(double[] weights, double[] inputs) {
    double weightedSum = 0;
    for(int i = 0; i < inputs.length; i++) {
        weightedSum += weights[i] * inputs[i];
    }

    return squash(weightedSum);
}

// the sigmoid function forces a value to be between 0 and 1
private double squash(double value) {
    return 1 / (1 + Math.exp(-value));
}
```

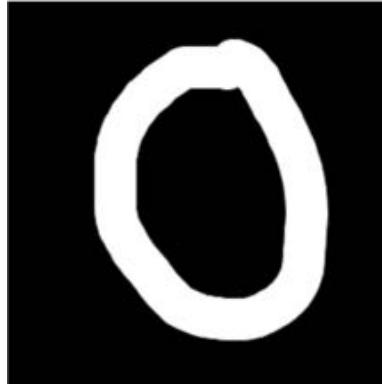


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Digit Recognition Example

Let's make feature vectors from pictures of numbers



input = [0, 0, 0, 0, . . . , 1, 0, 0, 1, . . . 0, 0, 1, 0]
label = 0



input = [0, 0, 1, 1, . . . , 0, 1, 1, 0, . . . 0, 1, 0, 0]
Piech, CS10 label = 1
mordUniversity



Computer Vision



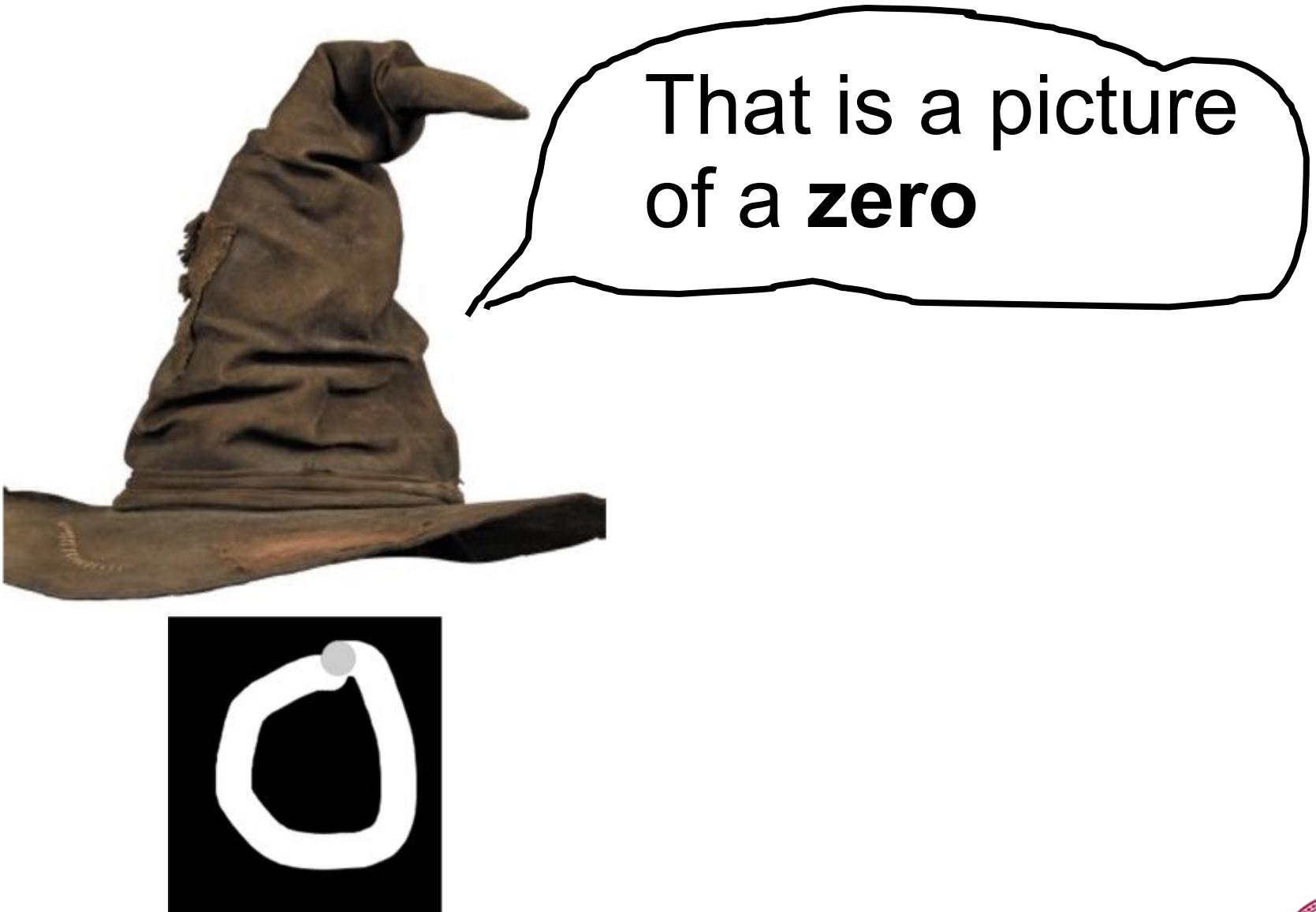
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Classification



Classification



Classification

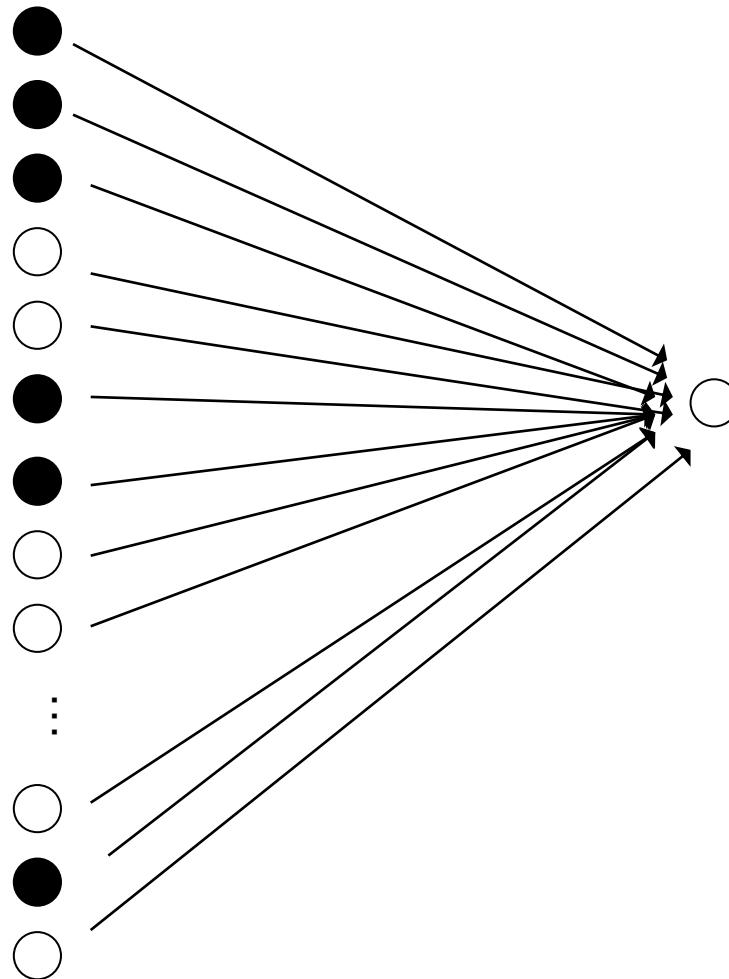
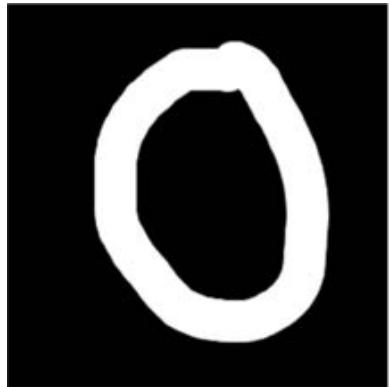


* It doesn't have to
be correct all of the
time



Can you do it?

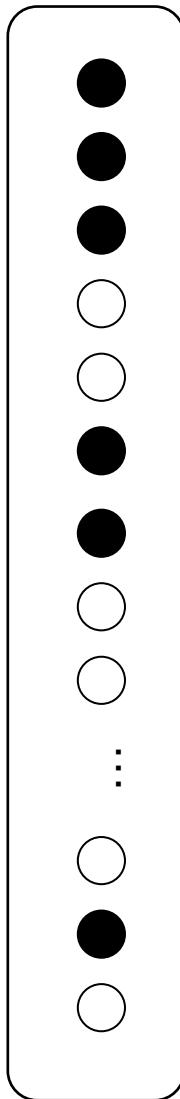
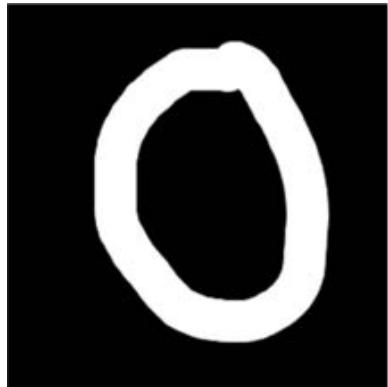
Single Neuron



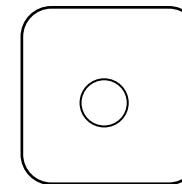
This means it
predicts a 0



Single Neuron



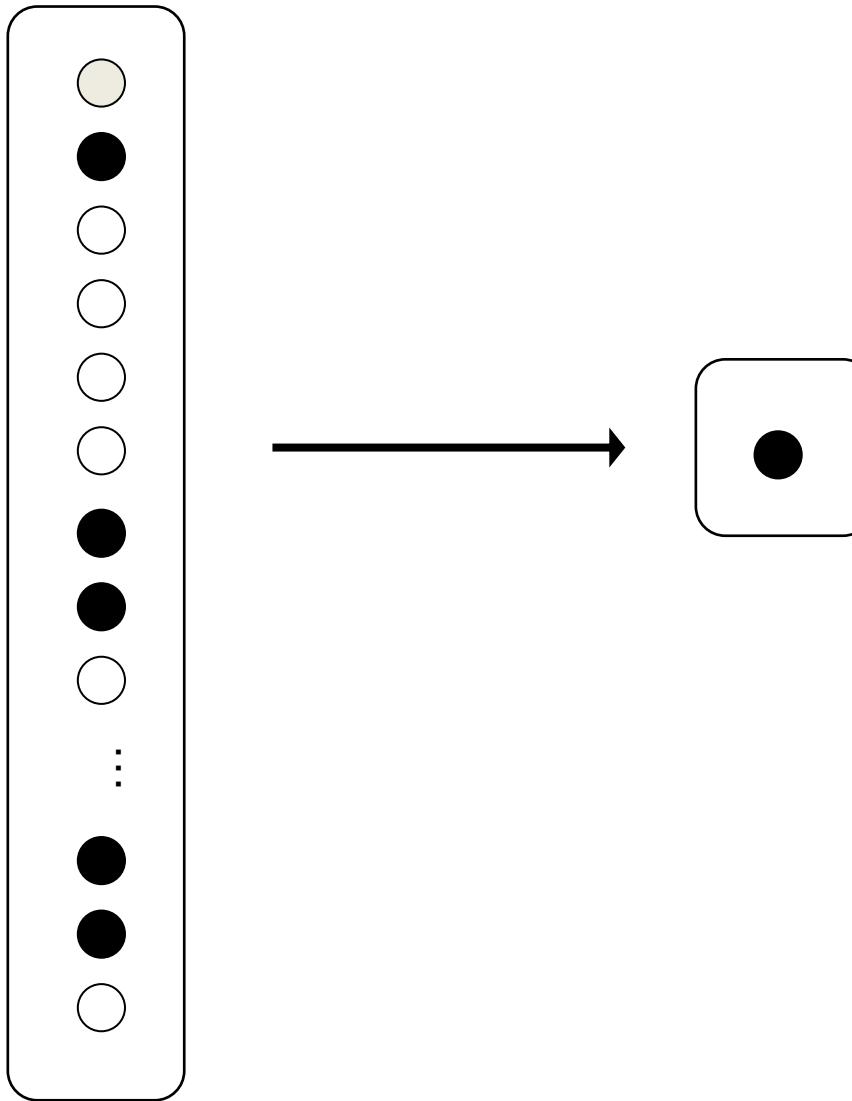
Indicates fully
connected



This means it
predicts a 0



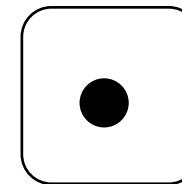
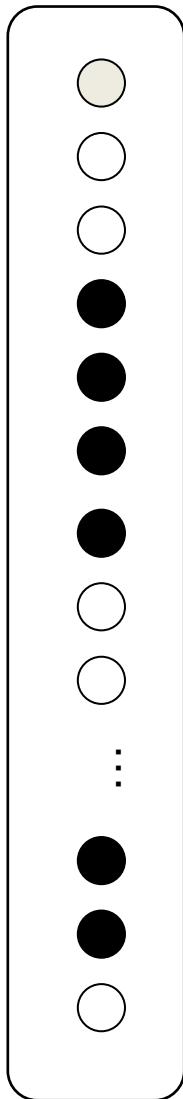
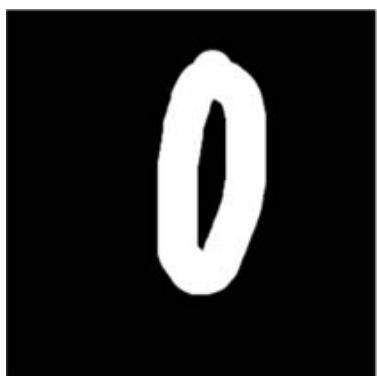
Single Neuron



This means it
predicts a 1



Not So Good

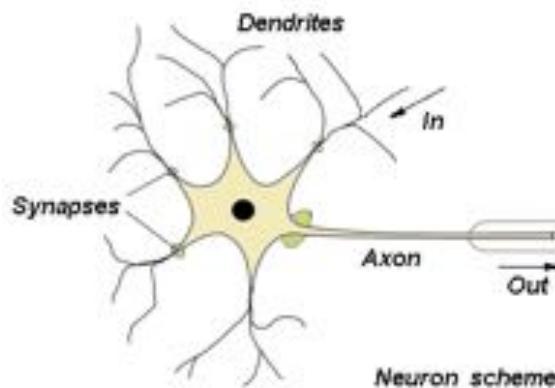


This means it
predicts a 1

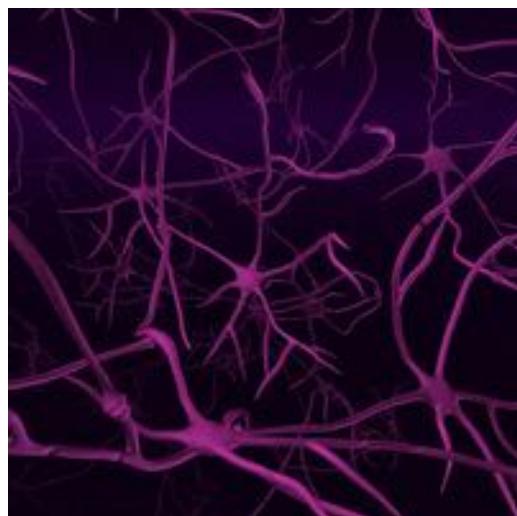


Biological Basis for Neural Networks

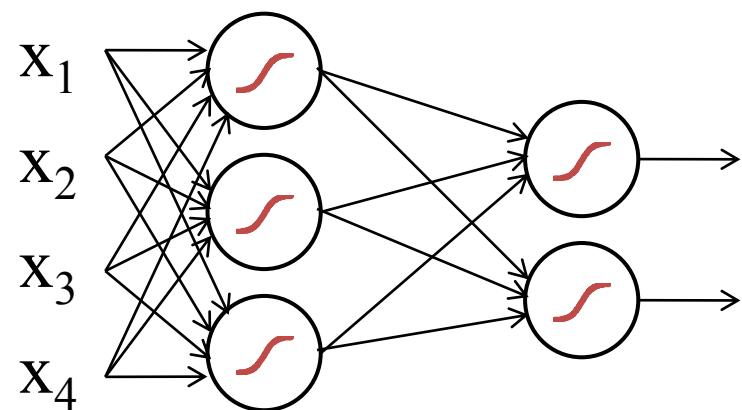
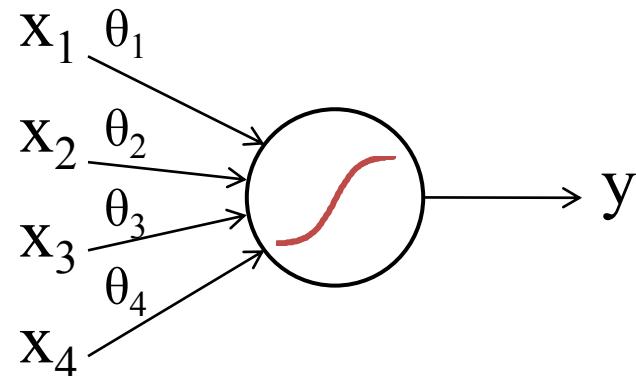
- A neuron



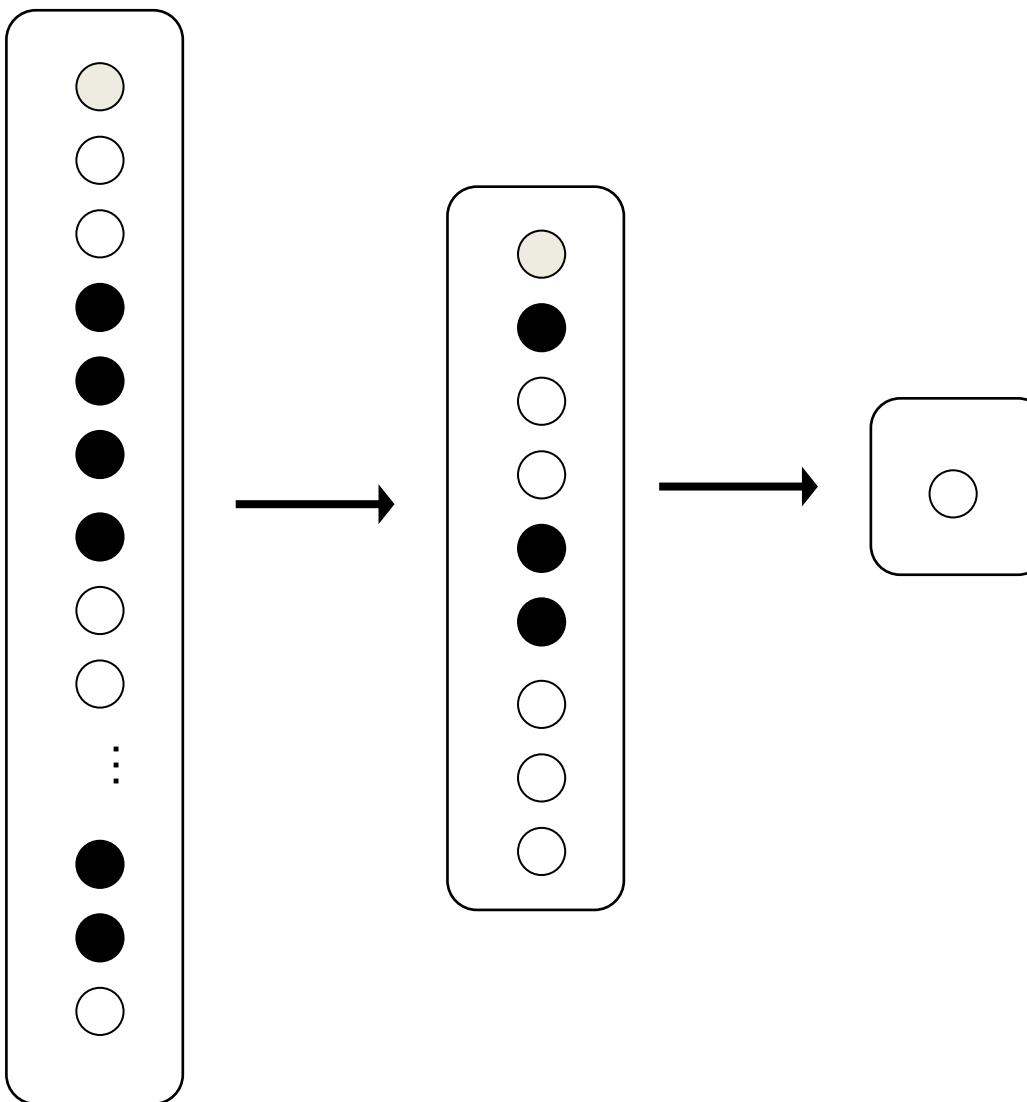
- Your brain



Actually, it's probably someone else's brain



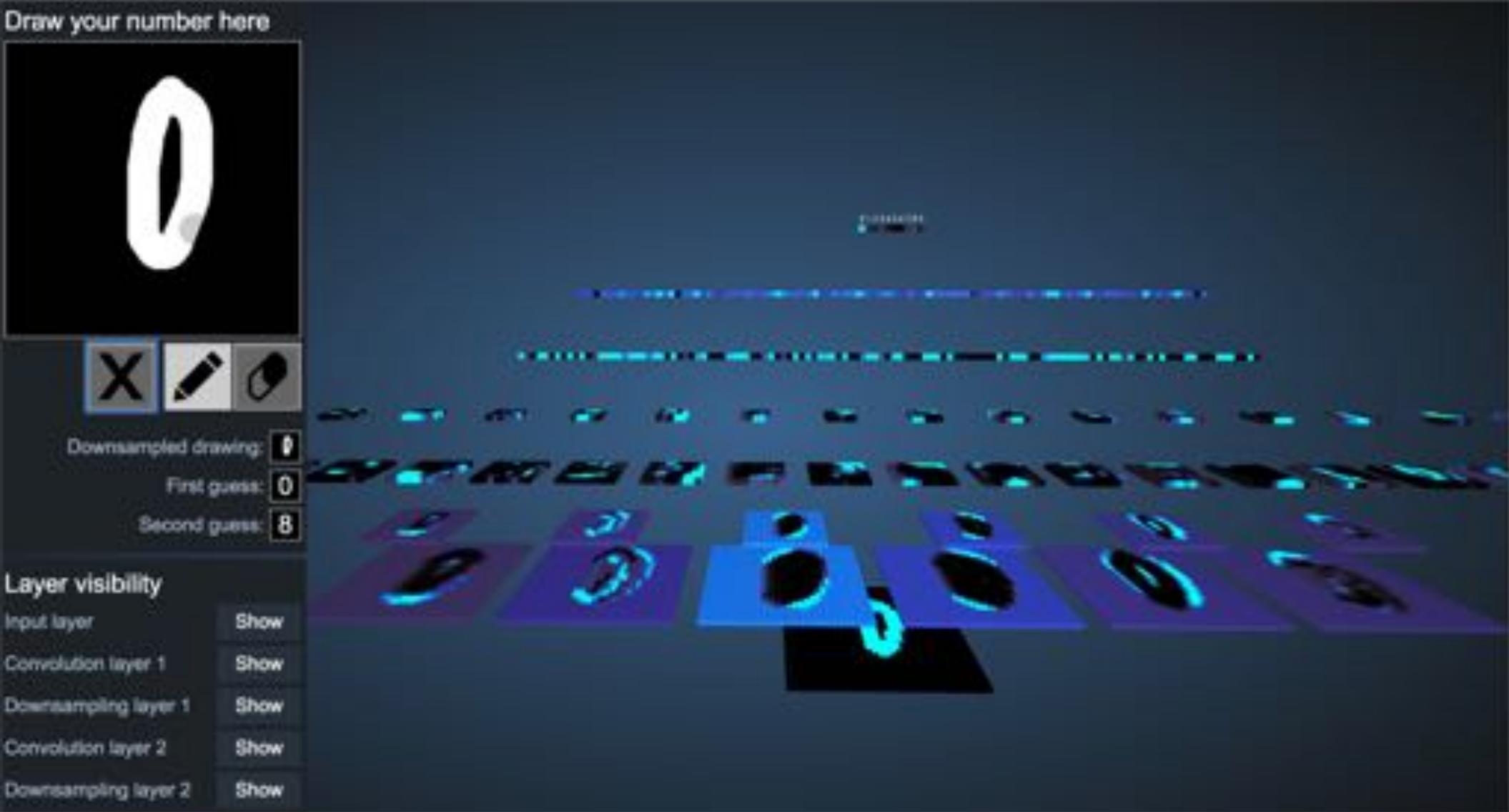
We Can Put Neurons Together



This means it
predicts a 0



Demonstration



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

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What Does This Look Like in Code?

The screenshot shows the Eclipse IDE interface with two code editors open. The left editor contains `NeuralNetwork.java` and the right editor contains `Neuron.java`. Both files are part of a Java project named "Neuron".

```
workspace - Java - Neuron/Neuron.java - Eclipse
```

```
NeuronNetwork.java
```

```
1 public class NeuralNetwork extends ConsoleProgram {  
2  
3     private static final int N_INPUTS = 1824;  
4     private static final int N_LAYER1 = 20;  
5  
6     private ArrayList<Neuron> layer1 = null;  
7     private Neuron prediction = null;  
8  
9     public void run() {  
10        loadNeuralNetwork();  
11  
12        // make predictions  
13        GImage birdImage = new GImage("bird6.png");  
14        GImage planeImage = new GImage("airplane4.png");  
15  
16        makePrediction(birdImage);  
17        makePrediction(planeImage);  
18    }  
19  
20    private void makePrediction(GImage img) {  
21        // turn the image into inputs  
22        ArrayList<Double> inputs = new ArrayList<Double>();  
23        int[] pixelArray = img.getPixelArray();  
24        for(int r = 0; r < pixelArray.length; r++) {  
25            for(int c = 0; c < pixelArray[0].length; c++) {  
26                Color color = new Color(pixelArray[r][c]);  
27                double greyScale = getGrey(color);  
28                inputs.add(greyScale);  
29            }  
30        }  
31    }  
32}
```

```
Neuron.java
```

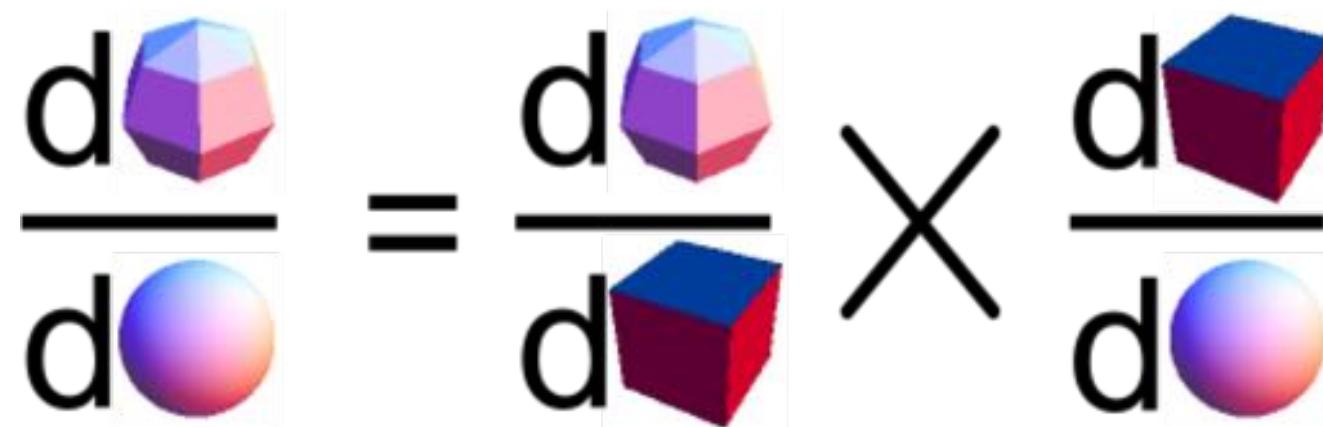
```
1 public class Neuron extends GraphicsProgram {  
2  
3     private ArrayList<Double> weights = null;  
4  
5     public Neuron(String fileName, int n) {  
6         loadWeightsFromFile(fileName, n);  
7     }  
8  
9     public double activate(ArrayList<Double> inputs) {  
10        double weightedSum = 0.0;  
11        for(int i = 0; i < inputs.size(); i++) {  
12            weightedSum += inputs.get(i) * weights.get(i);  
13        }  
14        return sigmoid(weightedSum);  
15    }  
16  
17    private double sigmoid(double x) {  
18        return 1.0 / (1.0 + Math.exp(-x));  
19    }  
20  
21    private void loadWeightsFromFile(String fileName, int n) {  
22        weights = new ArrayList<Double>();  
23        try {  
24            BufferedReader rd = new BufferedReader(new FileInput...  
25            while(true) {  
26                String line = rd.readLine();  
27                if(line == null) break;  
28                weights.add(Double.parseDouble(line));  
29            }  
30        } catch (Exception e) {  
31            System.out.println("Error reading file " + fileName);  
32        }  
33    }  
34}
```



Aside: decomposition

How do we get those weights?

Chain Rule Down the Network

$$\frac{d}{d\theta} = \frac{d}{d\phi} \times \frac{d}{d\theta}$$




Learning Weights

$$LL(\theta) = y \log \sigma(\theta^T \mathbf{x}) + (1 - y) \log[1 - \sigma(\theta^T \mathbf{x})]$$

$$\begin{aligned}\frac{\partial LL(\theta)}{\partial \theta_j} &= \frac{\partial}{\partial \theta_j} y \log \sigma(\theta^T \mathbf{x}) + \frac{\partial}{\partial \theta_j} (1 - y) \log[1 - \sigma(\theta^T \mathbf{x})] \\ &= \left[\frac{y}{\sigma(\theta^T \mathbf{x})} - \frac{1 - y}{1 - \sigma(\theta^T \mathbf{x})} \right] \frac{\partial}{\partial \theta_j} \sigma(\theta^T \mathbf{x}) \\ &= \left[\frac{y}{\sigma(\theta^T \mathbf{x})} - \frac{1 - y}{1 - \sigma(\theta^T \mathbf{x})} \right] \frac{\partial}{\partial \theta_j} \sigma(\theta^T \mathbf{x}) \\ &= \left[\frac{y - \sigma(\theta^T \mathbf{x})}{\sigma(\theta^T \mathbf{x})[1 - \sigma(\theta^T \mathbf{x})]} \right] \sigma(\theta^T \mathbf{x})[1 - \sigma(\theta^T \mathbf{x})] x_j \\ &= [y - \sigma(\theta^T \mathbf{x})] x_j\end{aligned}$$



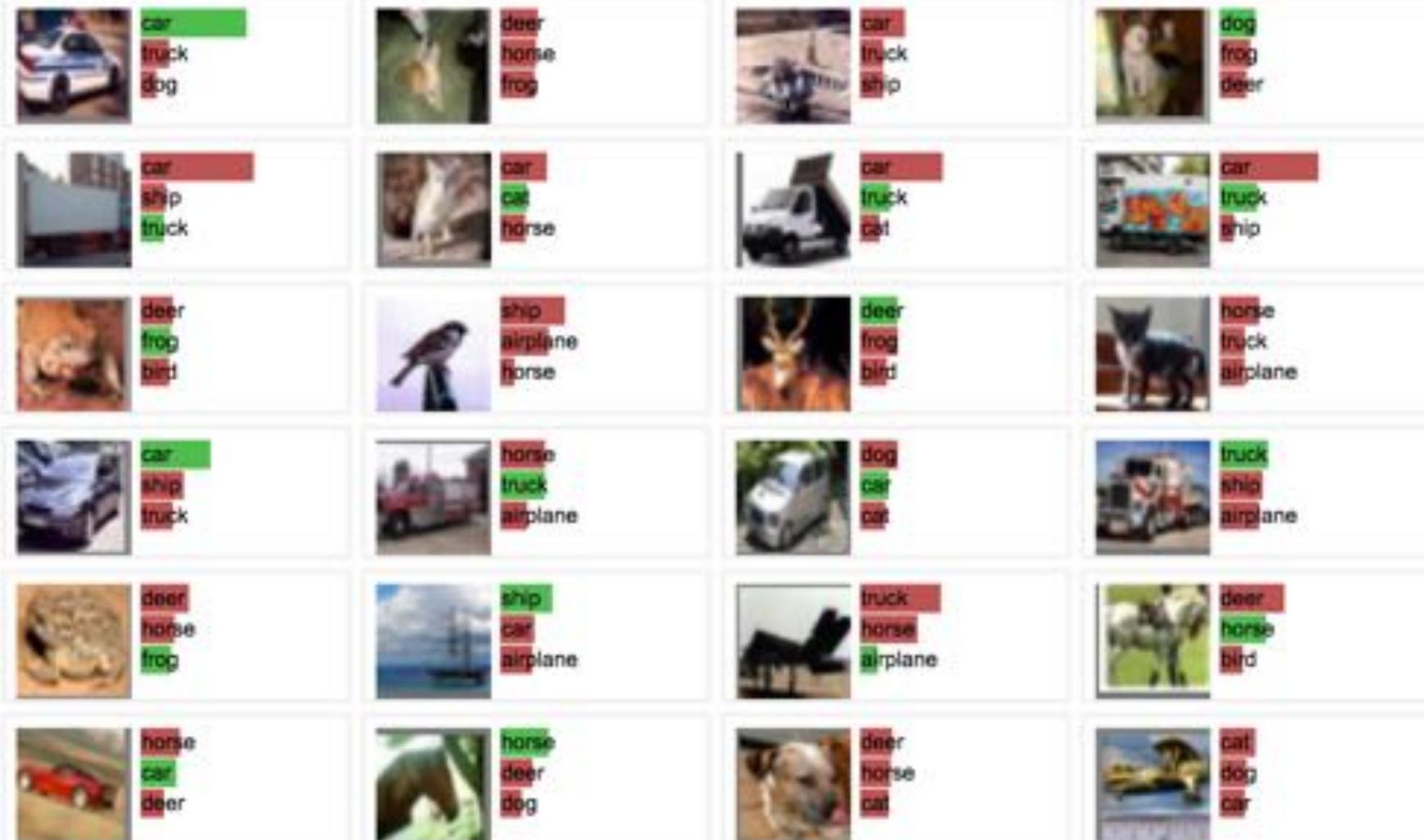
**Artificial Neurons: One of the greatest
decompositions of our lifetimes**

`model.calculatePartialDerivative(data)`

model.update(data)

Let's Train!

test accuracy based on last 200 test images: 0.2894736842105263



<http://cs.stanford.edu/people/karpathy/convnetjs/demo/classify2d.html>



Like lego pieces

Visualize the Weights



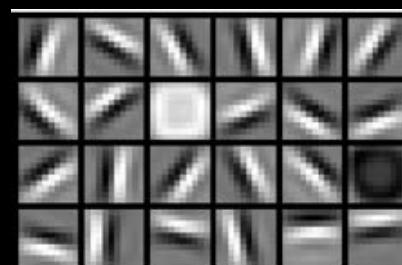
Training set: Aligned
images of faces.



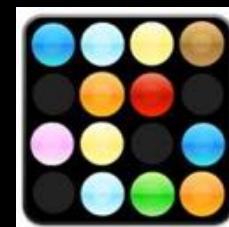
object models



object parts
(combination
of edges)

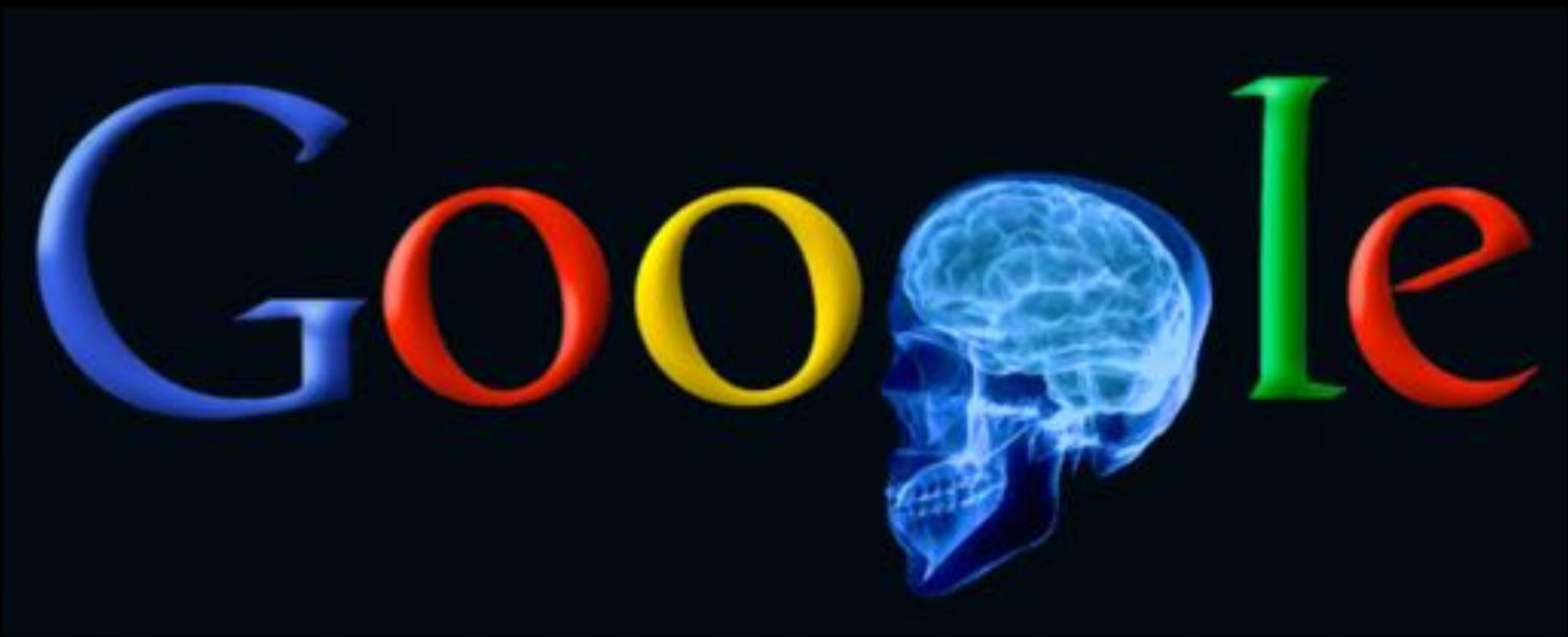


edges



pixels

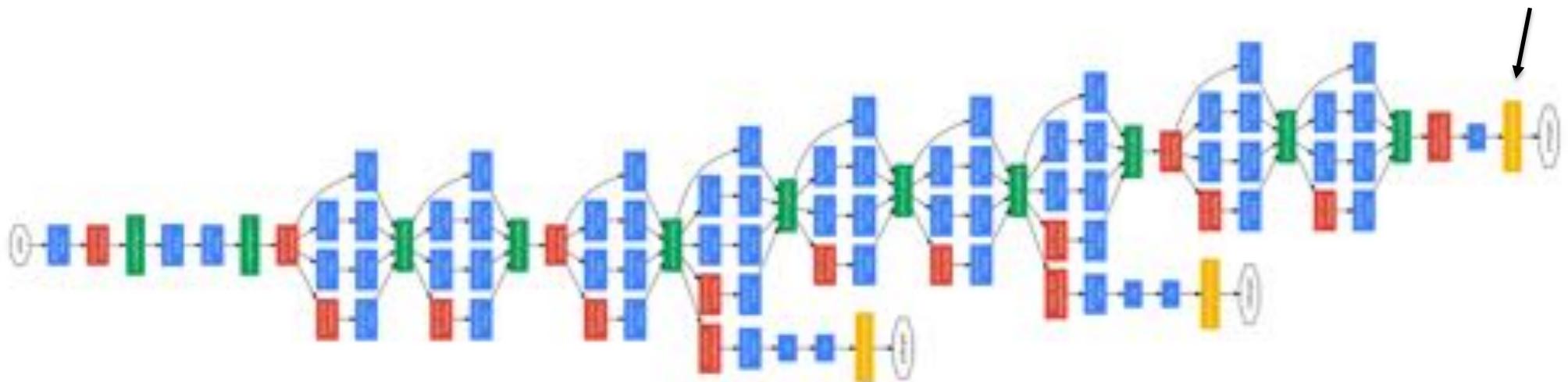
GoogLeNet Brain



1 Trillion Artificial Neurons

GoogLeNet Brain Graph

Multiple,
Multi class output



22 layers deep



The Face Neuron



Top stimuli from the test set



Optimal stimulus
by numerical optimization

The Cat Neuron



Top stimuli from the test set



Optimal stimulus
by numerical optimization

Hire the smartest people in the world



Invent cat detector

Best Neuron Stimuli

Neuron 1



Neuron 2



Neuron 3



Neuron 4



Neuron 5



Best Neuron Stimuli

Neuron 6



Neuron 7



Neuron 8



Neuron 9



ImageNet Classification

22,000 categories

14,000,000 images

Hand-engineered features (SIFT, HOG, LBP),
Spatial pyramid, SparseCoding/Compression

22,000 is a lot!

...

smoothhound, smoothhound shark, *Mustelus mustelus*

American smooth dogfish, *Mustelus canis*

Florida smoothhound, *Mustelus norrisi*

whitetip shark, reef whitetip shark, *Triaenodon obesus*

Atlantic spiny dogfish, *Squalus acanthias*

Pacific spiny dogfish, *Squalus suckleyi*

hammerhead, hammerhead shark

smooth hammerhead, *Sphyrna zygaena*

smalleye hammerhead, *Sphyrna tudes*

shovelhead, bonnethead, bonnet shark, *Sphyrna tiburo*

angel shark, angelfish, *Squatina squatina*, monkfish

electric ray, crampfish, numbfish, torpedo

smalltooth sawfish, *Pristis pectinatus*

guitarfish

roughtail stingray, *Dasyatis centroura*

butterfly ray

eagle ray

spotted eagle ray, spotted ray, *Aetobatus narinari*

cownose ray, cow-nosed ray, *Rhinoptera bonasus*

manta, manta ray, devilfish

Atlantic manta, *Manta birostris*

devil ray, *Mobula hypostoma*

grey skate, gray skate, *Raja batis*

little skate, *Raja erinacea*

...

Stingray



Mantaray



0.005%

Random guess

1.5%

Pre Neural Networks

?

GoogLeNet

0.005%

Random guess

1.5%

Pre Neural Networks

43.9%

GoogLeNet

0.005%

Random guess

1.5%

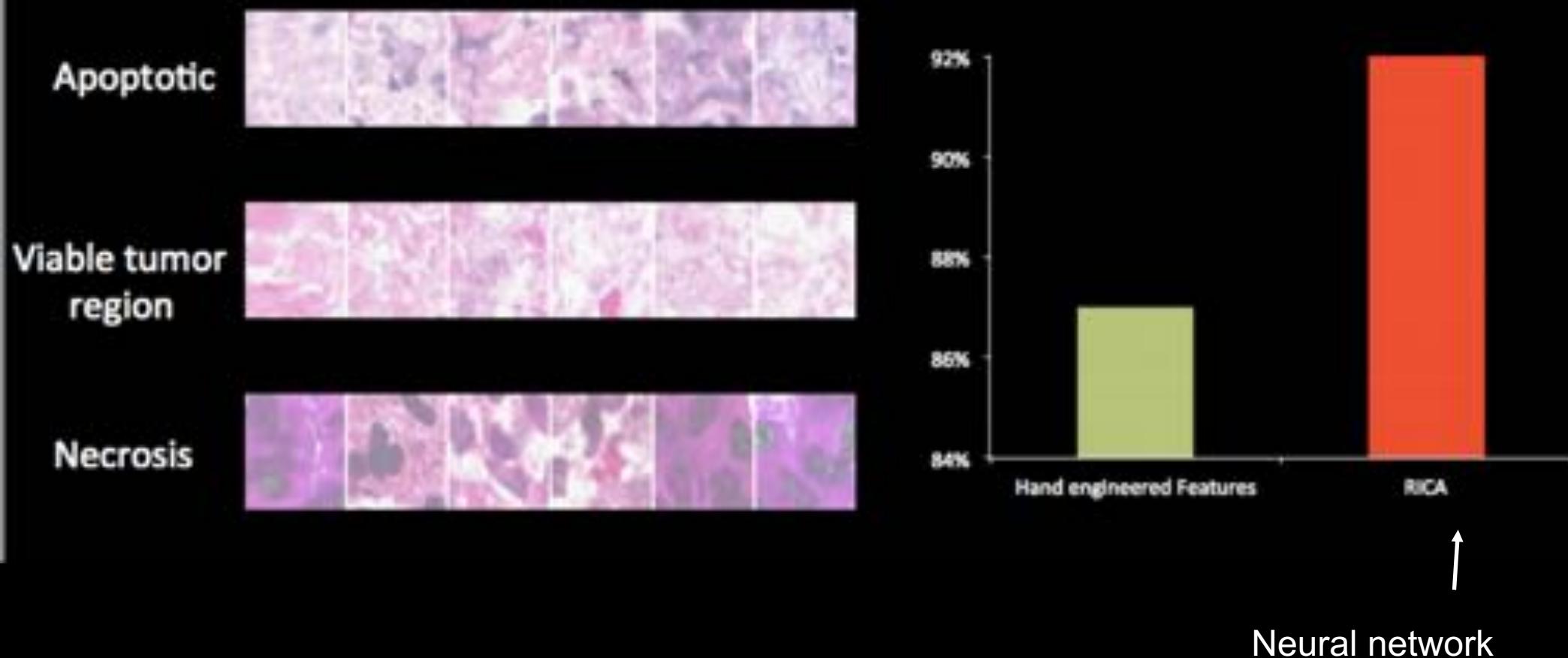
Pre Neural Networks

82.7%

NASNet



Vision has Social Implications



Where is this useful?

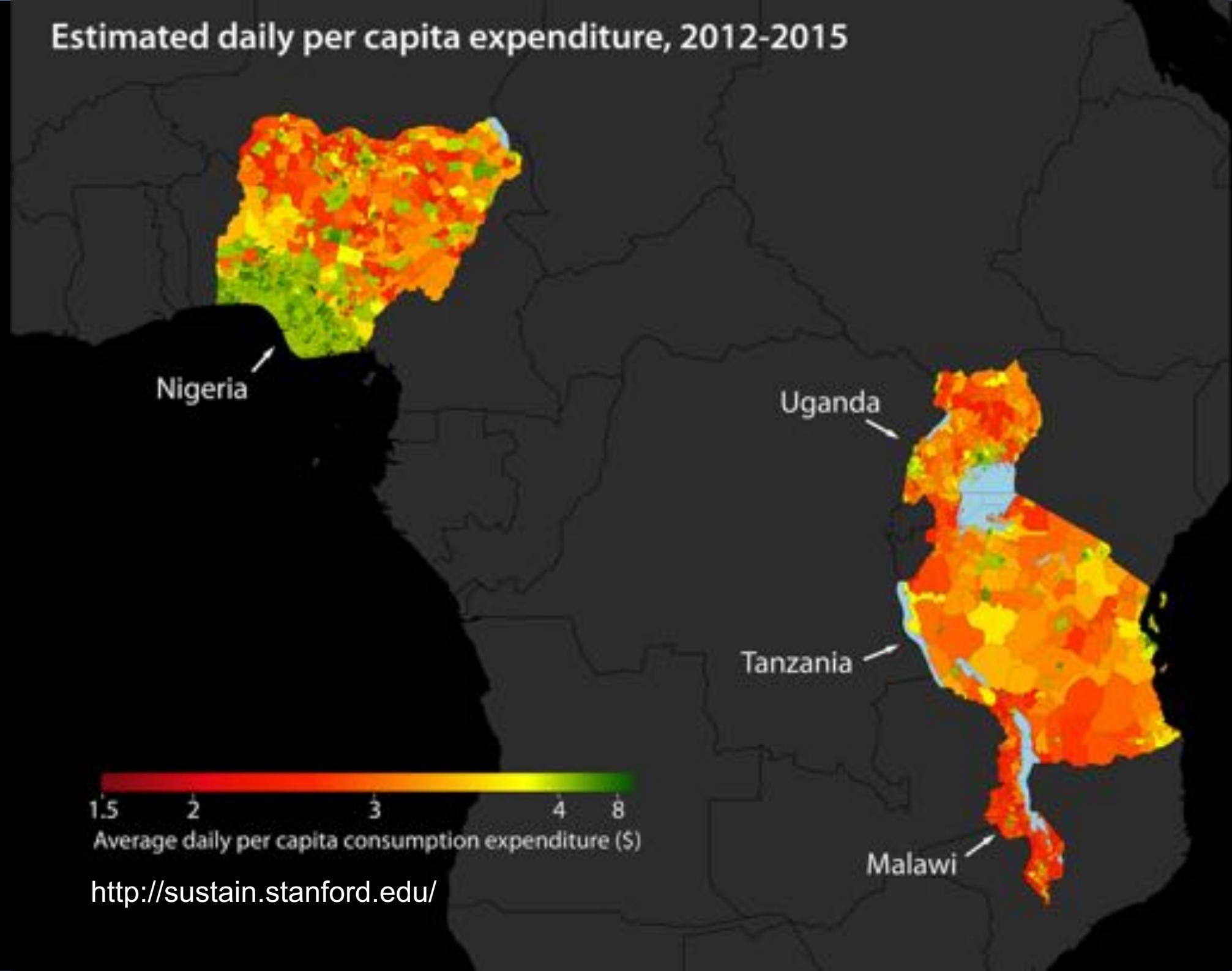


A machine learning algorithm performs **better than** the best dermatologists.

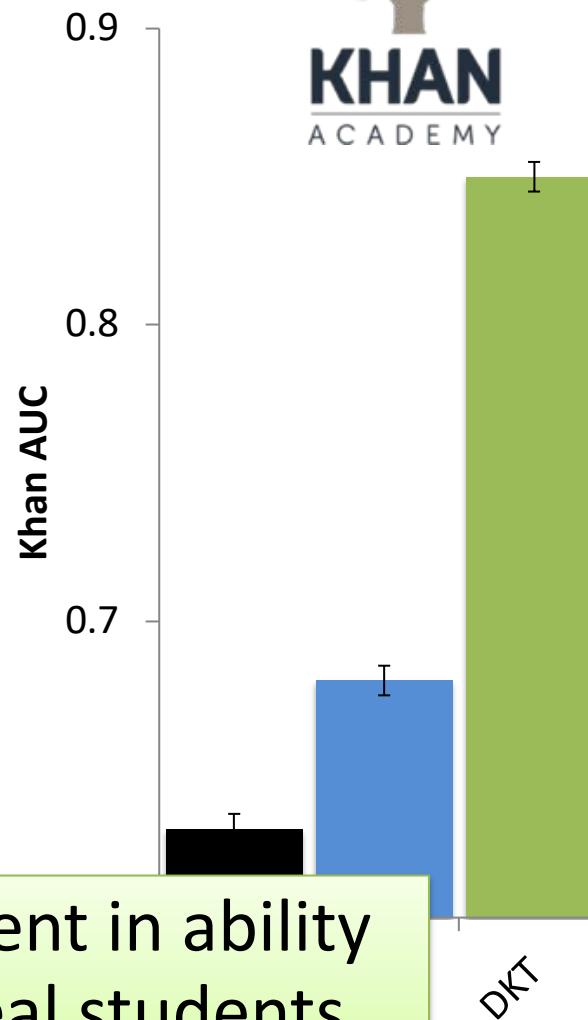
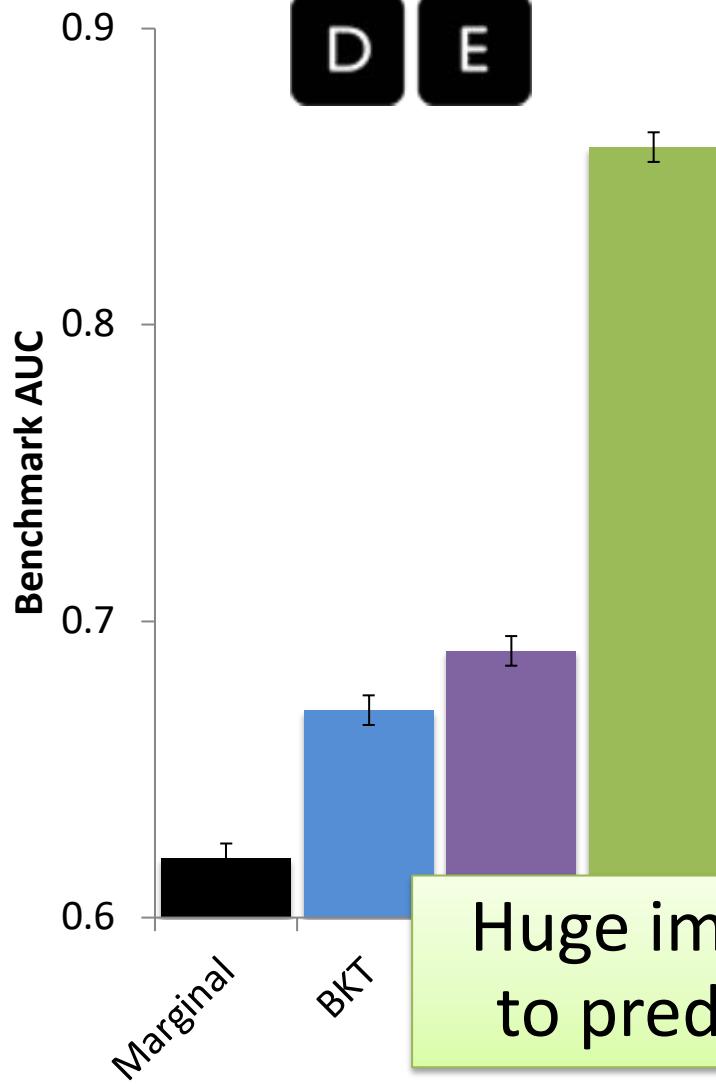
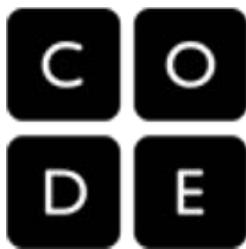
Developed this year, at Stanford.

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

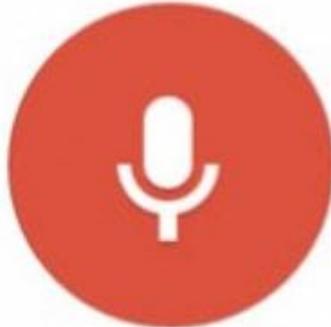


Tl;dr our brain is constantly decomposing

Told Vision Was 30 Years Out



Told Speech Was 30 Years Out



Almost perfect...

Piech, CS106A, Stanford University



What a time to be alive



Ethics in AI

The end

Export to JAR

- **JAR: Java Archive.** A compressed binary of a Java program.
 - The typical way to **distribute a Java app as a single file.**
 - Essentially just a ZIP file with Java
- Making a JAR of your project in E
 - File → Export ... → Java → **Runnable JAR File**
- *see handout on course web site*

