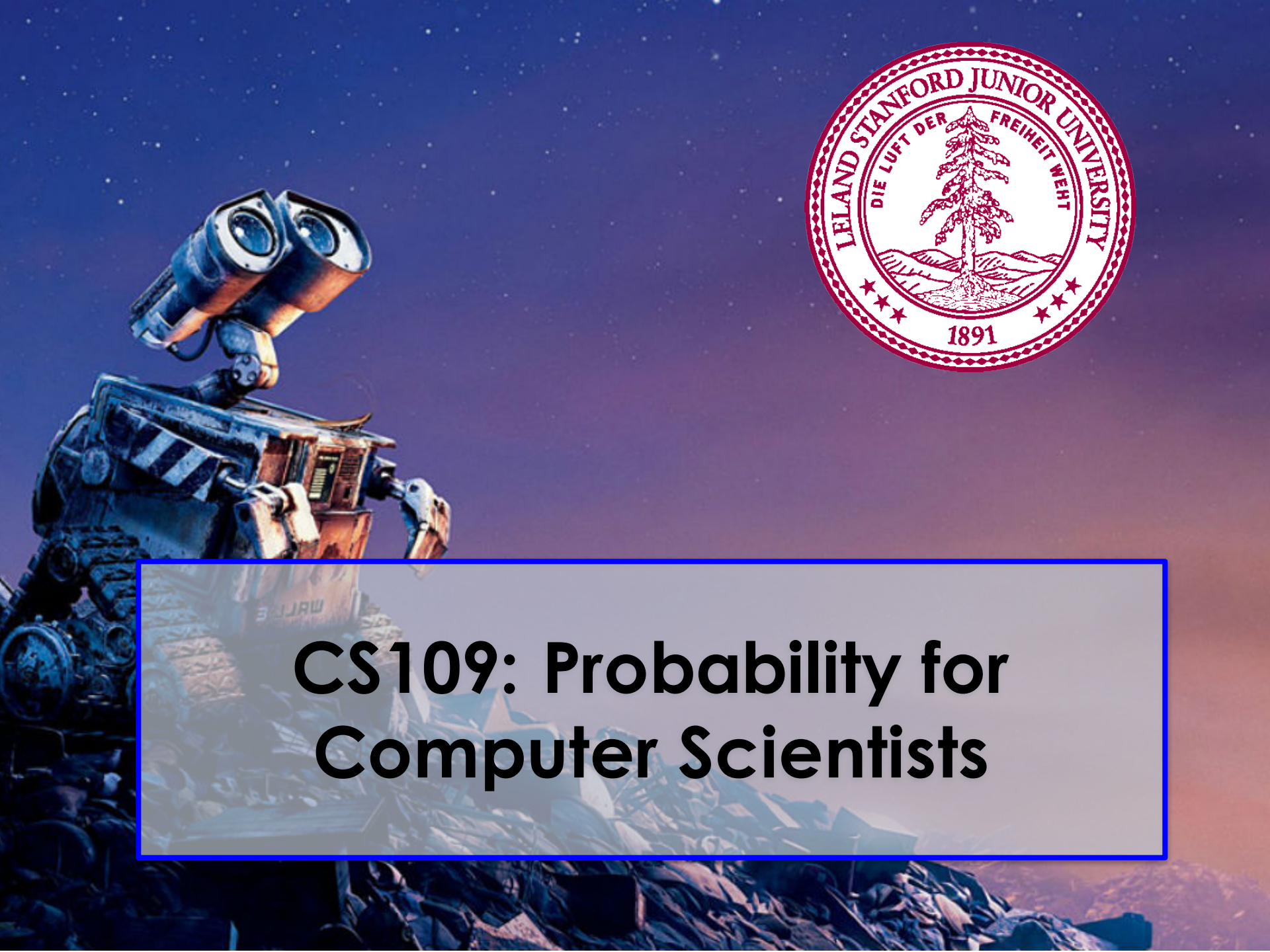


A background image of the WALL-E robot from the Pixar movie "WALL-E". WALL-E is a small, boxy robot with large, expressive eyes, standing on a rocky, desolate landscape under a starry night sky. The robot is positioned on the left side of the frame, looking towards the right.

CS109: Probability for Computer Scientists

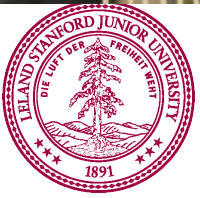
Chris Piech

My parents are interesting folks

I originally concentrated in graphics and worked at Pixar

- Childhood: Nairobi, Kenya
- High School: Kuala Lumpur, Malaysia
- Stanford University Ph.D. in Deep Learning
- Research lab on AI for Social Good
- Researcher in the MSR cognitive science lab

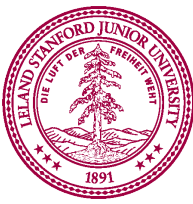
The problem I really want to solve is to make high quality more education accessible



I Took the First CS109 Class



Back when I looked like this 😊



Teaching Assistants

Several outstanding TAs:

William Monroe (Head TA)

William Chen

Govinda Dasu

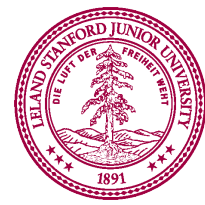
Luke Johnston

Winnie Lin

Michela Meister

Priyanka Nigam

Benjamin Ulmer

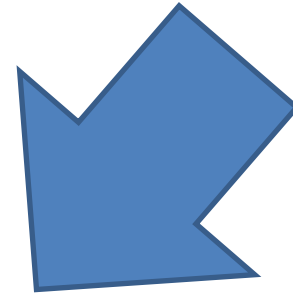
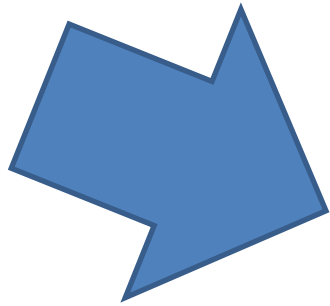


Who are you?

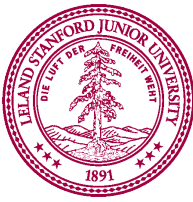
Course mechanics

(this is a light version. Please read the handout for details).

Essential Information



cs109.stanford.edu



Are you in the right place?

Prereqs

What you really need:

CS106B/X:

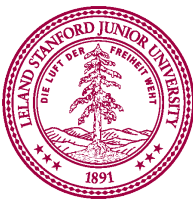
- Recursion
- Hash Tables
- Binary Trees
- Programming

CS103:

- Proof techniques (induction)
- Set theory

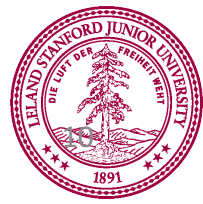
Math 51 or CME 100

- Multivariate differentiation
- Multivariate integration
- Basic facility with linear algebra (vectors)



Staff Contact

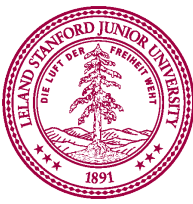
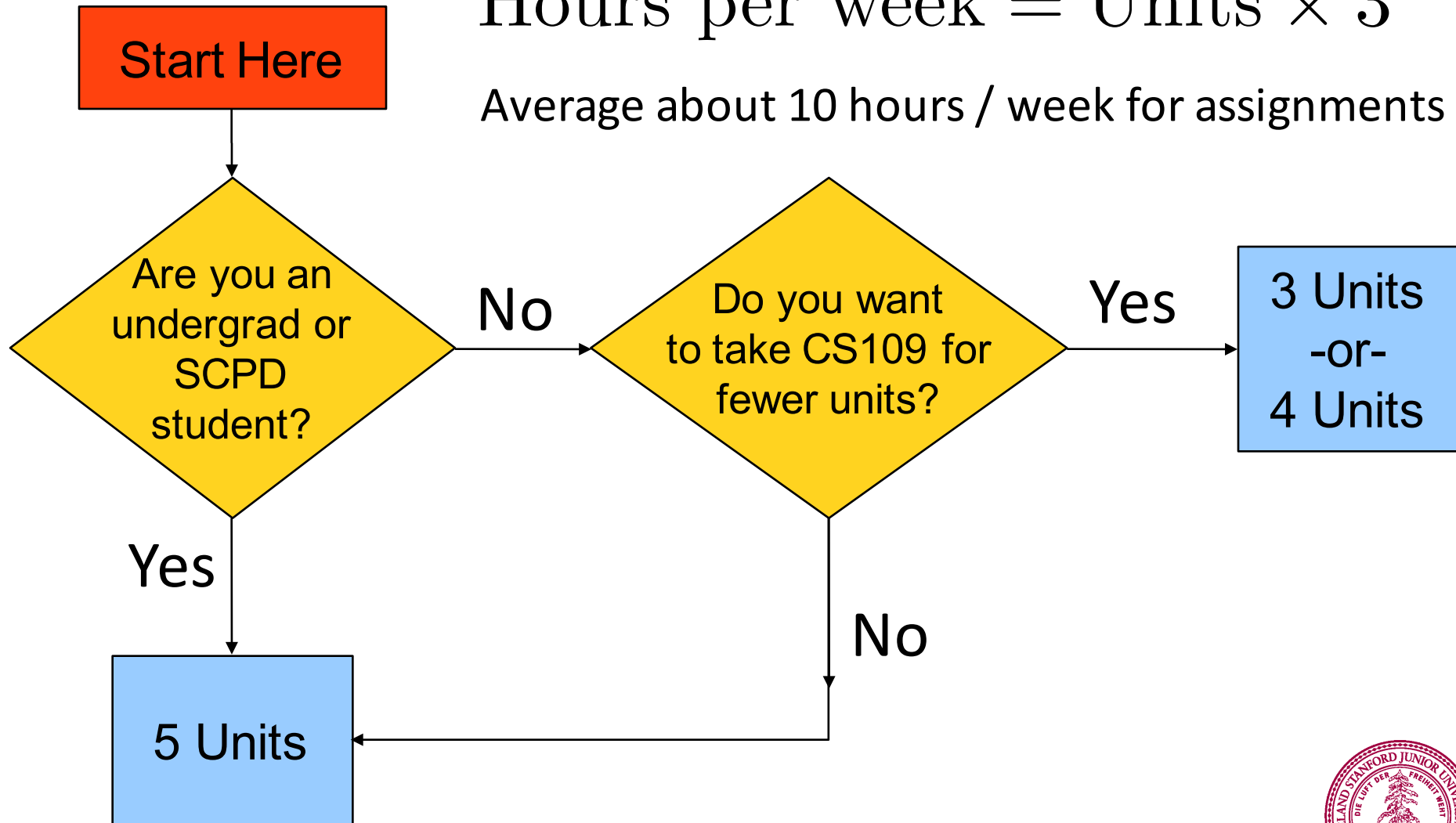
- Post to Piazza
- Go to Working Office Hours
- Email cs109@cs.stanford.edu
- Email Chris or go to his office for course level issues.



CS109 Units

$$\text{Hours per week} = \text{Units} \times 3$$

Average about 10 hours / week for assignments



Video Taped



Class Breakdown

45%

6 Assignments

20%

Midterm

Tuesday May 9th, 7-9pm

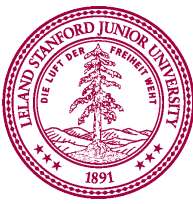
35%

Final

Tuesday June 13th, 8:30-11:30am

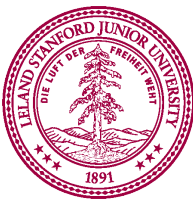
5%

Lecture Participation



Late Days

2



The Student Honor Code

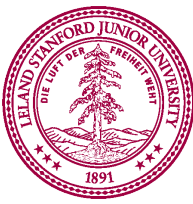
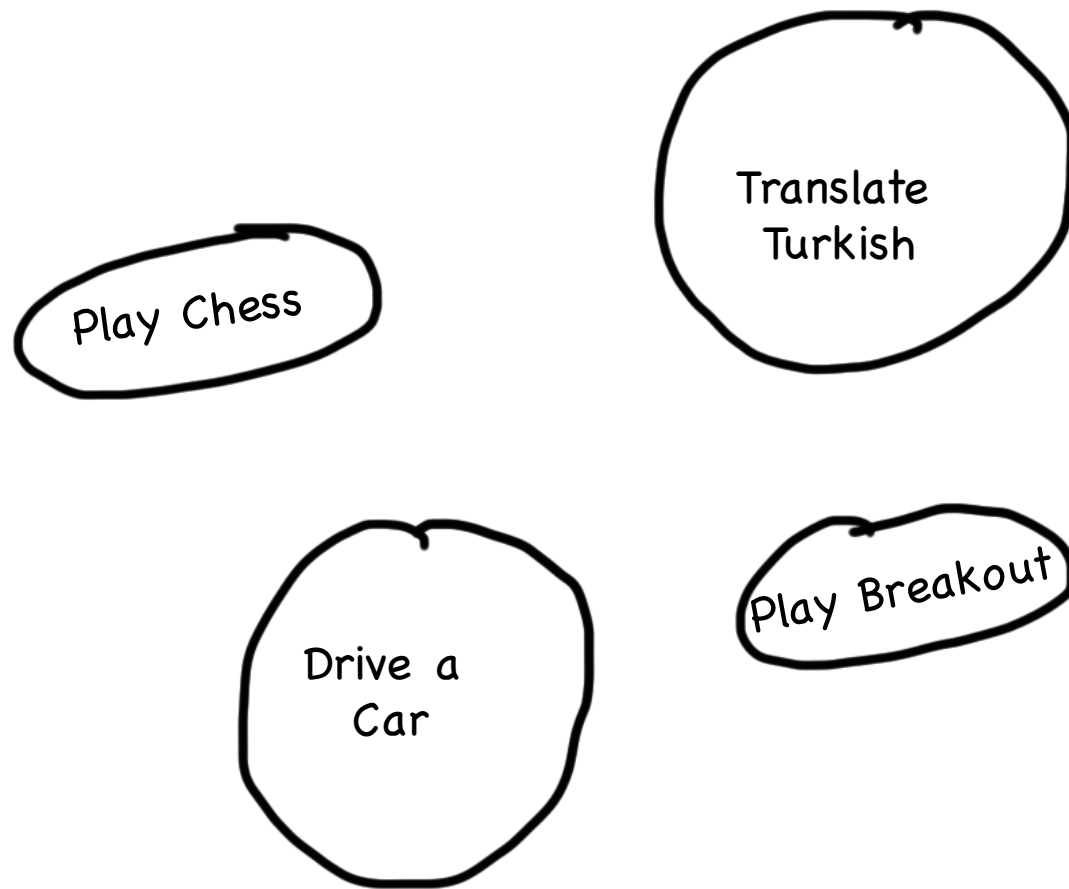
Story of Modern AI

Modern AI or, How we learned to combine probability and programming

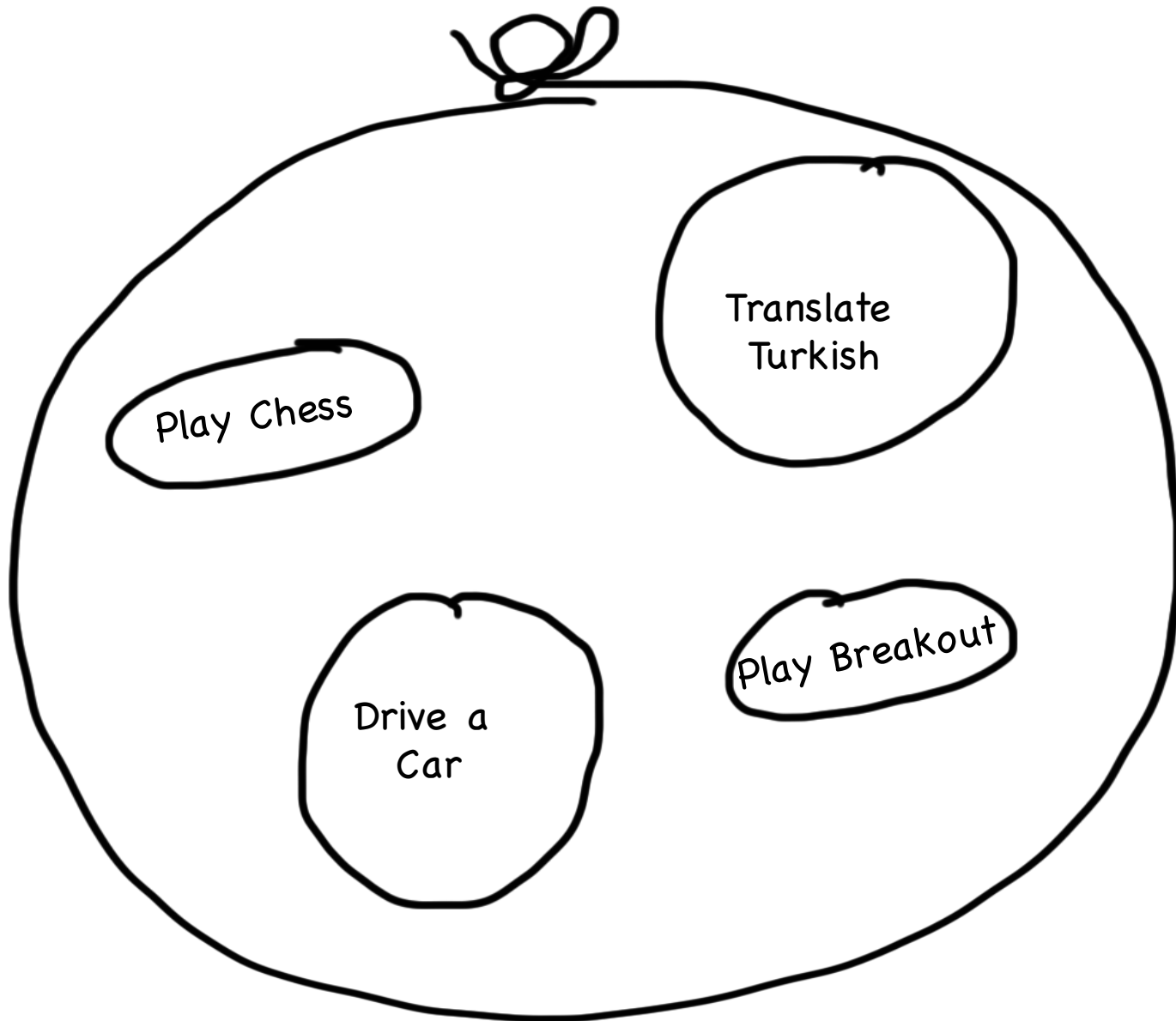
Brief History



Narrow Intelligence



General Intelligence

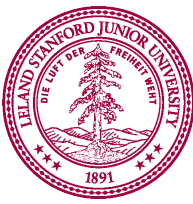
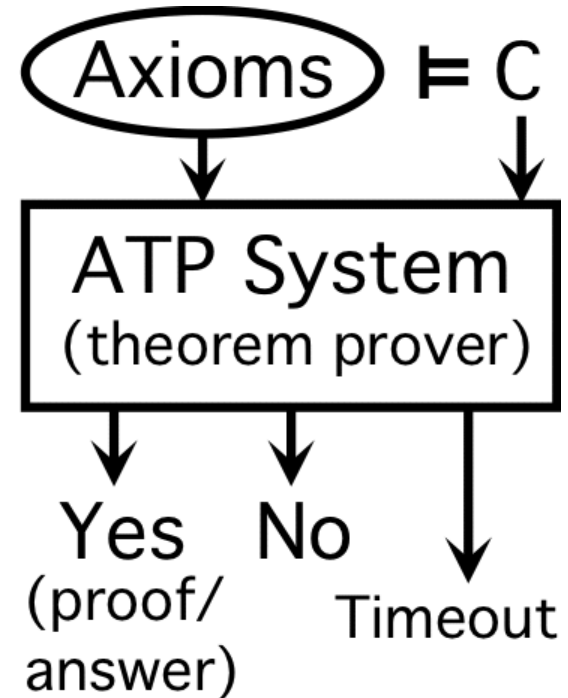


Early Optimism 1950

1952

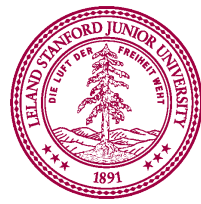


1955



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.

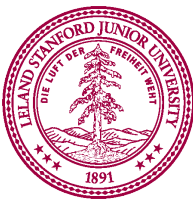


(Russian)



The vodka is good but the meat is rotten.

The world is too complex



BRACE YOURSELVES



WINTER IS COMING

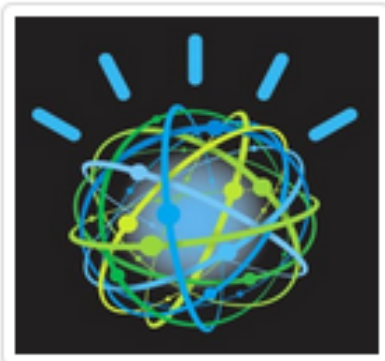
Big Milestones Pt 1



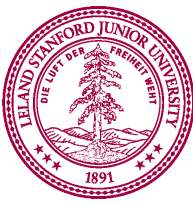
1997 Deep Blue



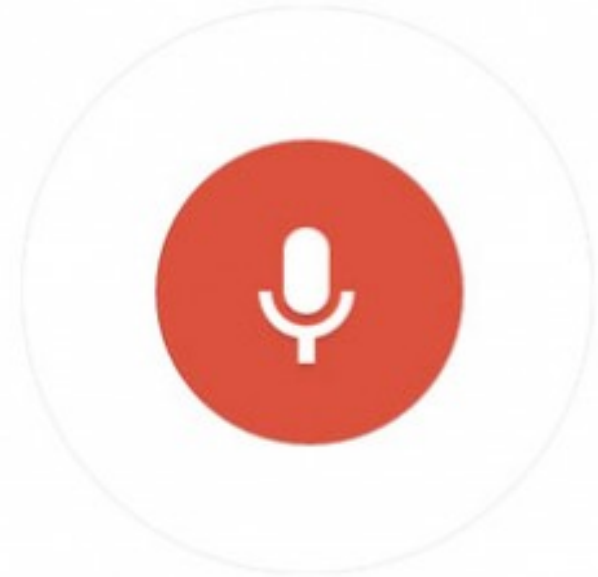
2005 Stanley



2011 Watson



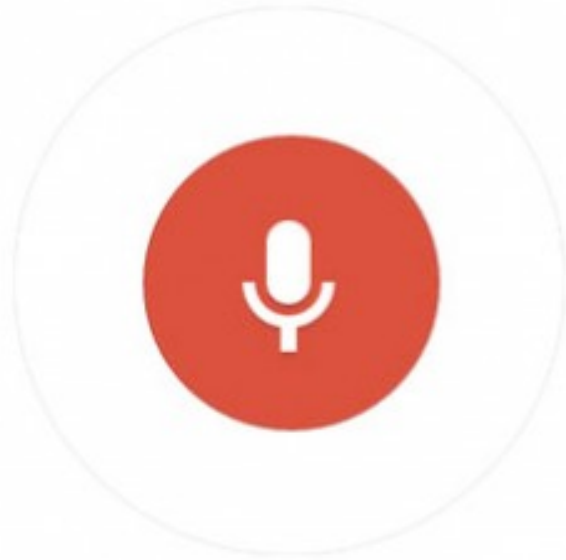
Big Milestones Pt 2



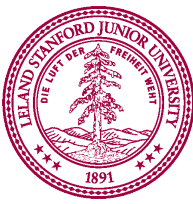
Told Vision Was 30 Years Out



Told Speech Was 30 Years Out



Almost perfect...



Huge Progress



Passed the Turing Test?

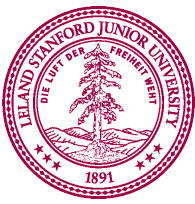


EUGENE GOOSTMAN
THE WEIRDEST CREATURE IN THE WORLD

33% of judges thought Eugene was a human

When only given 5 minutes...

Piech, CS106A, Stanford University



What is going on?

Two Great Ideas

1. Probability from Examples

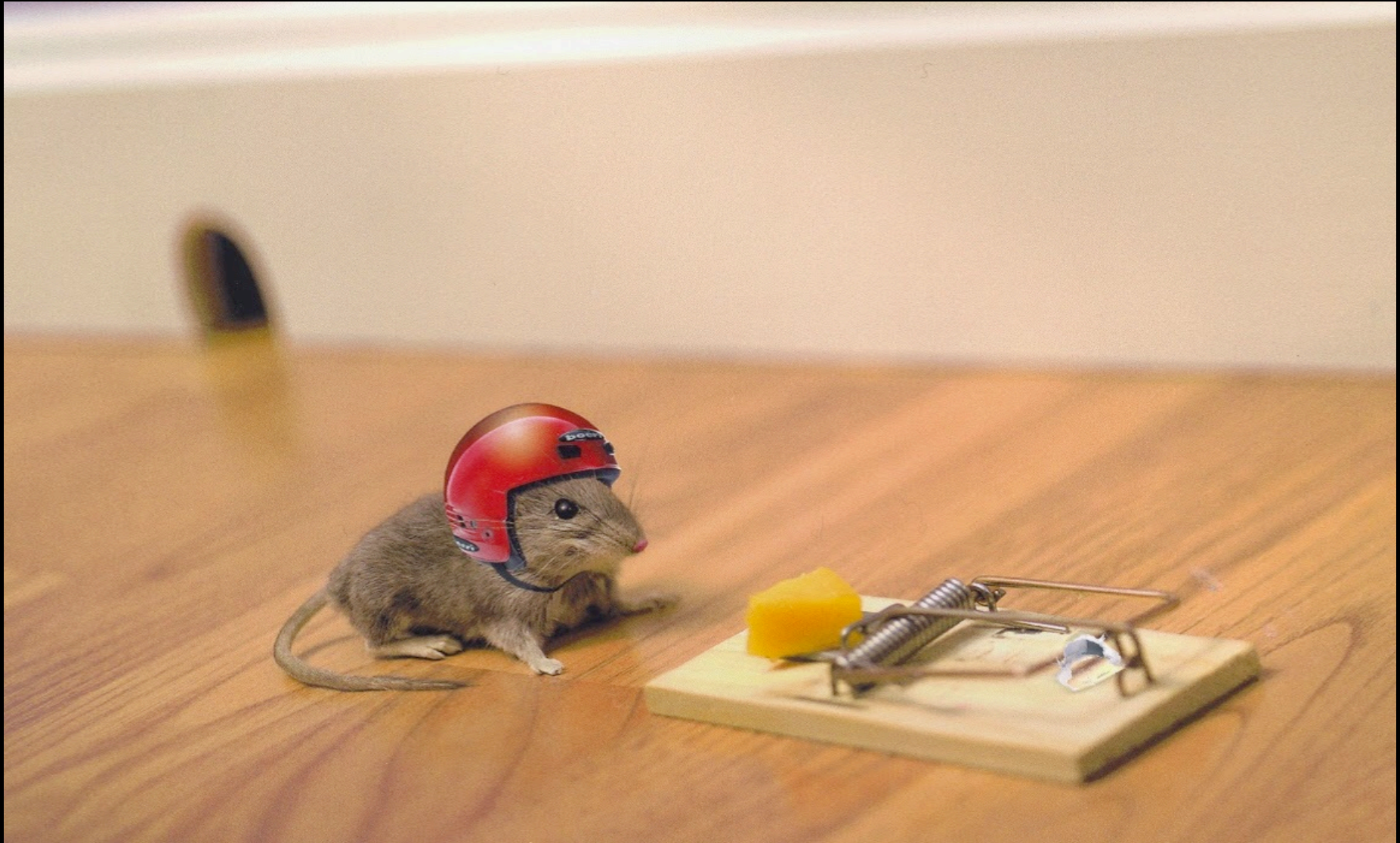
2. Artificial Neurons

Two Great Ideas

1. Probability from Examples

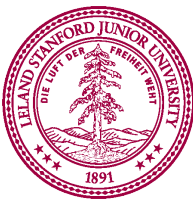
2. Artificial Neurons

1. Probability From Examples

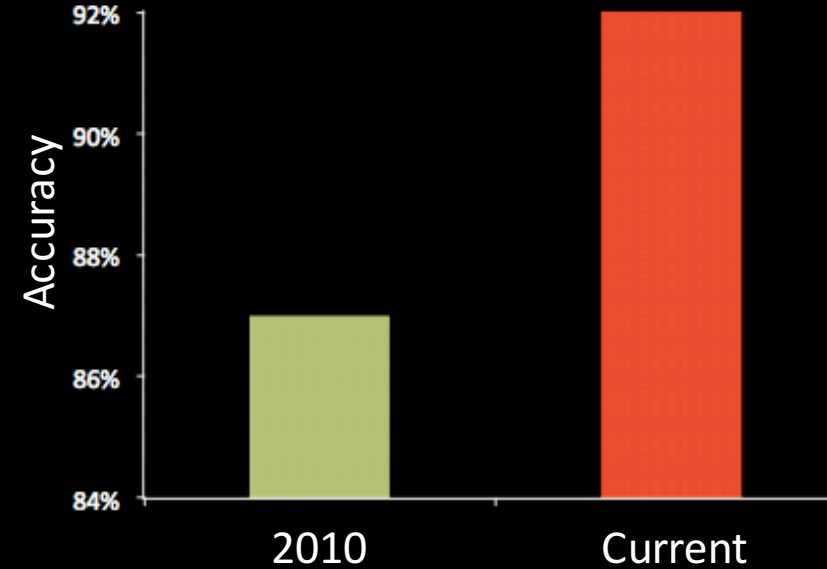
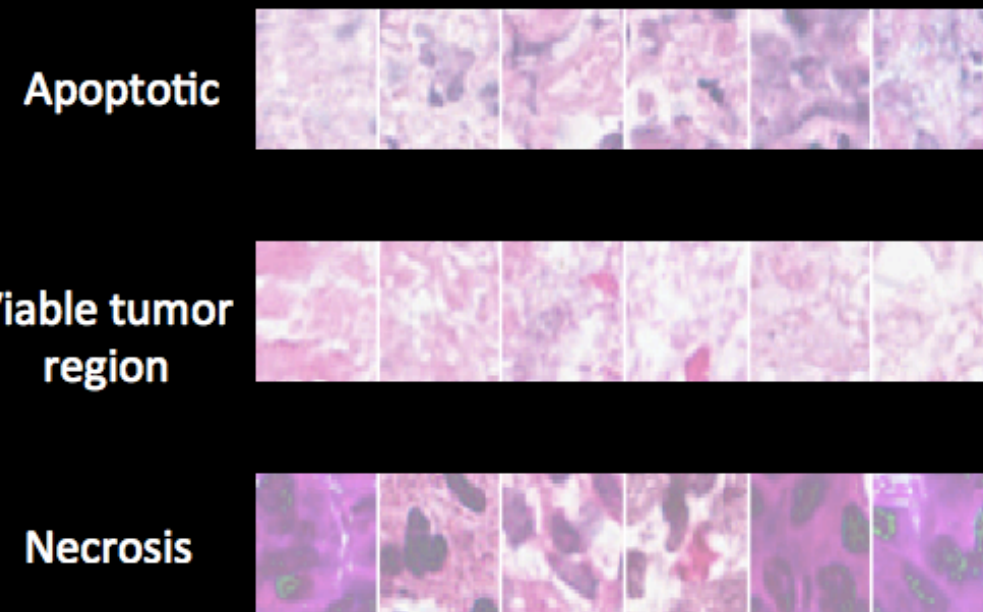


Machine Learning

Basically just a rebranding of statistics
and probability.



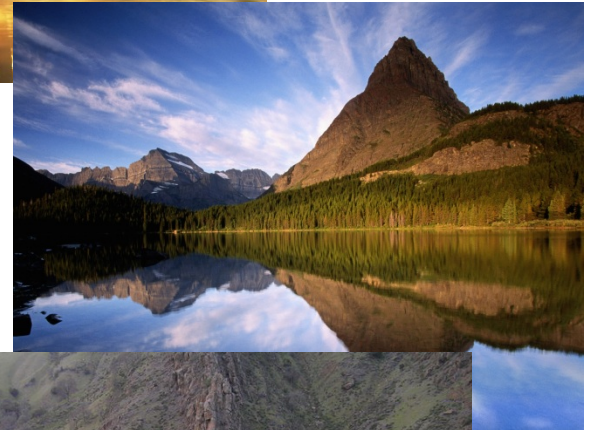
Vision: Detect Cancer



Vision: Detect Digits



Vision: Detect Faces



Vision is Hard

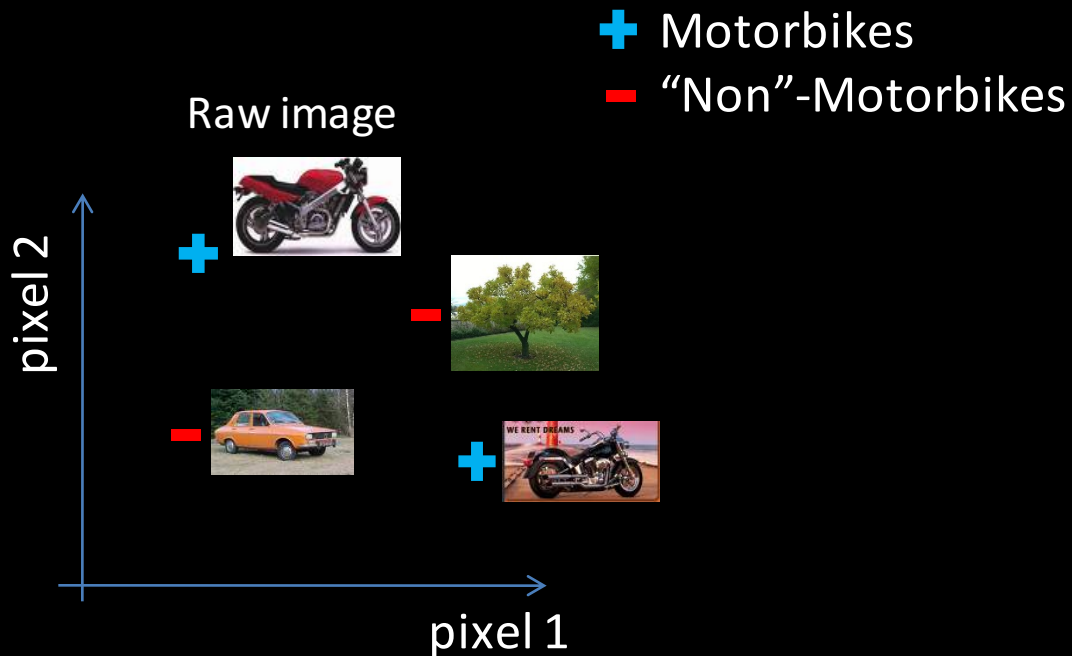
You see this:



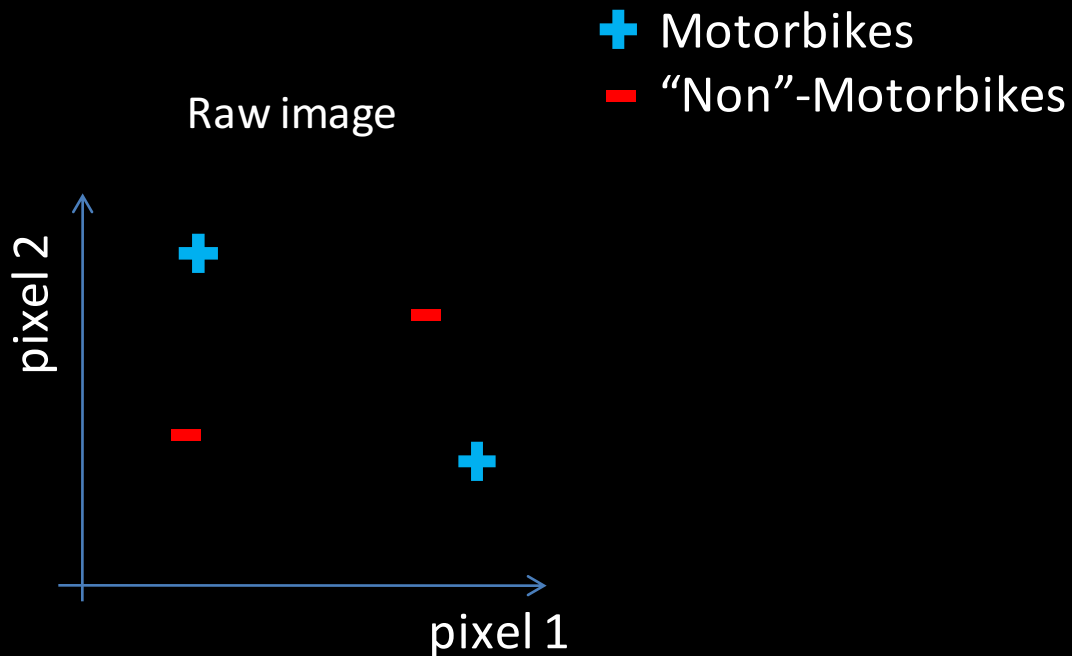
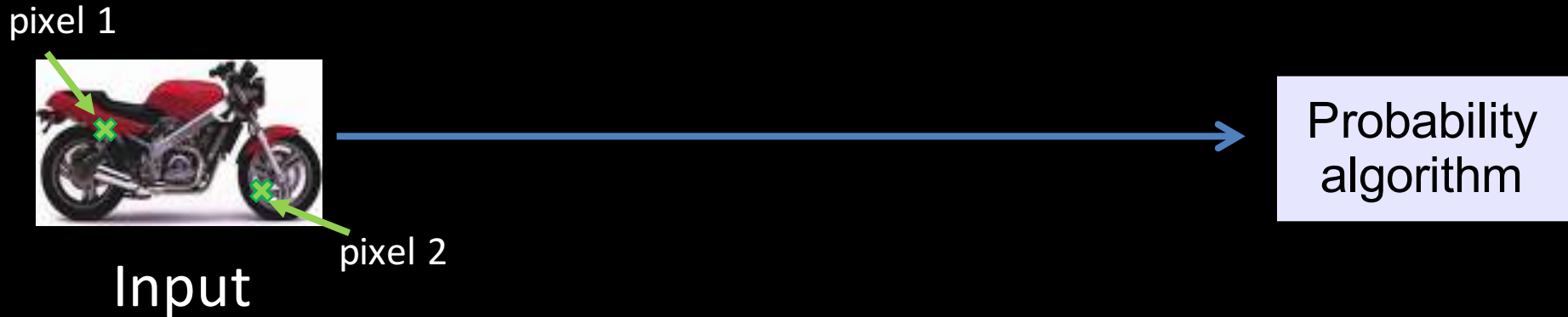
But the camera sees this:

194	210	201	212	199	213	215	195	178	158	182	209
180	189	190	221	209	205	191	167	147	115	129	163
114	126	140	188	176	165	152	140	170	106	78	88
87	103	115	154	143	142	149	153	173	101	57	57
102	112	106	131	122	138	152	147	128	84	58	66
94	95	79	104	105	124	129	113	107	87	69	67
68	71	69	98	89	92	98	95	89	88	76	67
41	56	68	99	63	45	60	82	58	76	75	65
20	43	69	75	56	41	51	73	55	70	63	44
50	50	57	69	75	75	73	74	53	68	59	37
72	59	53	66	84	92	84	74	57	72	63	42
67	61	58	65	75	78	76	73	59	75	69	50

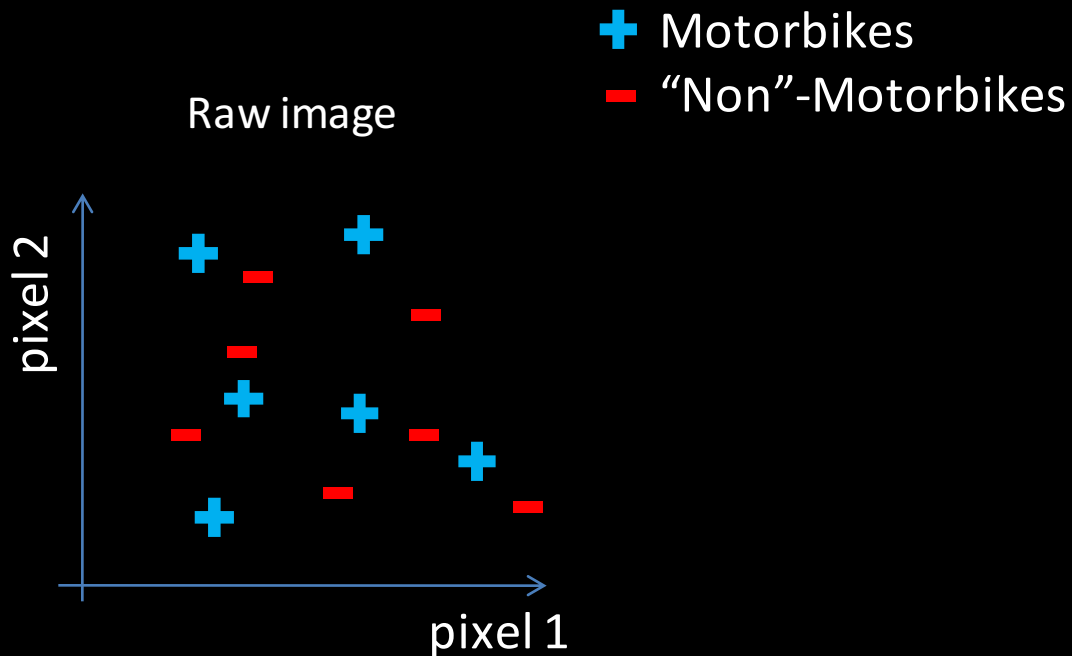
Machine Learning



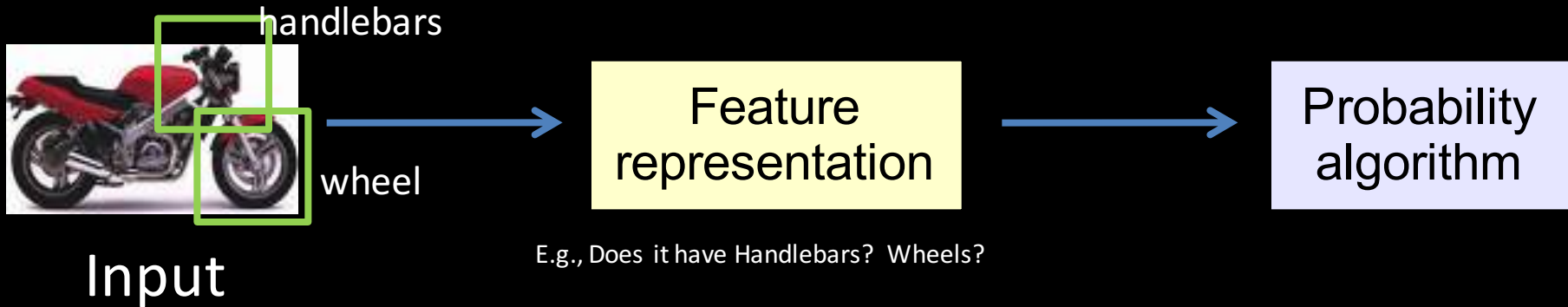
Machine Learning



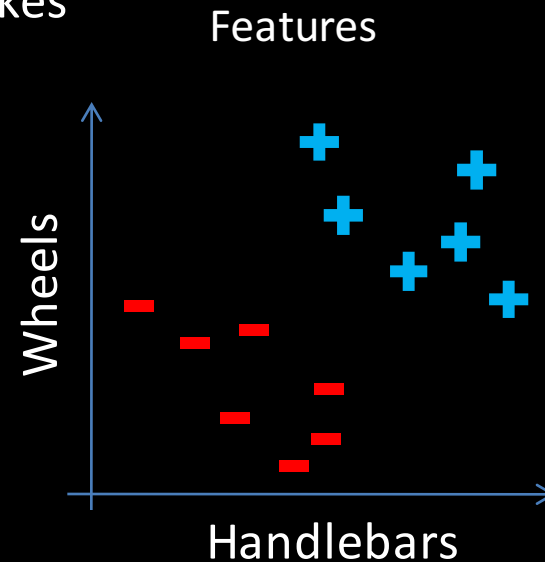
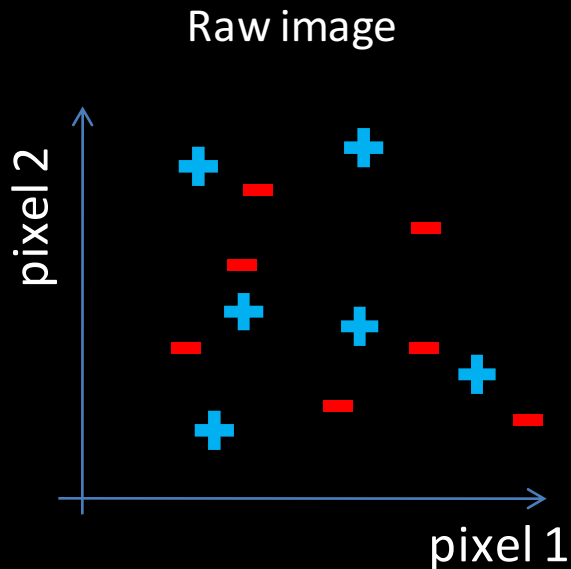
Machine Learning



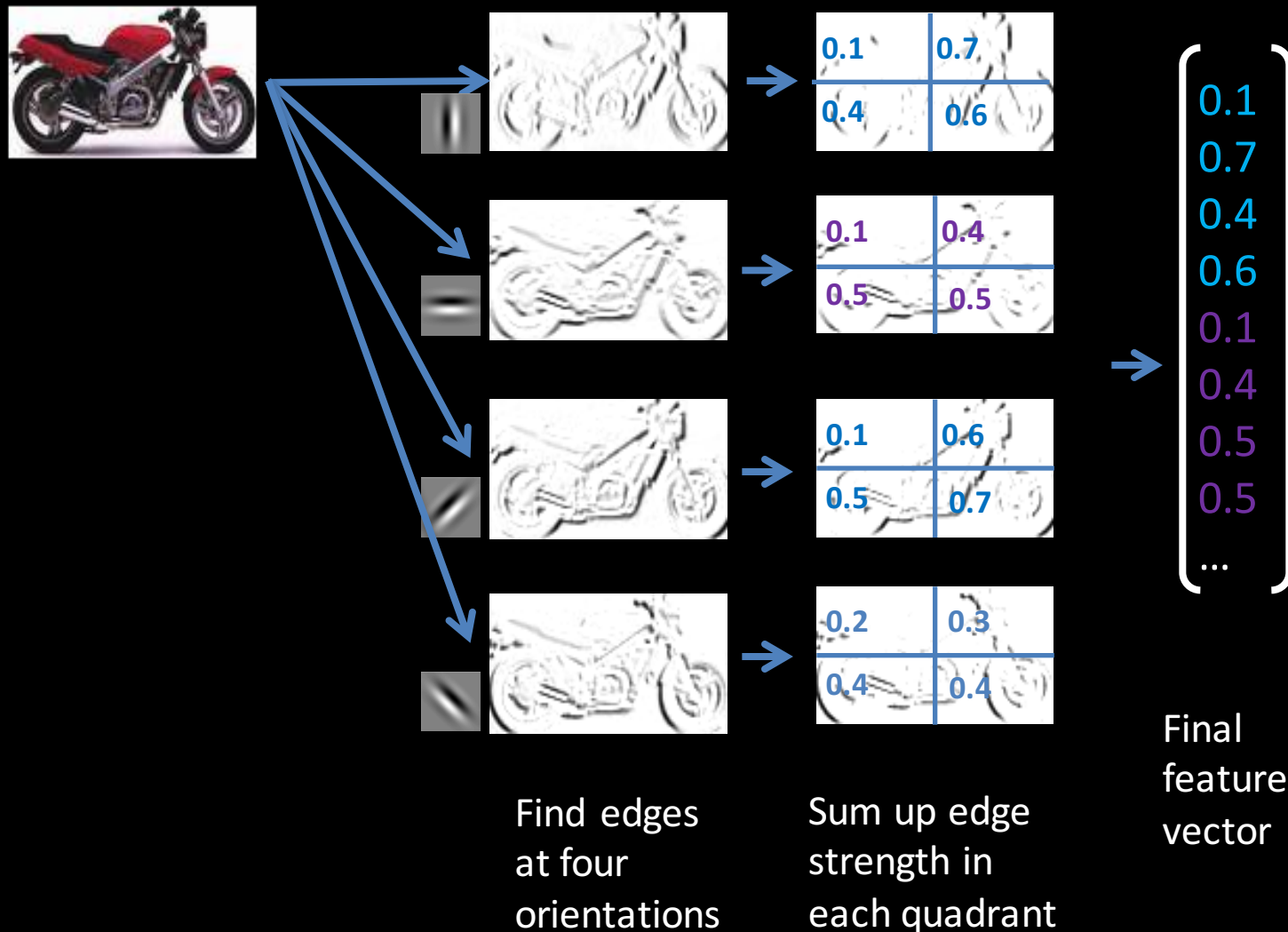
Machine Learning



+ Motorbikes
- "Non"-Motorbikes



Human Created Features



Some Great Thinkers



Daphne Koller



Mehran Sahami

Straight ML Not Perfect...

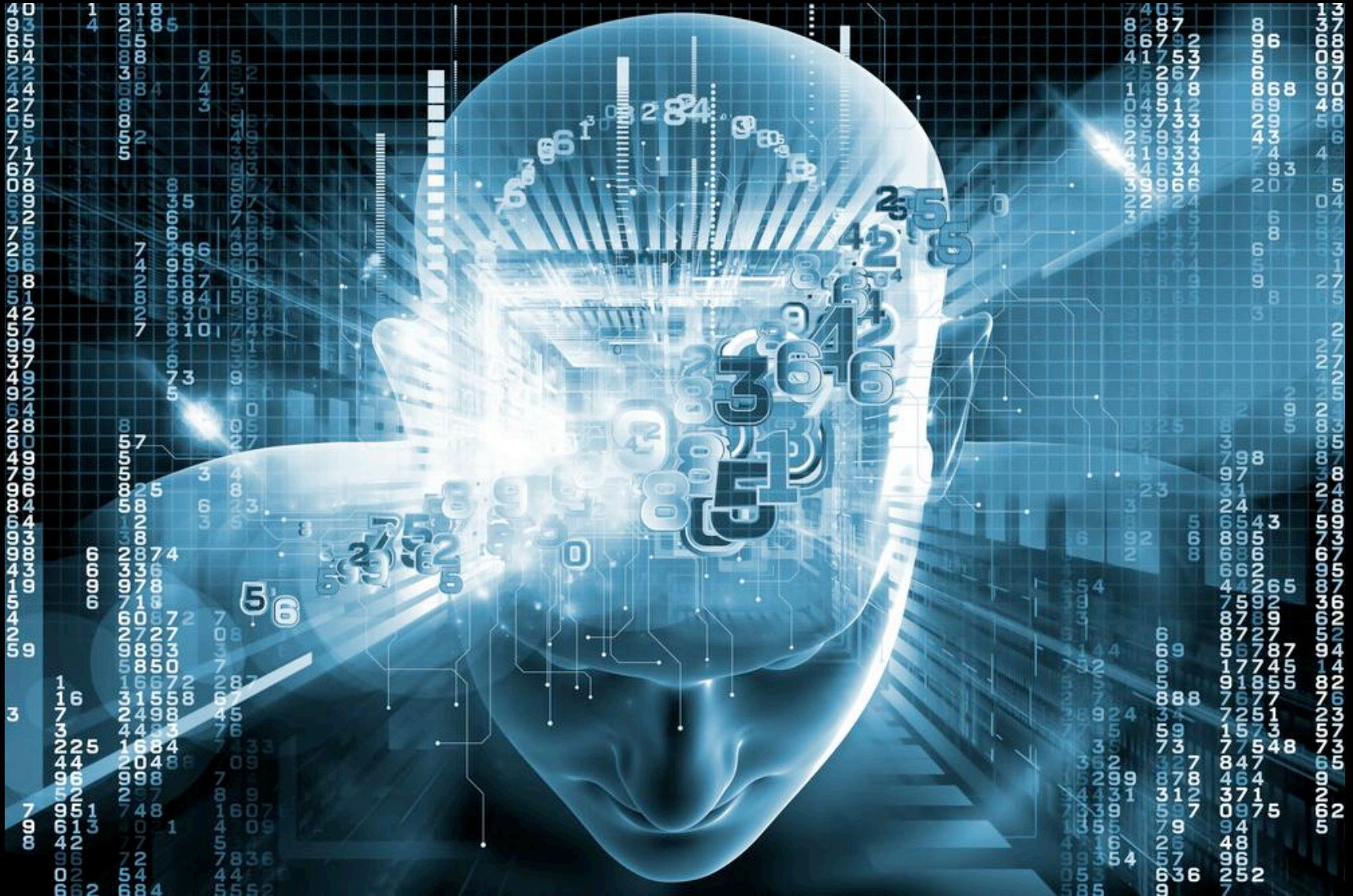


Two Great Ideas

1. Probability from Examples

2. Artificial Neurons

2. Artificial Neurons



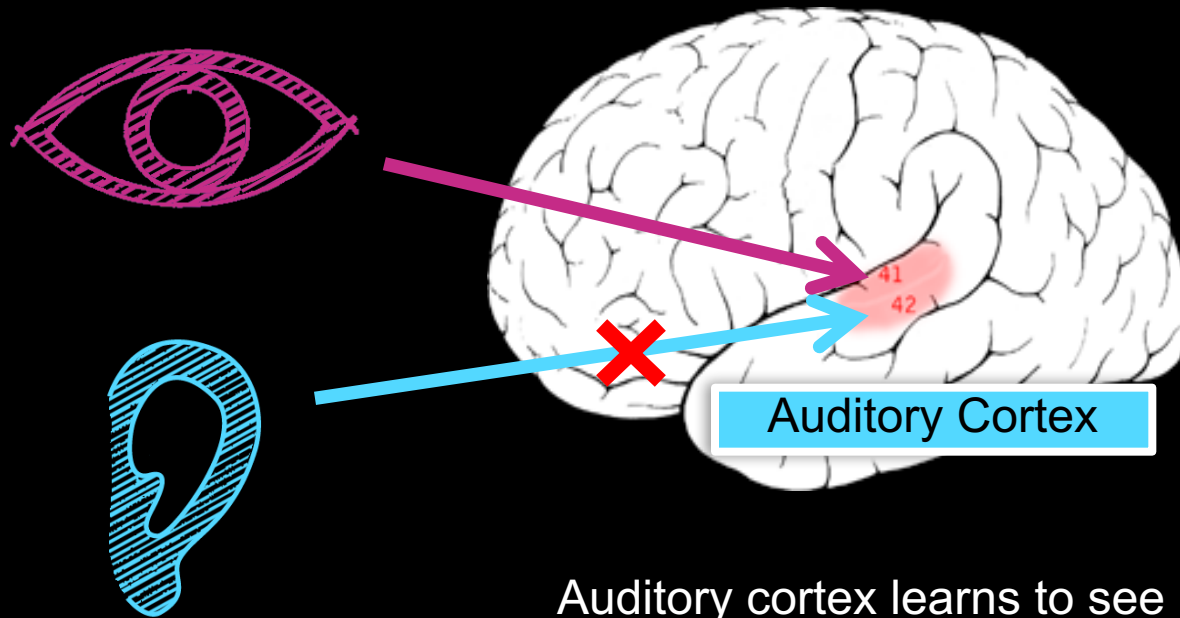
One Algorithm Hypothesis

Much of perception in the brain can be explained with a single learning algorithm.



[Andrew Ng]

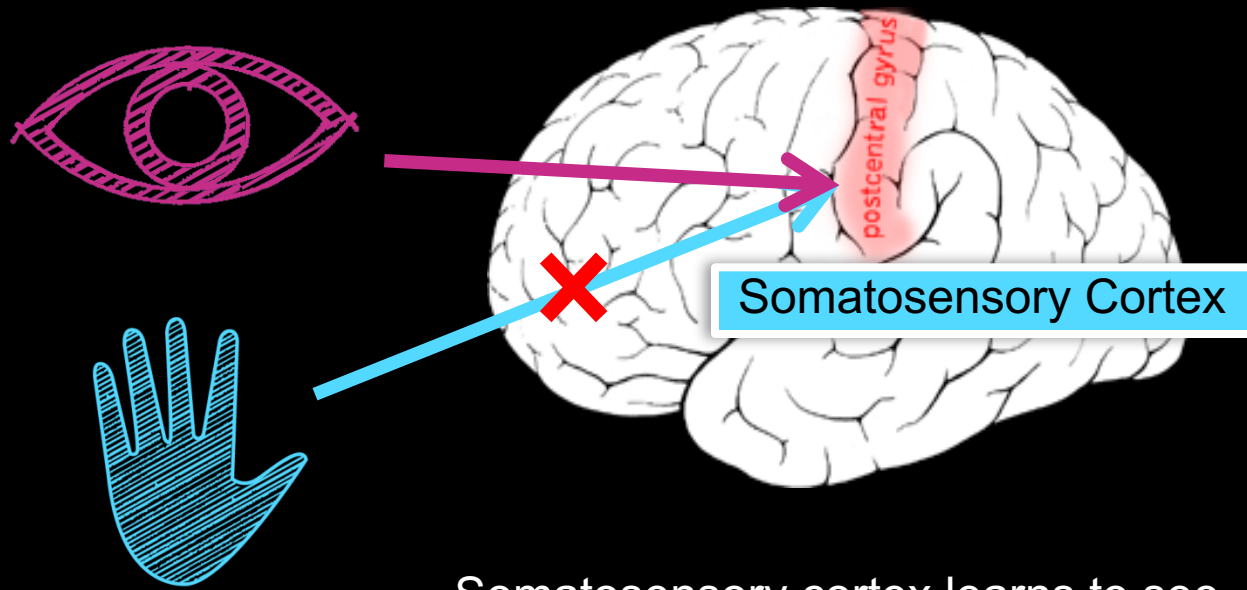
One Algorithm Hypothesis



[Roe et al., 1992]

[Andrew Ng]

One Algorithm Hypothesis



Somatosensory cortex learns to see

[Metin & Frost, 1989]

[Andrew Ng]

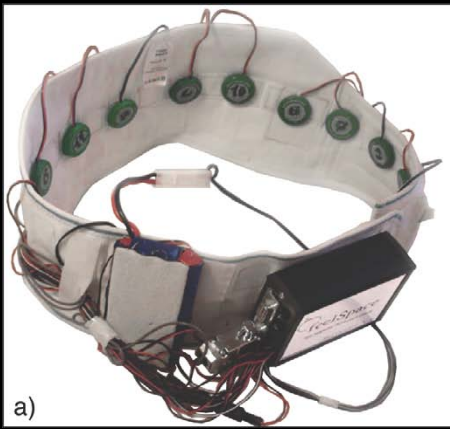
Sensor Representations



Seeing with your tongue



Human echolocation (sonar)



a)



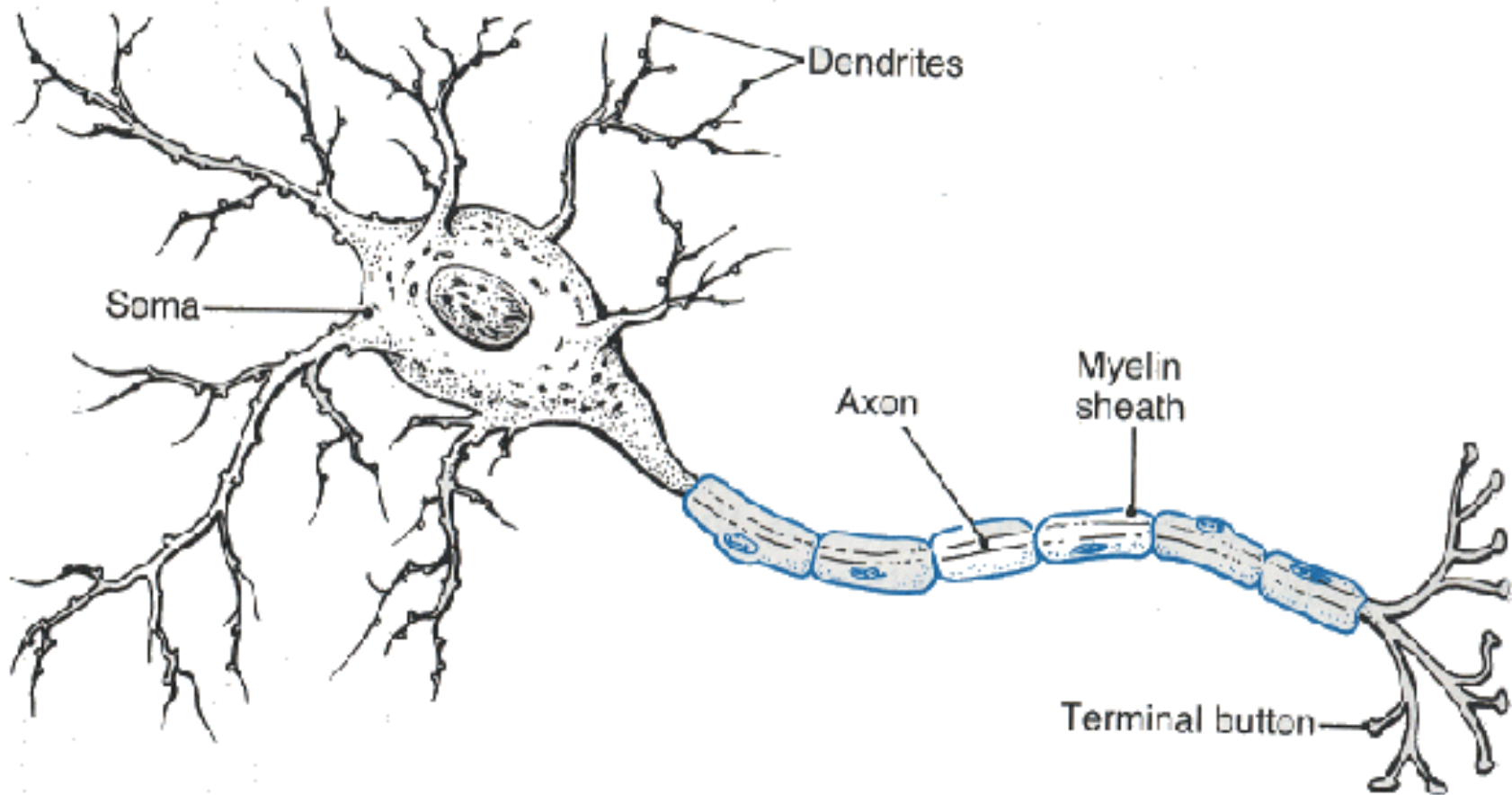
Haptic belt: Direction sense



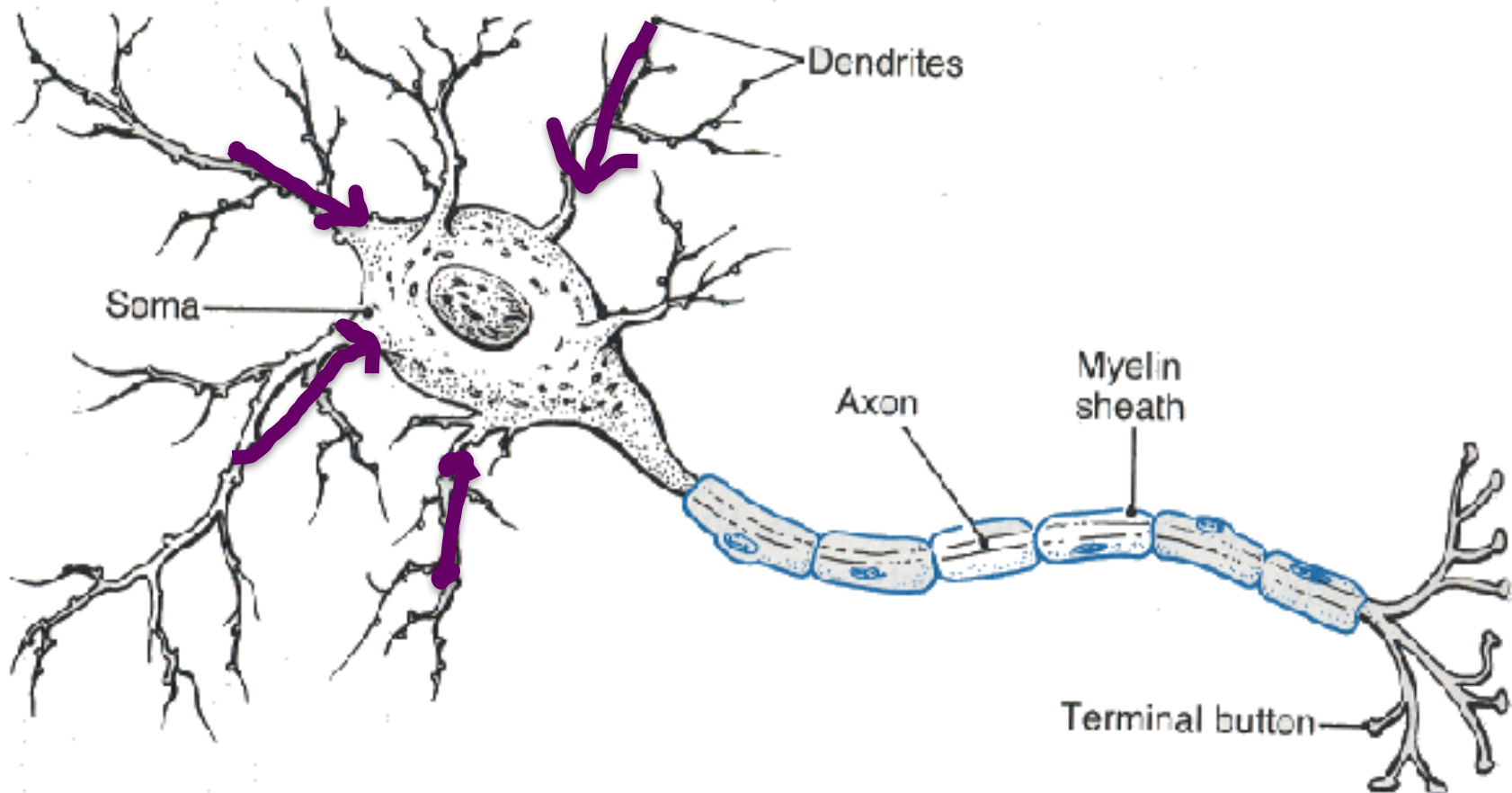
a

Implanting a 3rd eye

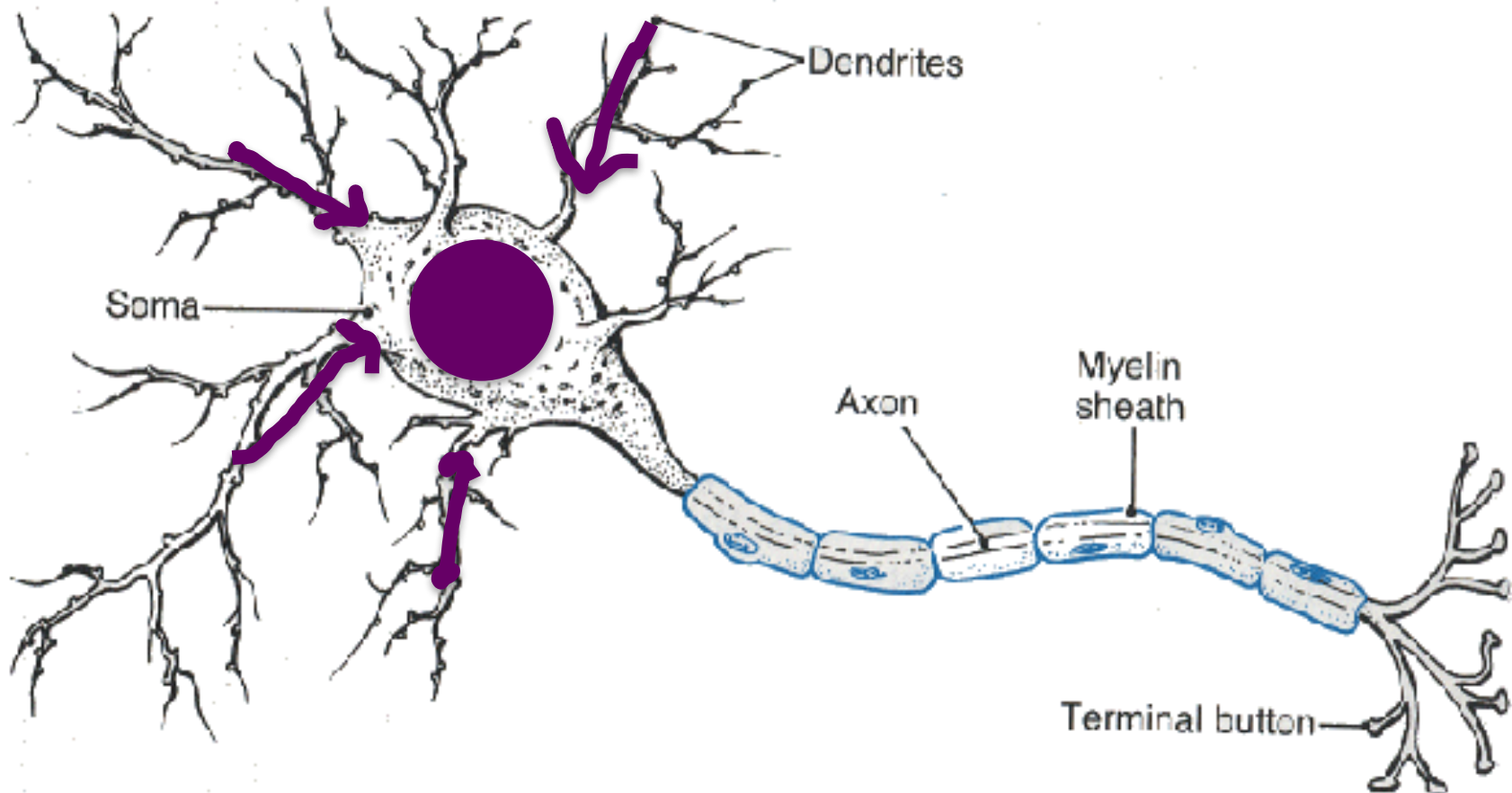
Neuron



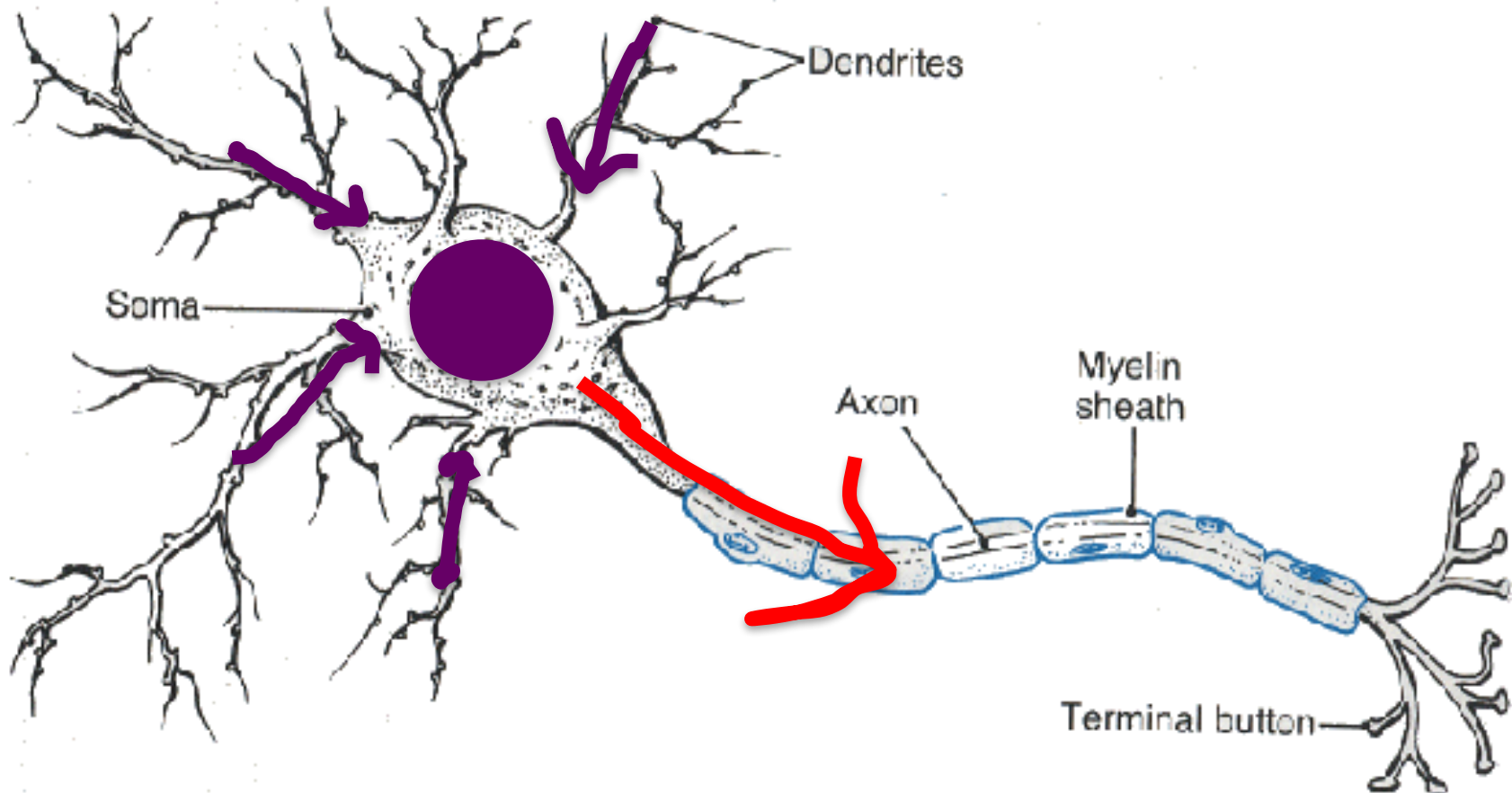
Neuron



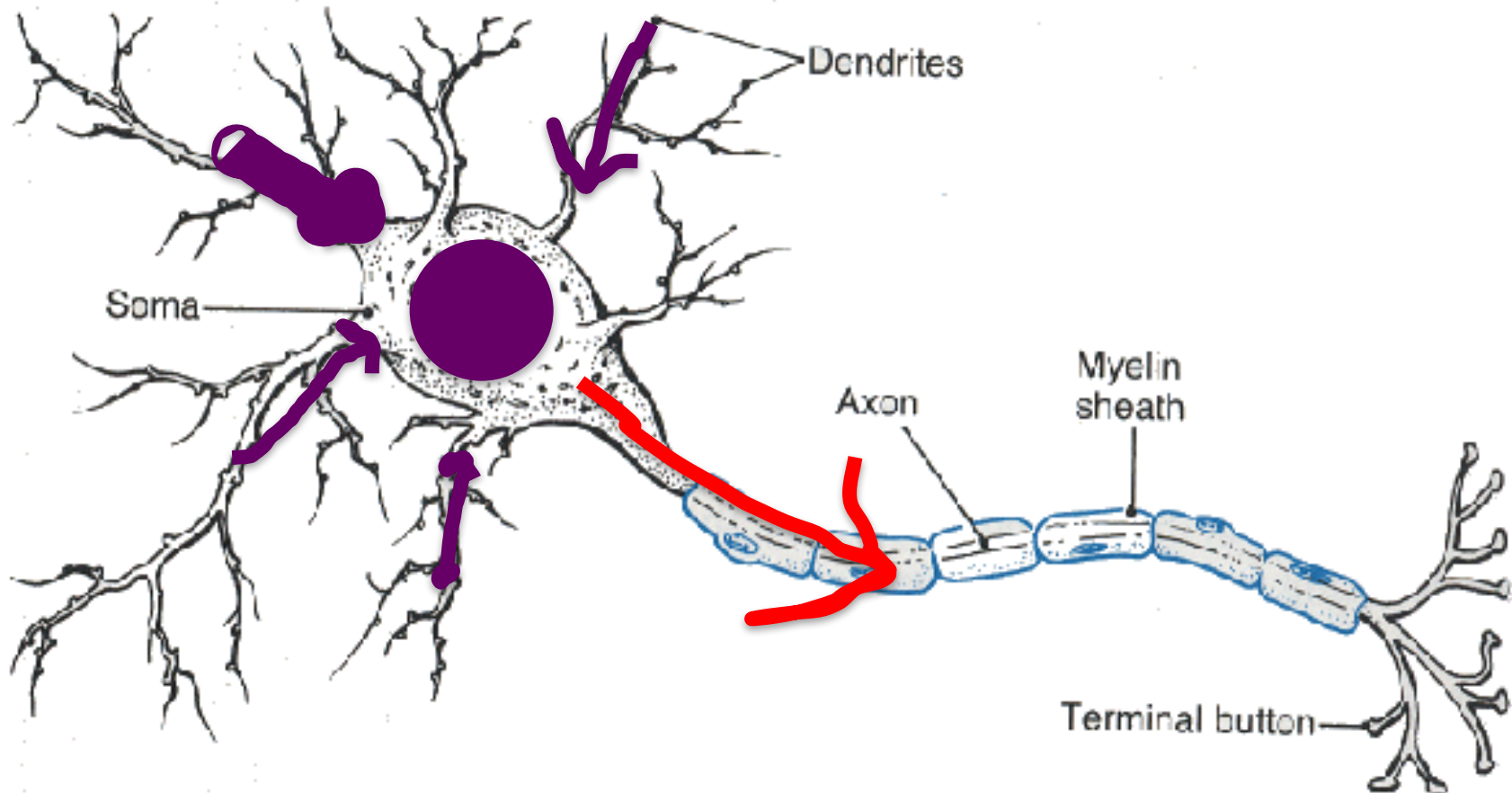
Neuron



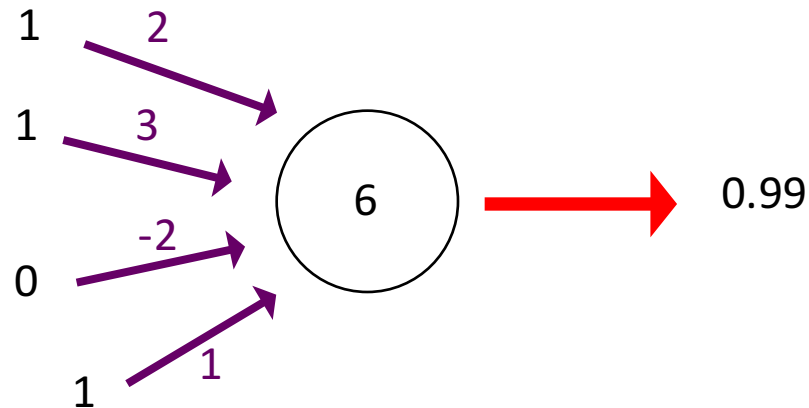
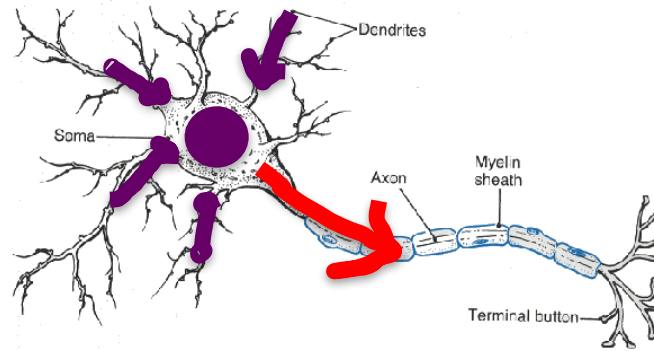
Neuron



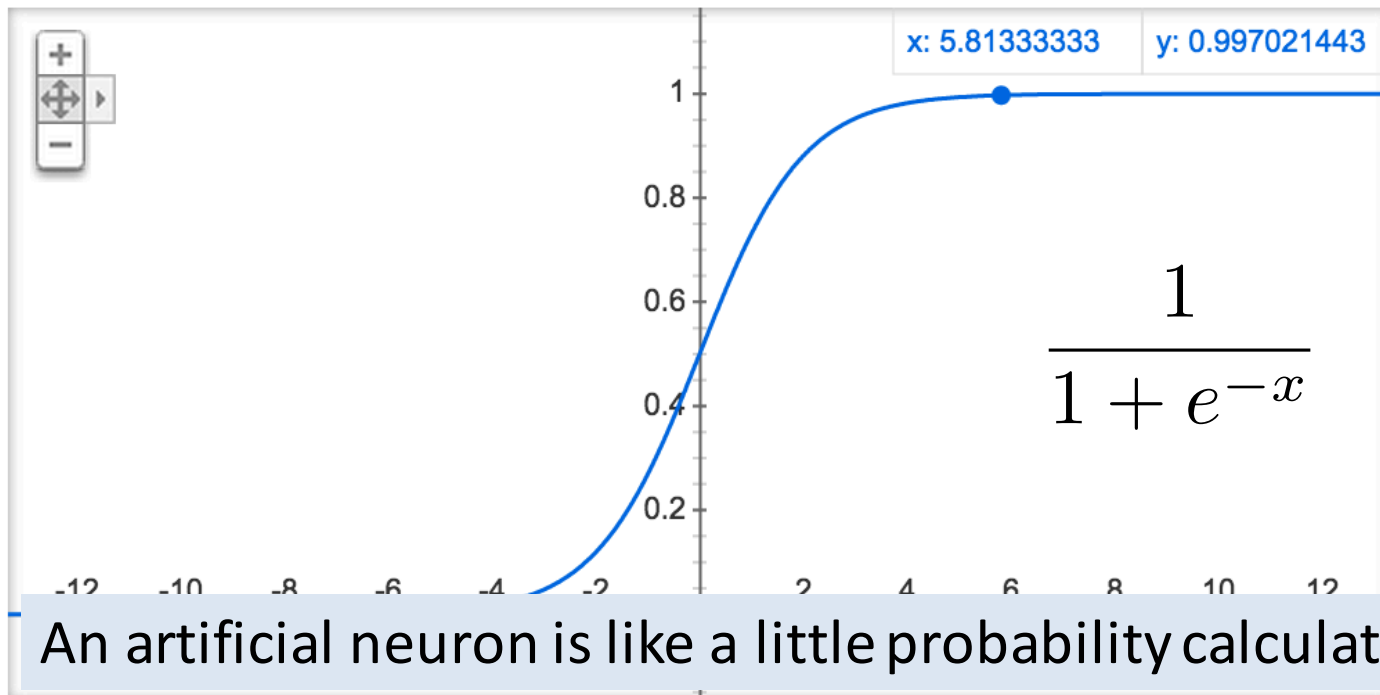
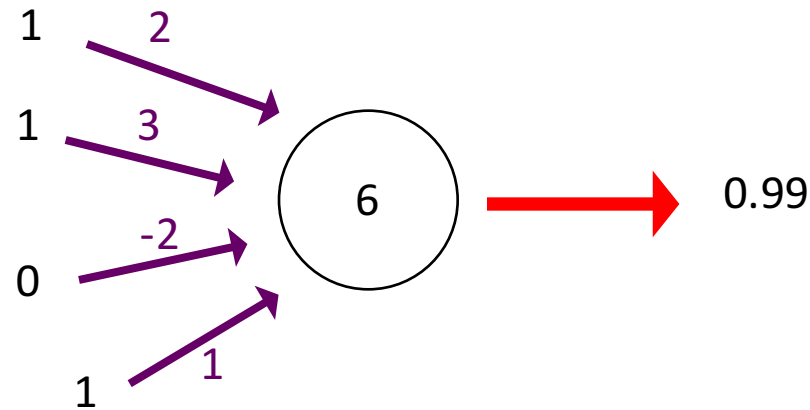
Some Inputs are More Important



Artificial Neuron



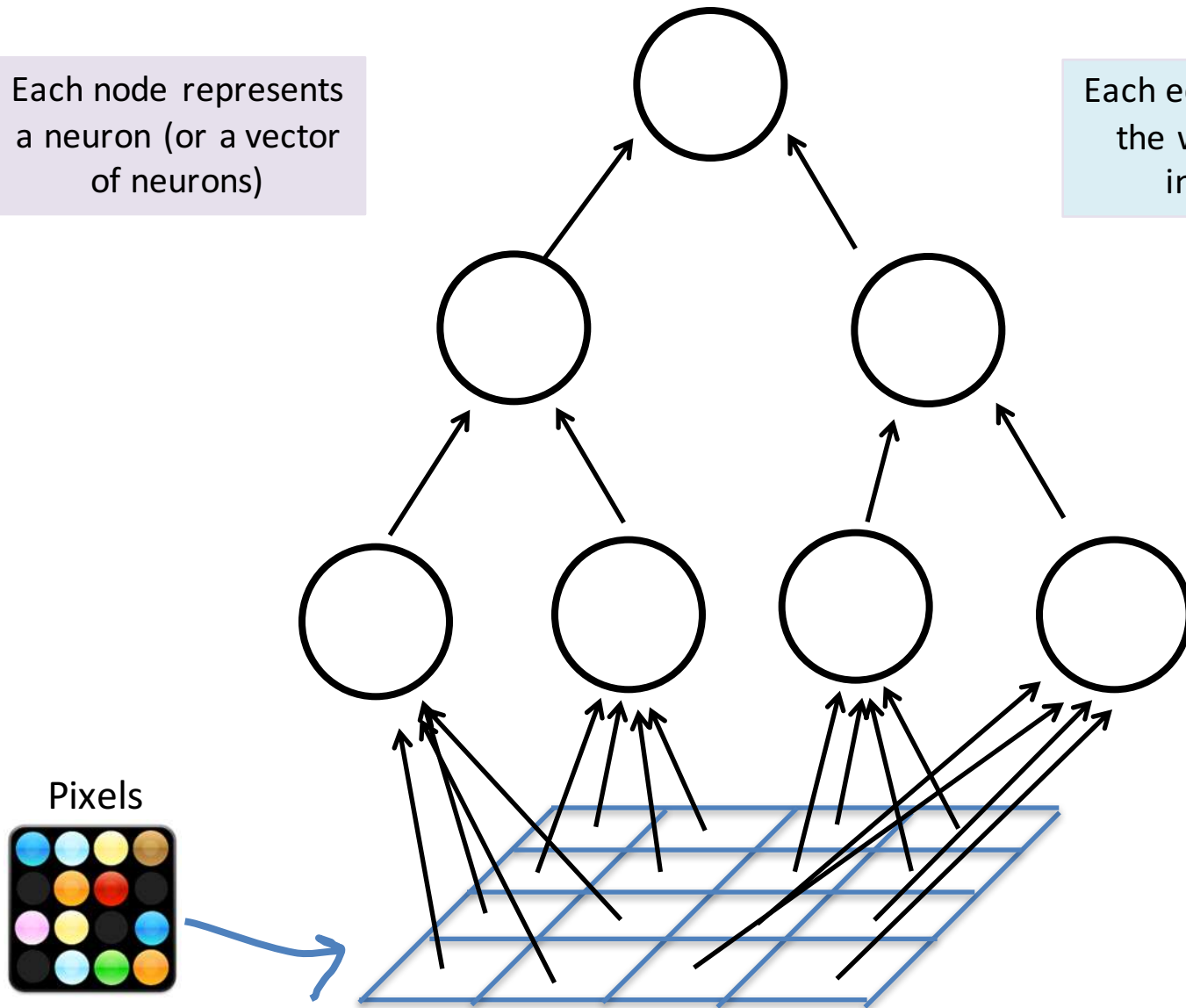
Sigmoid Function



Neural Network

Each node represents
a neuron (or a vector
of neurons)

Each edge represents
the weight of the
interaction

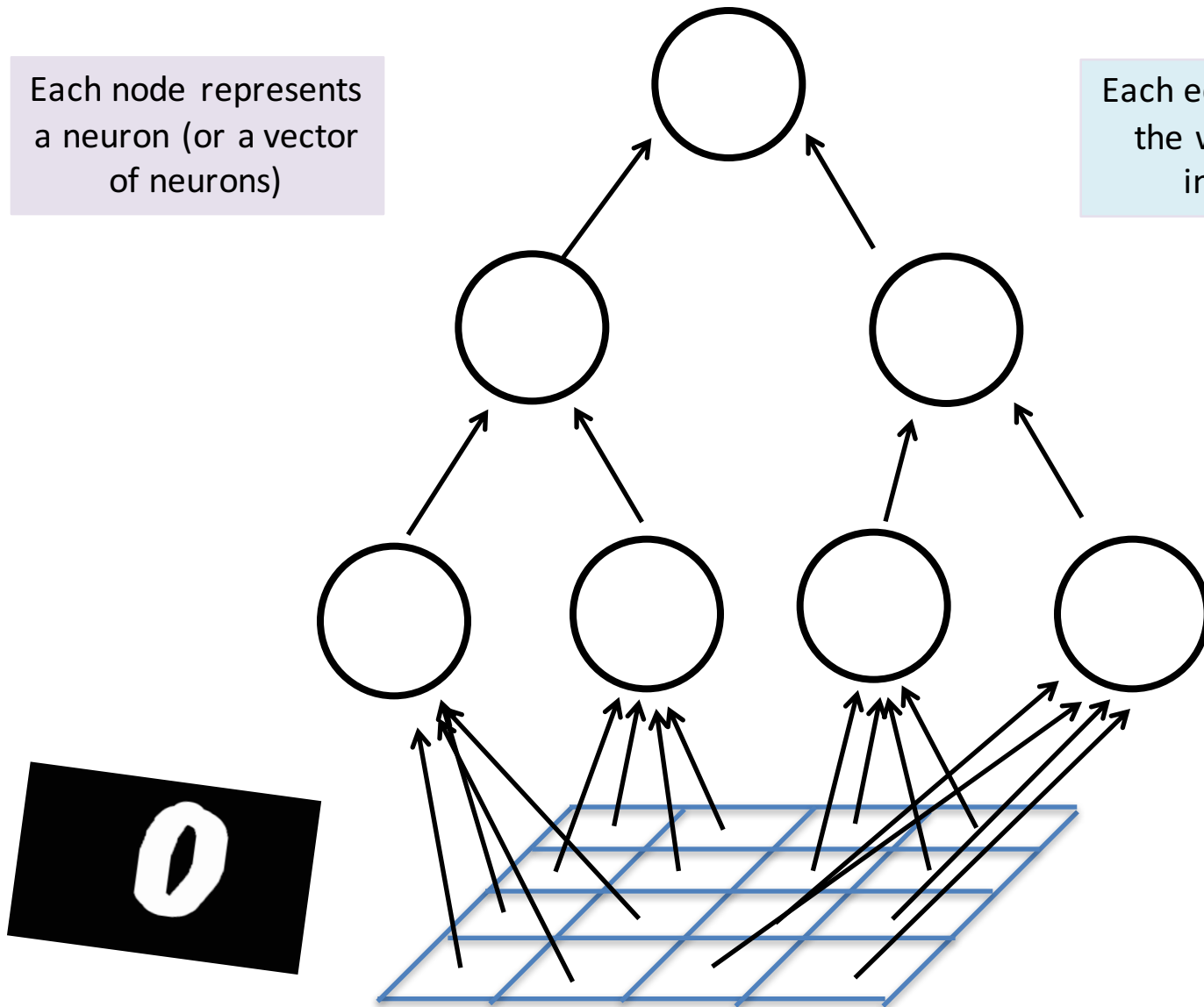


Forward Pass...

Forward Pass

Each node represents
a neuron (or a vector
of neurons)

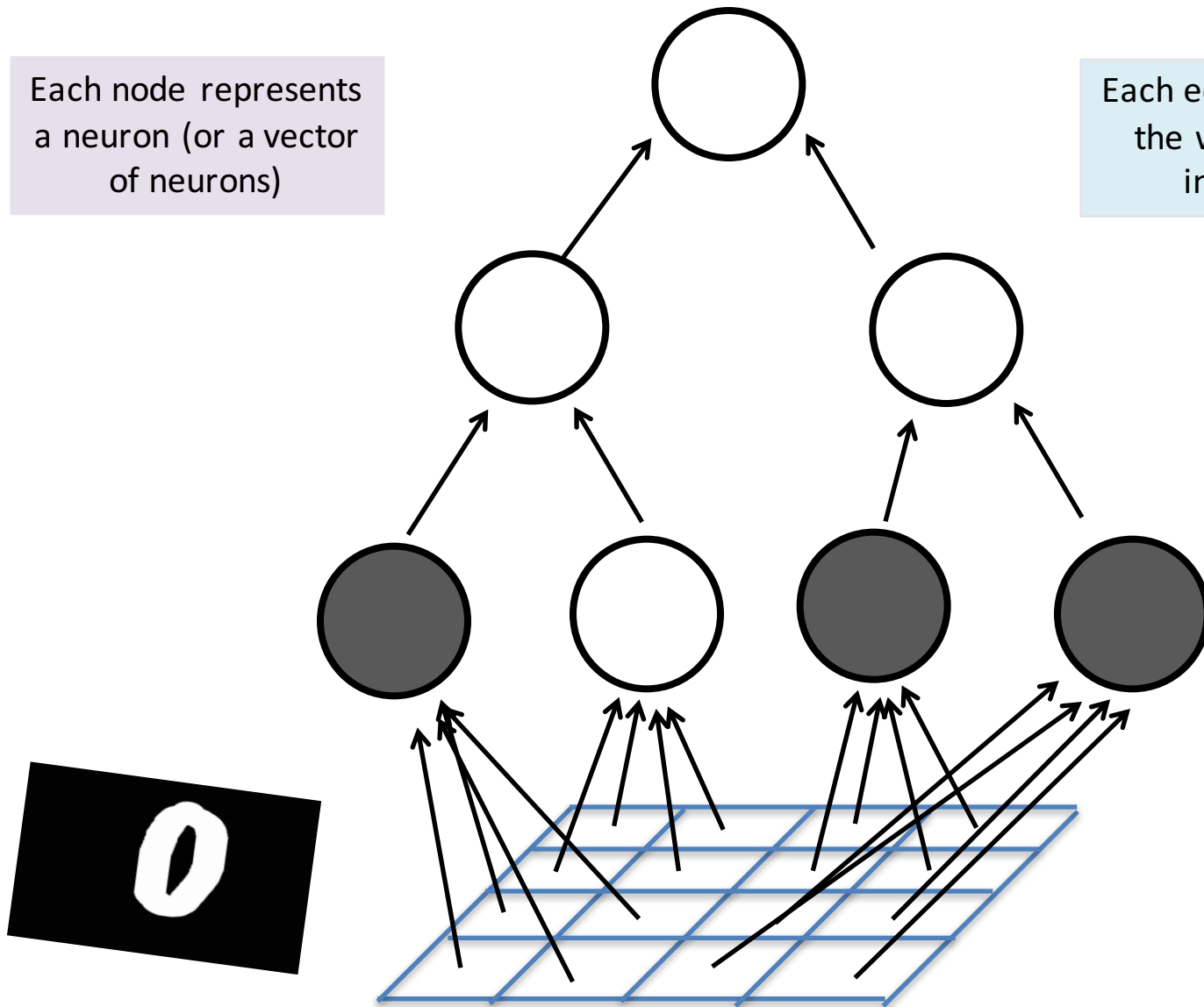
Each edge represents
the weight of the
interaction



Forward Pass

Each node represents
a neuron (or a vector
of neurons)

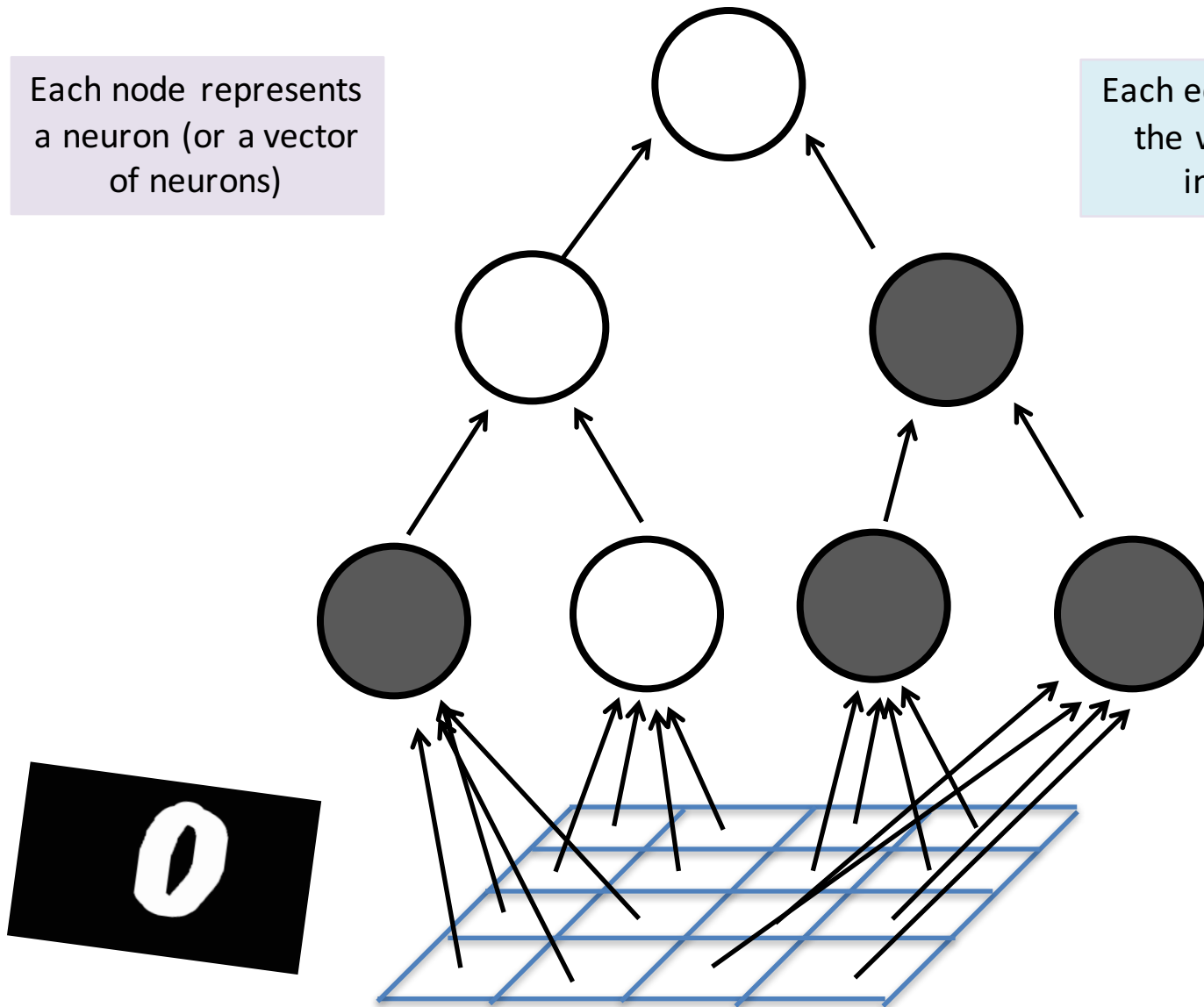
Each edge represents
the weight of the
interaction



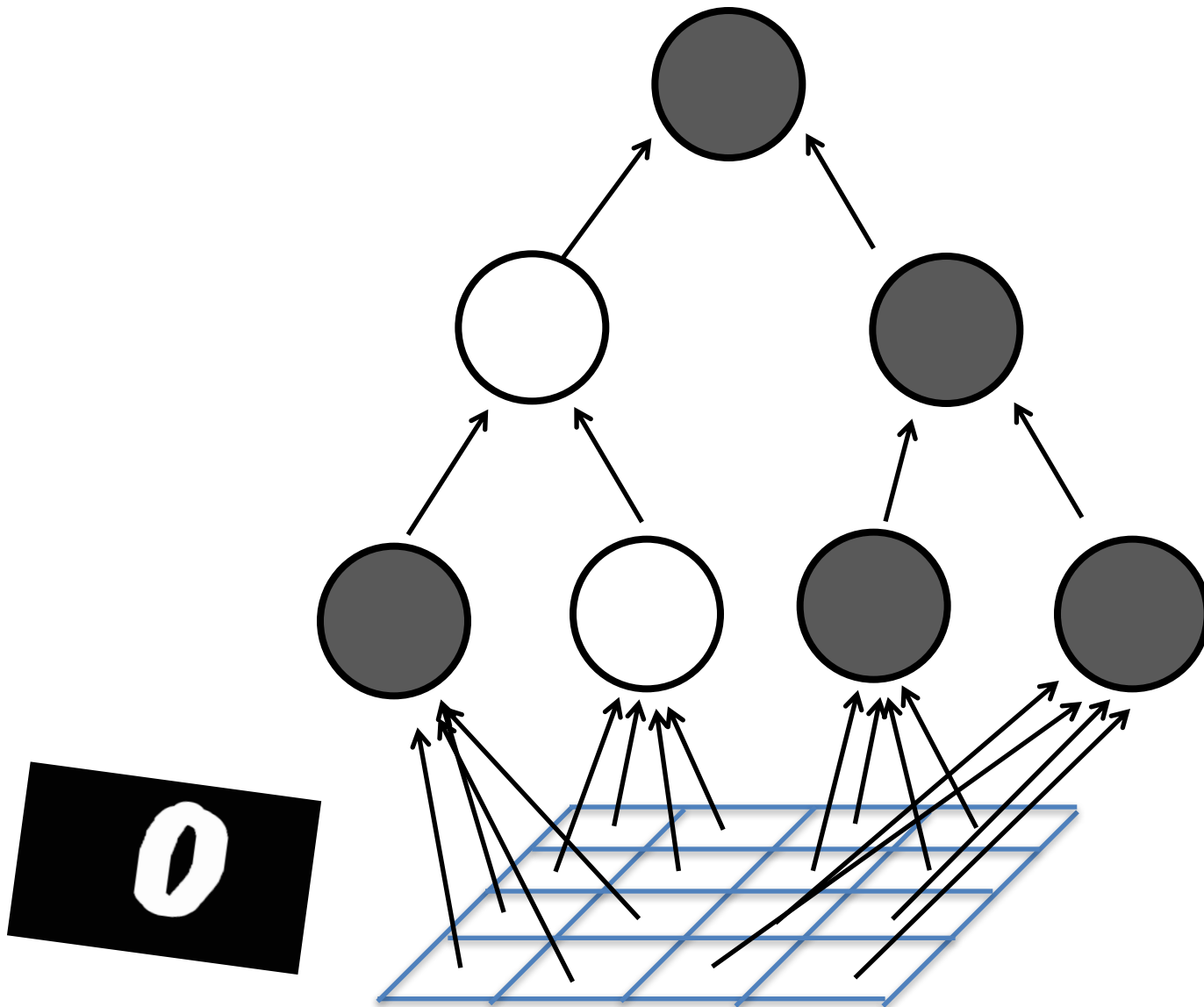
Forward Pass

Each node represents
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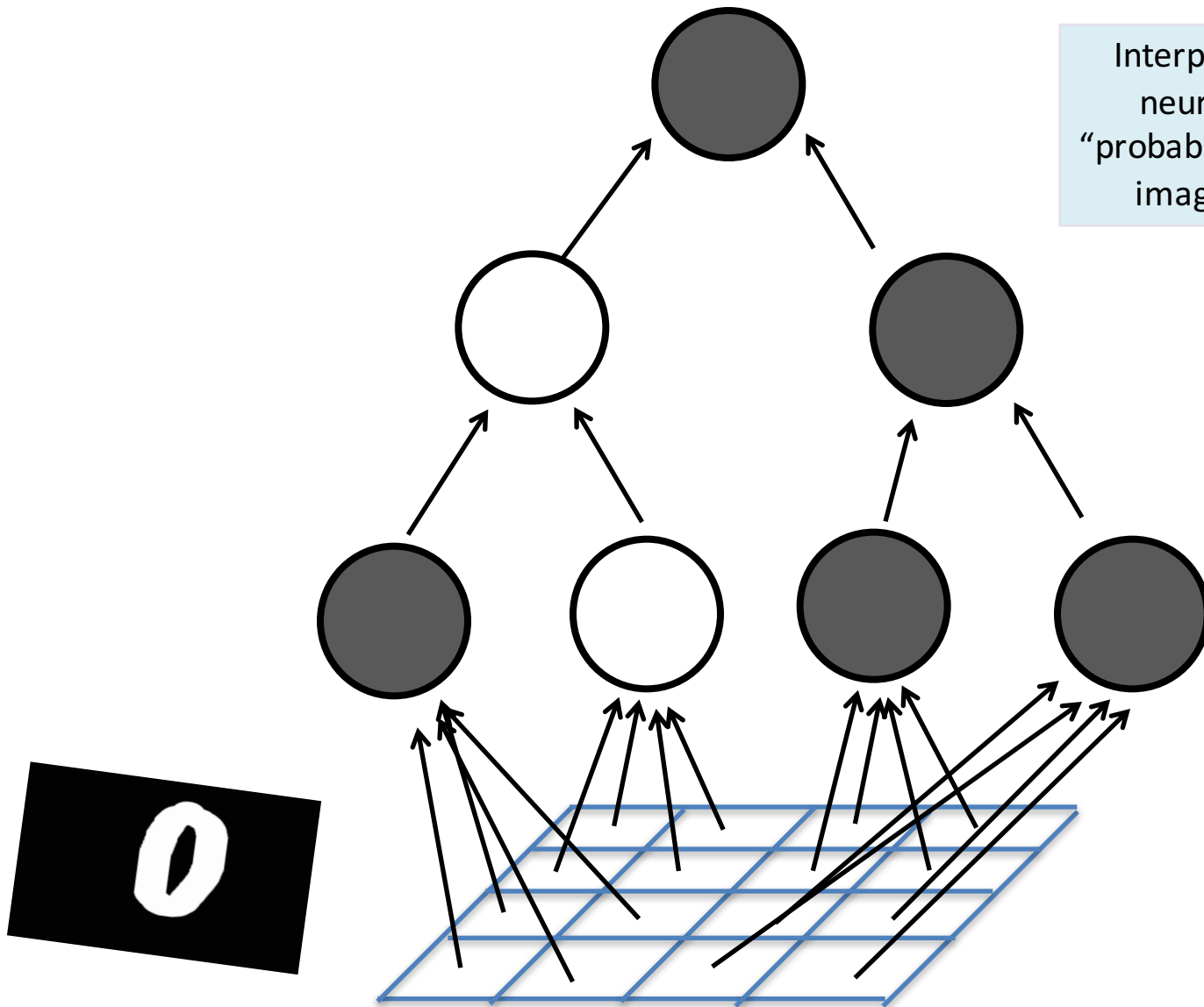


Forward Pass

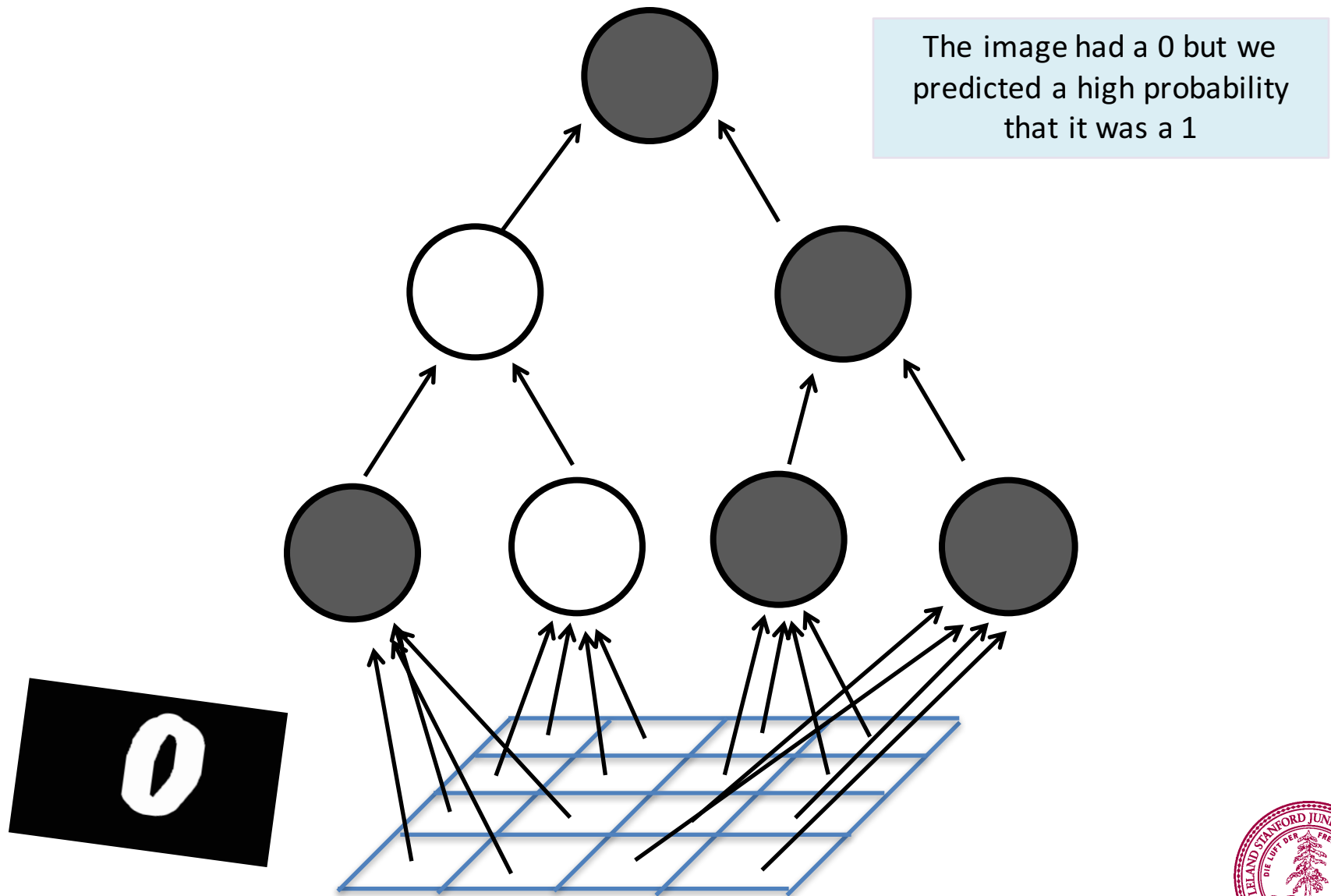


Forward Pass

Interpret the last neuron as the “probability” that the image is of a 1



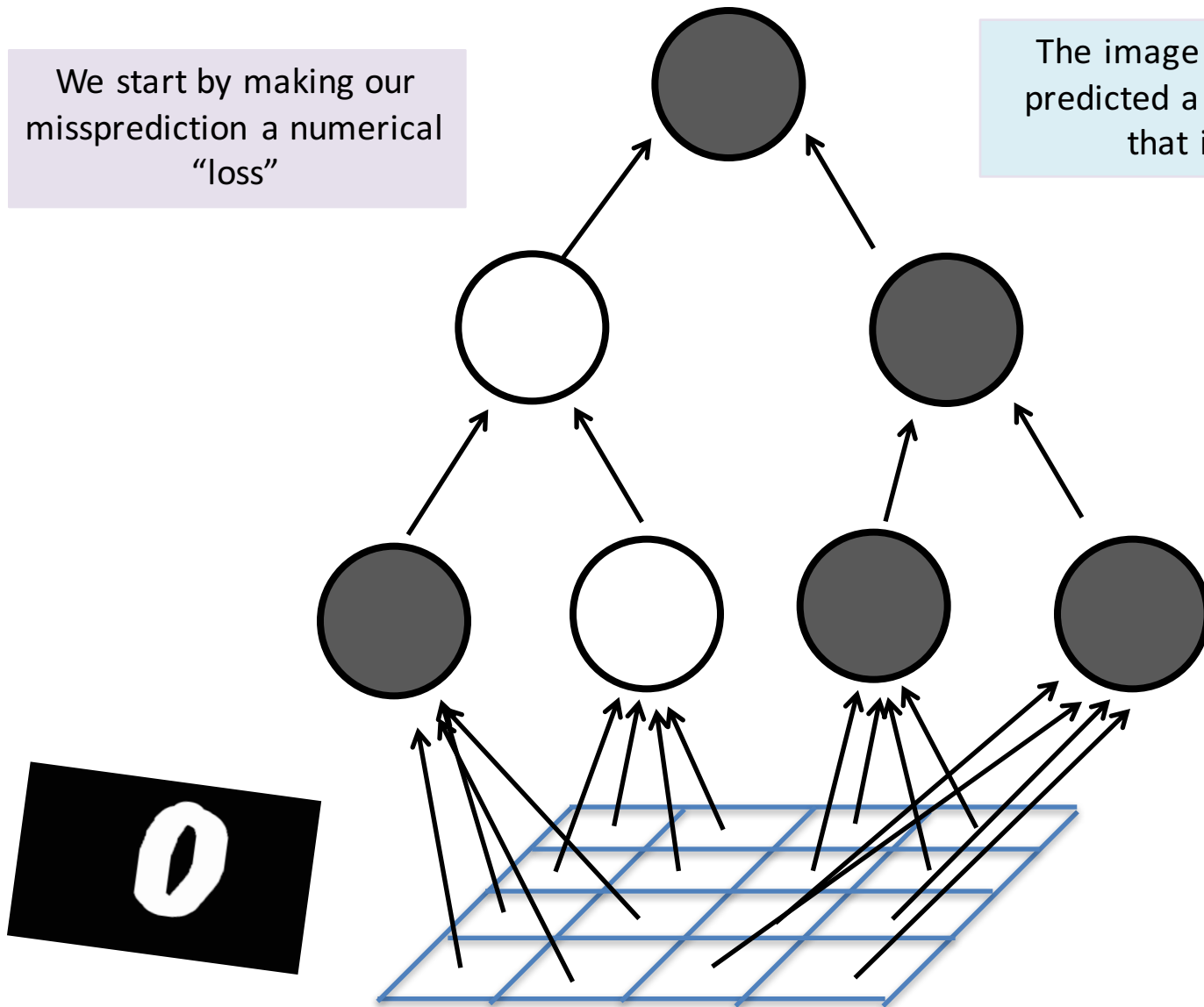
Backward Pass



Backward Pass

We start by making our missprediction a numerical “loss”

The image had a 0 but we predicted a high probability that it was a 1

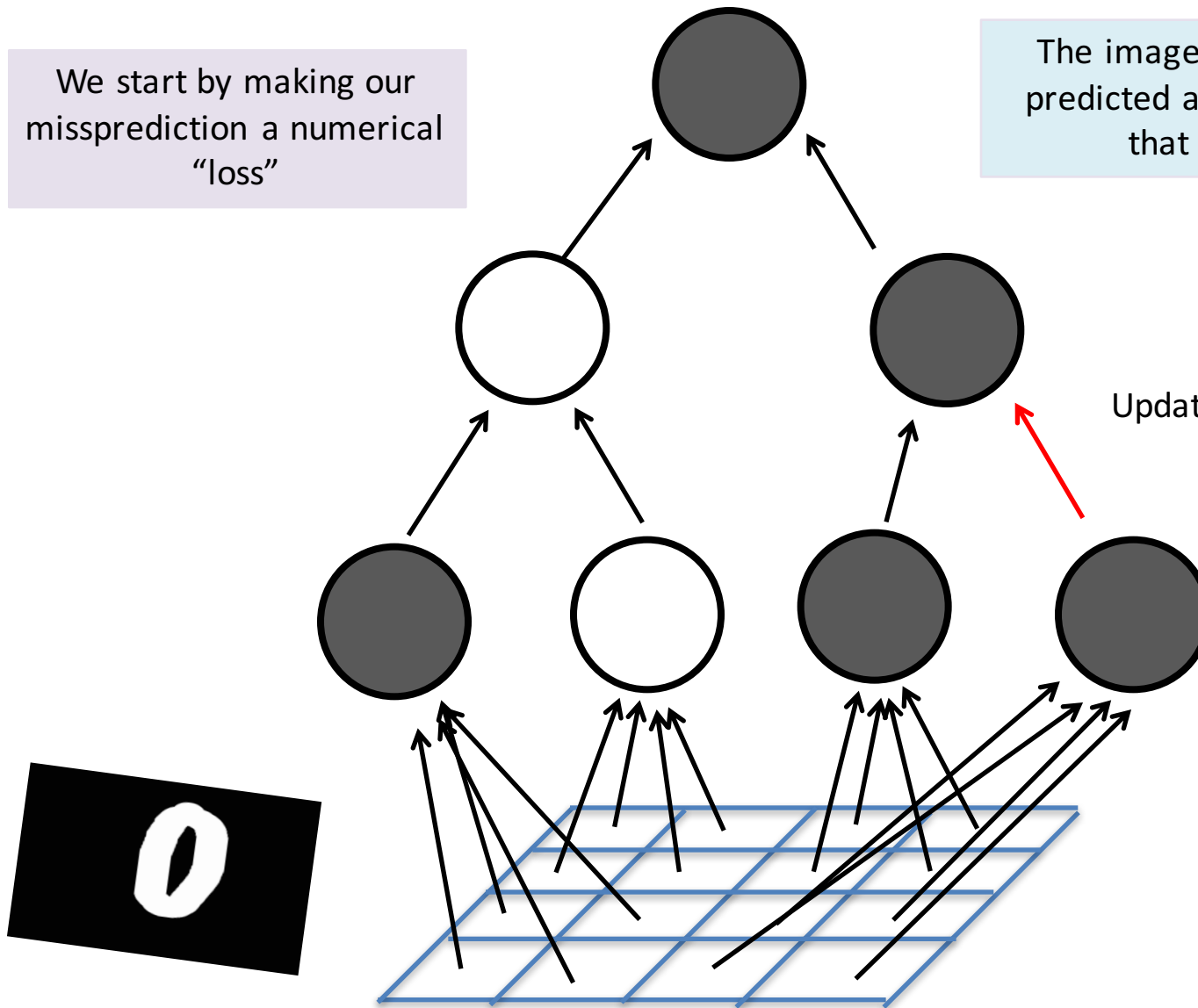


Backward Pass

We start by making our missprediction a numerical "loss"

The image had a 0 but we predicted a high probability that it was a 1

Update each connection



Chose weights that maximize the probability of the right answers

$$P(Y = 1|X = \mathbf{x}) = \hat{y} \qquad \hat{y} = \sigma \left(\sum_{j=0}^{m_h} \mathbf{h}_j \theta_j^{(\hat{y})} \right)$$

For one datum

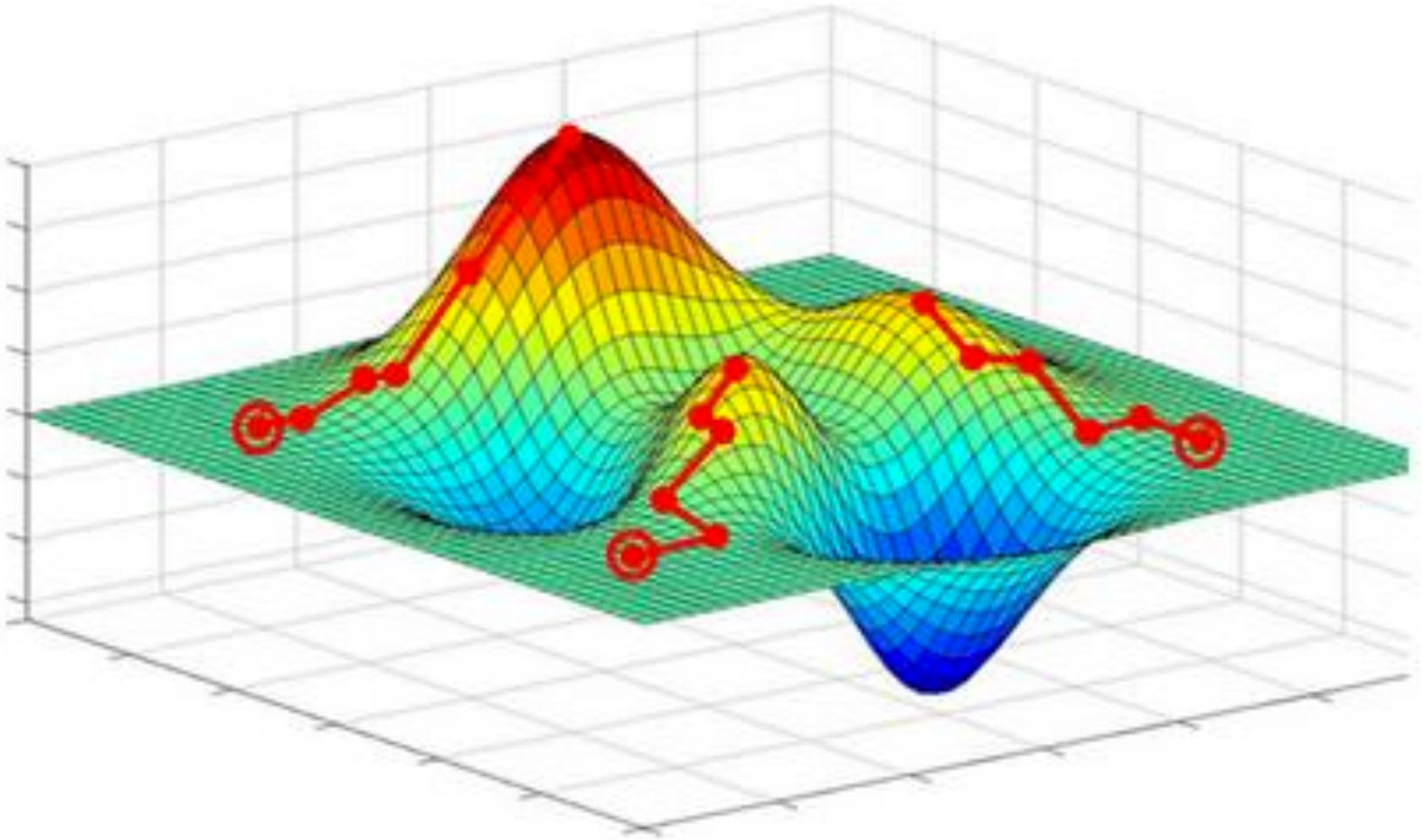
$$P(Y = y|X = \mathbf{X}) = (\hat{y})^y (1 - \hat{y})^{1-y}$$

For IID data

$$\begin{aligned} L(\theta) &= \prod_{i=1}^n P(Y = y^{(i)}|X = \mathbf{x}^{(i)}) \\ &= \prod_{i=1}^n (\hat{y}^{(i)})^{y^{(i)}} \cdot \left[1 - (\hat{y}^{(i)}) \right]^{(1-y^{(i)})} \end{aligned}$$

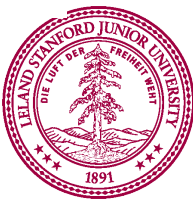


Gradient Ascent



Walk uphill and you will find a local maxima
(if your step size is small enough)

Piech, CS106A, Stanford University



Gradient of output layer params

$$\frac{\partial L}{\partial \theta_i^{(\hat{y})}} = \frac{\partial L}{\partial \hat{y}} \cdot \frac{\partial \hat{y}}{\partial \theta_i^{(\hat{y})}}$$

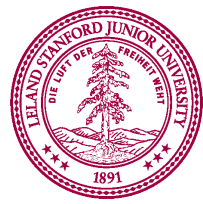
$$\hat{y} = \sigma \left(\sum_{j=0}^{m_h} \mathbf{h}_j \theta_j^{(\hat{y})} \right)$$

$$\frac{\partial \hat{y}}{\partial \theta_i^{(\hat{y})}} = \sigma \left(\sum_{j=0}^{m_h} \mathbf{h}_j \theta_j^{(\hat{y})} \right) \left[1 - \sigma \left(\sum_{j=0}^{m_h} \mathbf{h}_j \theta_j^{(\hat{y})} \right) \right] \cdot \frac{\partial}{\partial \theta_i^{(\hat{y})}} \sum_{j=0}^{m_h} \mathbf{h}_j \theta_j^{(\hat{y})}$$

$$= \hat{y} [1 - \hat{y}] \cdot \frac{\partial}{\partial \theta_i^{(\hat{y})}} \sum_{j=0}^{m_h} \mathbf{h}_j \theta_j^{(\hat{y})}$$

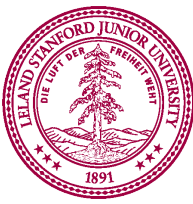
$$= \hat{y} [1 - \hat{y}] \cdot h_i$$

That looks scarier than it is



Chain Rule Down the Network

$$\frac{d \text{ [dodecahedron]}}{d \text{ [sphere]}} = \frac{d \text{ [dodecahedron]}}{d \text{ [cube]}} \times \frac{d \text{ [cube]}}{d \text{ [sphere]}}$$



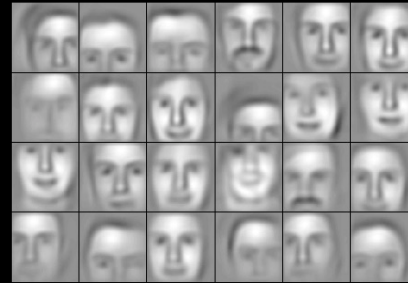
Where you will be by the end of class

When you train,
something really neat happens

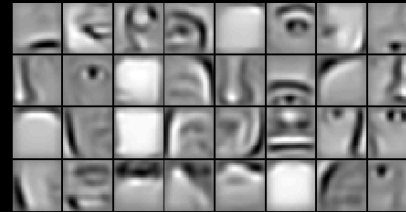
Visualize the Weights



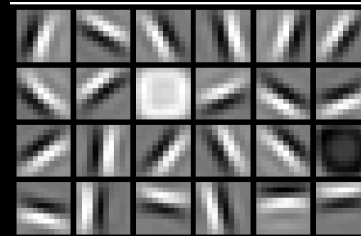
Training set: Aligned images of faces.



object models



object parts
(combination
of edges)

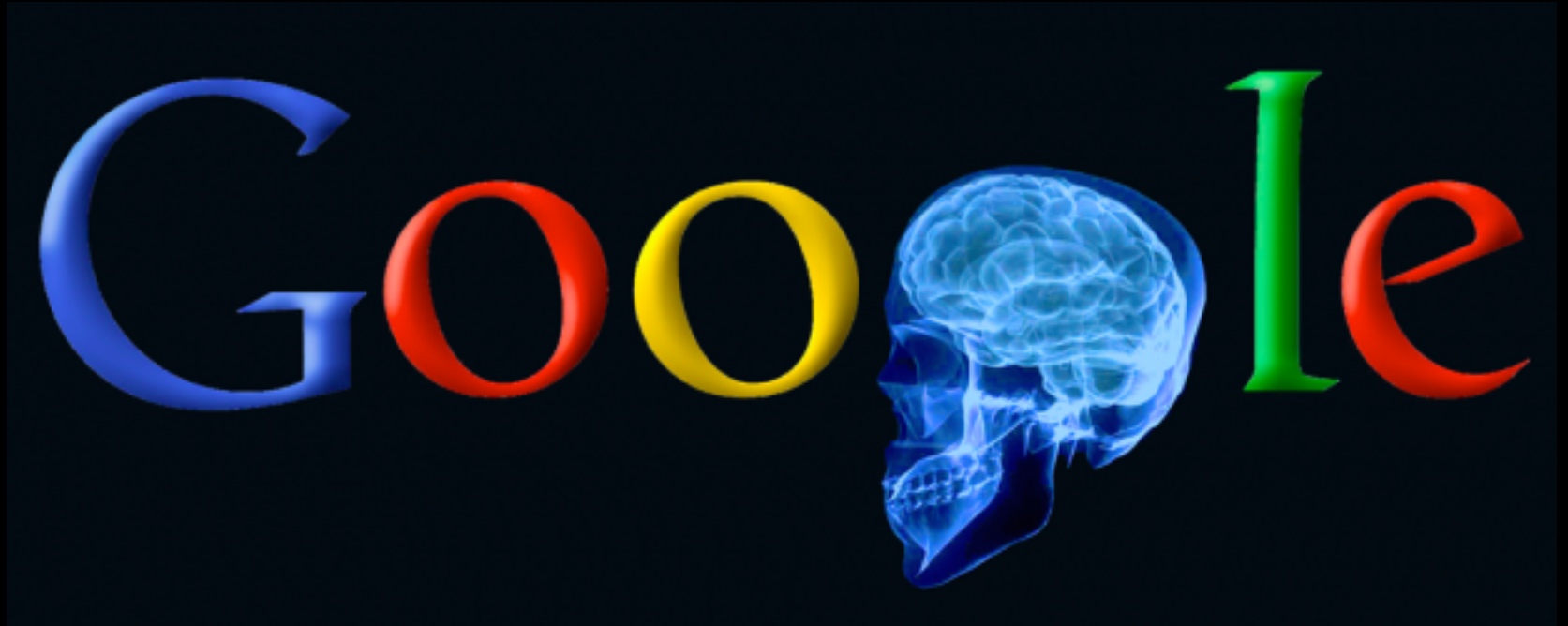


edges

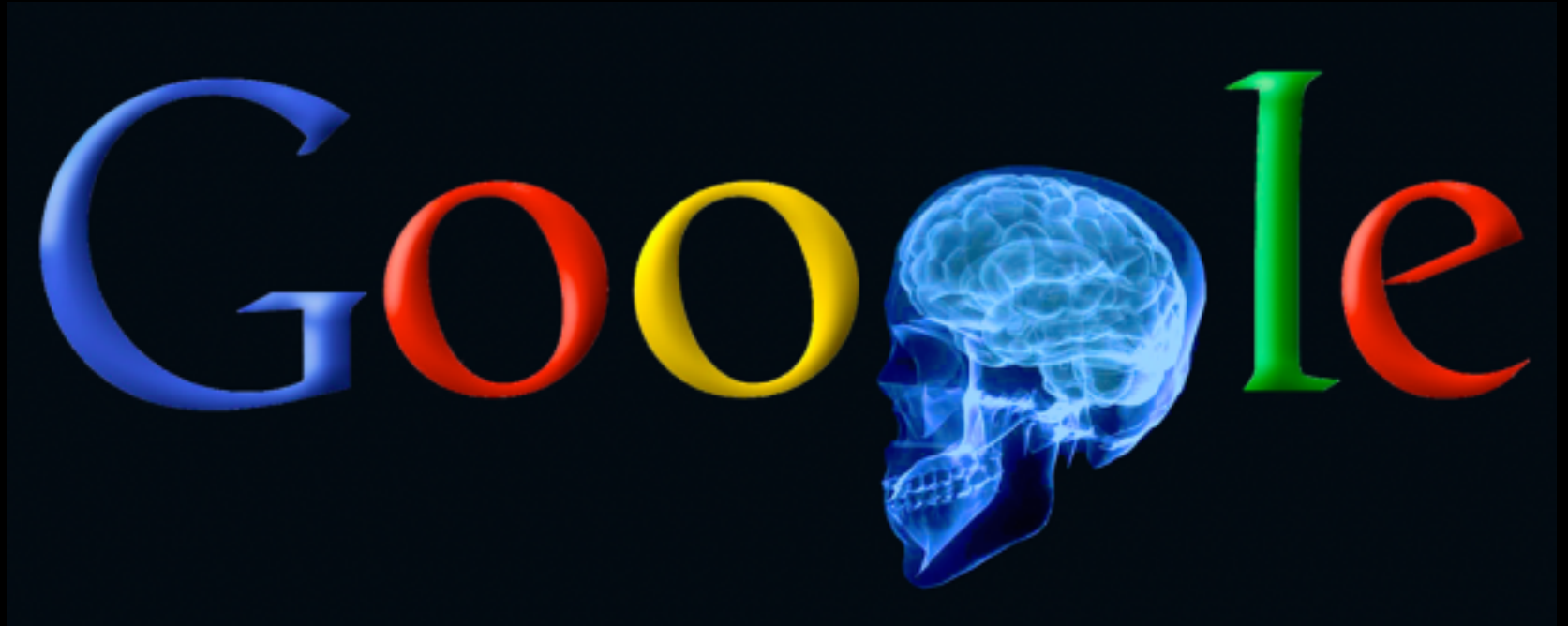


pixels

Google Brain



Google Brain



1 Trillion Artificial Neurons

A Neuron That Fires When It Sees Cats



Top stimuli from the test set



Optimal stimulus
by numerical optimization

We're essentially

a cat detection company



Other Neurons

Neuron 1



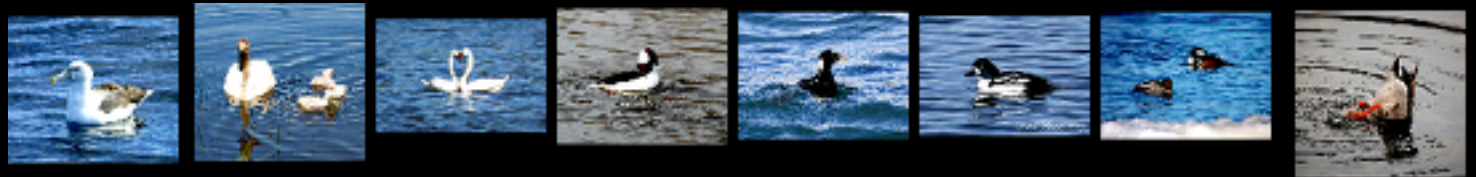
Neuron 2



Neuron 3



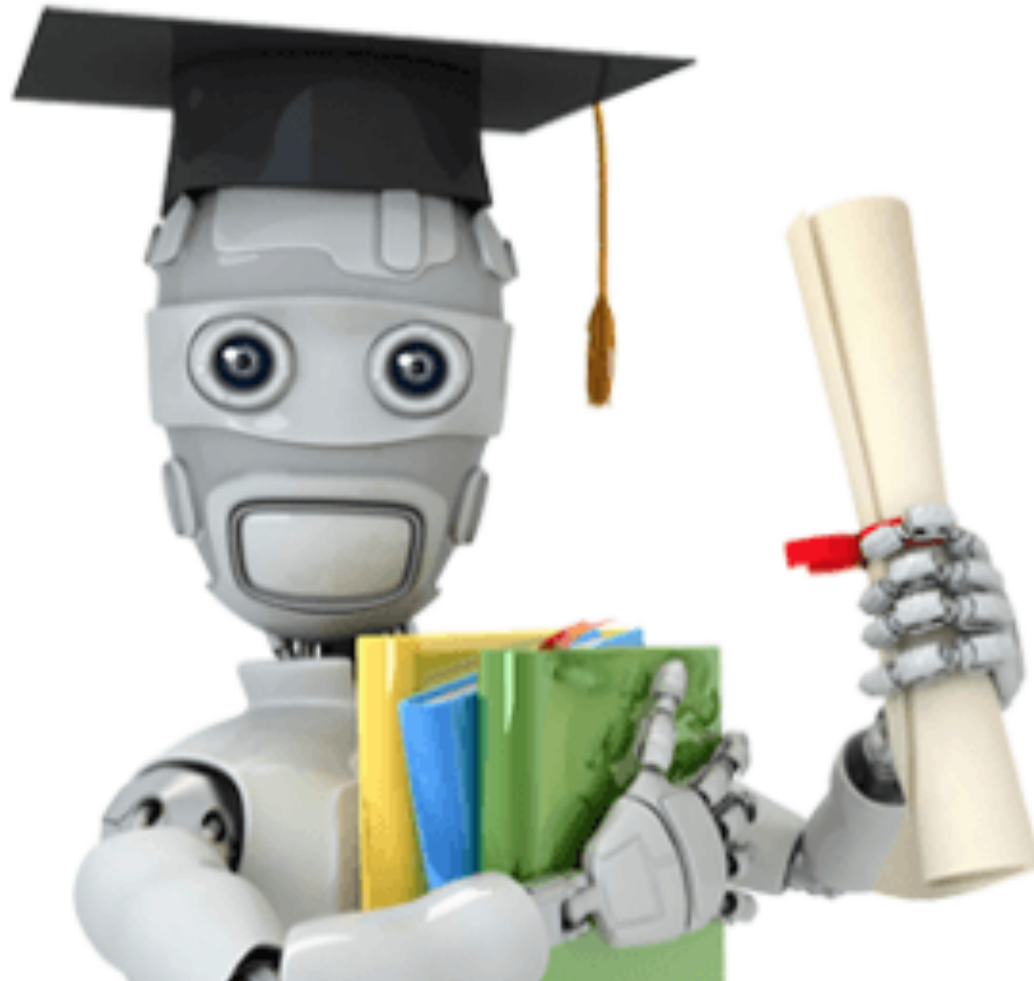
Neuron 4



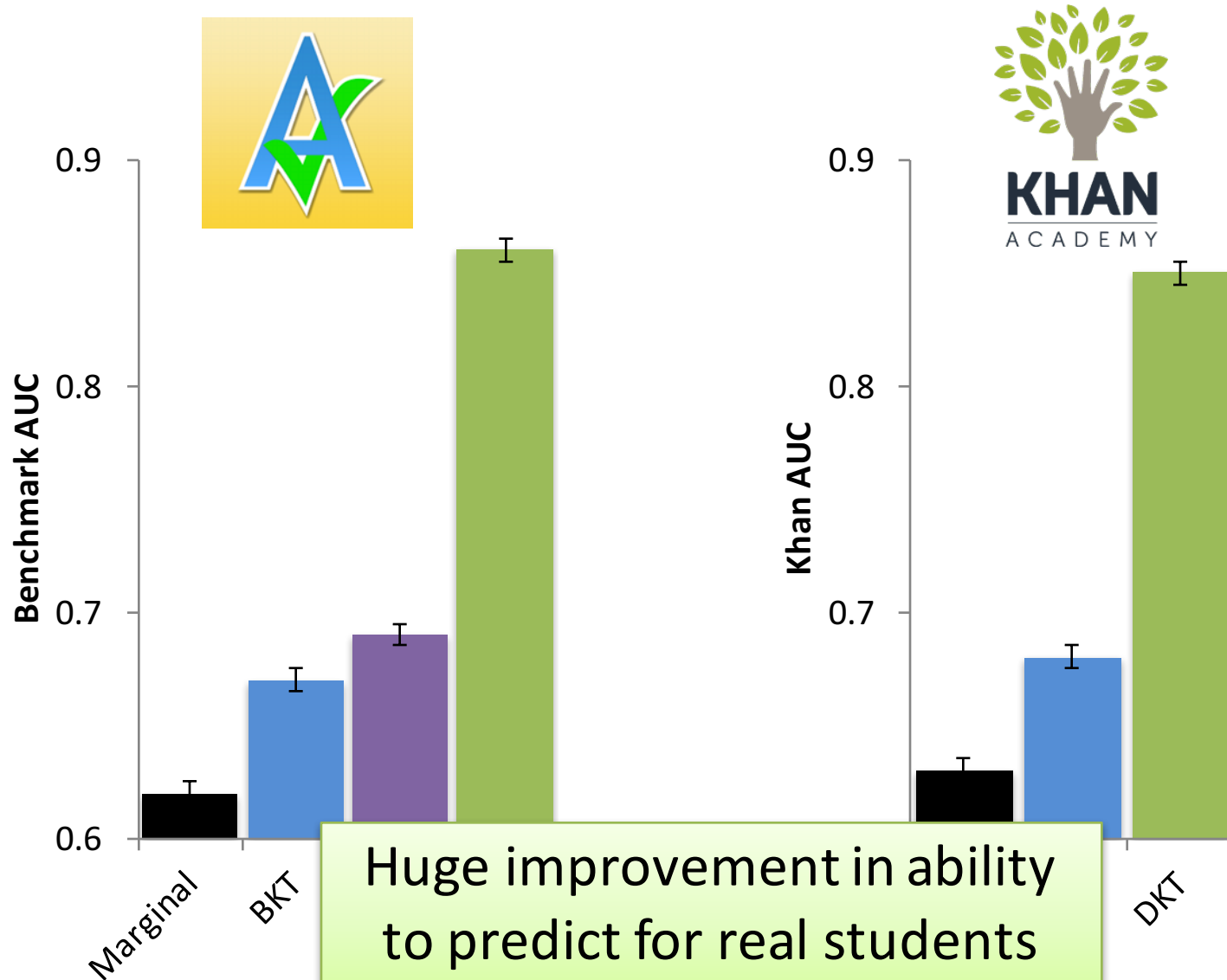
Neuron 5



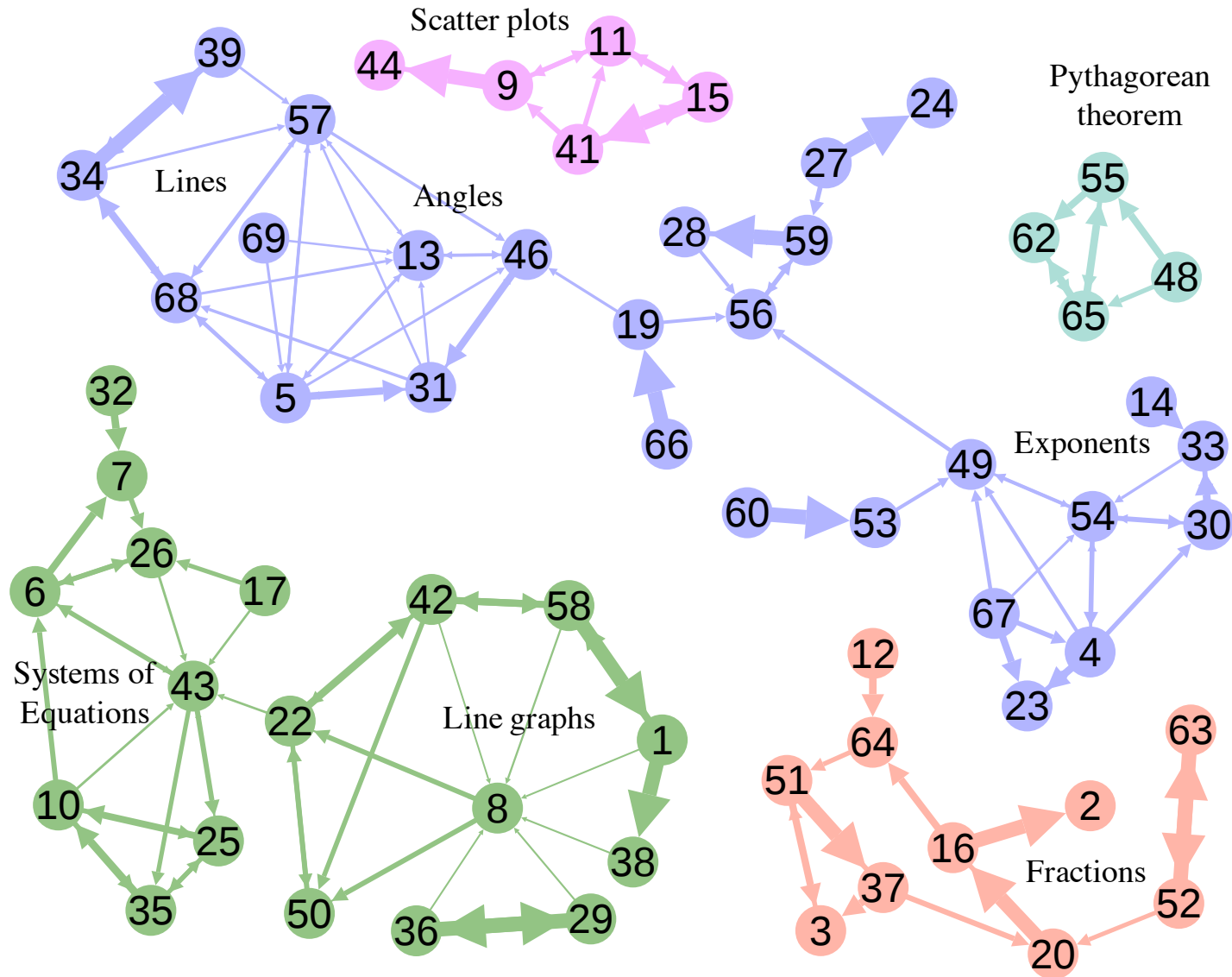
Autonomous Tutor



Prediction Results



Autonomous Tutor



Not once, but twice, AI was revolutionized by people who understood probability theory.

End of Story

Except it isn't the end of the story...

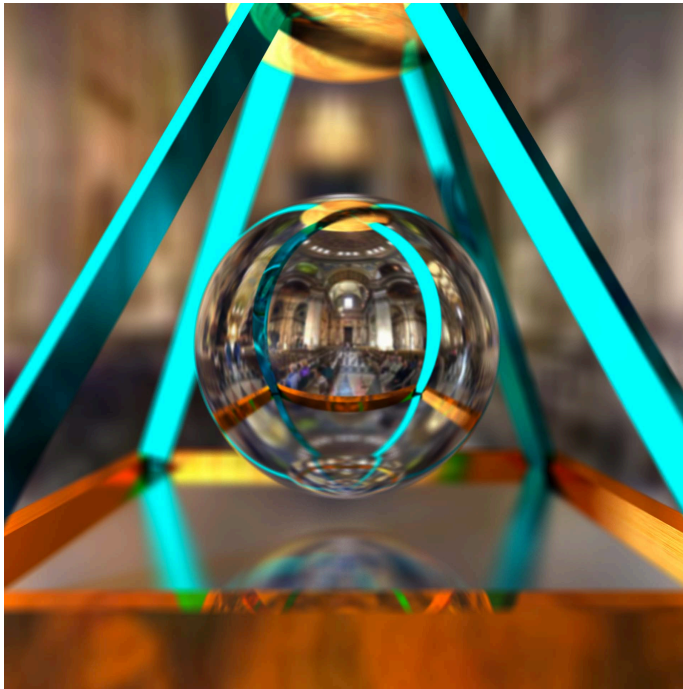
Probability is more than just machine learning

Abundance of Important Problems

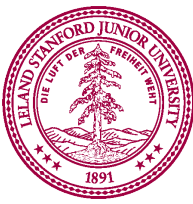
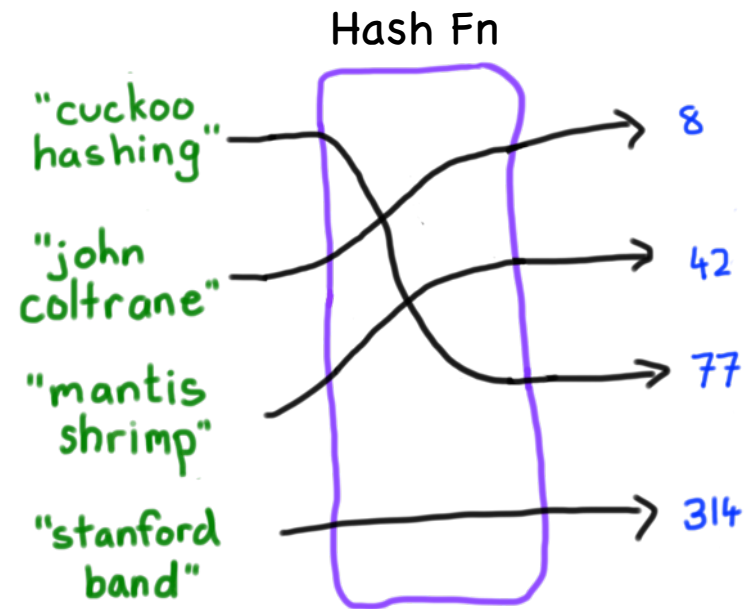


Algorithms and Probability

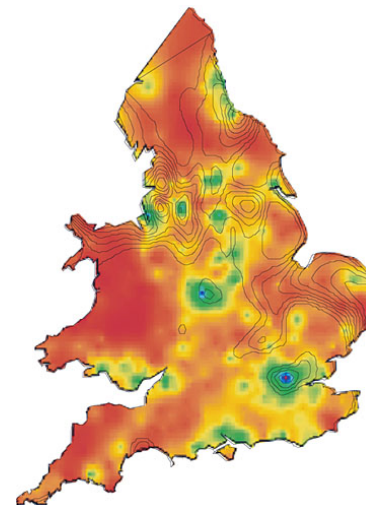
Eg Raytracing



Eg HashMaps



Medicine and Probability



Autocomplete

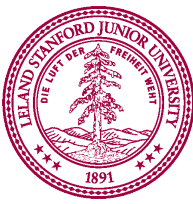


dinosaurs we

[Advanced Search](#)
[Language Tools](#)

- dinosaurs websites for kids
- dinosaurs we're back
- dinosaurs webcomic
- dinosaurs webquest
- dinosaurs were made up by the cia to discourage time travel
- dinosaurs website
- dinosaurs went extinct
- dinosaurs weight
- dinosaurs we are scientists
- dinosaurs weed episode

Google Search I'm Feeling Lucky



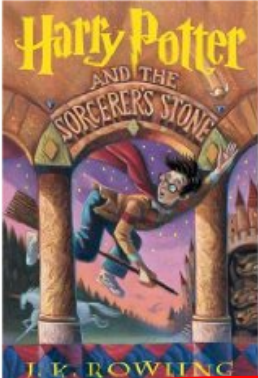
Probability in Practice

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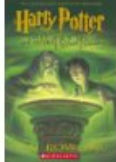
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Page 1 of 20


Harry Potter and the Prisoner of Azkaban (Book 3) by J.K. Rowling
★★★★★ (2,599) \$16.49


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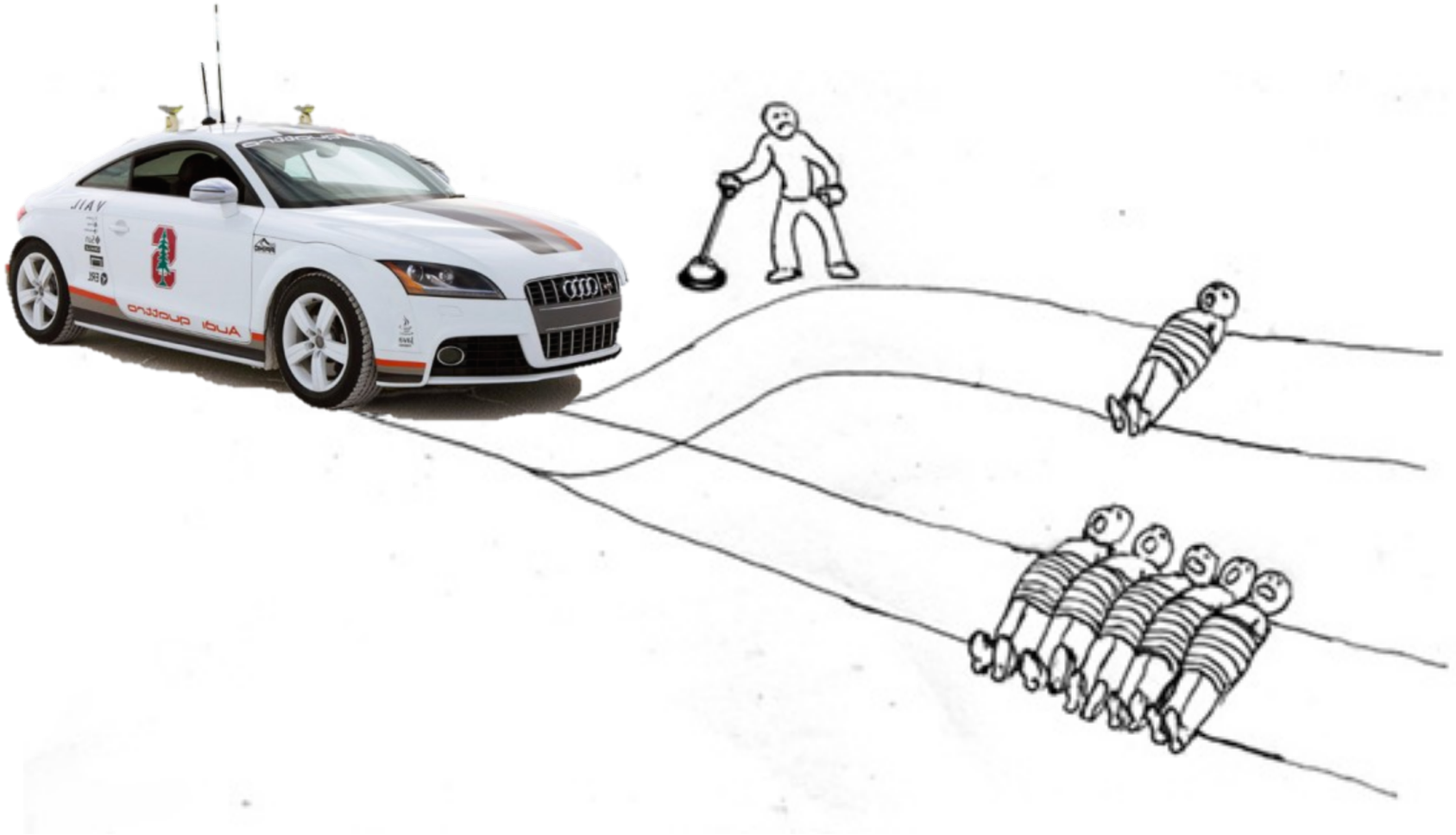

Harry Potter and the Order of the Phoenix (Book 5) by J. K. Rowling
★★★★★ (5,876) \$10.18


Harry Potter and the Half-Blood Prince (Book 6) by J.K. Rowling
★★★★★ (3,597) \$10.18

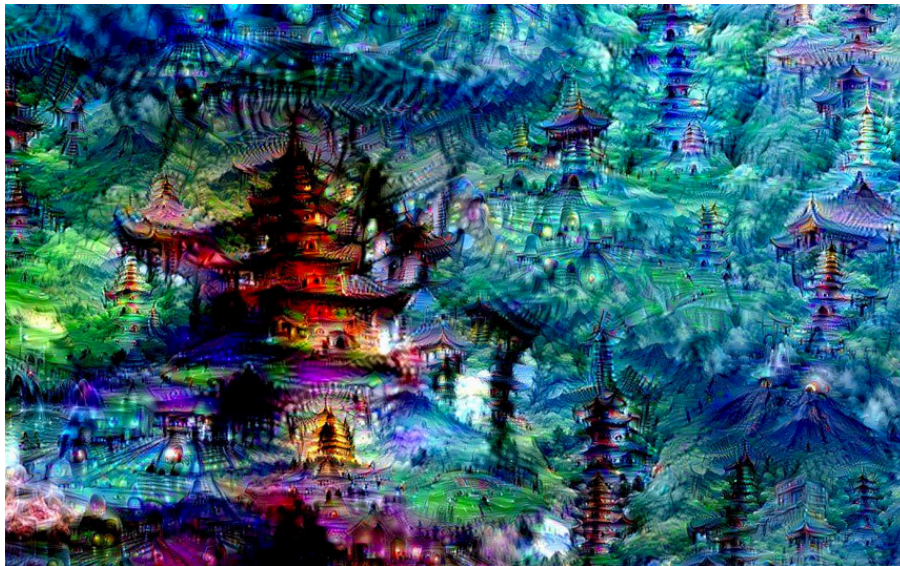
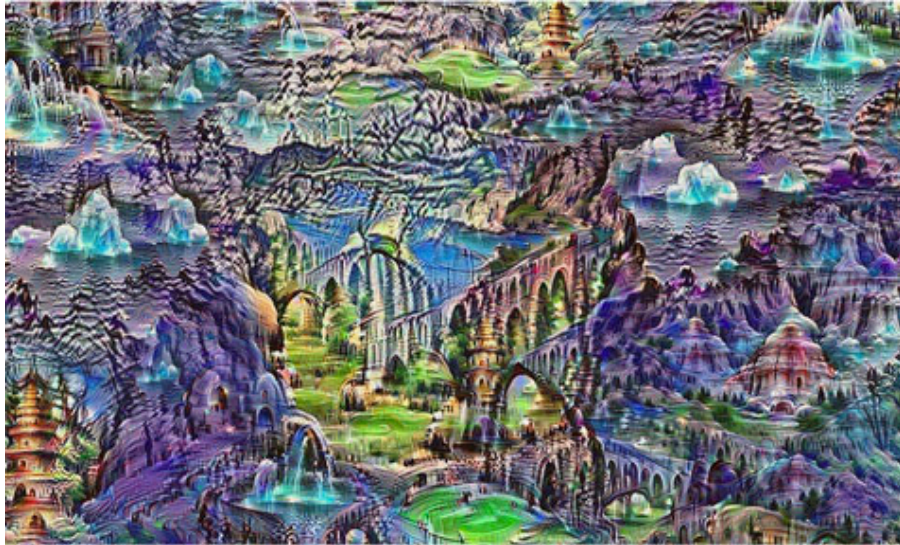

The Tales of Beedle the Bard, Collector's Ed... by J. K. Rowling
★★★★★ (176)



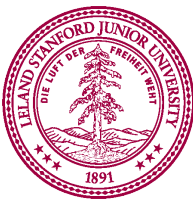
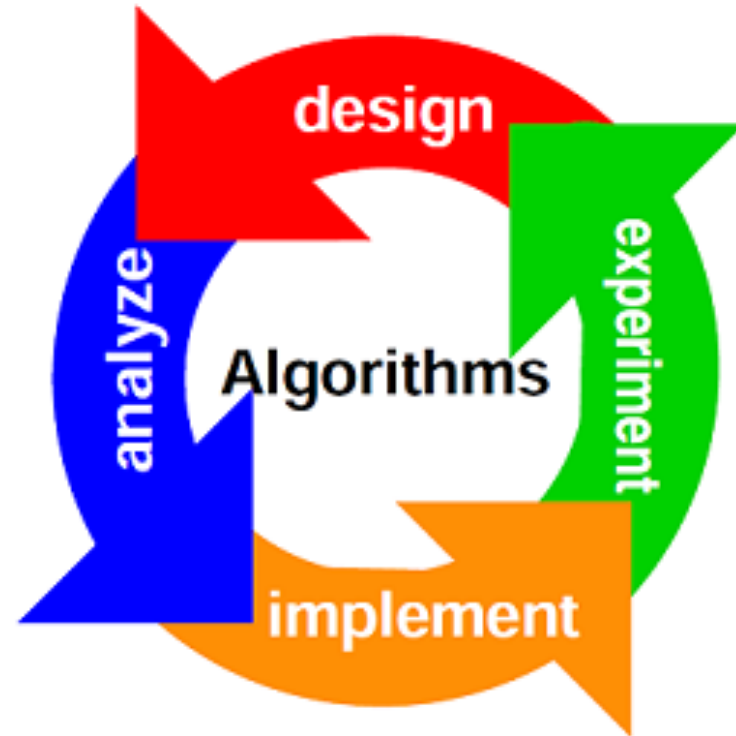
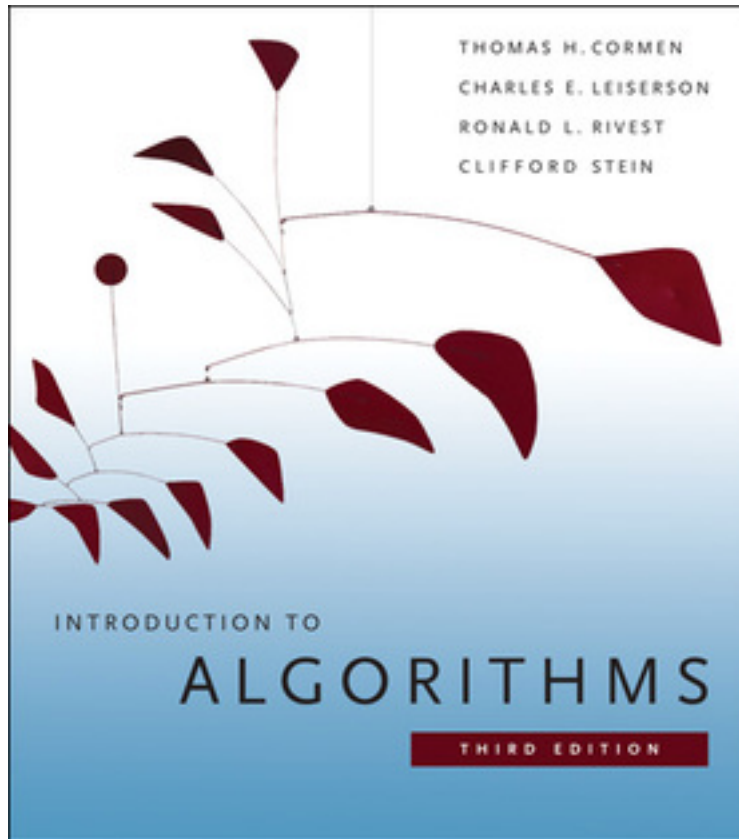
Philosophy and Probability



Art and Probability



Probabilistic Analysis of Algorithms



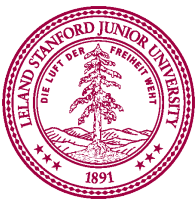
#1 Most Desired Skill in Industry

Microsoft's competitive advantage, [Bill Gates] responded, was its expertise in "Bayesian [probabilistic] networks."

(from Los Angeles Times, Oct. 28, 1996)

"The sexy job in the next 10 years will be statisticians."

-Hal Varian, Chief Economist at Google
(from New York Times, August 6, 2009)

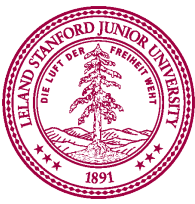


#1 Most Desired Skill in Industry

“I believe over the next decade computing will become even more ubiquitous and intelligence will become ambient. The coevolution of software and new hardware form factors will intermediate and digitize — many of the things we do and experience in business, life and our world. This will be made possible by an ever-growing network of connected devices, incredible computing capacity from the cloud, insights from big data, and intelligence from machine learning. ”

-- Satya Nadella (CEO, Microsoft)

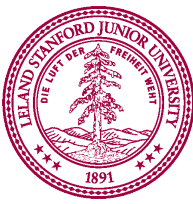
Email to all employees on first day as CEO (Feb. 04, 2014)



#1 Most Desired Skill in Academia

Most CS PhD students list their highest desiderata upon graduation as:

“Better understanding of probability”



Foundation for your future

But its not always intuitive

Zika Test



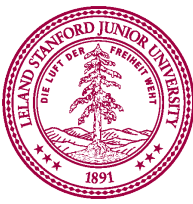
Positive Zika.

What is the probability of zika?

-
- *0.08% of people have zika*
 - *90% positive rate for people with zika*
 - *7% positive rate for people without zika*

The right answer is 1%

* Errata: this previously said 9%, which was a typo ☺



Probability = Important + Needs Study

Delayed gratification

What is CS109?

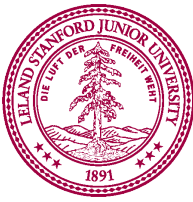
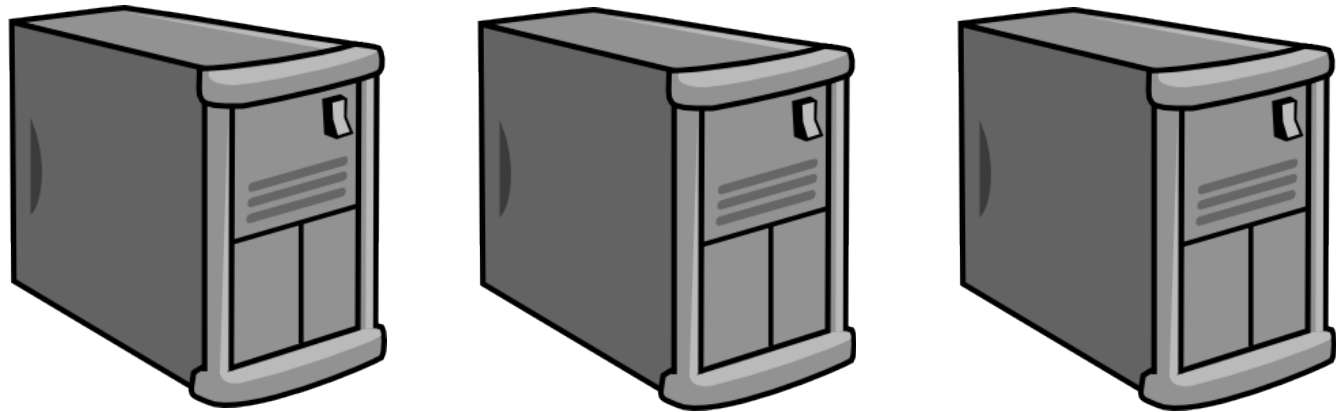
Traditional View of Probability



CS View of Probability

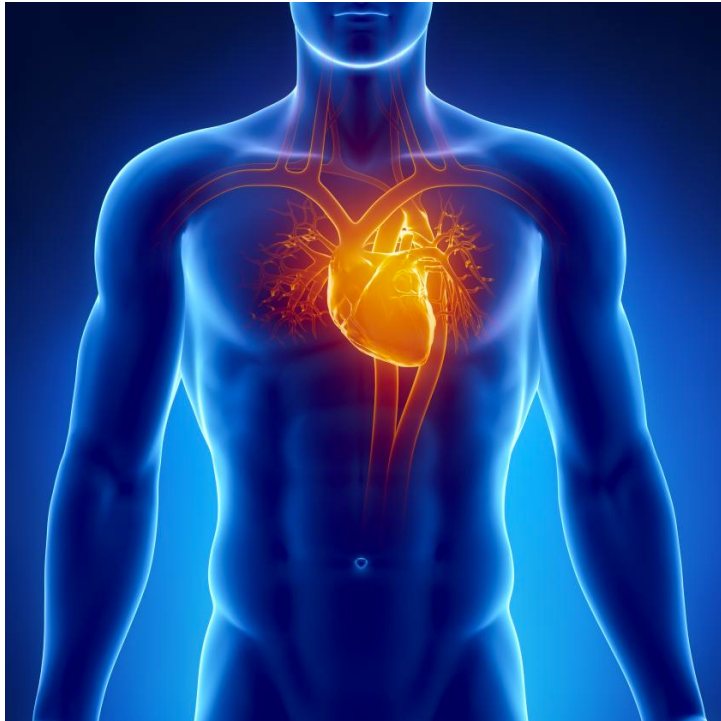
<http://www.site.com>

Give you the tools necessary to build and understand probabilistic CS algorithms.



CS View of Probability

Heart



Ancestry



23andMe

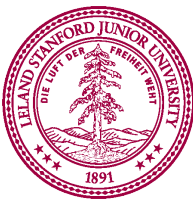
Netflix



CS View of Probability



Piech, CS106A, Stanford University



CS View of Probability

Teach you how to write programs
that most people are not able to write.

Spring vs Winter?



Lets dive in...

Counting



Our Route

