



# CS109 Summary

Chris Piech

CS109, Stanford University

Sorting  
Choosing  
*Buckets*  
**Counting**

Mutual  
Exclusion  
Axioms  
Independence  
**Probability  
Principles**

Chain Rule  
*Bayes Theorem*  
**Conditional  
Probability**

PMF     $E[X]$      $\text{Var}[X]$   
**Random Variables**

Bernoulli    Binomial    Geometric  
**Discrete Parametric  
Distributions**

PDF    CDF    Normal    Normal  
Approx  
**Continuous Parametric  
Distributions**

Joint table    Convolution  
Great  
Expectation  
Beta  
Conditional  
Covariance  
**Multivariate  
Distributions**

Unbiased  
Estimate  
Bootstrapping  
**Sampling**

CLT    Bounds  
**Central Theorems**

MLE    MAP  
*Gradient  
Ascent*  
**Parameter  
Estimation**

Naïve Bayes  
Logistic  
Regression  
Deep  
Learning  
**Machine Learning**

IMAGINE THAT YOU'RE DRAWING  
AT RANDOM FROM AN URN  
CONTAINING FIFTEEN BALLS—  
SIX RED AND NINE BLACK.

OK. I REACH IN AND...  
...MY GRANDFATHER'S  
ASHES?!? OH GOD!

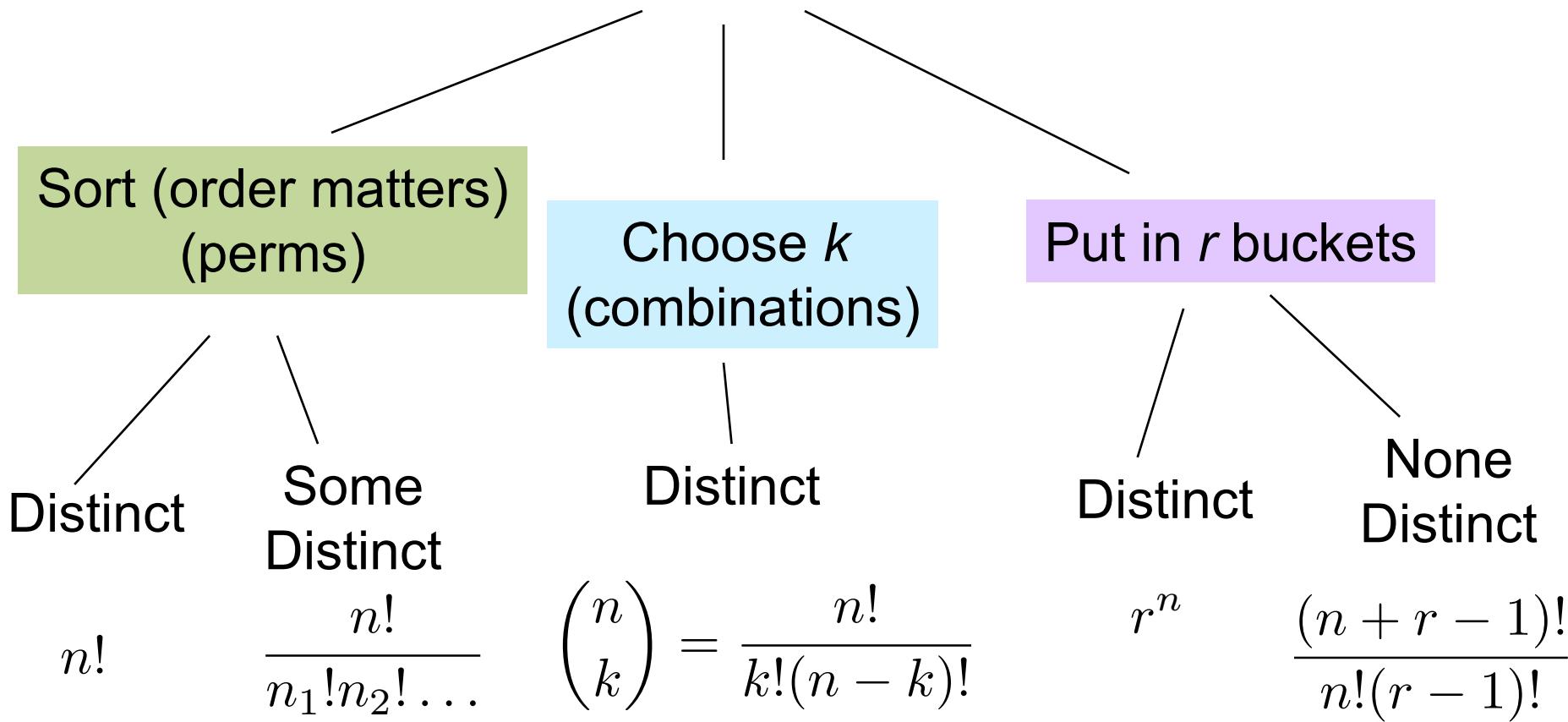
I...WHAT?

WHY WOULD YOU  
DO THIS TO ME?!?



# Counting Rules

Counting operations on  $n$  objects



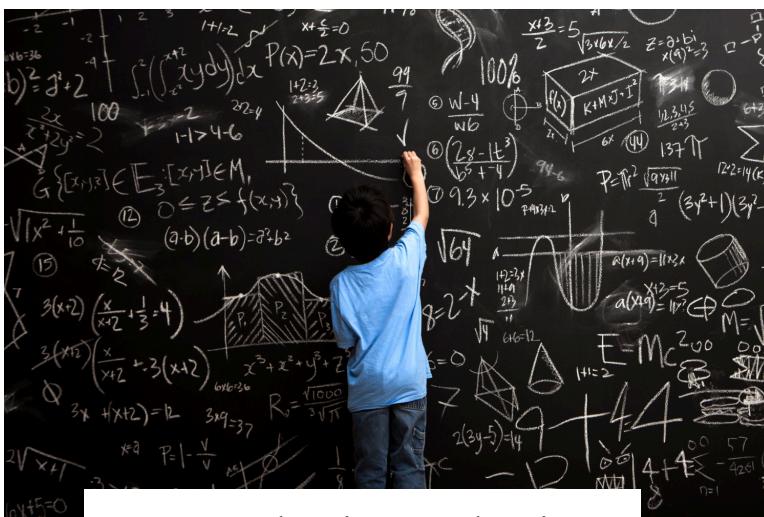
# Sources of Probability



1. Experimentation



2. Dataset



3. Analytic Solution



4. Expert Opinion

Hiecn, CS106A, Stanford University



# What is a Probability?

$$P(E) = \lim_{n \rightarrow \infty} \frac{n(E)}{n}$$



Everything in the world is either



a potato

or not a potato.

$$P(X) + P(X^C) = 1$$

# Sending Bit Strings

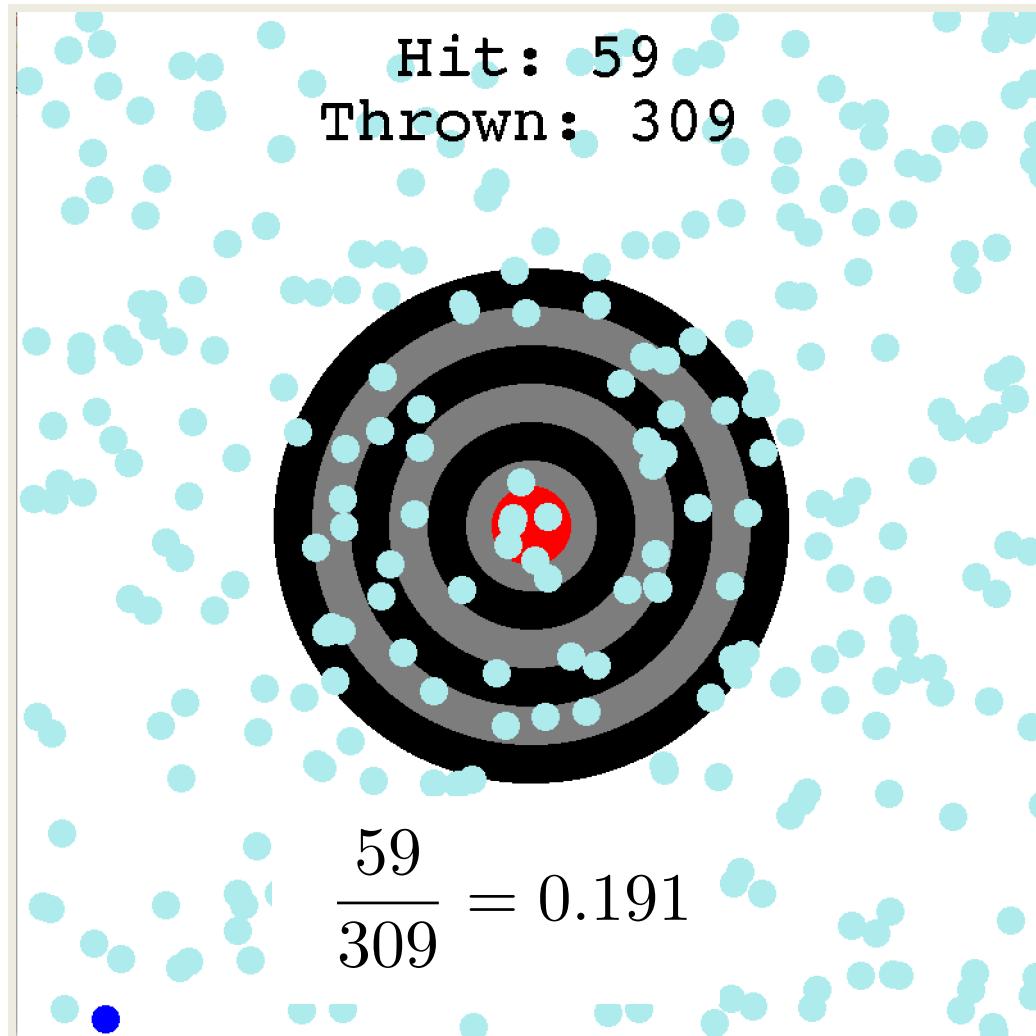
- Bit string with  $m$  0's and  $n$  1's sent on network
  - All distinct arrangements of bits equally likely
  - E = first bit received is a 1
  - F =  $k$  of first  $r$  bits received are 1's

$$P(E|F)?$$



\*Think of the bits as distinct so that all outcomes are equally likely

# Target Revisited



Screen size = 800x800  
Radius of target = 200

The dart is equally likely to land anywhere on the screen.

What is the probability of hitting the target?

$$|S| = 800^2$$

$$|E| = \pi 200^2$$

$$p(E) = \frac{\pi \cdot 200^2}{800^2} \approx 0.1963$$





WHEN YOU MEET YOUR BEST FRIEND

Somewhere you didn't expect to.



Trailing the dovetail shuffle to it's lair – Persi Diaconosis

# Netflix and Learn

What is the probability  
that a user will watch  
Life is Beautiful?

$$P(E)$$



$$P(E) = \lim_{n \rightarrow \infty} \frac{n(E)}{n} \approx \frac{\text{\#people who watched movie}}{\text{\#people on Netflix}}$$

$$P(E) = 10,234,231 / 50,923,123 = 0.20$$

# Netflix and Learn

What is the probability  
that a user will watch  
Life is Beautiful, given  
they watched Amelie?

$$P(E|F)$$



$$P(E|F) = \frac{P(EF)}{P(F)} = \frac{\text{\#people who watched both}}{\text{\#people who watched } F}$$

$P(E|F) = 0.42$   
Piech, CS106A, Stanford University



# Let's Make a Deal

- Game show with 3 doors: A, B, and C



- Behind one door is prize (equally likely to be any door)
- Behind other two doors is nothing
- We choose a door
- Then host opens 1 of other 2 doors, revealing nothing
- We are given option to change to other door
- Should we?
  - Note: If we don't switch,  $P(\text{win}) = 1/3$  (random)

# First Ever Sections!



**Lecturer: Chris Piech**  
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⌚ Wed 1-3pm  
📍 Gates 193



**Head TA: Ben Ulmer**  
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**TA: Ana-Maria Istrate**  
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**TA: Gobi Dasu**  
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**TA: Julia Daniel**  
✉ jdaniel7@stanford.edu



**TA: Luke Johnston**  
✉ lukej@stanford.edu



**TA: Elliot Chartock**  
✉ elboy@stanford.edu



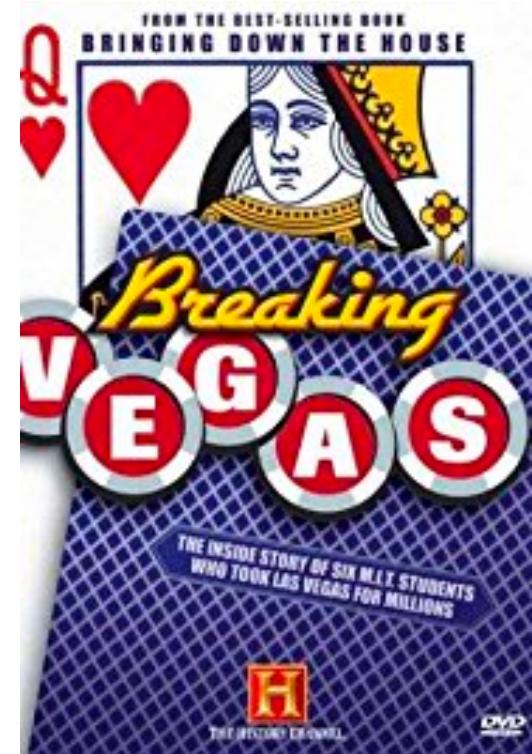
**TA: Eric Redondo**  
✉ eredondo@stanford.edu



**TA: Yuling Liu**  
✉ yulingl@stanford.edu



**TA: Brendan Corcoran**  
✉ bmc2016@stanford.edu



I'm not a robot

reCAPTCHA  
Privacy - Terms



reCAPTCHA  
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# Bayes Theorem

$$P(F|E) = \frac{P(E|F)P(F)}{P(E)}$$



# Zika Test



Positive Zika.

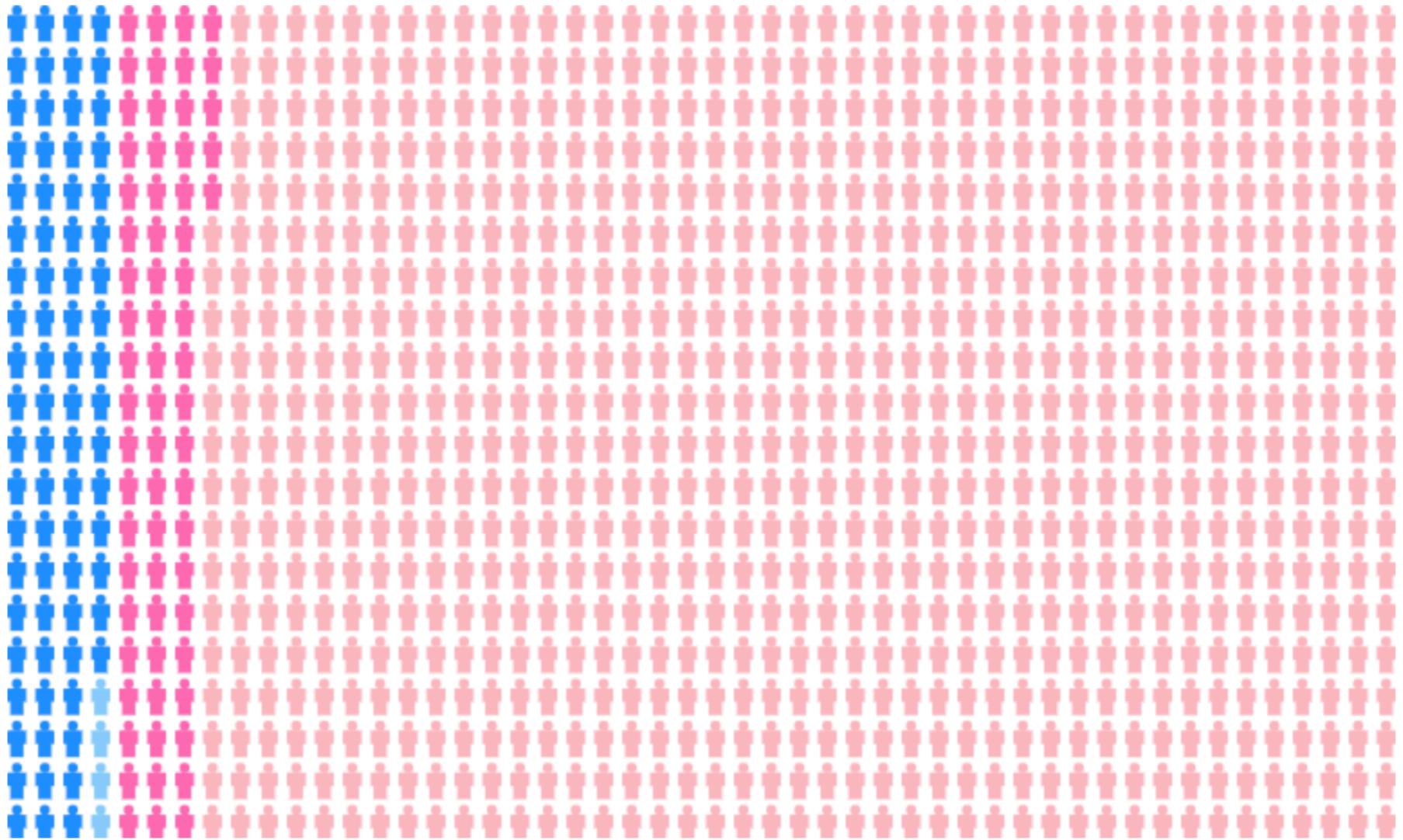
*What is the probability of zika?*

- 
- *0.1% of people have zika*
  - *90% positive rate for people with zika*
  - *7% positive rate for people without zika*

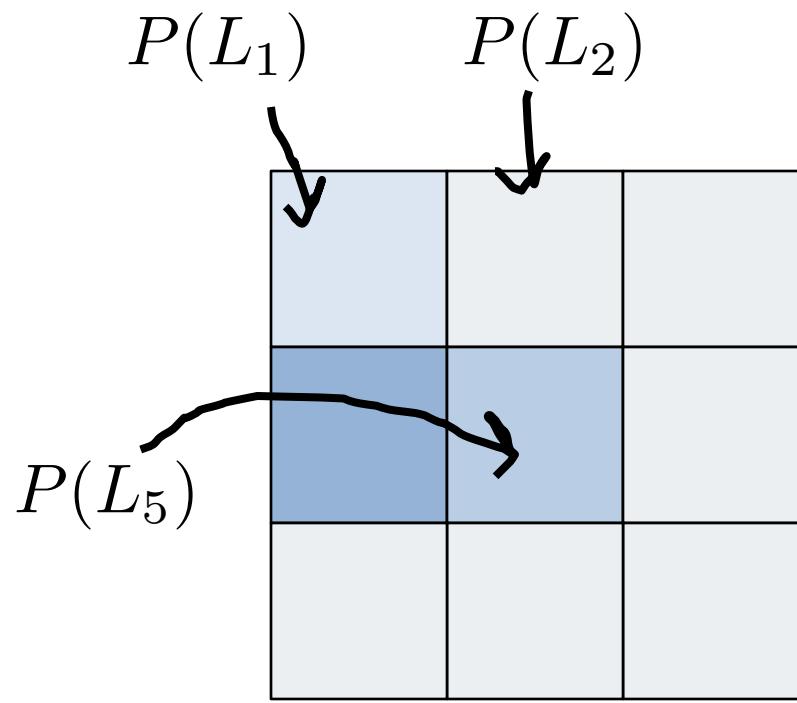
The right answer is 1%



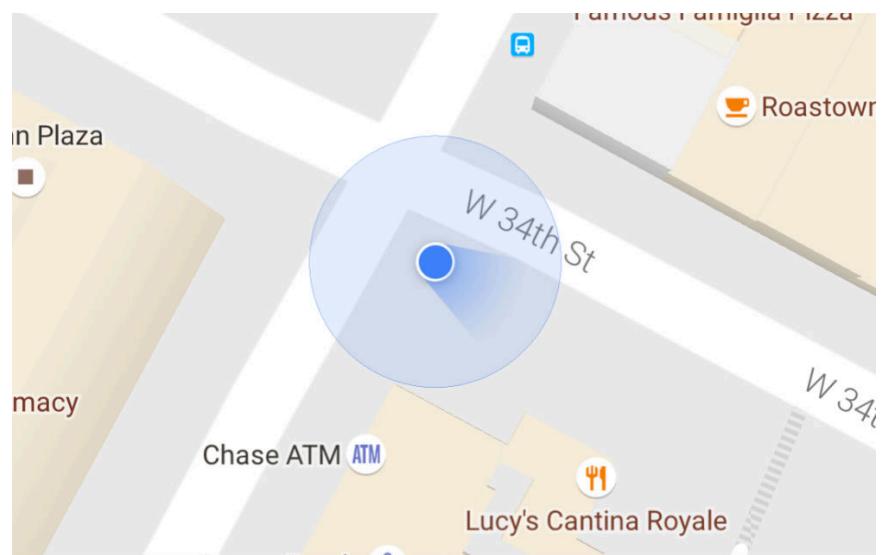
# Bayes Theorem Intuition



# Update Belief



Before Observation



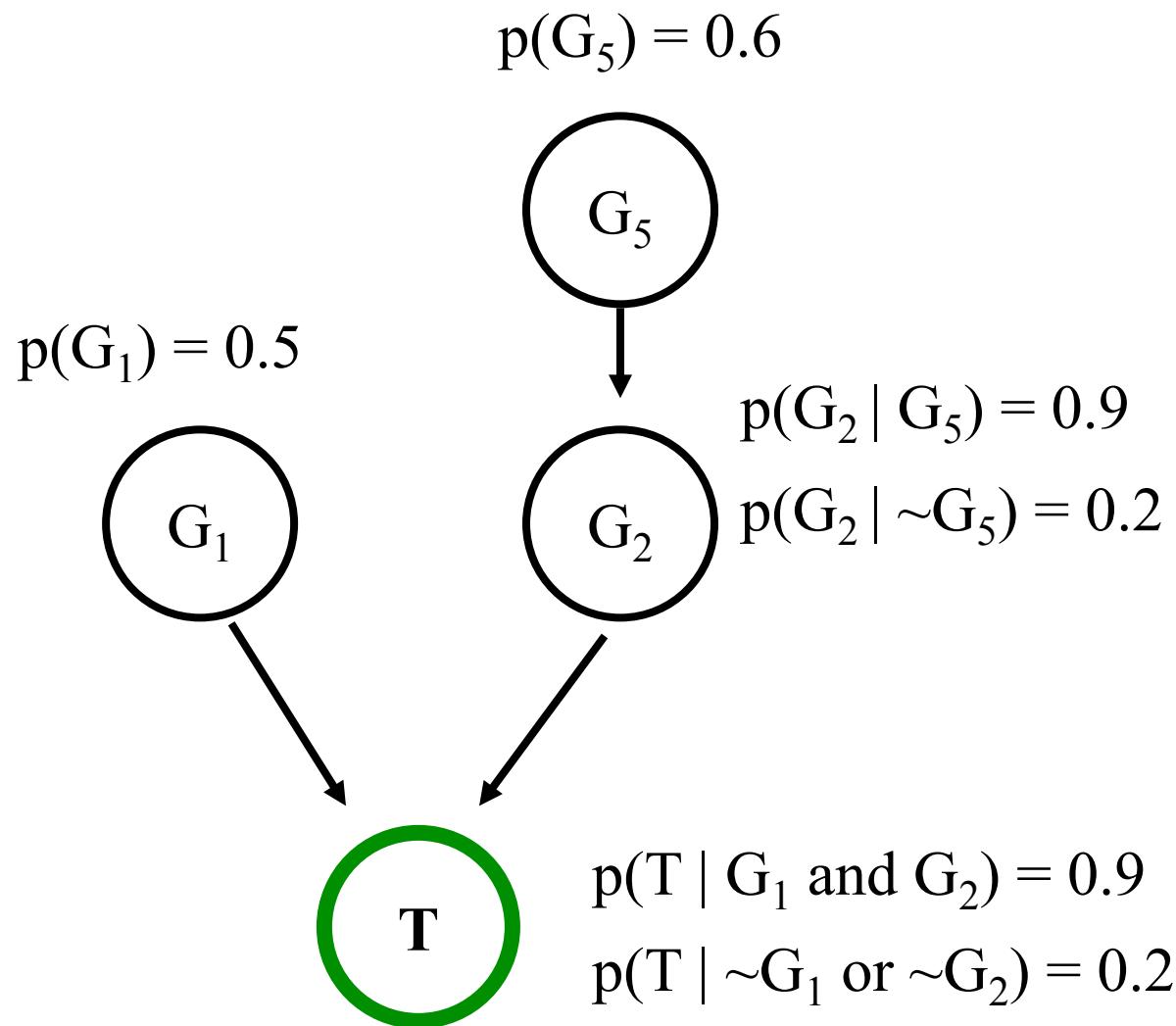
A photograph of the Statue of Liberty's head and upper torso. She is shown from the side, looking towards the right. Her right arm is raised high, holding a torch aloft. The statue is set against a clear, bright blue sky.

# Independence

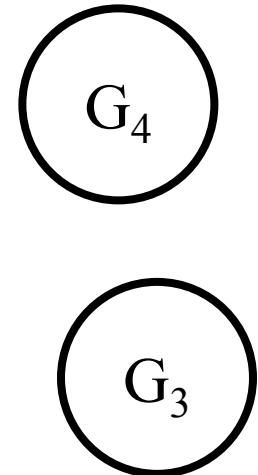
# Recall our Ebola Bats

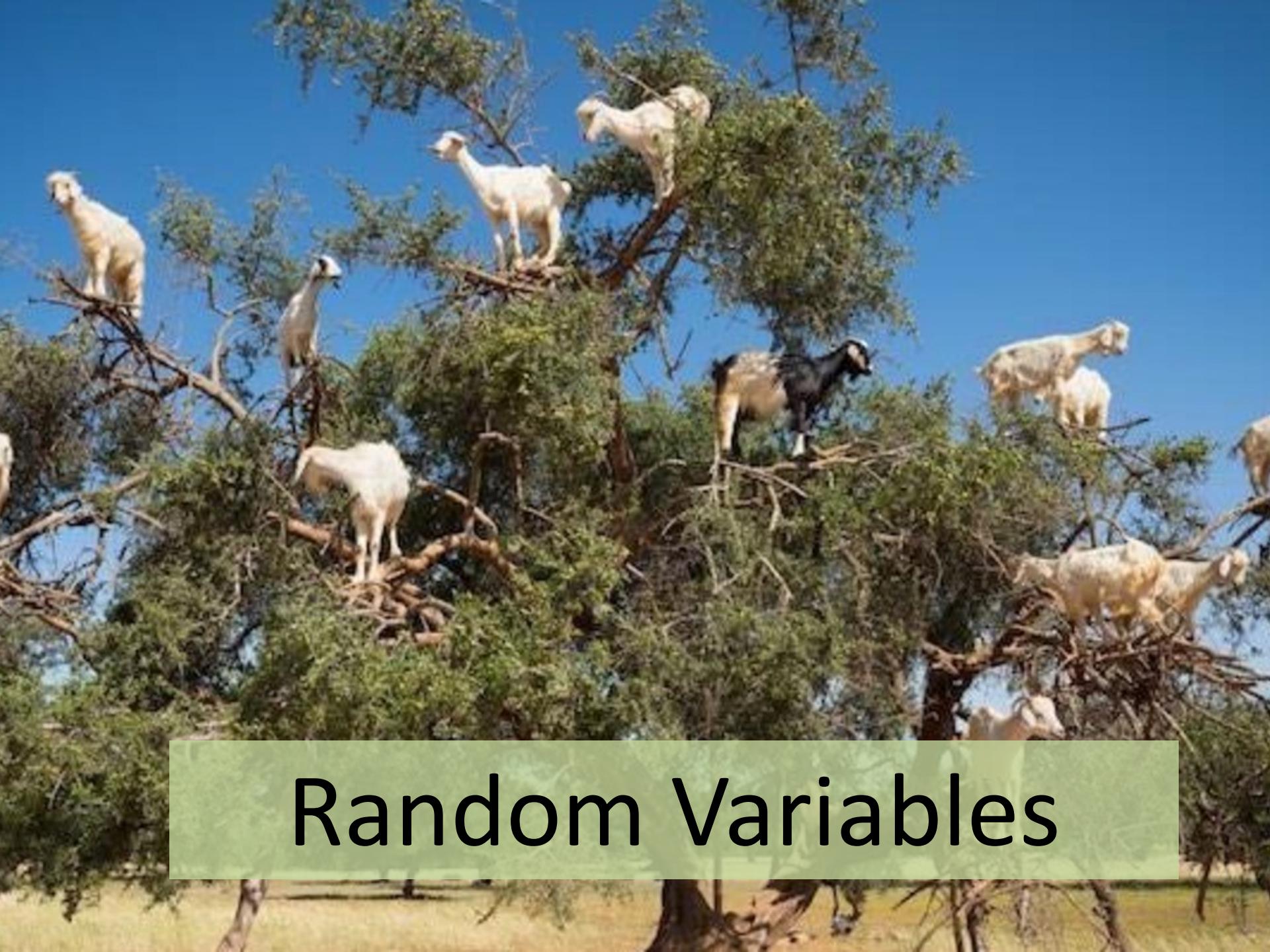


# Discovered Pattern



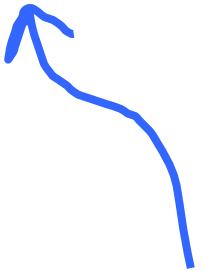
These genes  
don't impact  $T$





# Random Variables

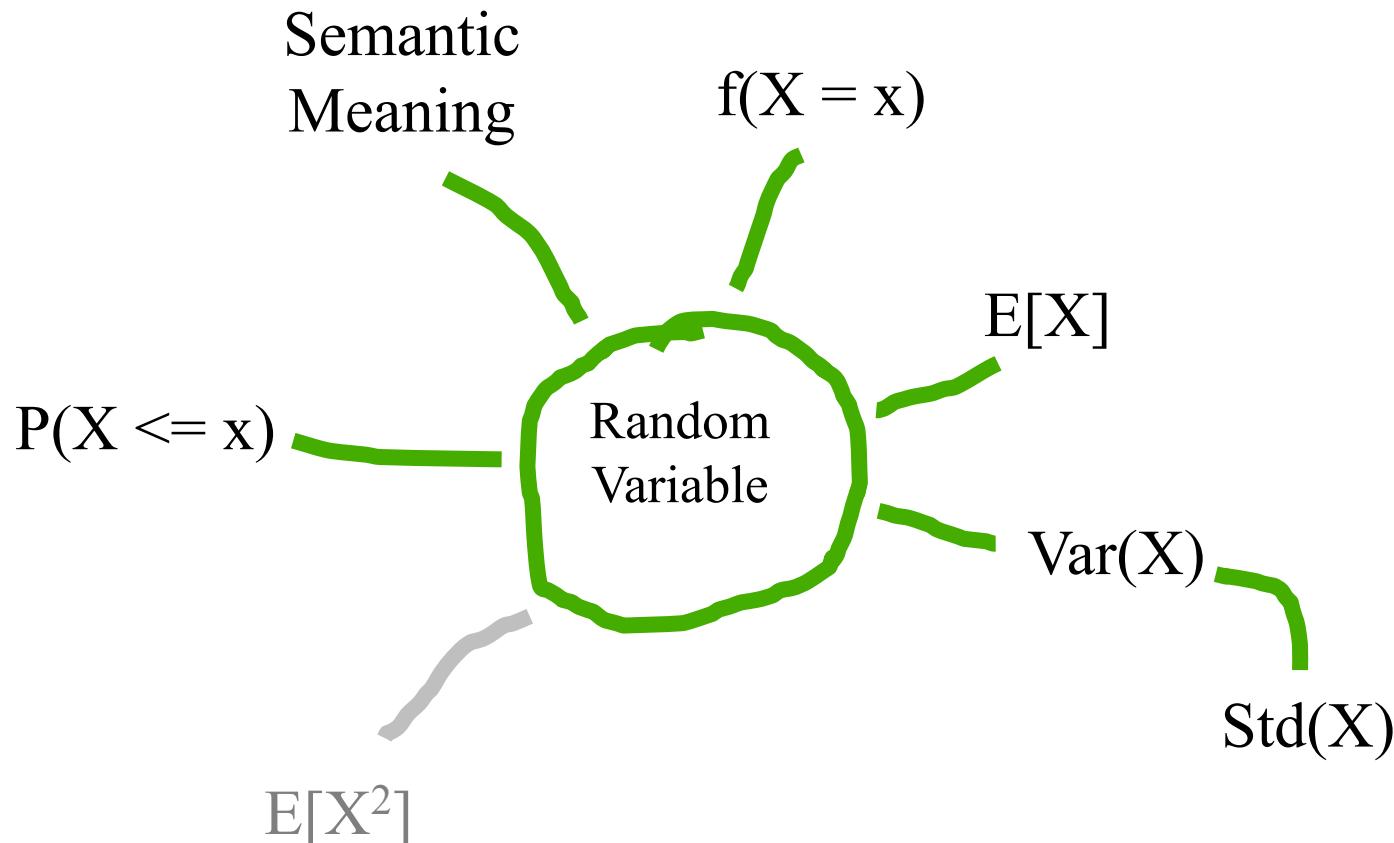
$$P(Y = k)$$



This is a function

For example Y is the number of heads in 5 coin flips

# Fundamental Properties



# Expectation

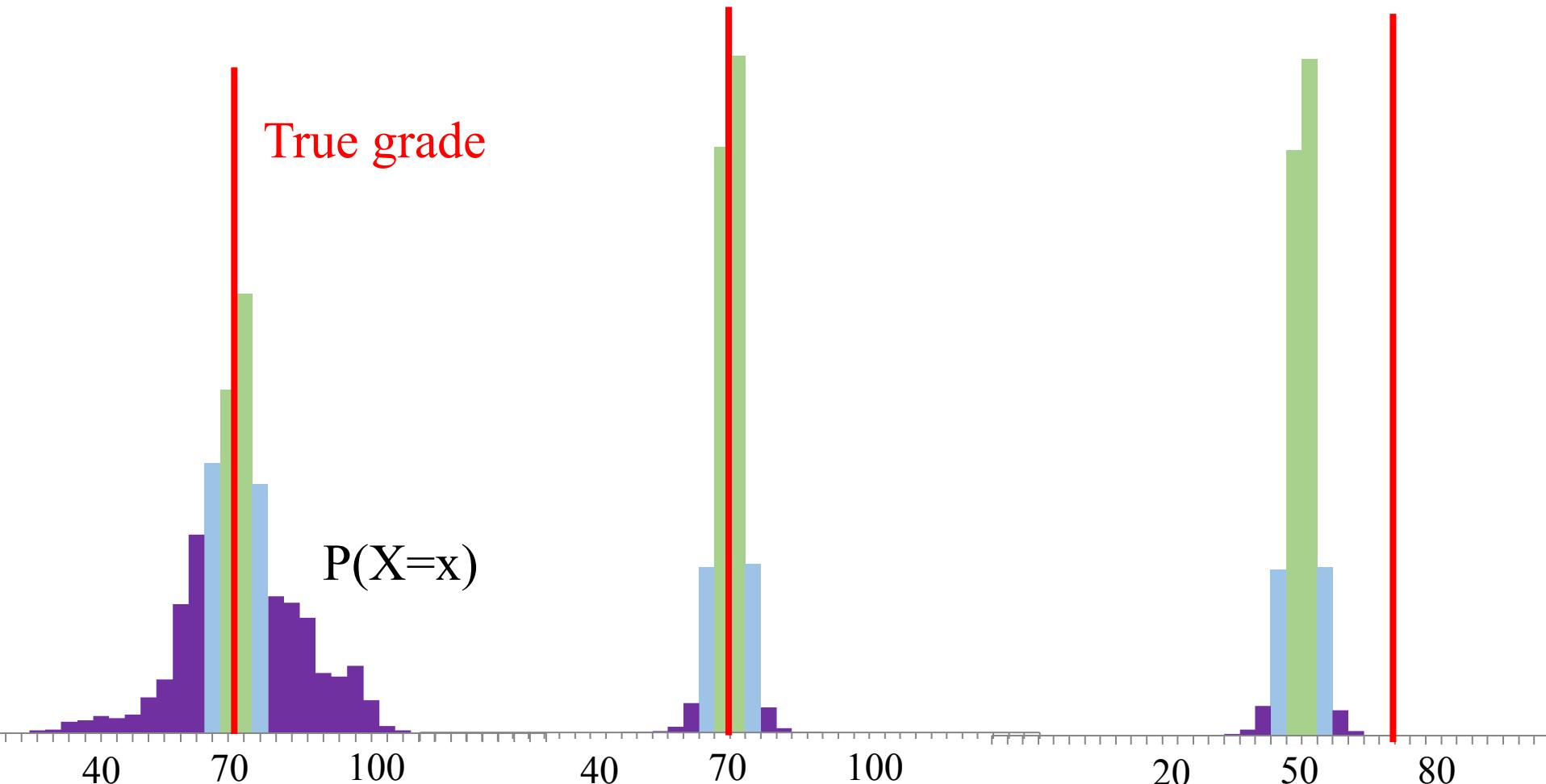
Big deal lemma: first  
stated without proof

$$E[X + Y] = E[X] + E[Y]$$

Generalized:  $E\left[\sum_{i=1}^n X_i\right] = \sum_{i=1}^n E[X_i]$

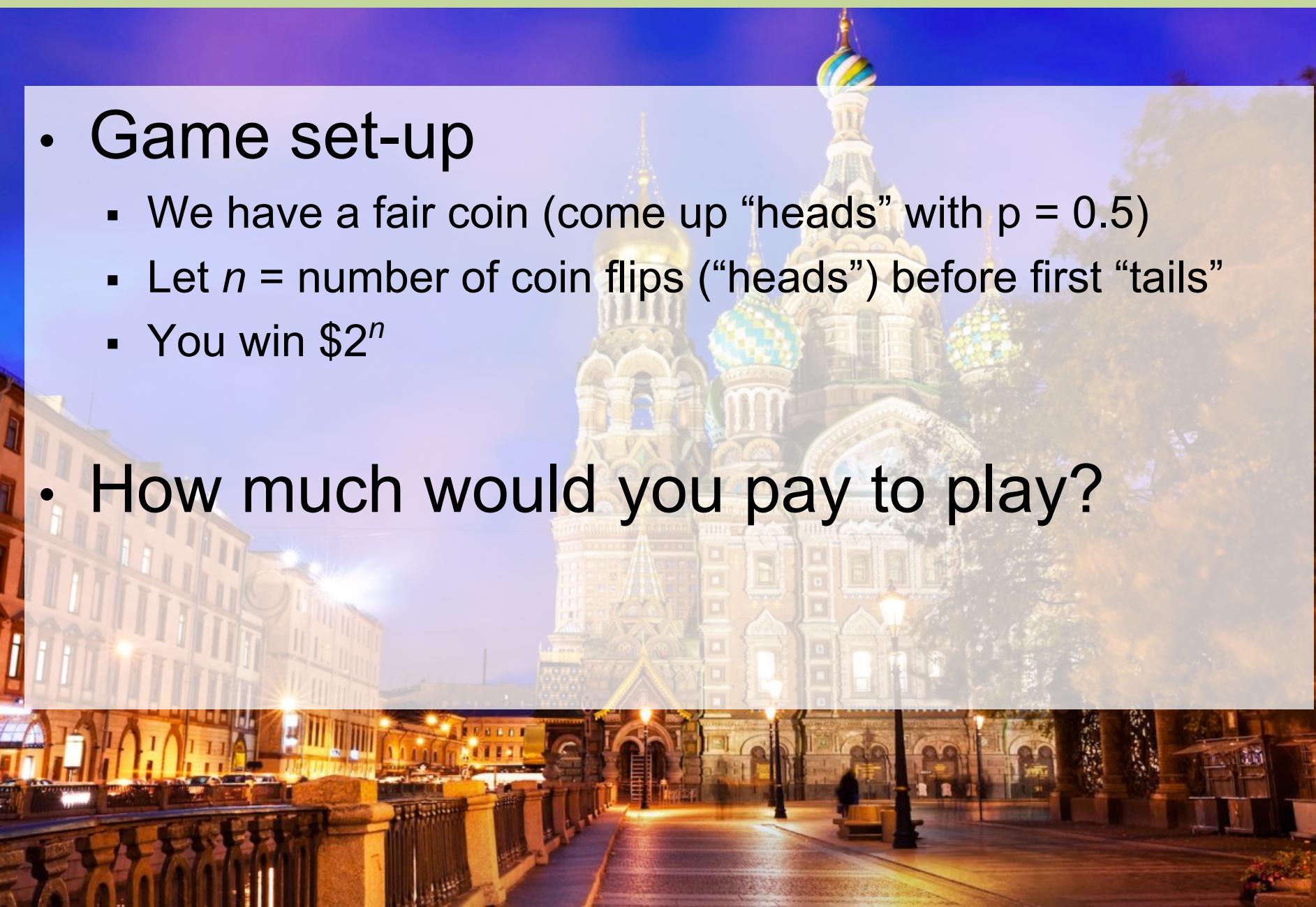
Holds regardless of dependency between  $X_i$ 's

$X$  is the score a peer grader gives to an assignment submission



# St Petersburg

- Game set-up
  - We have a fair coin (come up “heads” with  $p = 0.5$ )
  - Let  $n$  = number of coin flips (“heads”) before first “tails”
  - You win  $\$2^n$
- How much would you pay to play?



# Bernoulli



# Binomial

Our random variable

Num trials

Probability of success on each trial

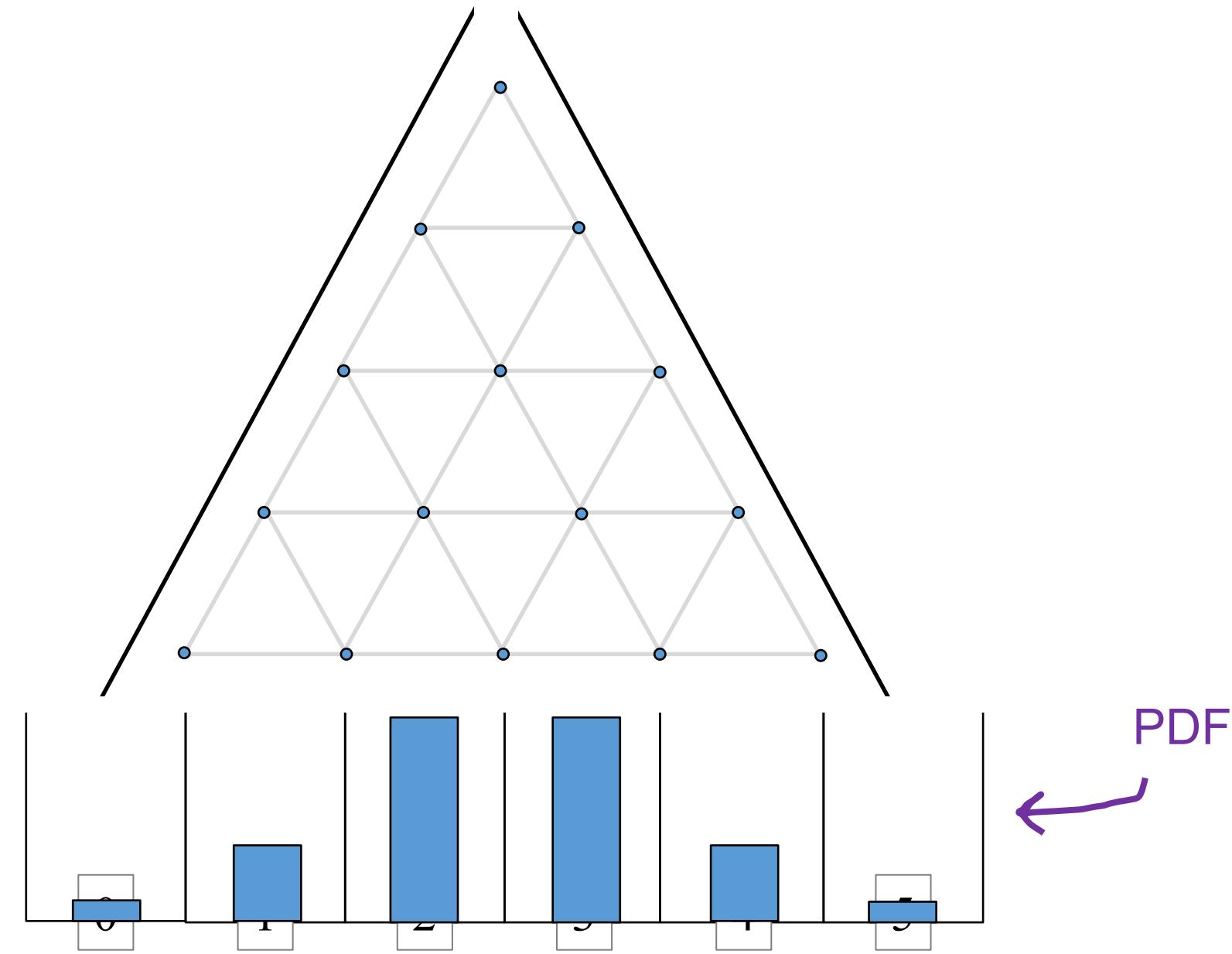
$$X \sim \text{Bin}(n, p)$$

Is distributed as a

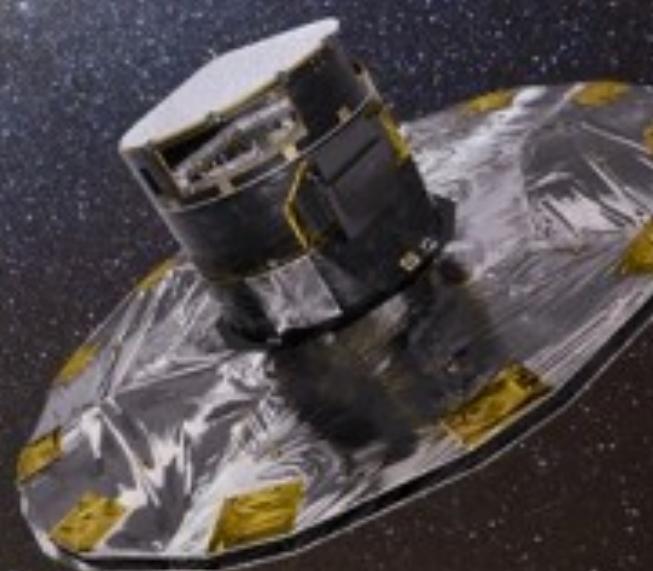
Binomial

With these parameters

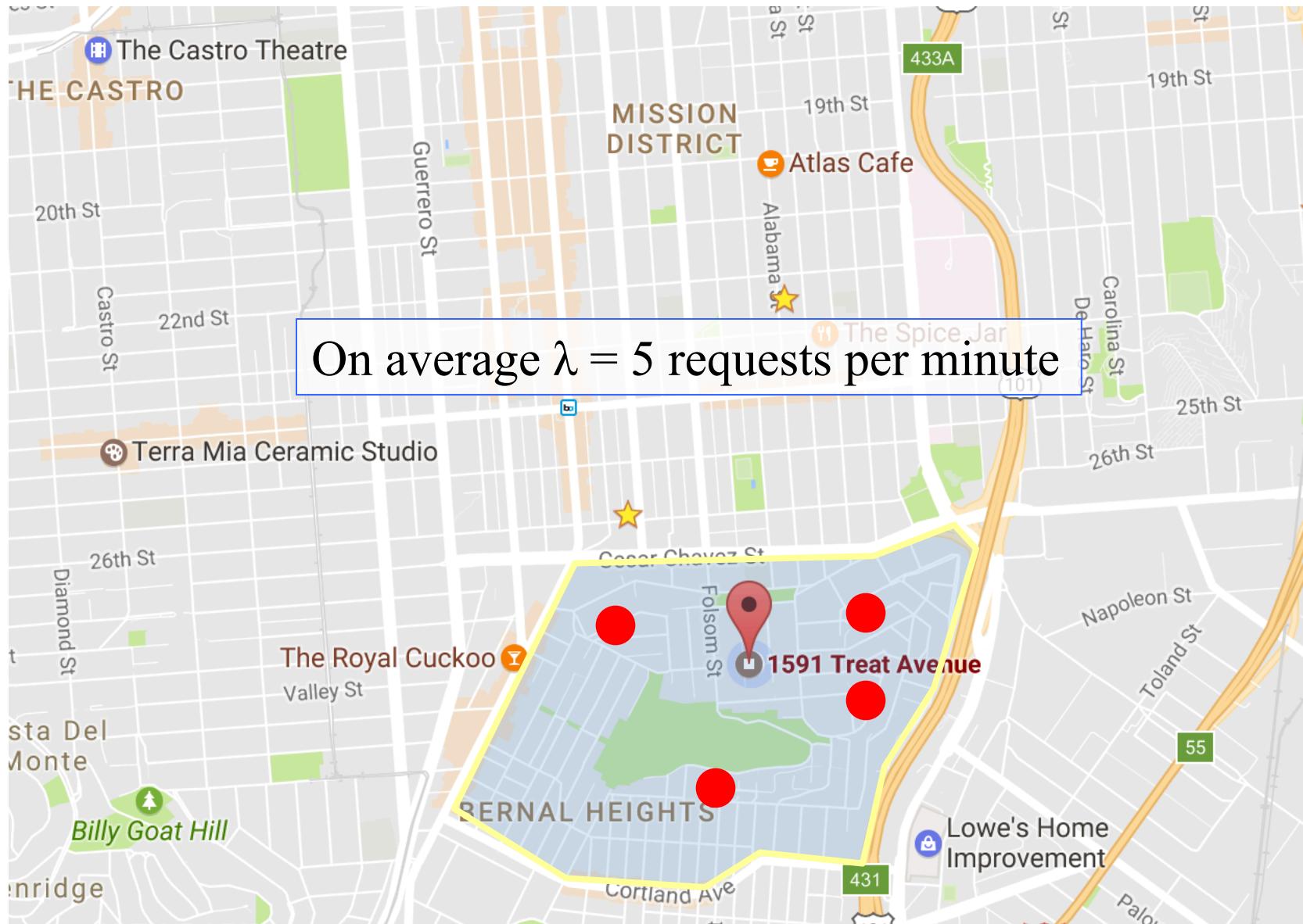
# Binomial



1001



# Poisson

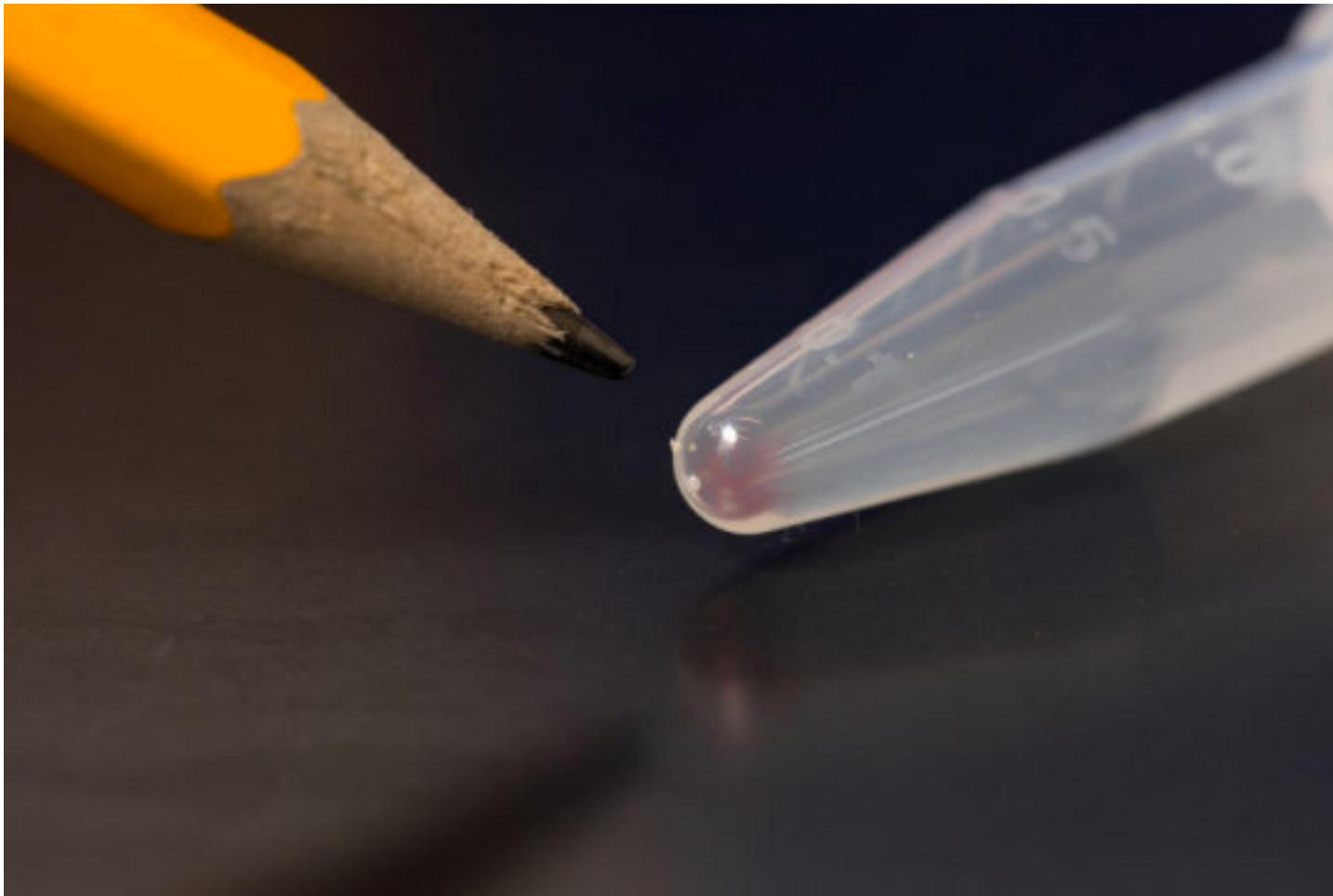


# Geometric

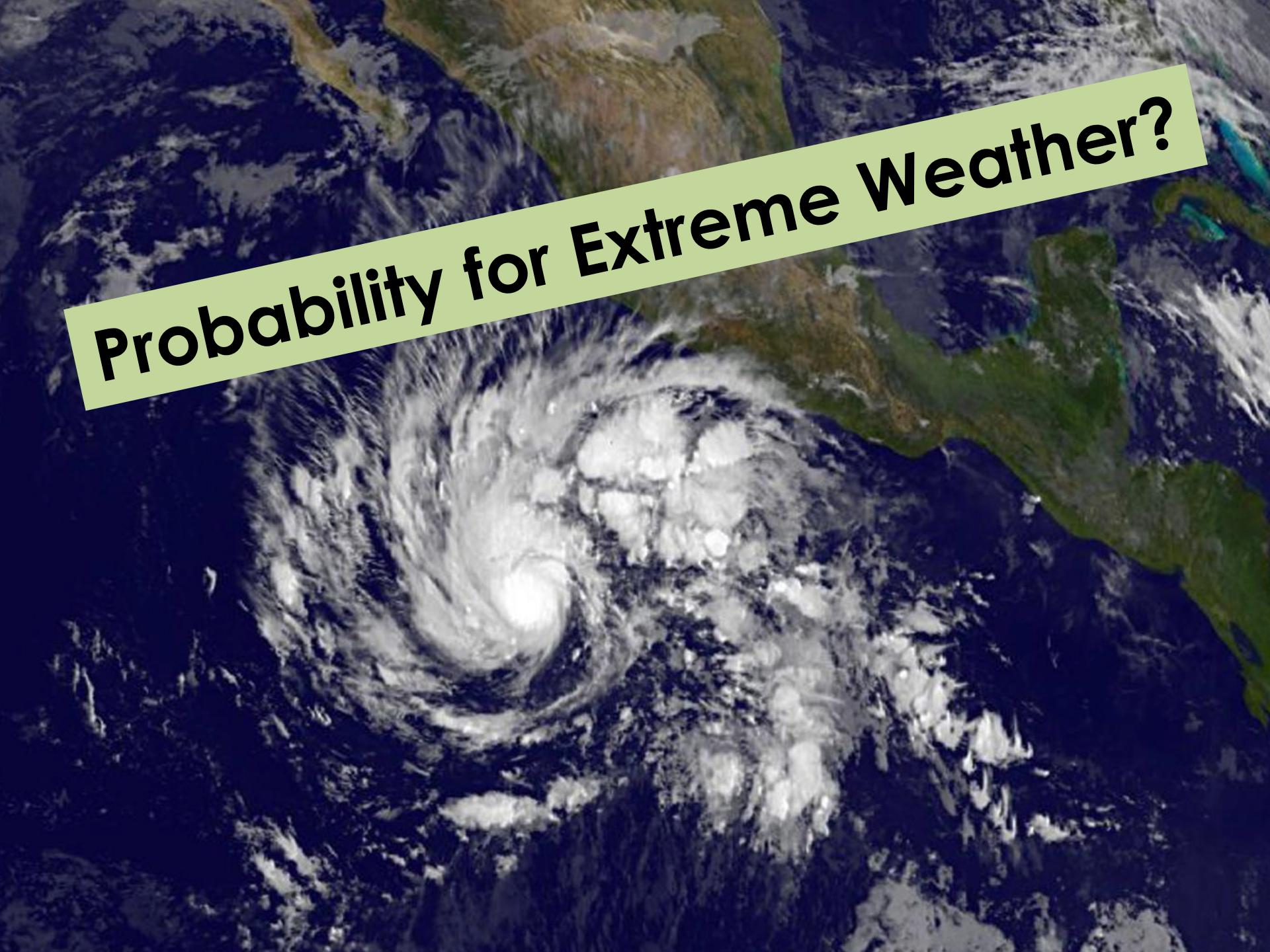
Sequence 1:

TTHHTHTTHTTTHTTTTHTTTHTHTHHT  
HTTHHTTHHHHTHHHTHTTHTHTTHTHHHTHHH  
HTHHTHHHTHTHTTHTTHHTHTHTHTHTTHT  
TTHHTHTTHTHTHTHTHTHTHTHHHTHTHT  
TTHHTHTHTHTHHHTTHTHTTTHTHHHT

# Storing Data on DNA



All the movies, images, emails and other digital data from more than 600 smartphones (10,000 gigabytes) can be stored in the faint pink smear of DNA at the end of this test tube.



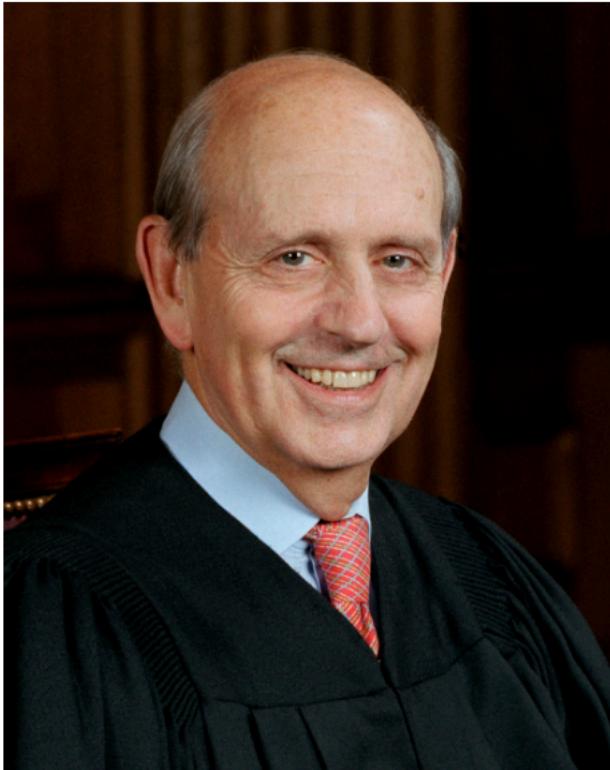
**Probability for Extreme Weather?**

# Bit Coin Mining

You “mine a bitcoin” if, for given data  $D$ , you find a number  $N$  such that  $\text{Hash}(D, N)$  produces a string that starts with  $g$  zeroes.



# Representative Juries



Simulate

Simulation:

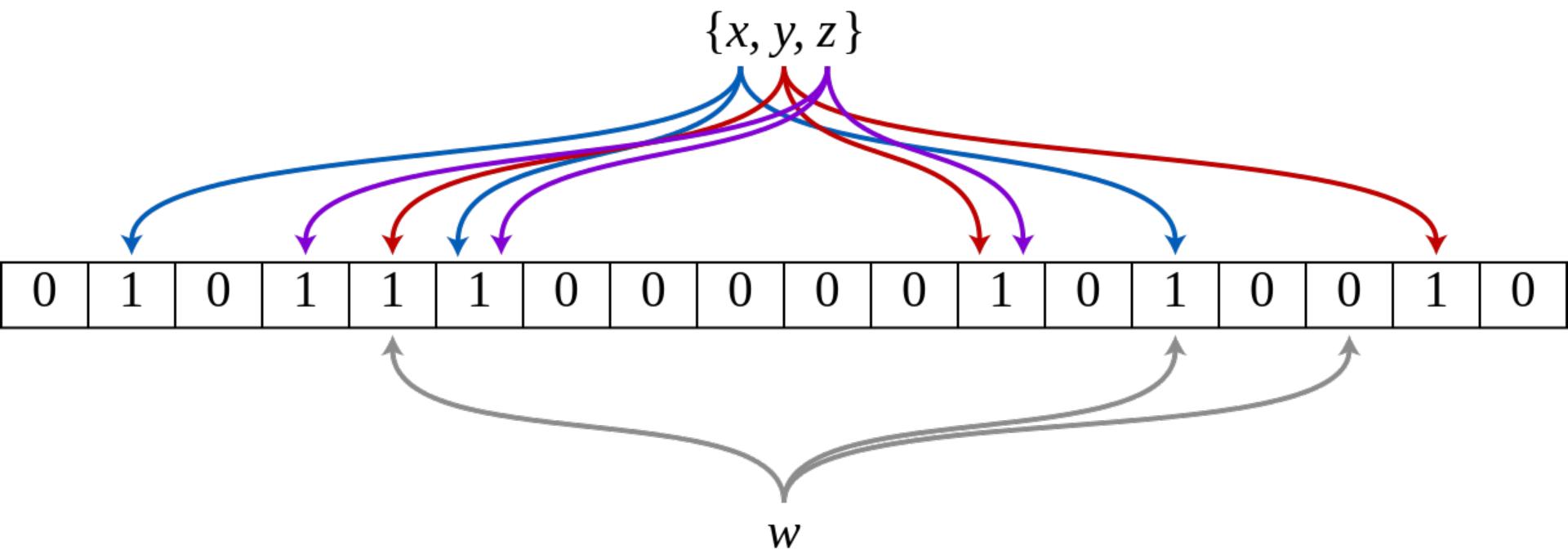


# Dating at Stanford

Each person you date has a 0.2 probability of being someone you spend your life with. What is the average number of people one will date? What is the standard deviation?

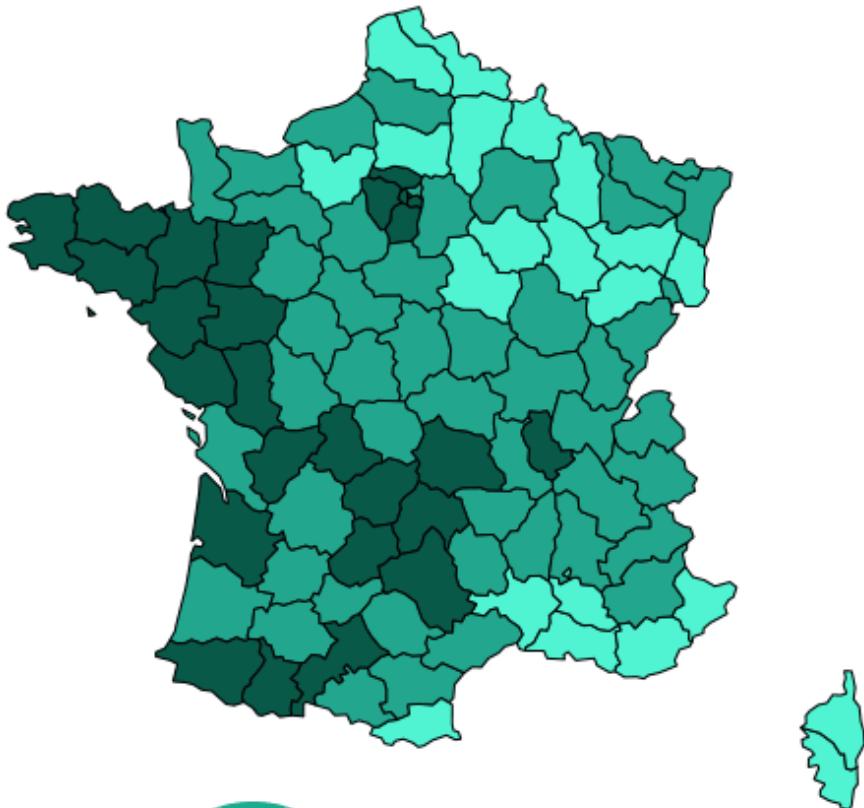


# Bloom Filter

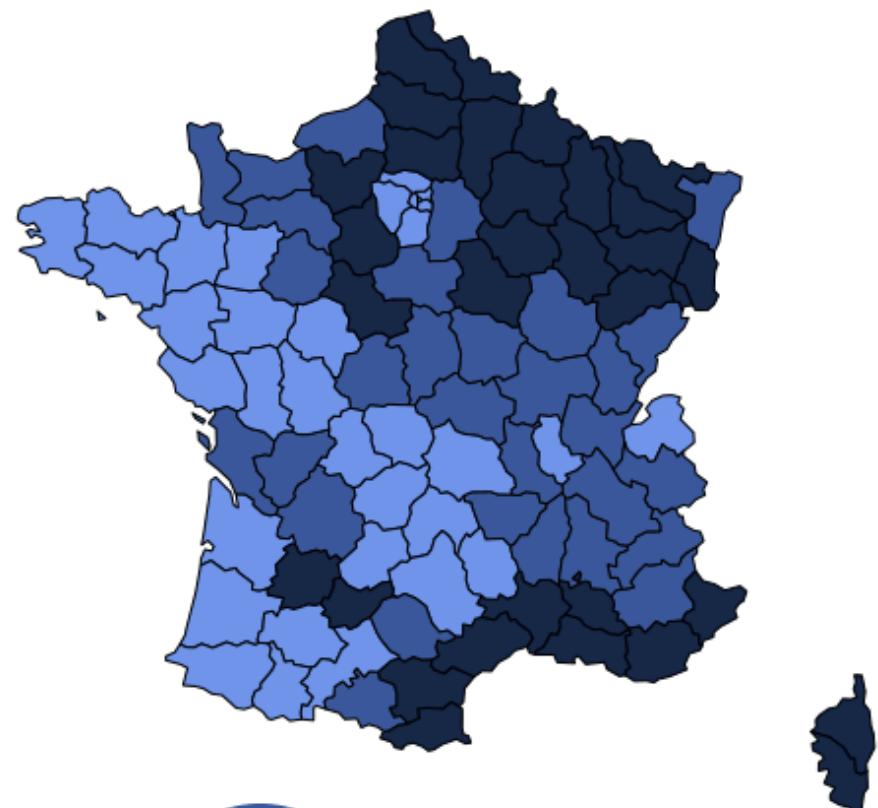


# Election Prediction

Macron polled strongly in the west, while Le Pen fared better in the north-east and south coast  
The two remaining presidential candidates' vote share mapped



- Less than 20%
- 20-25%
- More than 25%



- Less than 20%
- 20-25%
- More than 25%

random( ) ?

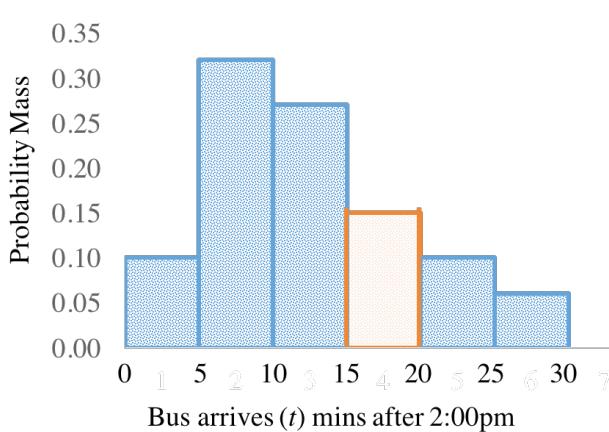
# Riding the Marguerite



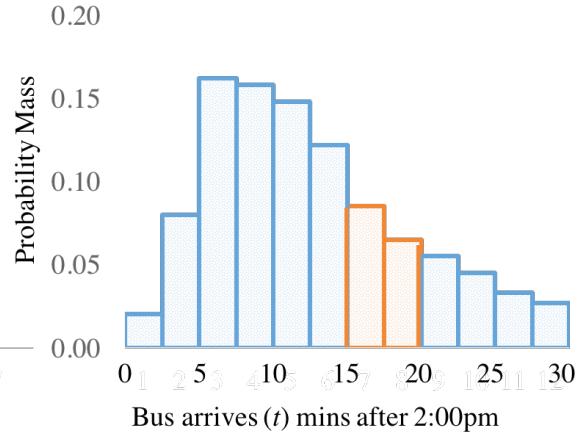
*You are running to the bus stop.  
You don't know exactly when  
the bus arrives. You arrive at  
2:20pm.*

What is  $P(\text{wait} < 5 \text{ min})$ ?

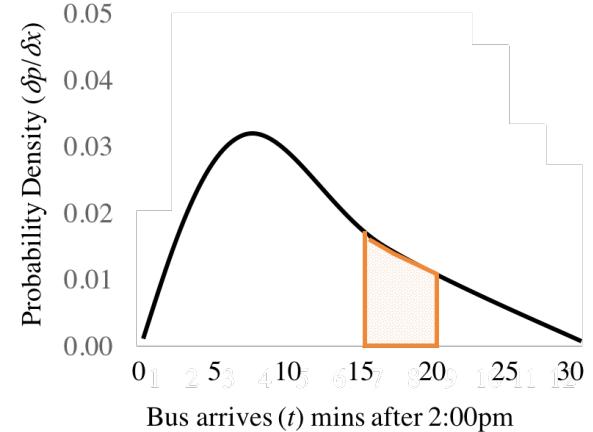
Discretize into 5 min chunks



Discretize into 2.5 min chunks



The limit at discretization size  $\rightarrow 0$



# Integrals



\*loving, not scary

What do you get if you  
integrate over a  
probability *density* function?

A probability!

# Simple Example



Consider a random  $5000 \times 5000$  matrix, where each element in the matrix is  $\text{Uniform}(0,1)$ . What is the probability that a selected eigenvalue ( $\lambda$ ) of the matrix is greater than 0?\*

\* With help from Wigner, Chris is going to rephrase this problem

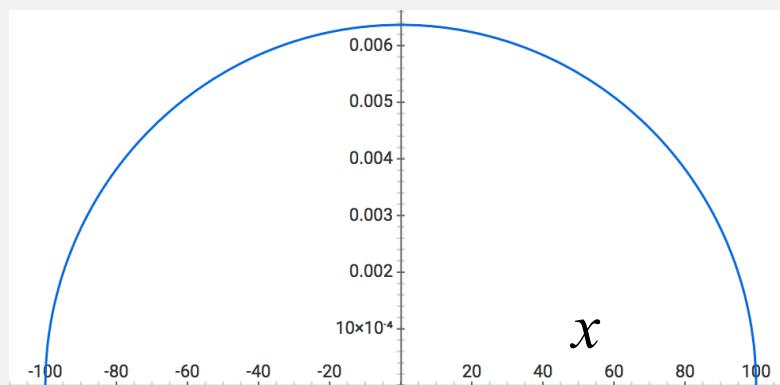
# Simple Example from Quantum Physics

Let  $X$  be a continuous random variable:

Theory

$$f(x) = \frac{1}{15708} \sqrt{100^2 - x^2}$$

$f(x)$



Practice

From simulations

$p(x)$

0.006  
0.005  
0.004  
0.003  
0.002  
0.001

$x$

-50 0 50 100

$$P(X > 0) = ?$$

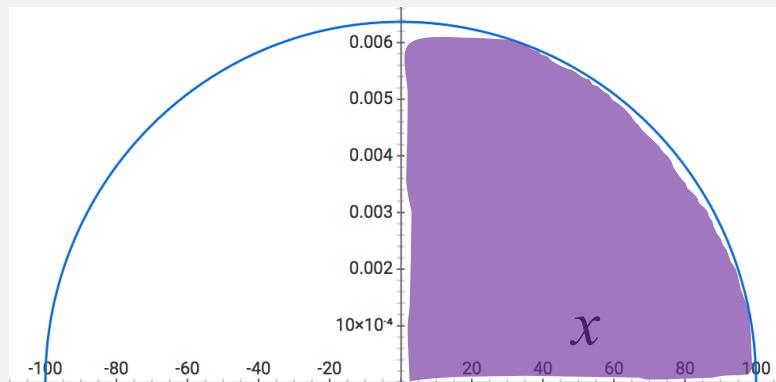
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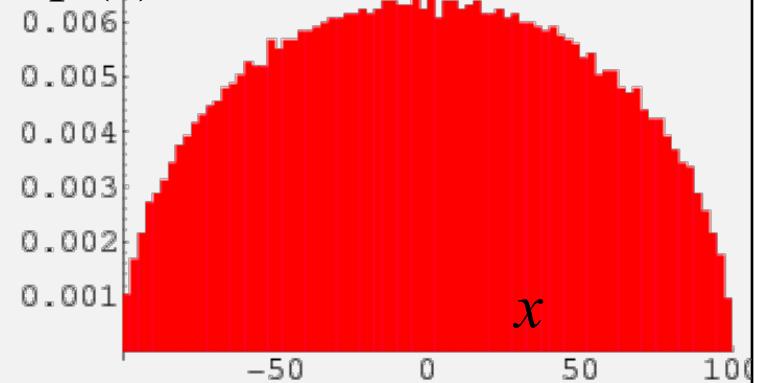
$f(x)$



Practice

From simulations

$p(x)$



Approach #1: Integrate over the PDF

$$P(X > 0) = \int_0^{100} f_X(x) dx$$

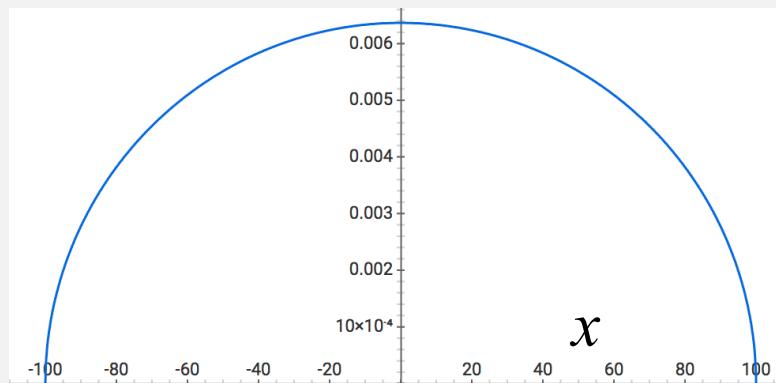
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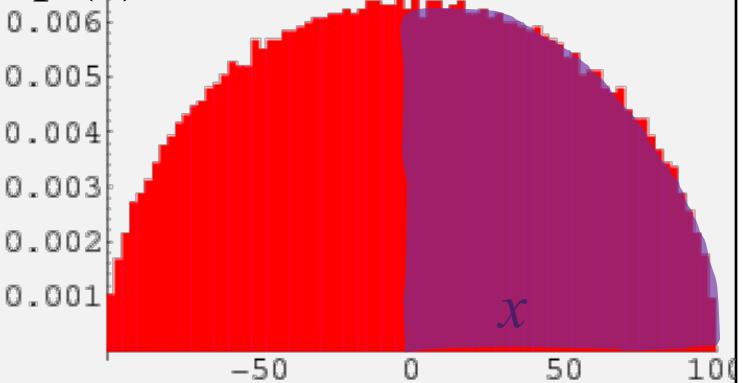
$f(x)$



Practice

From simulations

$p(x)$



Approach #2: Discrete Approximation

$$P(X > 0) \approx \sum_{i=0}^{100} P(X = i)$$

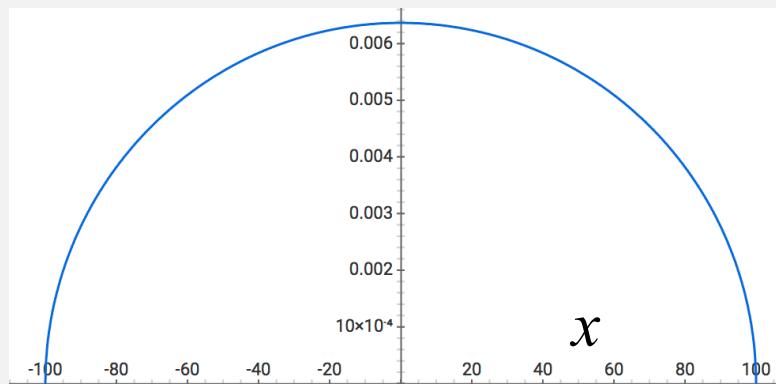
# Simple Example from Quantum Physics

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Theory

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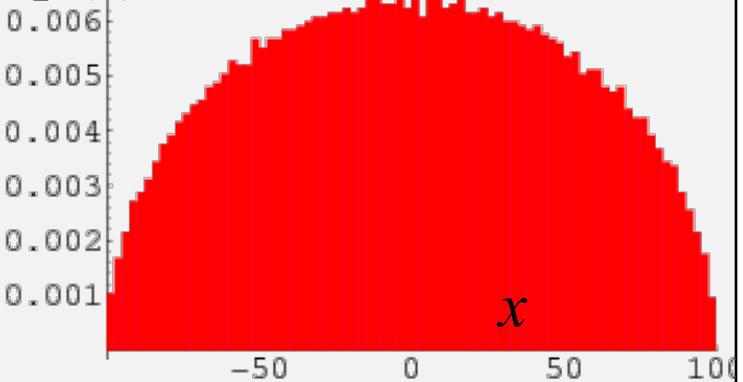
$f(x)$



Practice

From simulations

$p(x)$



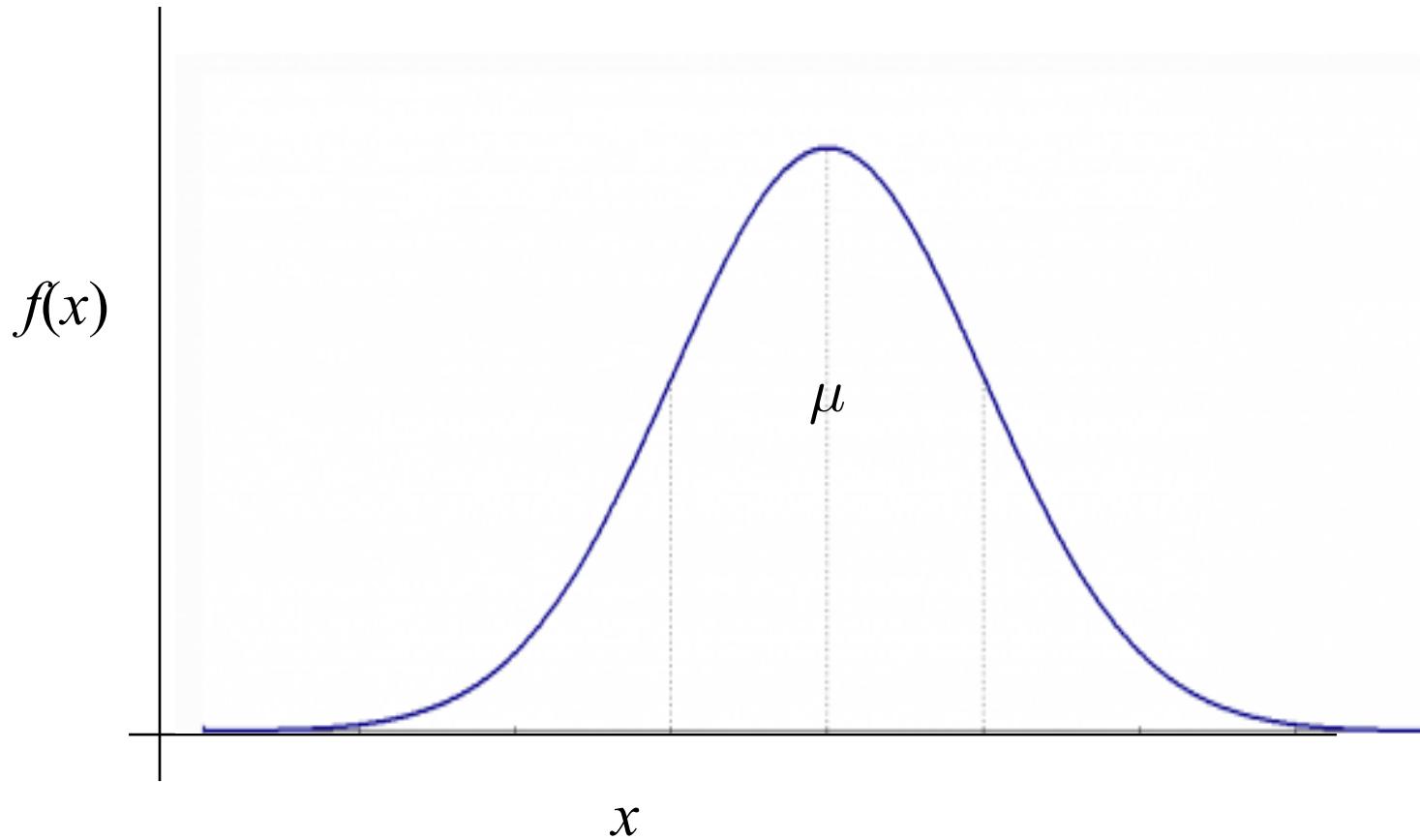
Approach #3: Know Semi-Circles

$$P(X > 0) = \frac{1}{2}$$

# Probability Density Function

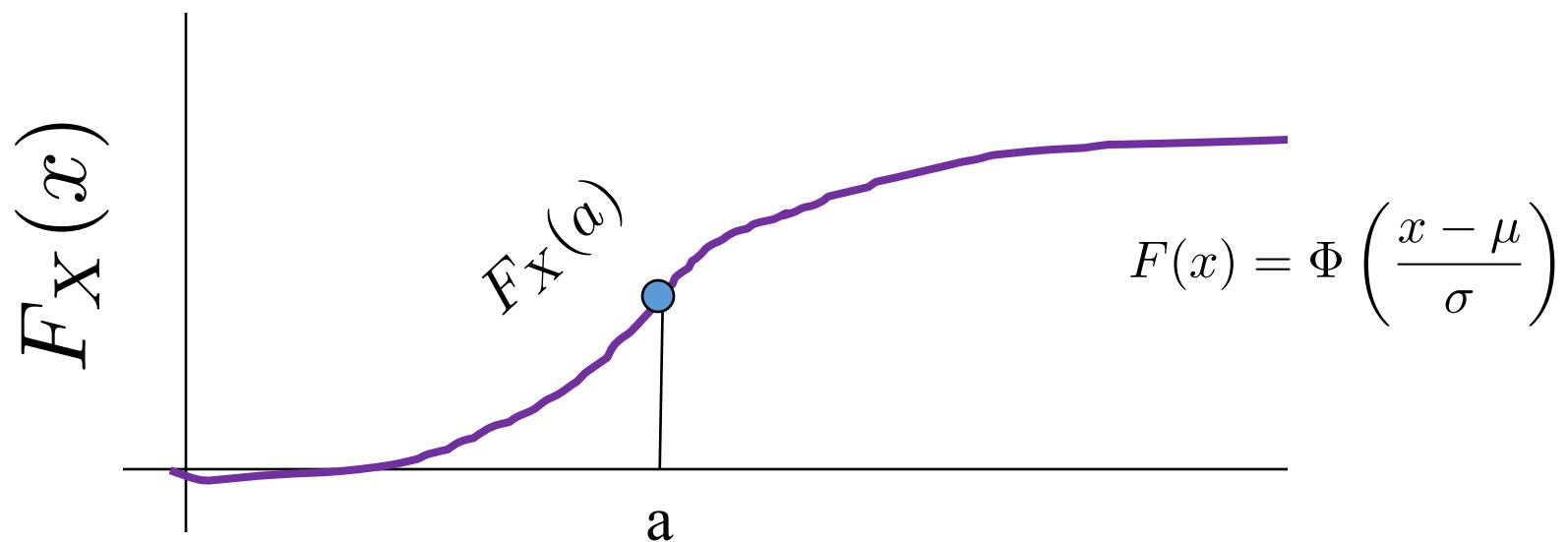
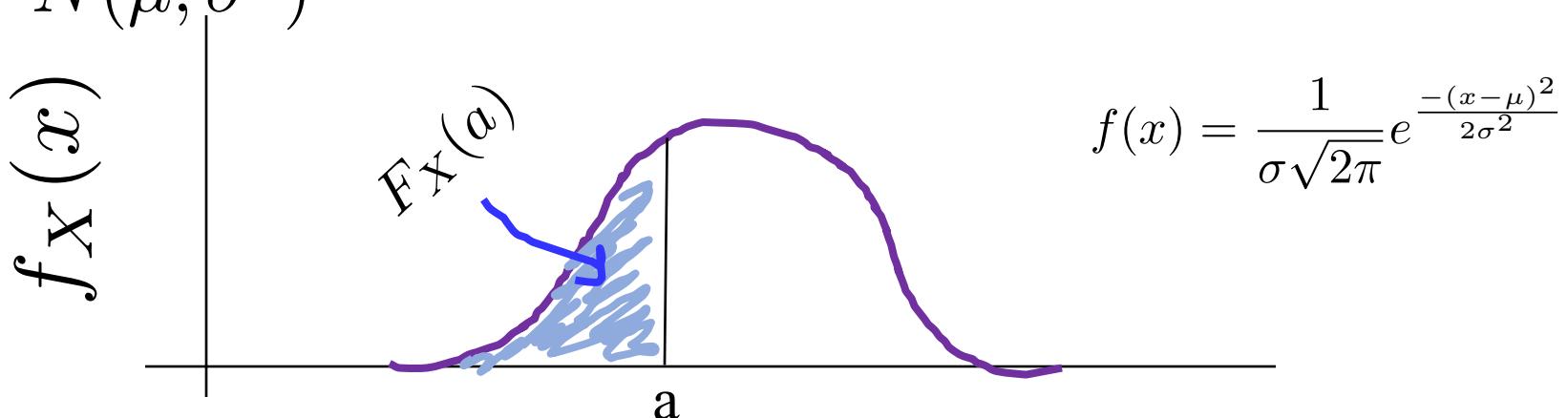
$$\mathcal{N}(\mu, \sigma^2)$$

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{\frac{-(x-\mu)^2}{2\sigma^2}}$$



# PDF and CDF of a Normal

$$X \sim N(\mu, \sigma^2)$$



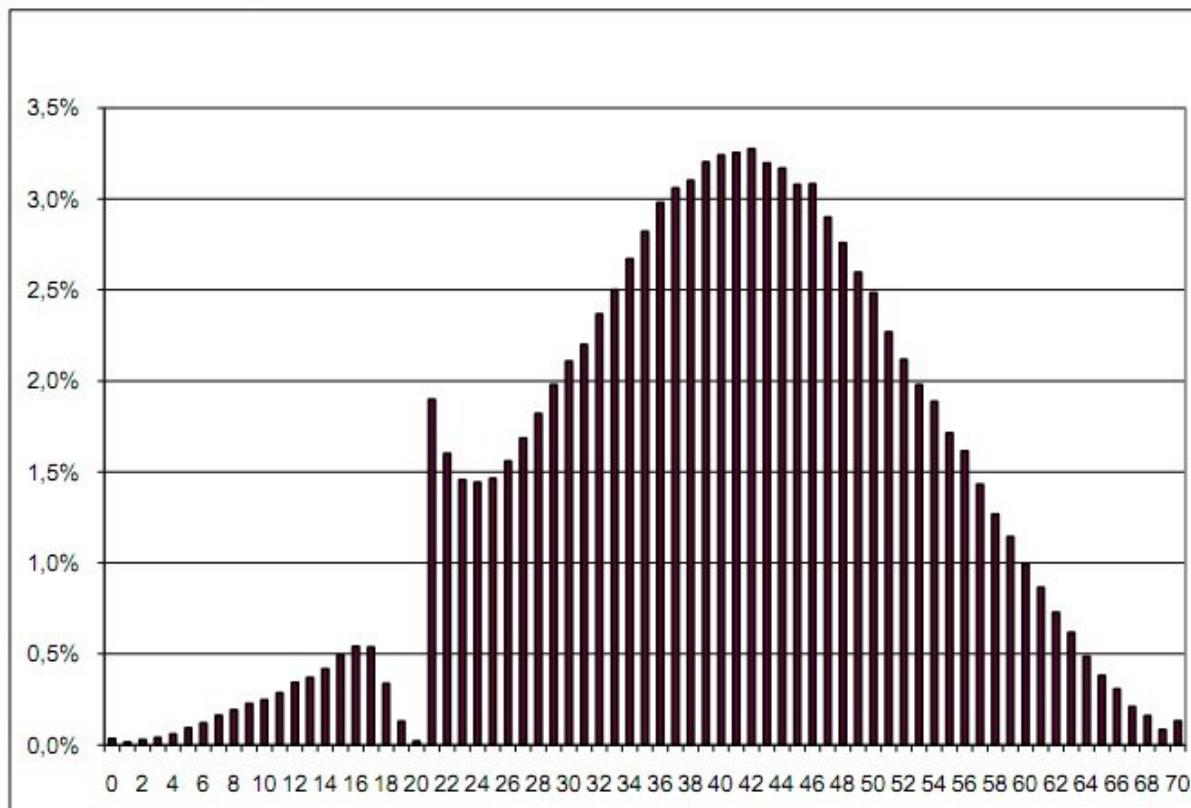
A CDF is the integral from  $-\infty$  to  $x$  of the PDF

# Altruism?

Scores for a standardized test that students in Poland are required to pass before moving on in school

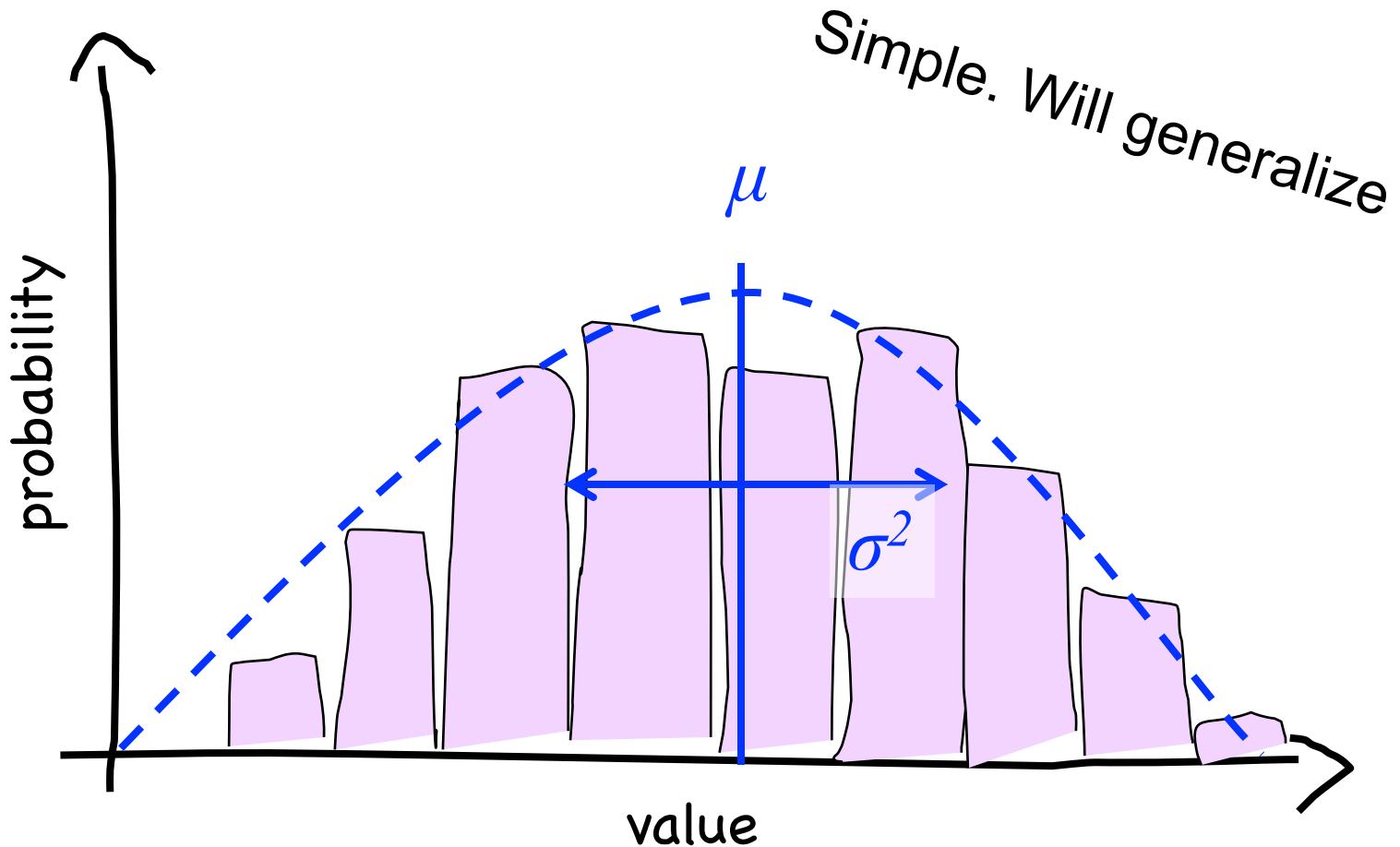
See if you can guess the minimum score to pass the test.

## 2.1. Poziom podstawowy



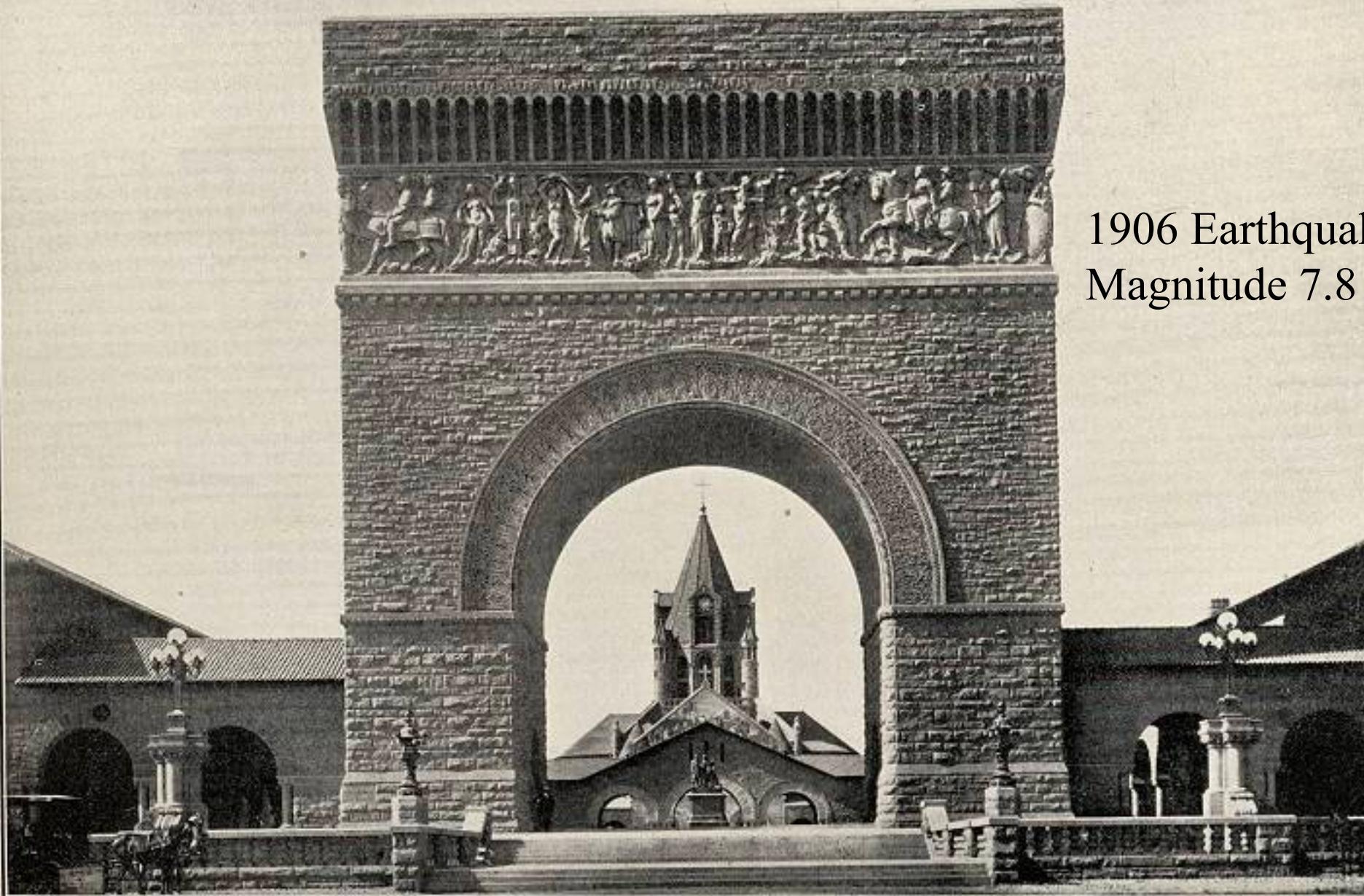
Wykres 1. Rozkład wyników na poziomie podstawowym

# Simplicity is Humble



\* A Gaussian maximizes entropy for a given mean and variance

1906 Earthquake  
Magnitude 7.8



ILL. No. 65. MEMORIAL ARCH, WITH CHURCH IN BACKGROUND, STANFORD UNIVERSITY, SHOWING TYPES OF CARVED WORK WITH THE SANDSTONE.

# Will the Warriors Win?



What is the probability that the Warriors beat the Blazers?

How do you model zero sum games?

# ELO Ratings

How it works:

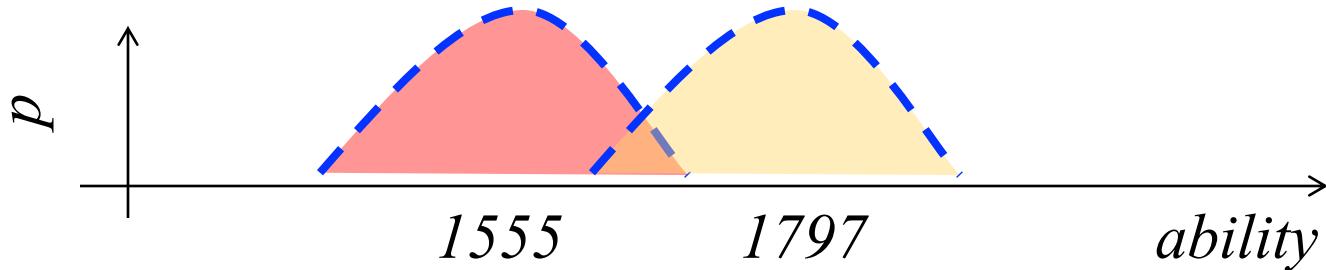
- Each team has an “ELO” score  $S$ , calculated based on their past performance.
- Each game, the team has ability  $A \sim N(S, 200^2)$
- The team with the higher sampled ability wins.



Arpad Elo

$$A_B \sim \mathcal{N}(1555, 200^2)$$

$$A_W \sim \mathcal{N}(1797, 200^2)$$



$$P(\text{Warriors win}) = P(A_W > A_B)$$

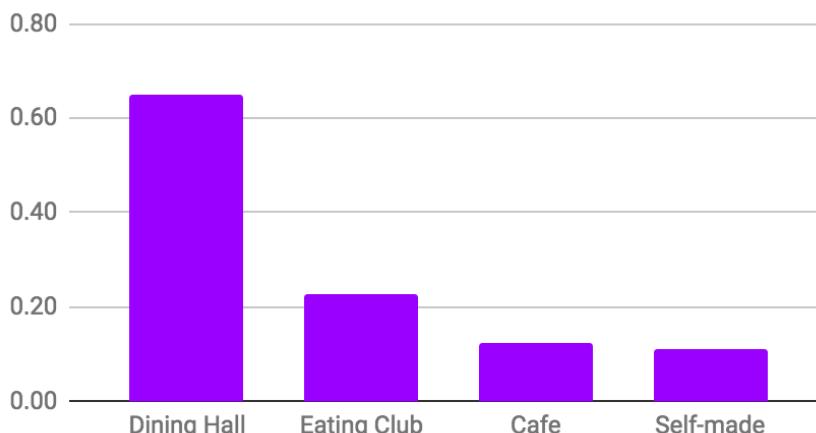
A photograph of three children playing outdoors at sunset. One child is in the foreground on the left, jumping with arms raised. Two other children are in the background, one slightly behind the other. They are silhouetted against a bright, cloudy sky. The foreground shows some low-lying plants.

# Joint Distributions

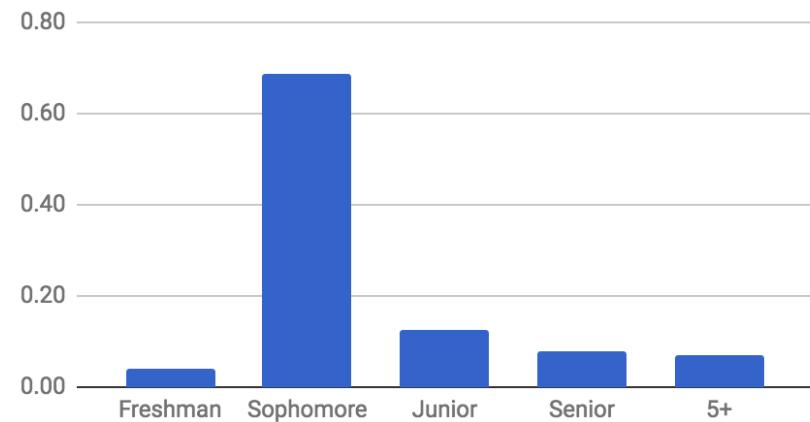
# Joint Probability Table

Joint Probability Table					
	Dining Hall	Eating Club	Cafe	Self-made	Marginal Year
Freshman	0.02	0.00	0.02	0.00	0.04
Sophomore	0.51	0.15	0.03	0.03	0.69
Junior	0.08	0.02	0.02	0.02	0.13
Senior	0.02	0.05	0.01	0.01	0.08
5+	0.02	0.01	0.05	0.05	0.07
Marginal Status	0.65	0.23	0.13	0.11	

Marginal Lunch Probability



Marginal Year



# It's Complicated Demo



Relationship Status:

Interested in:

Looking for:

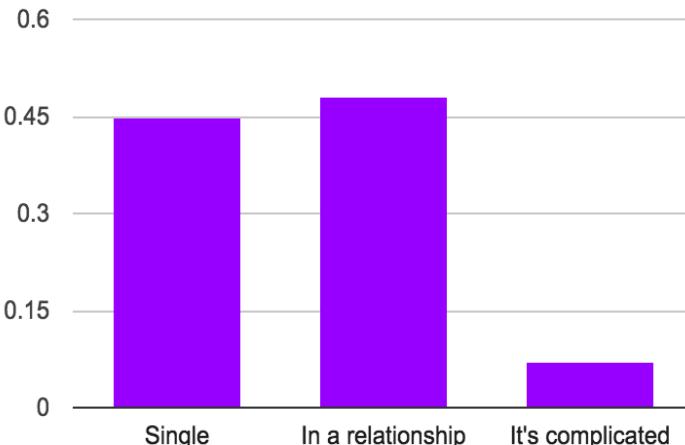
Single  
In a Relationship  
Engaged  
Married  
**It's Complicated**  
In an Open Relationship  
Widowed

Go to this URL: <https://goo.gl/jCMY18>

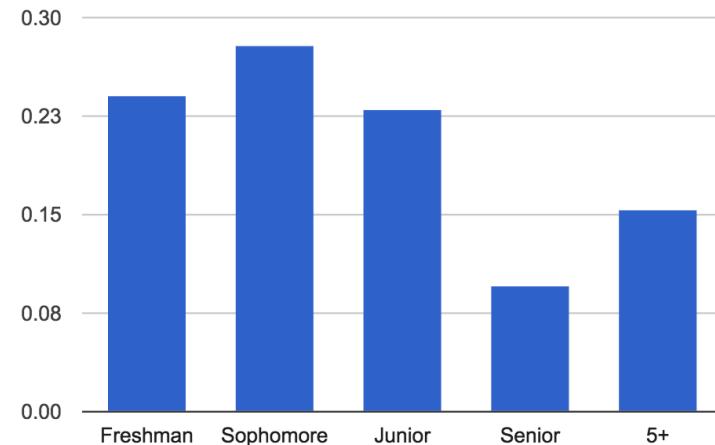
# Probability Table

Joint Probability Table				
	Single	In a relationship	It's complicated	Marginal Year
Freshman	0.13	0.09	0.02	0.24
Sophomore	0.16	0.10	0.02	0.28
Junior	0.12	0.10	0.02	0.23
Senior	0.01	0.09	0.00	0.10
5+	0.03	0.12	0.01	0.15
<b>Marginal Status</b>	0.45	0.48	0.07	

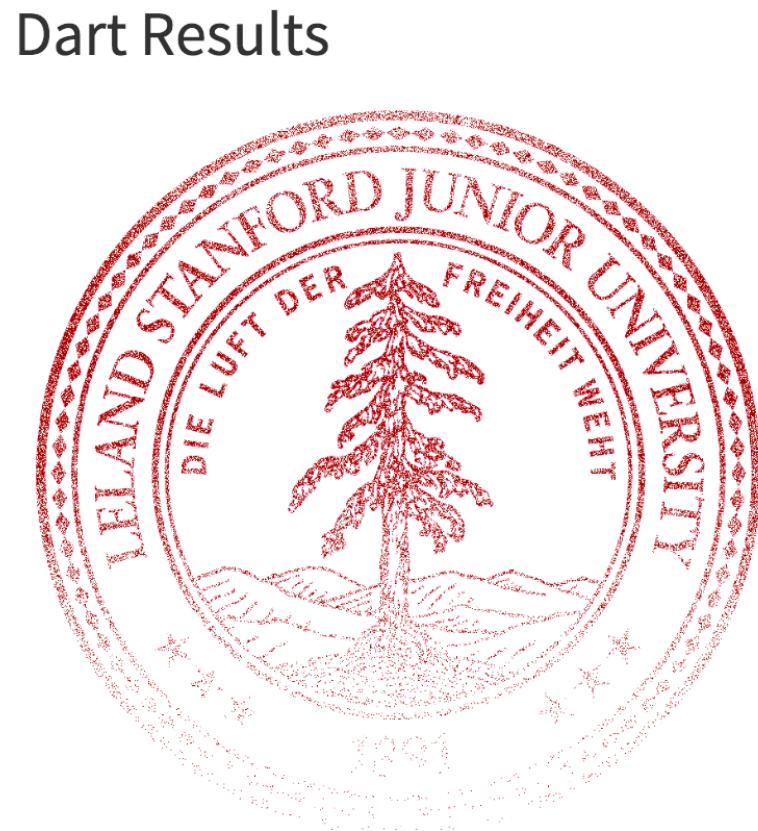
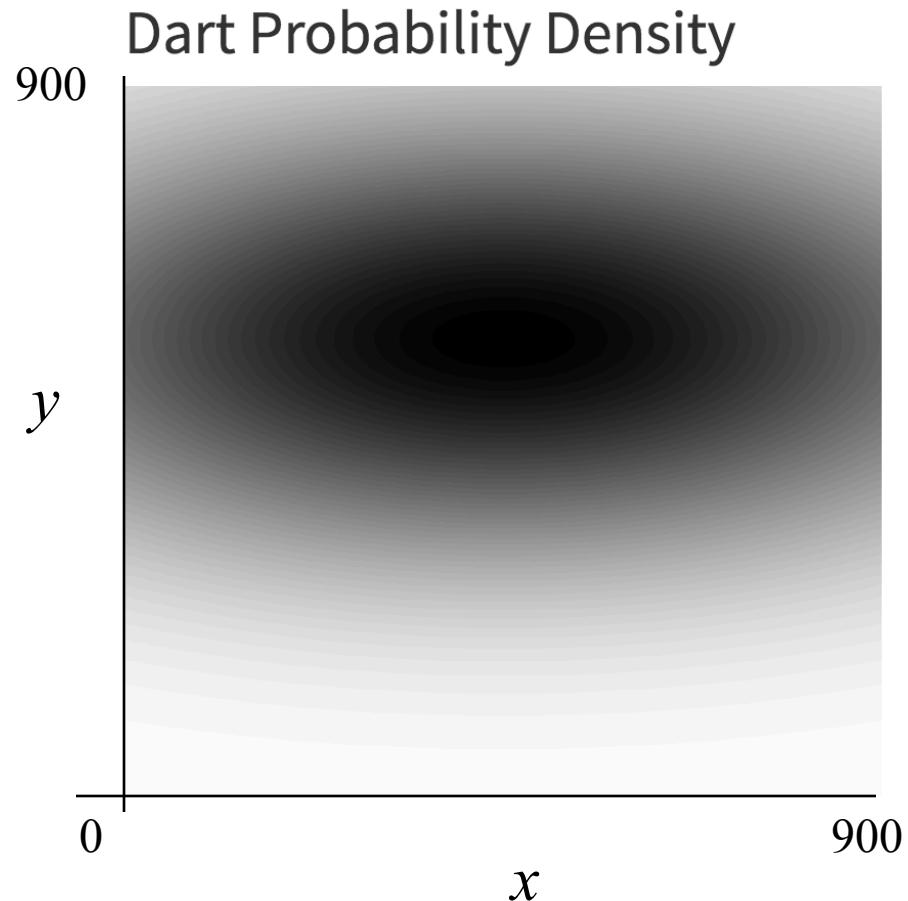
## Marginal Status Probability



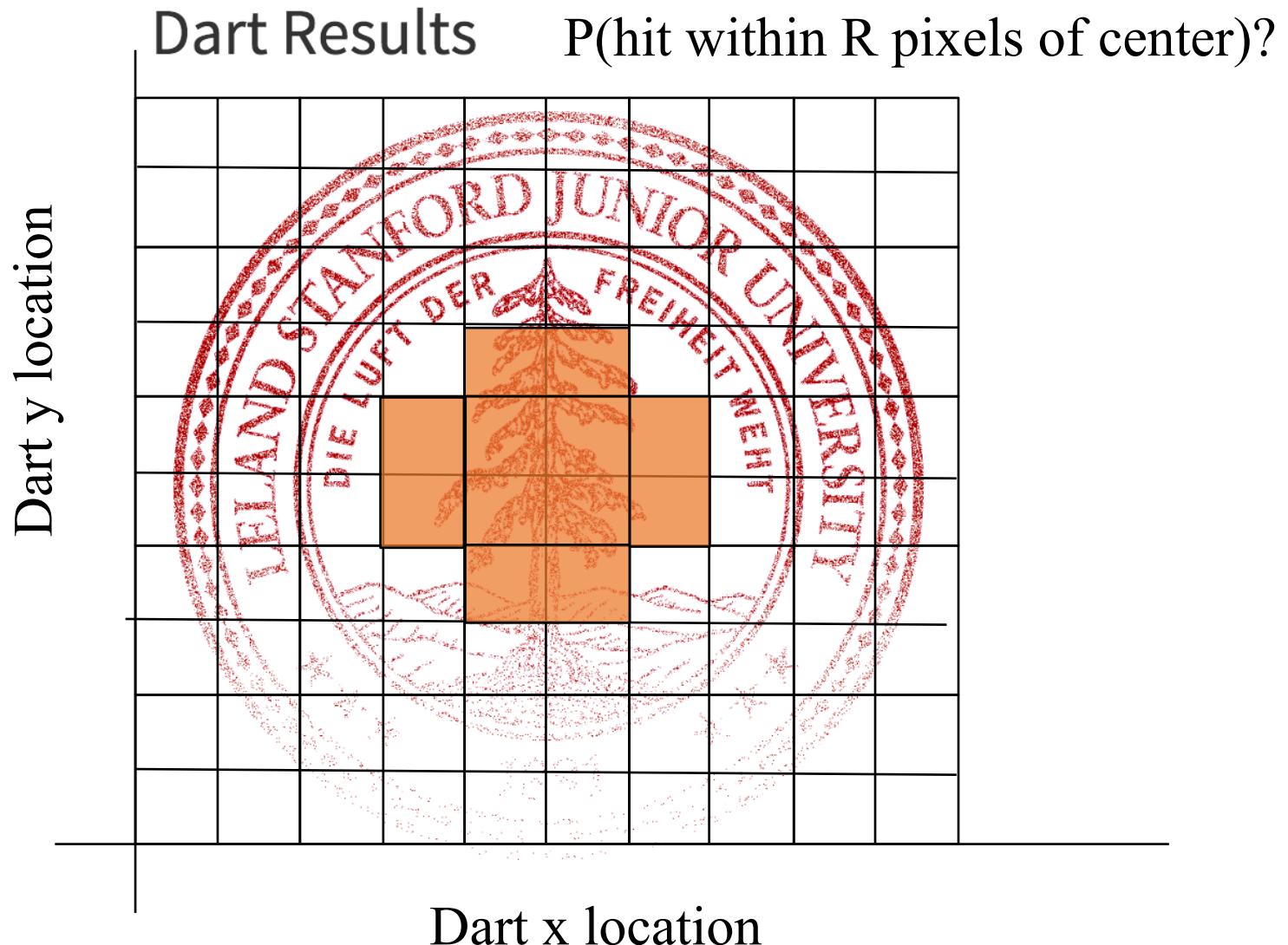
## Marginal Year Probability



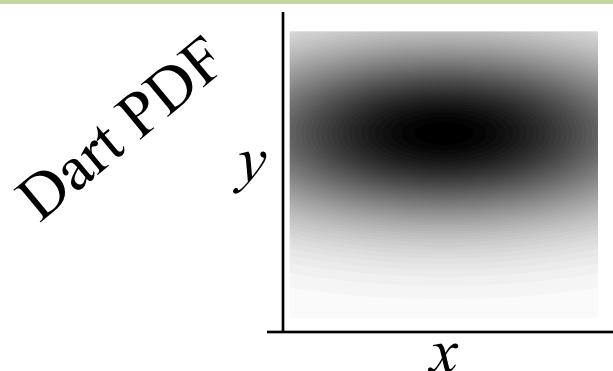
# Joint Dart Distribution



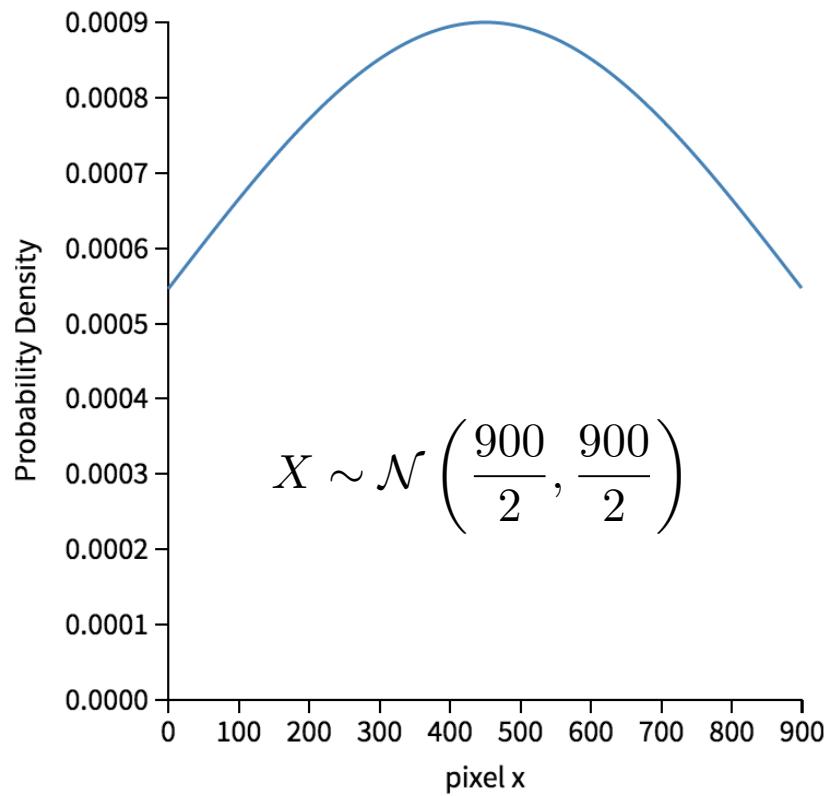
# Joint Dart Distribution



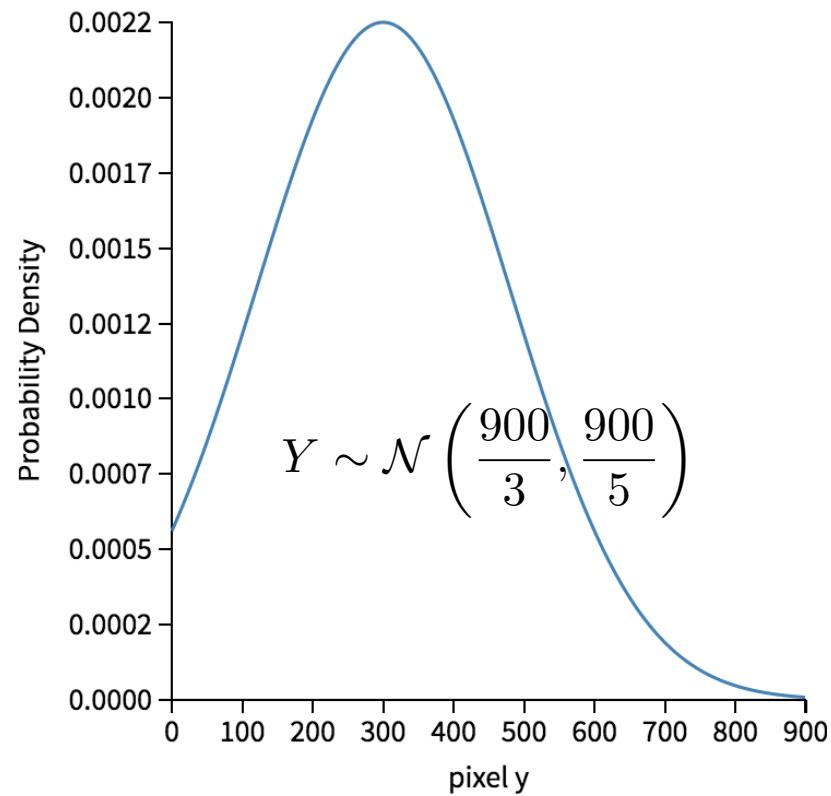
# Darts!



X-Pixel Marginal



Y-Pixel Marginal



# Gaussian Blur



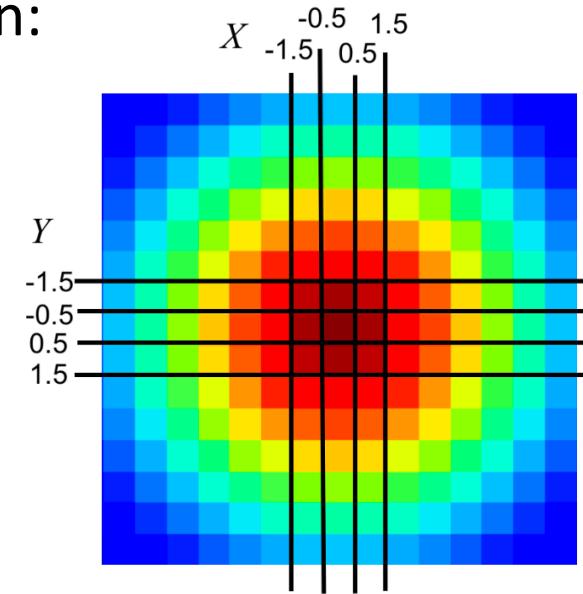
In image processing, a Gaussian blur is the result of blurring an image by a Gaussian function. It is a widely used effect in graphics software, typically with the standard deviation based on a joint probability distribution:

## Joint PDF

$$f_{X,Y}(x, y) = \frac{1}{2\pi \cdot 3^2} e^{-\frac{x^2+y^2}{2 \cdot 3^2}}$$

## Joint CDF

$$F_{X,Y}(x, y) = \Phi\left(\frac{x}{3}\right) \cdot \Phi\left(\frac{y}{3}\right)$$



Used to generate this weight matrix

# Multinomial

Example document:

“Pay for Viagra with a credit-card. Viagra is great.  
So are credit-cards. Risk free Viagra. Click for free.”

$$n = 18$$

$$P \left( \begin{array}{l} \text{Viagra} = 2 \\ \text{Free} = 2 \\ \text{Risk} = 1 \\ \text{Credit-card: } 2 \\ \dots \\ \text{For} = 2 \end{array} \mid \text{spam} \right) = \frac{n!}{2!2!\dots2!} p_{\text{viagra}}^2 p_{\text{free}}^2 \dots p_{\text{for}}^2$$

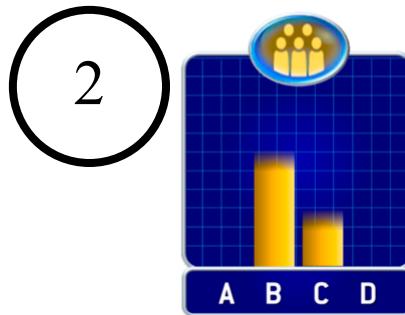
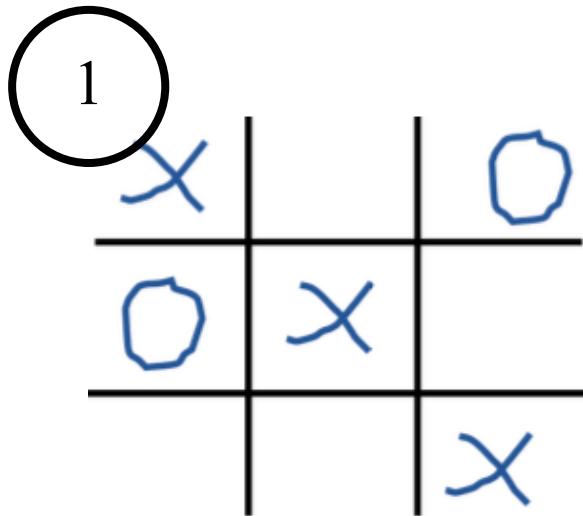
Probability of seeing this document | spam

It's a Multinomial!

The probability of a word in spam email being viagra

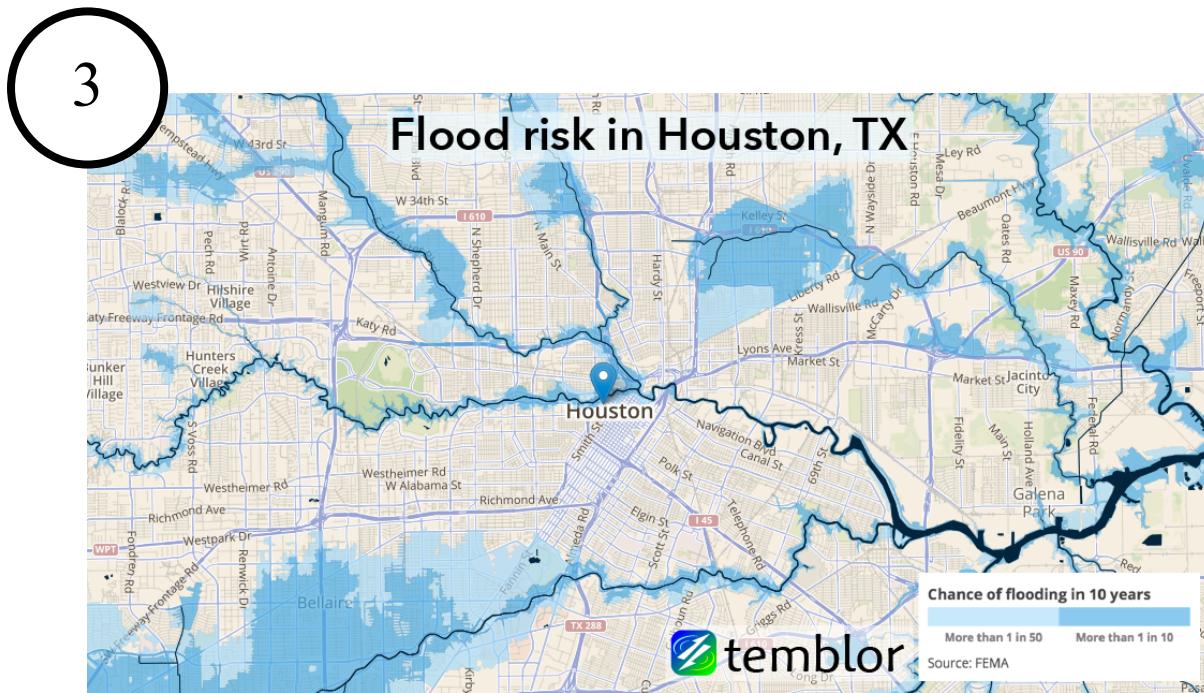


# Midterm (part 1)



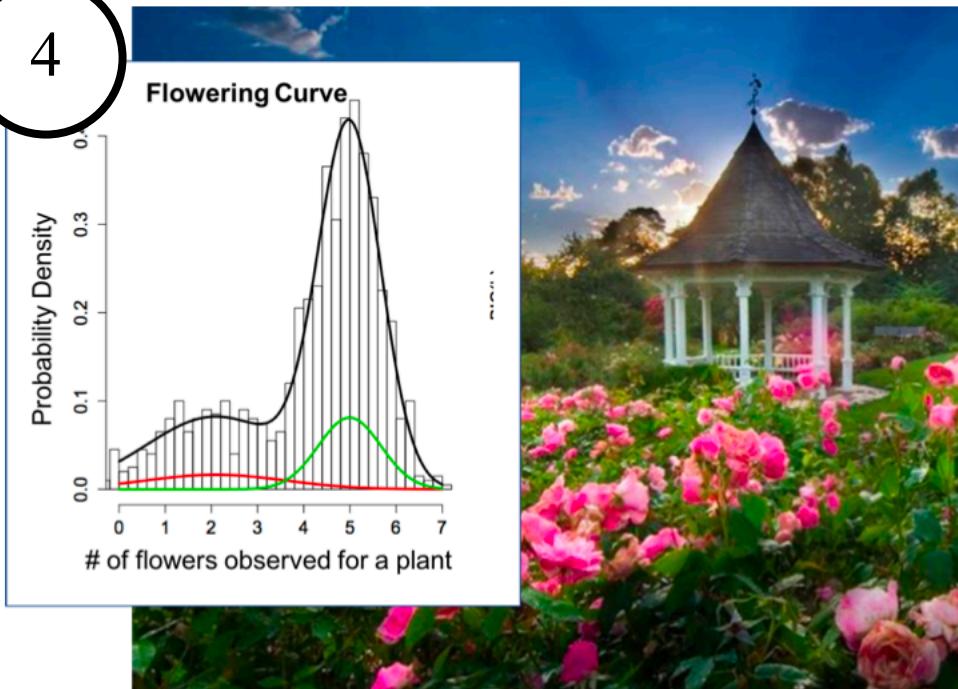
In 1999, what animal was taken off the U.S. Endangered species list after 29 years?

- A: Peregrine Falcon  
B: Humpback Whale  
C: D:

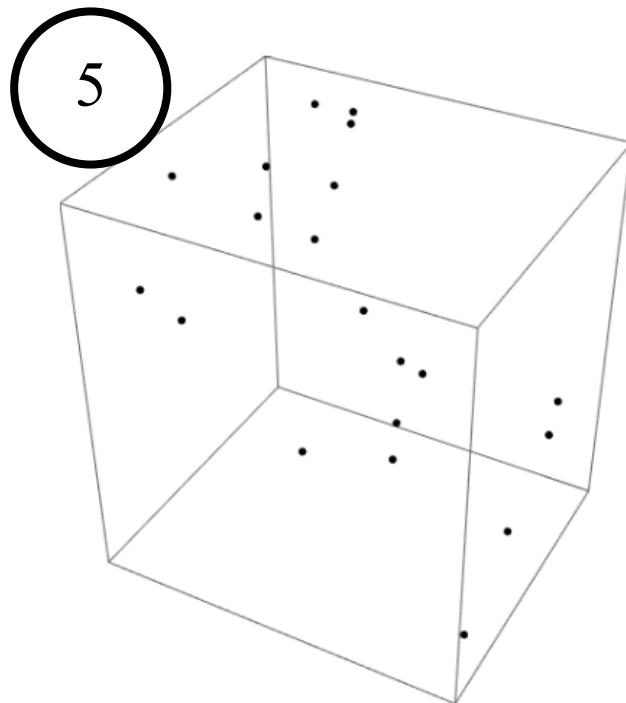


# Midterm (part 2)

4



5



6

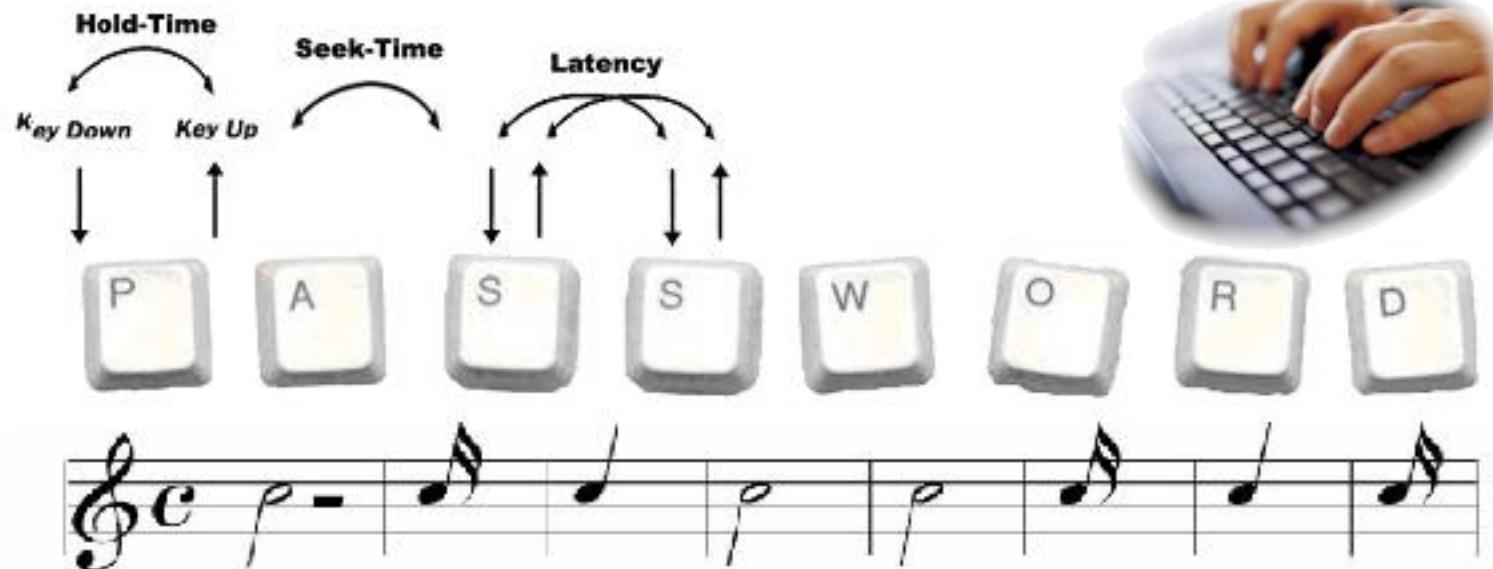
Generate  $N$  values  $(X_1, X_2, \dots, X_N)$  uniformly sampled over a range  $(a, b)$ . We can approximate the integral of a function  $h$  over  $(a, b)$  as:

$$\int_a^b h(x) dx \approx \frac{(b-a)}{N} \sum_{i=1}^N h(X_i)$$

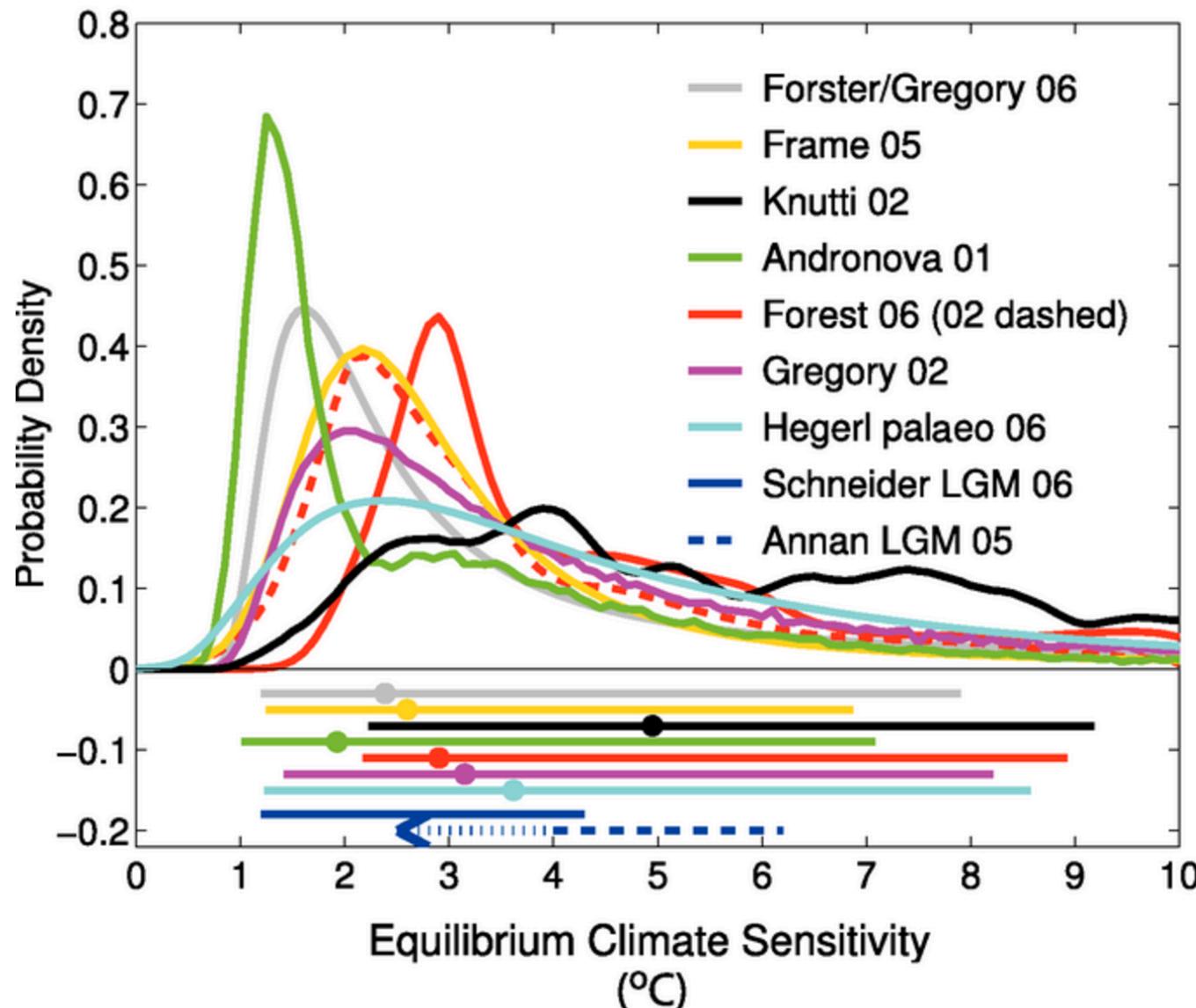
**Monte Carlo Integration**



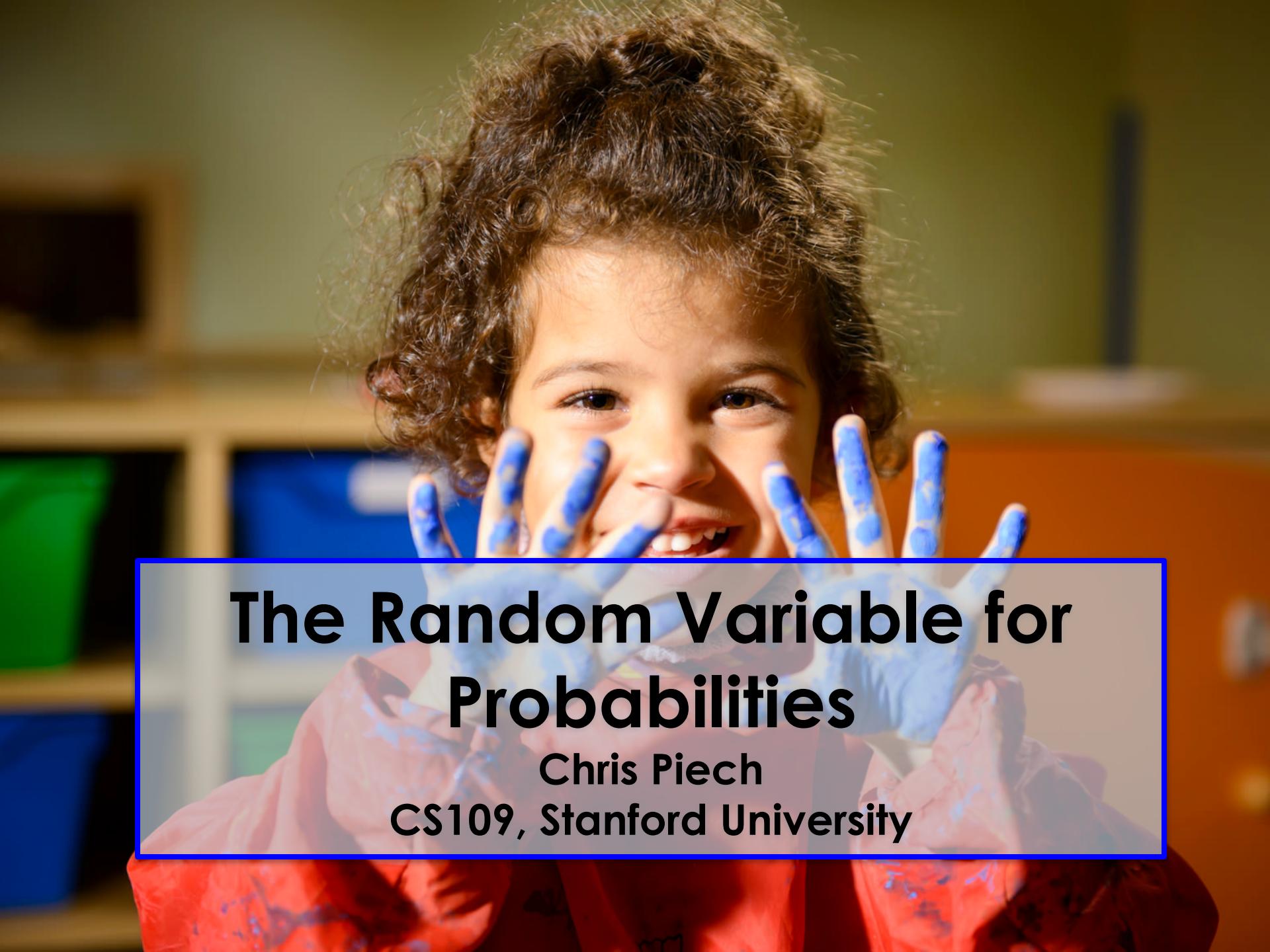
# Biometric Keystroke



# Climate Sensitivity







# The Random Variable for Probabilities

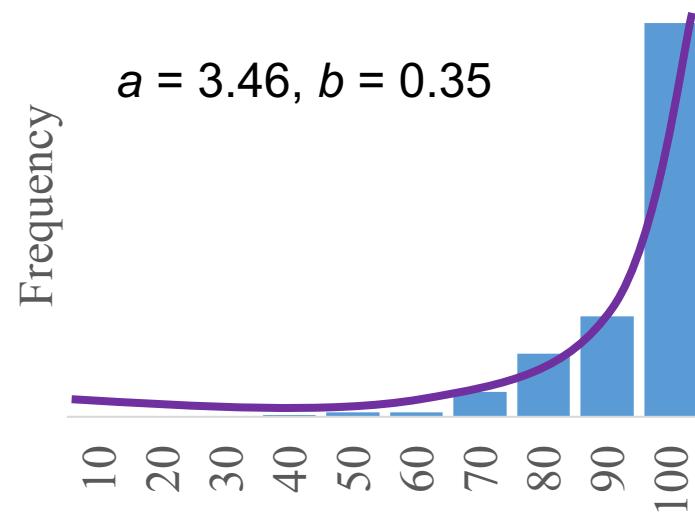
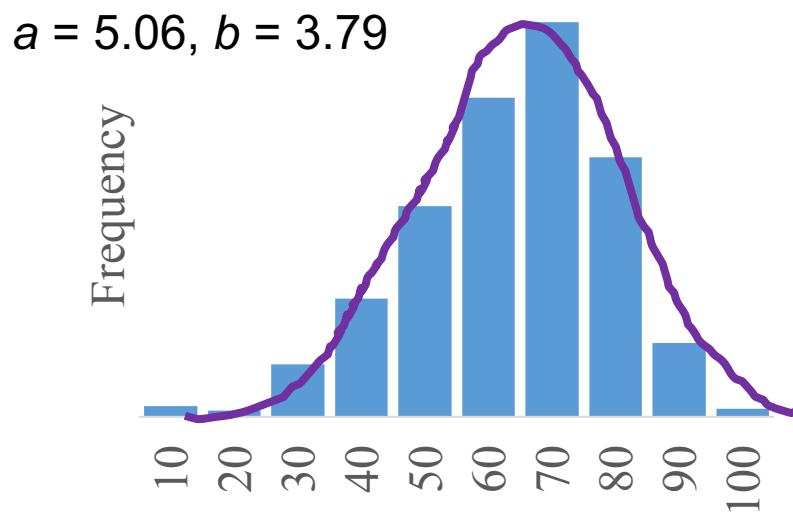
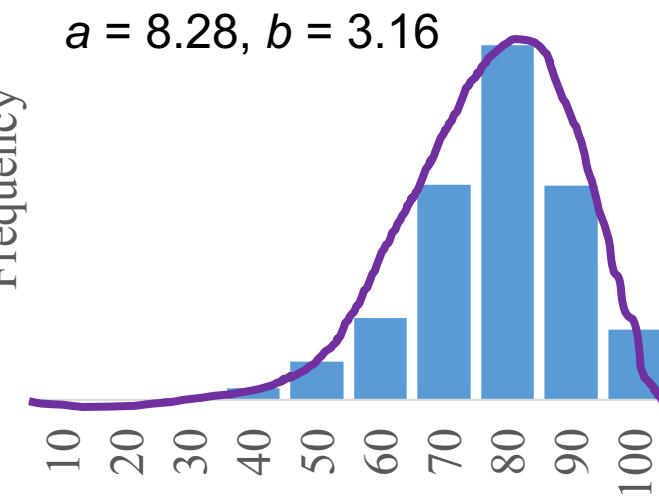
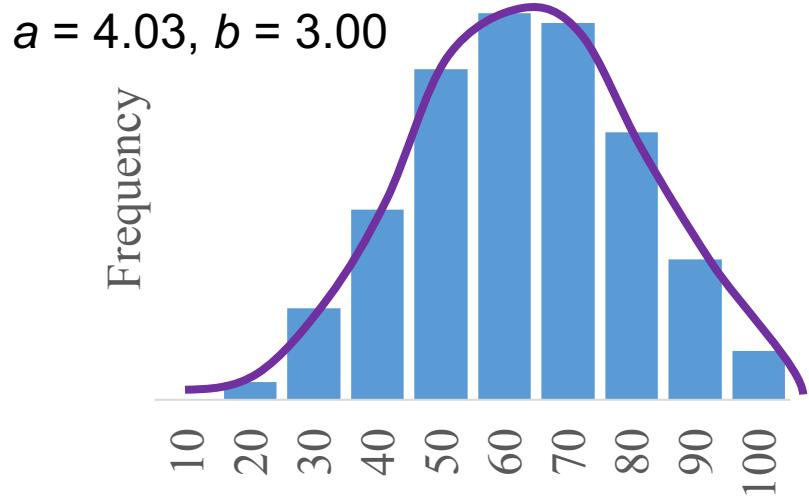
Chris Piech  
CS109, Stanford University

# Flip a Coin With Unknown Probability



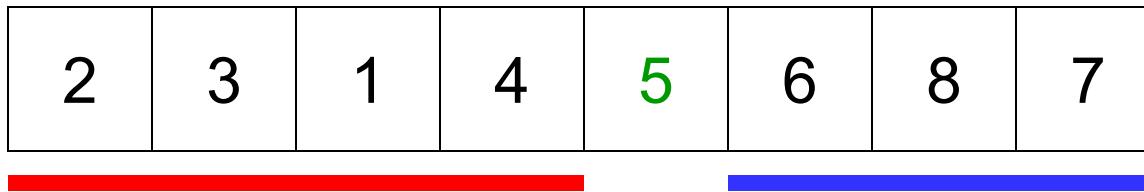
Demo

# Assignment Grades



We have 2055 assignment distributions from grade scope

# Recursive Insight



Partition array so:

- everything smaller than pivot is on left
- everything greater than or equal to pivot is on right
- pivot is in-between

# Machine Learning Example

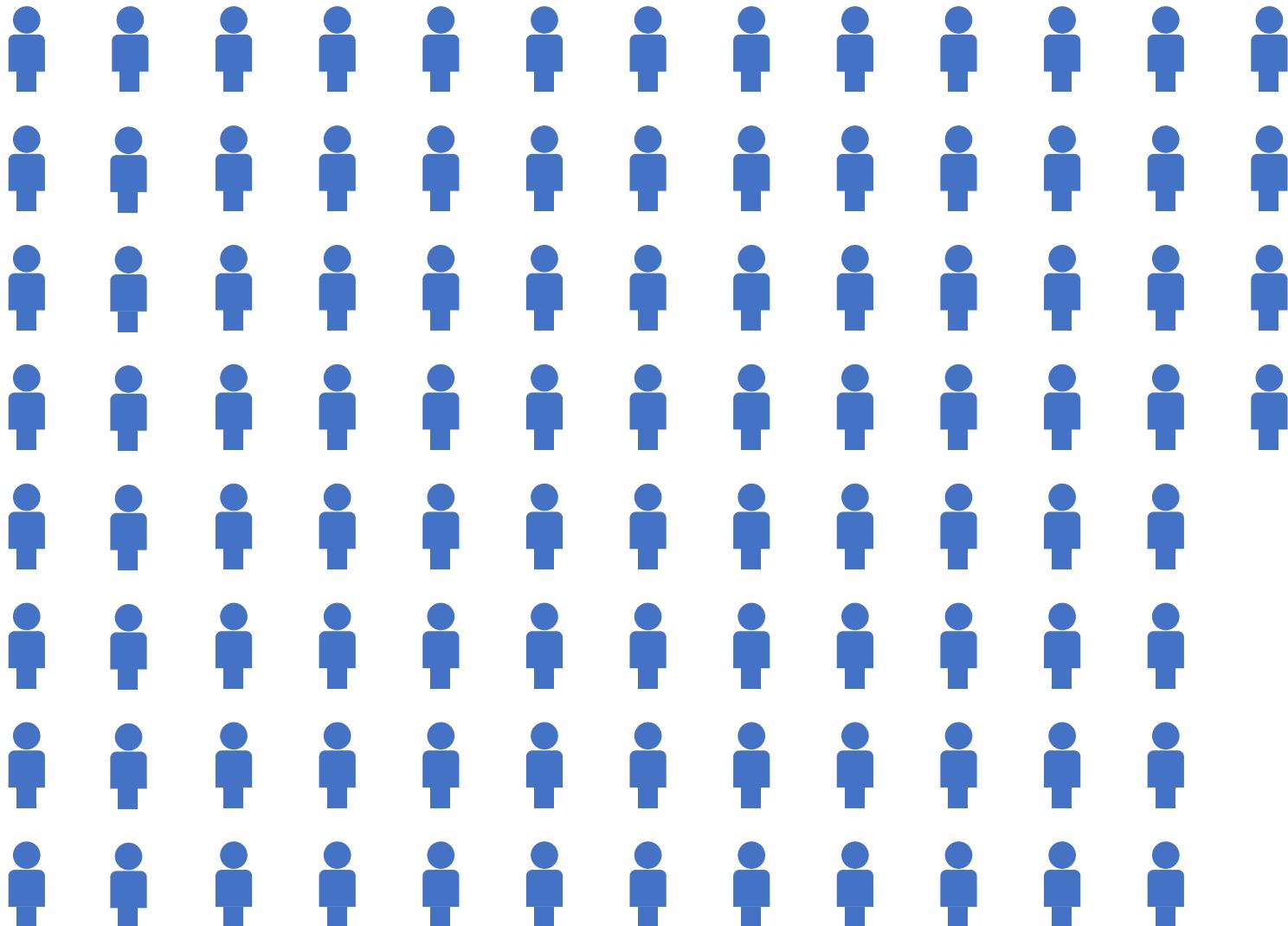
- You want to know the true mean and variance of happiness in Buthan
  - But you can't ask everyone.
  - Randomly sample 200 people.
  - Your data looks like this:



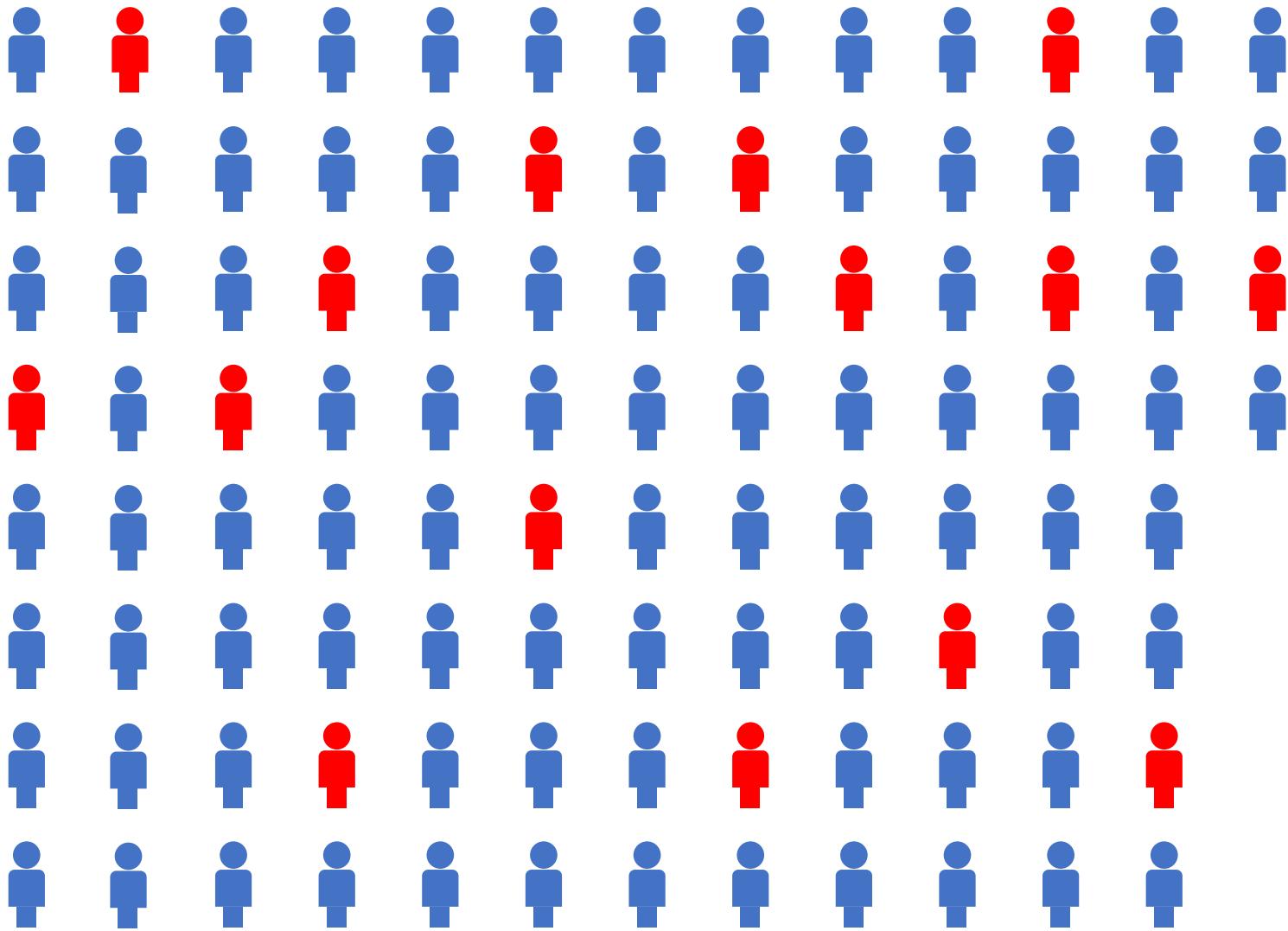
$$\text{Happiness} = \{72, 85, 79, 91, 68, \dots, 71\}$$

- The mean of all of those numbers is 83. Is that the true average happiness of Bhutanese people?

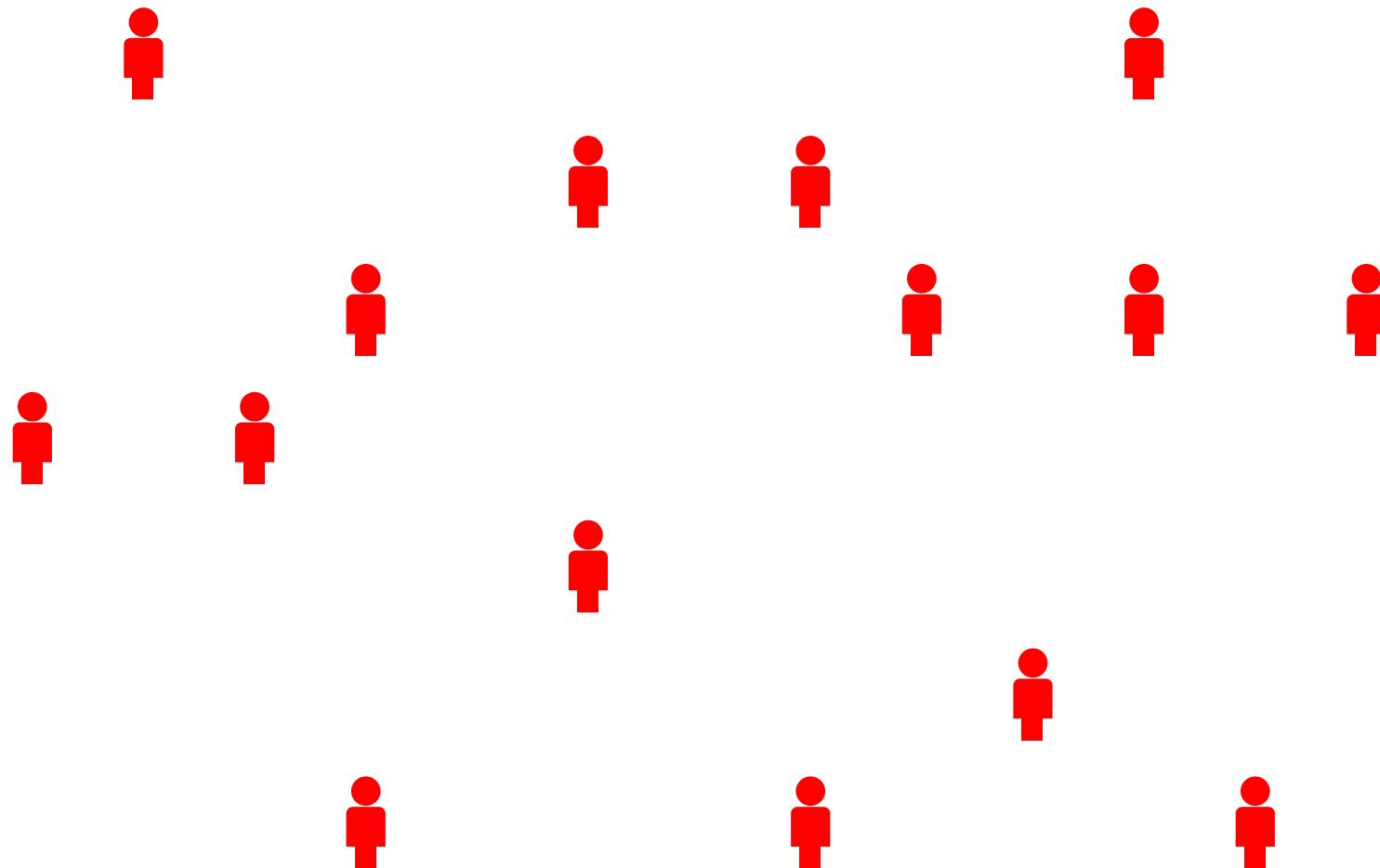
# Population



# Sample



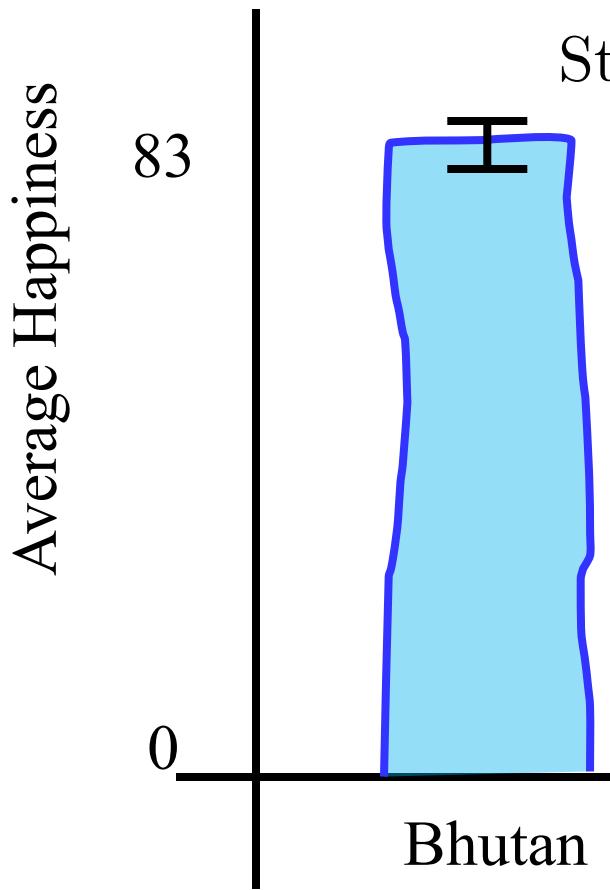
# Sample



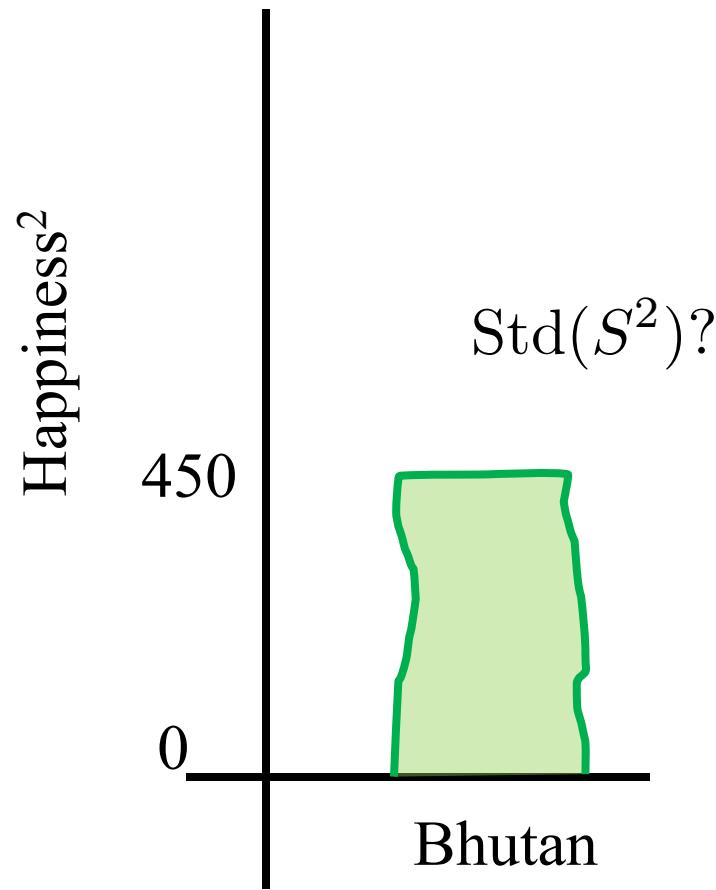
Collect one (or more) numbers from each person

# Sample Mean

Average Happiness



Variance of Happiness



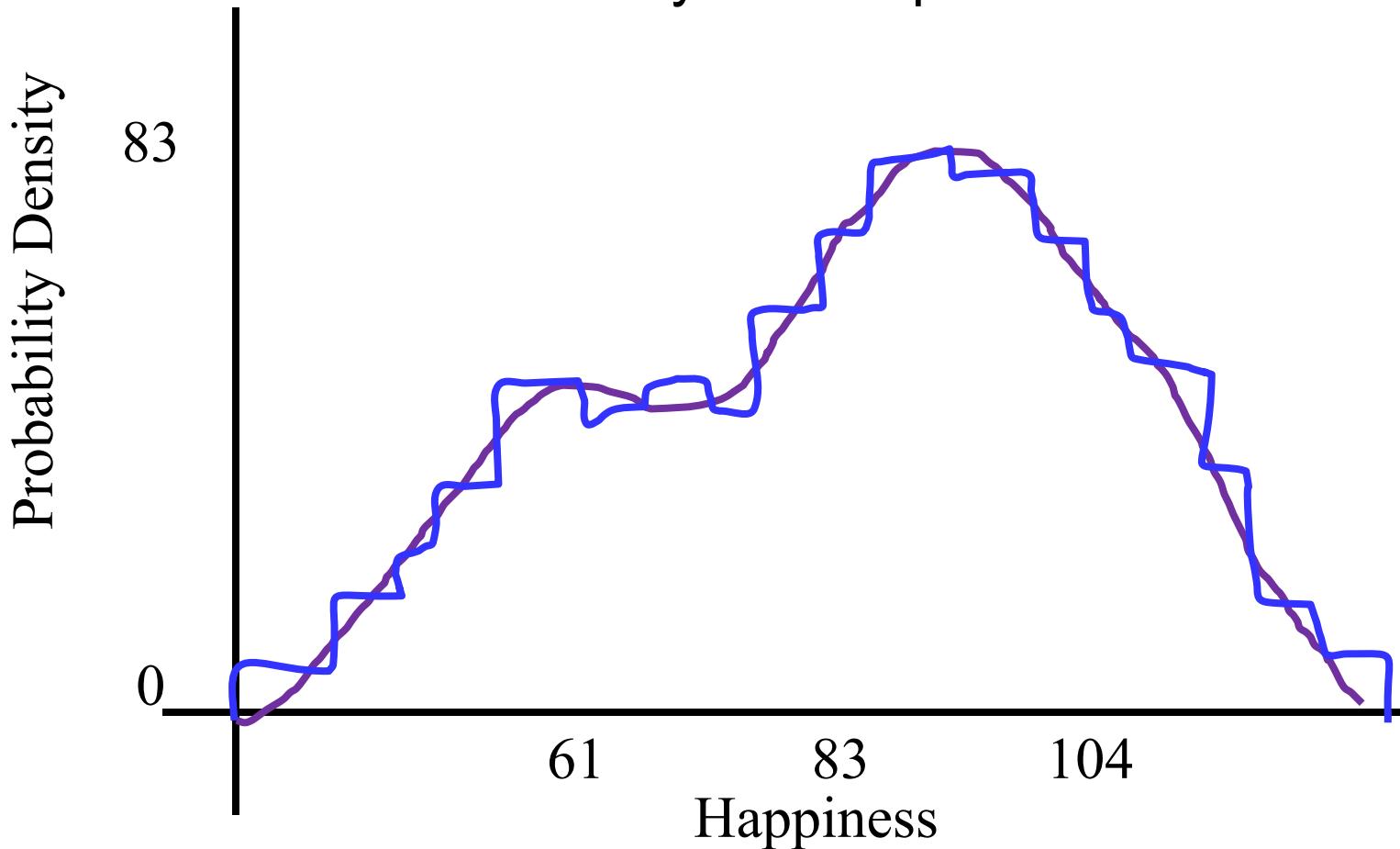
Claim: The average happiness of Bhutan is  $83 \pm 2$



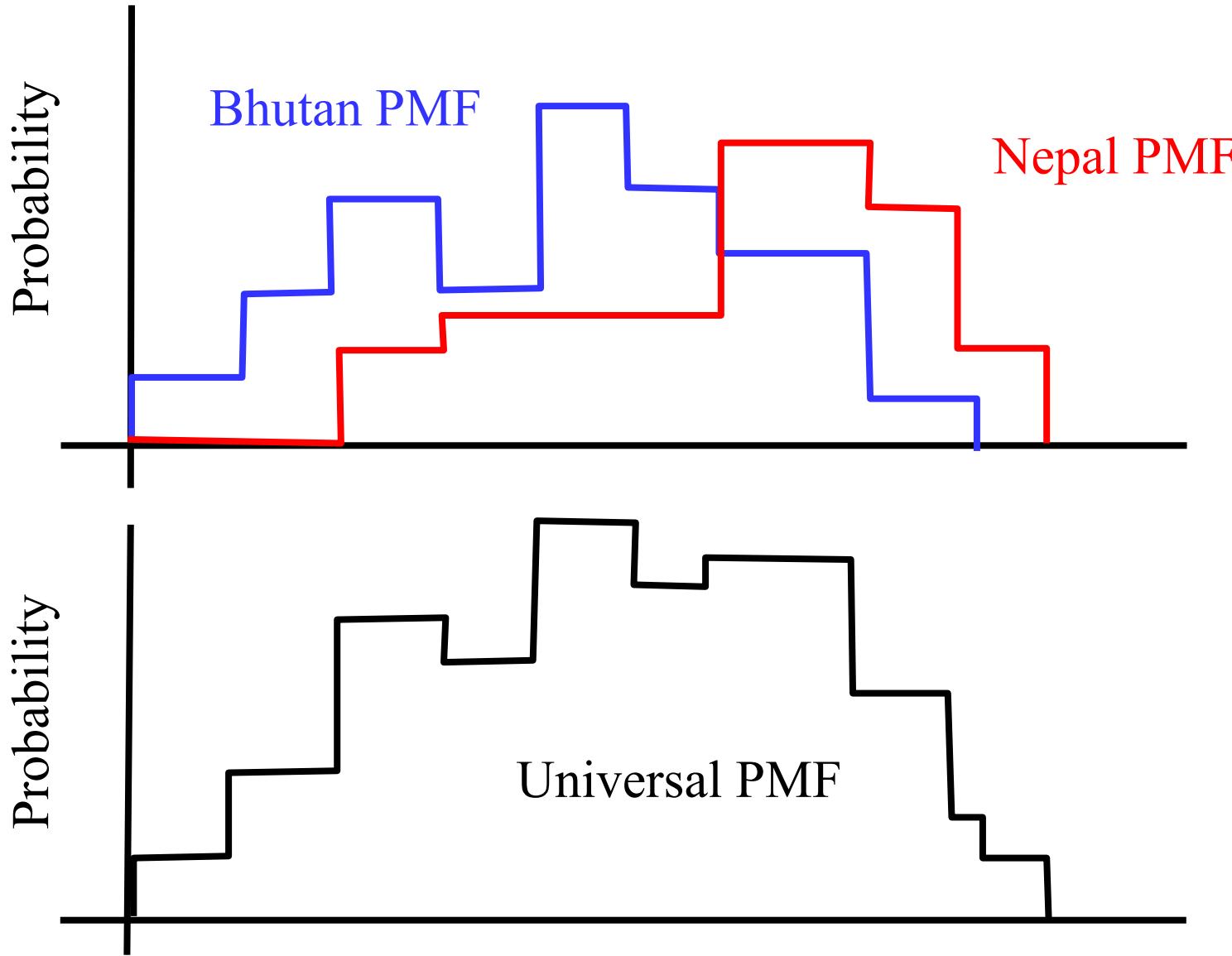
Bootstrap

# Key Insight

You can estimate the PMF of the underlying distribution, from your sample.



# Universal Sample



# Inequalities

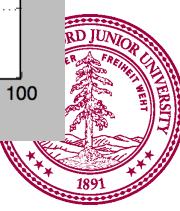
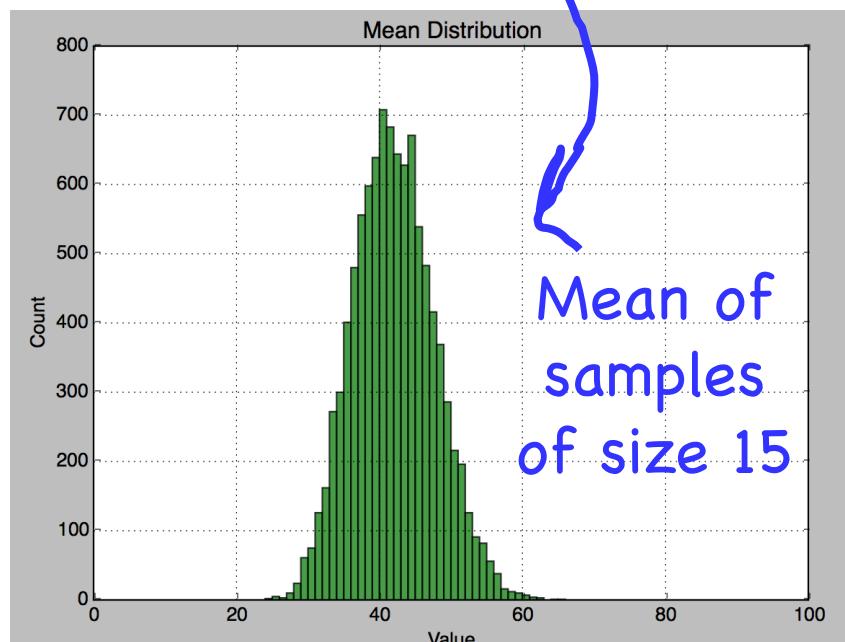
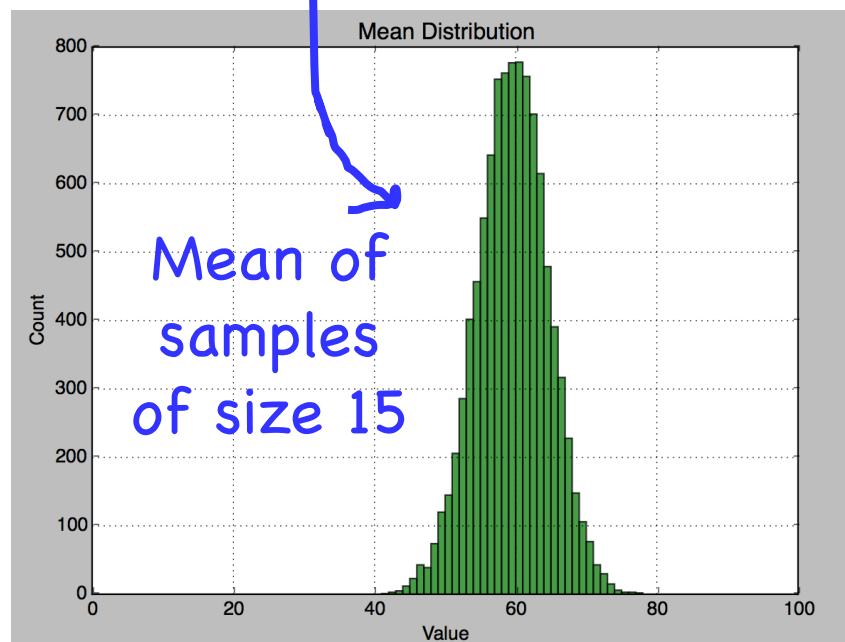
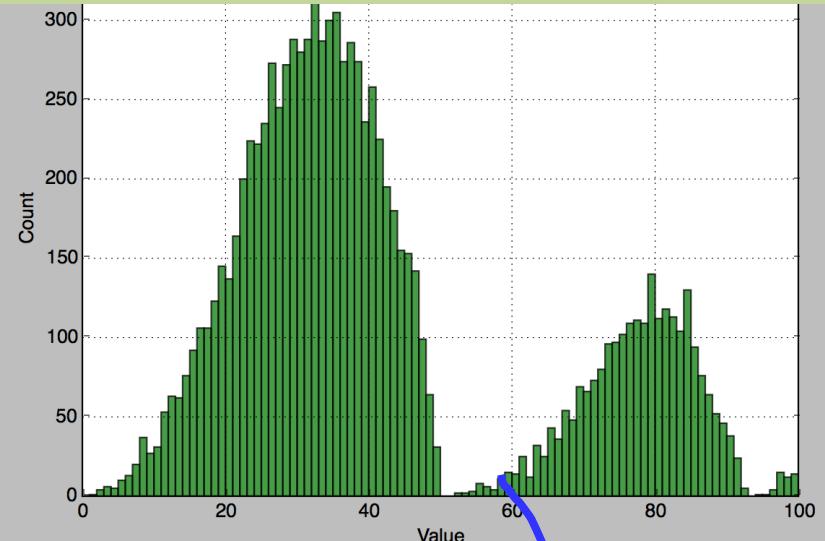
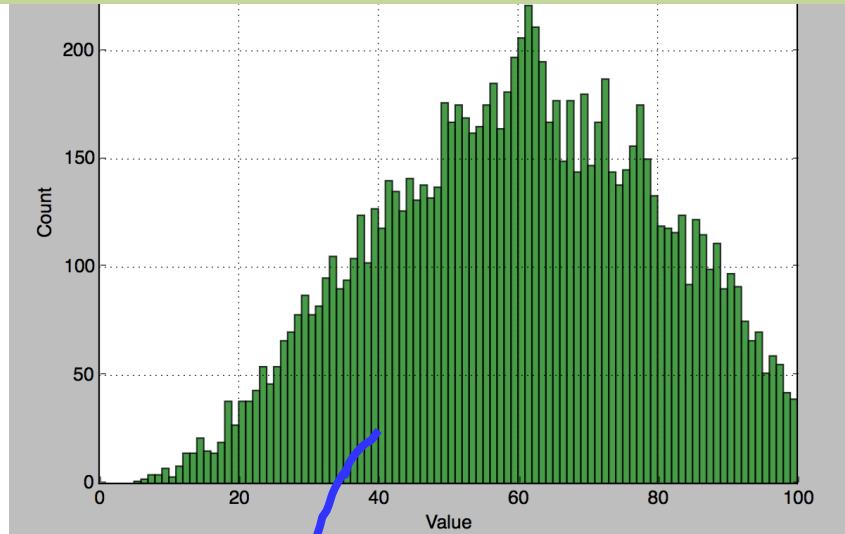


# Mystery: Why is Binomial Normal?

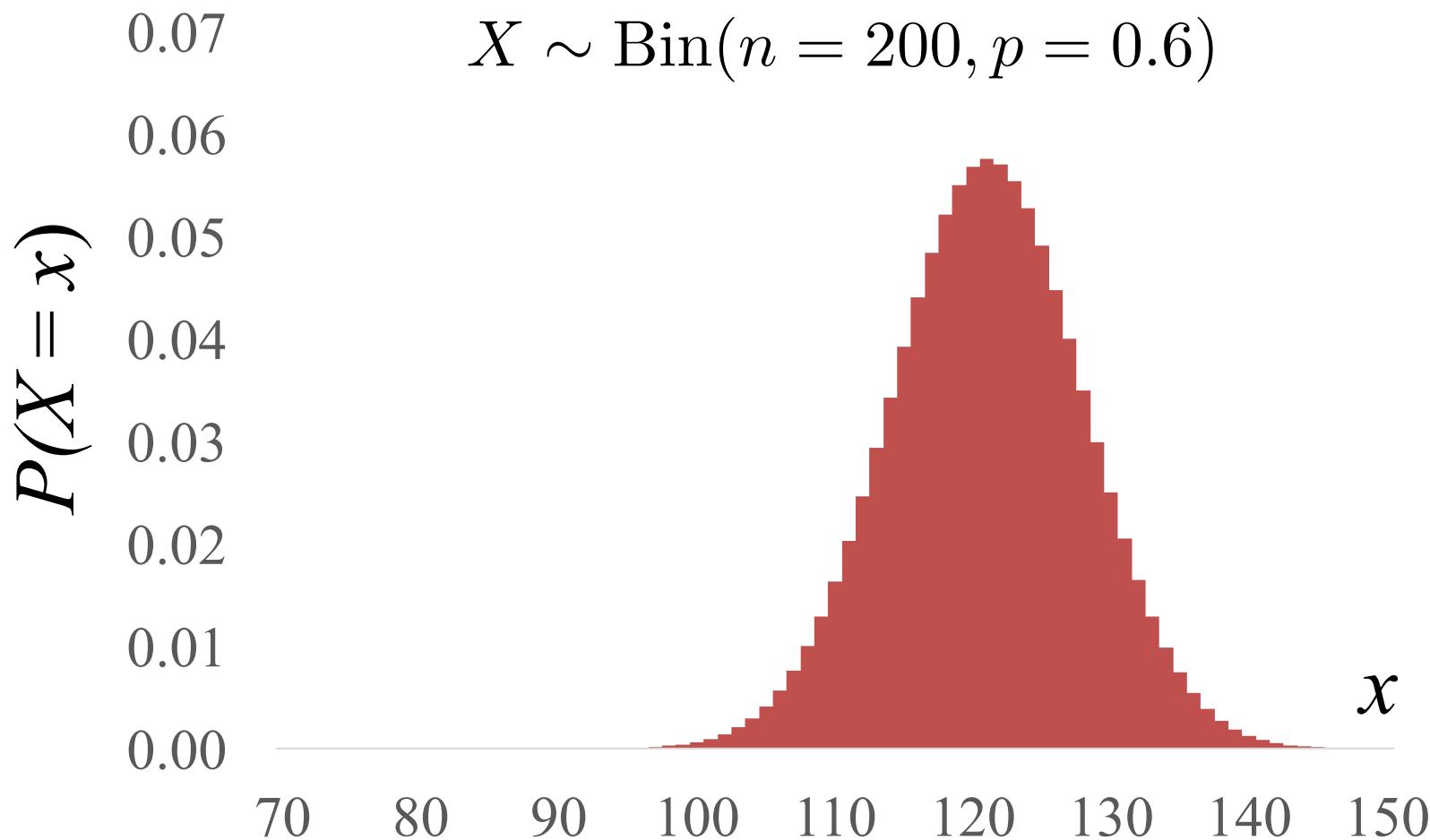
Mystery: Why is the sum of IID uniforms  
normal?

Mystery: Why is the mean of  
IID vars normal?

# C.L.T. Explains This

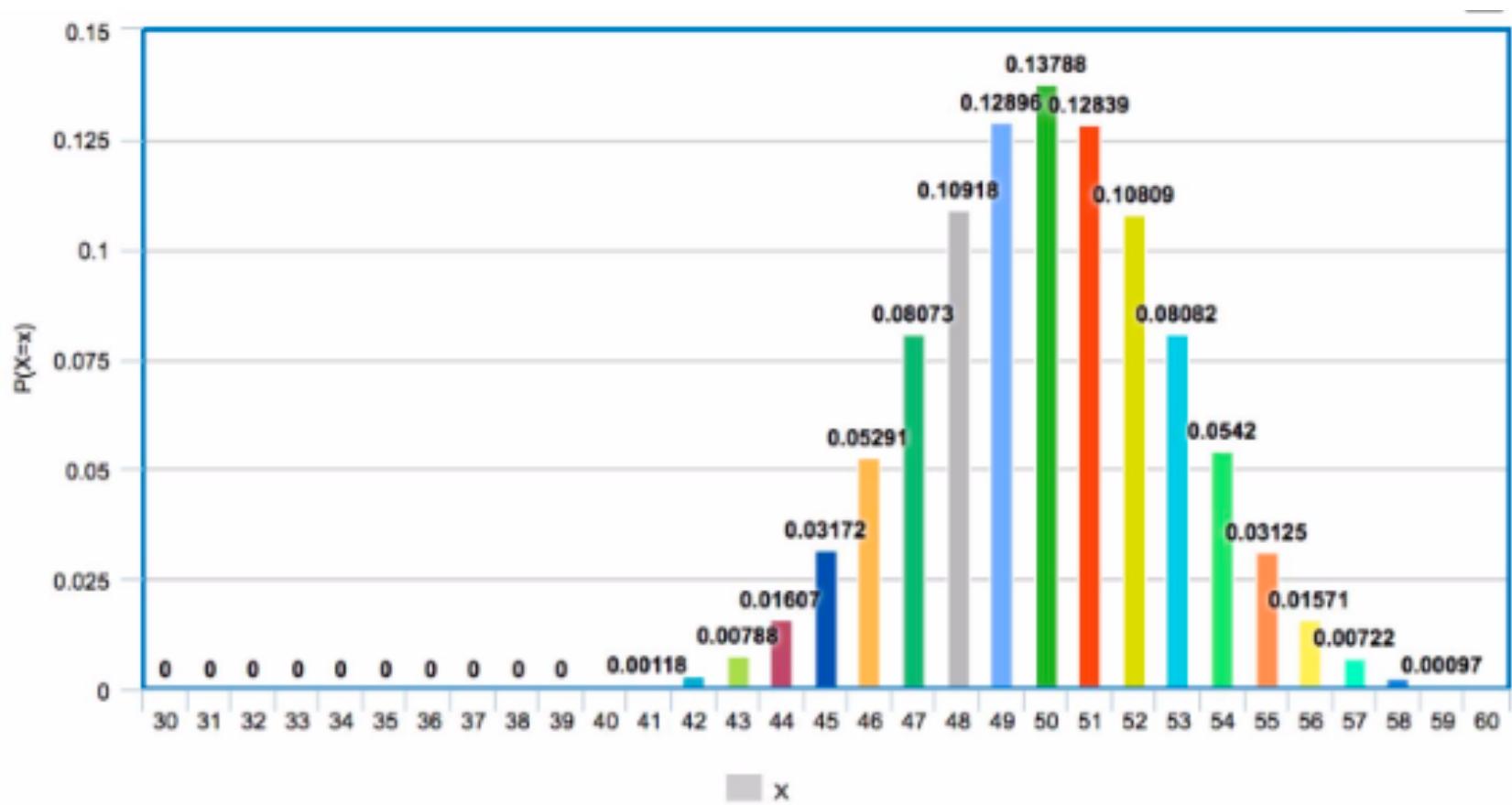


# C.L.T. Explains This

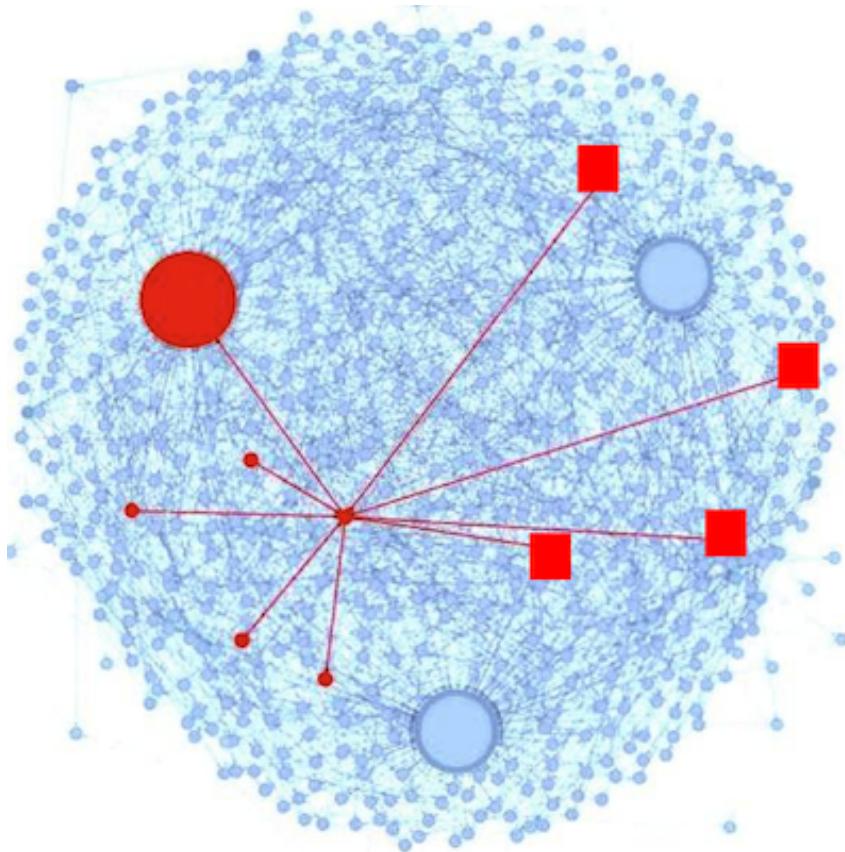


# C.L.T. Explains This

Problem set 5: What is the sum of IID uniforms?



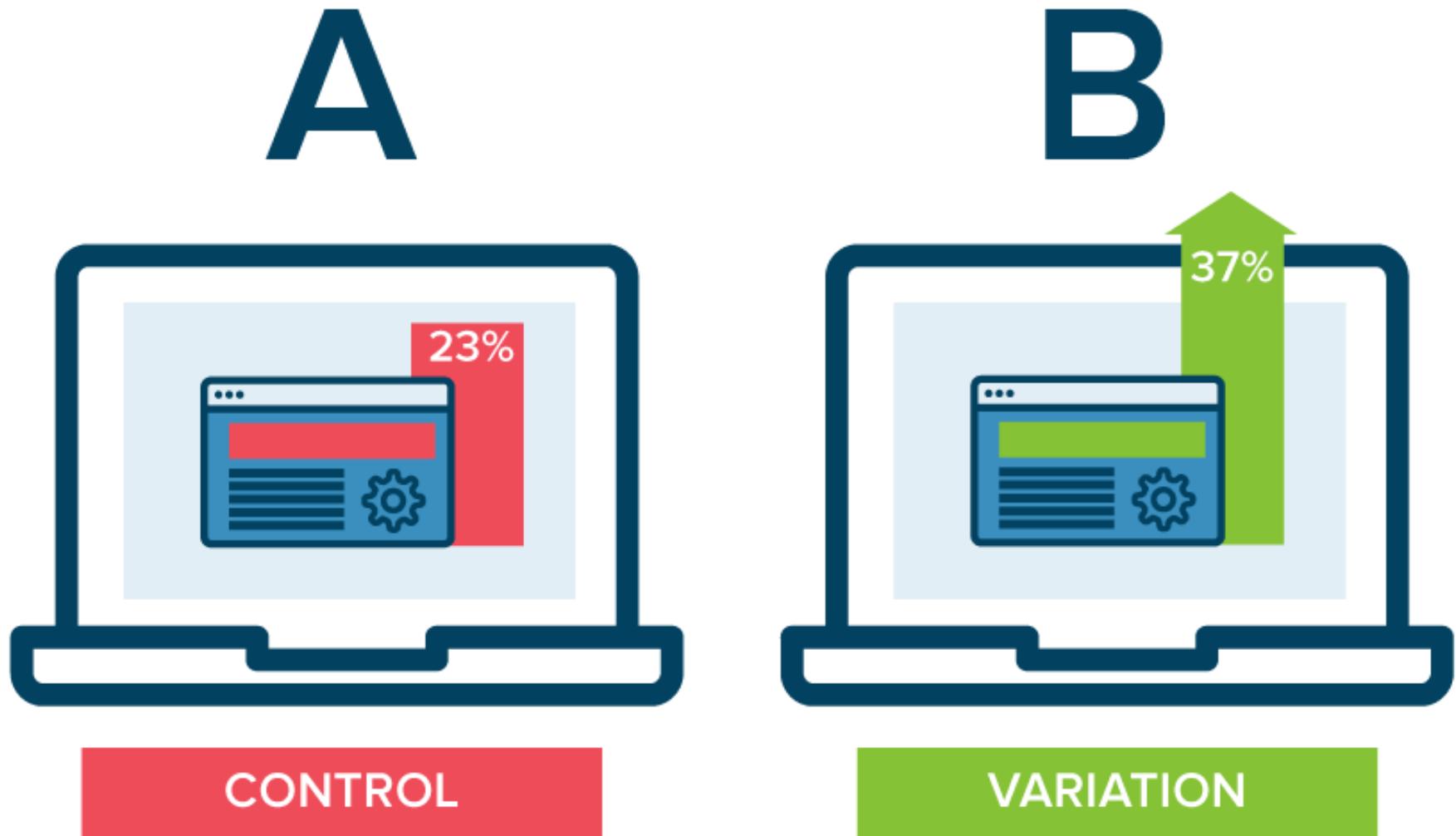
# Peer Grading



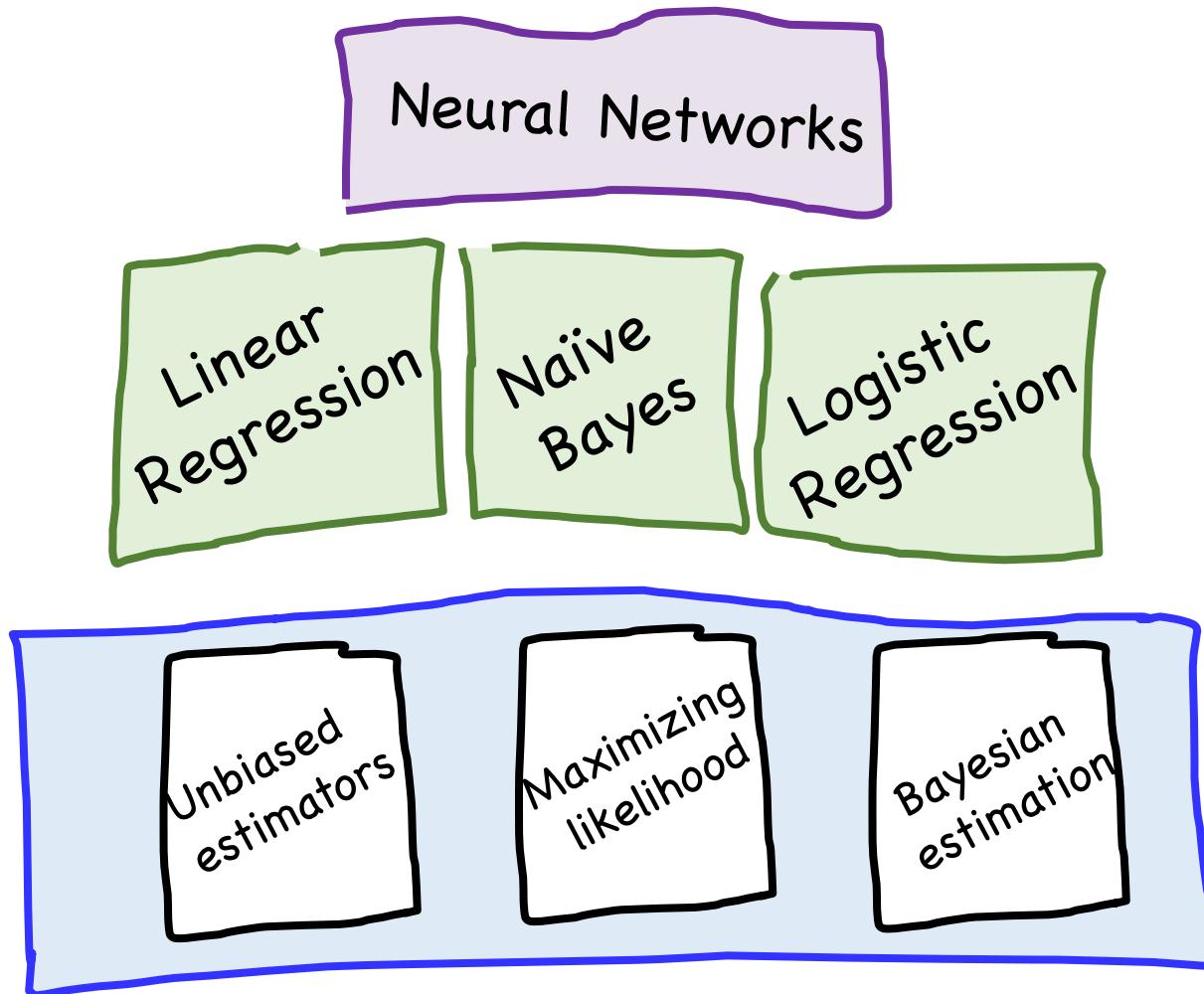
Peer Grading on Coursera  
HCI.

31,067 peer grades for  
3,607 students.

# A/B Testing

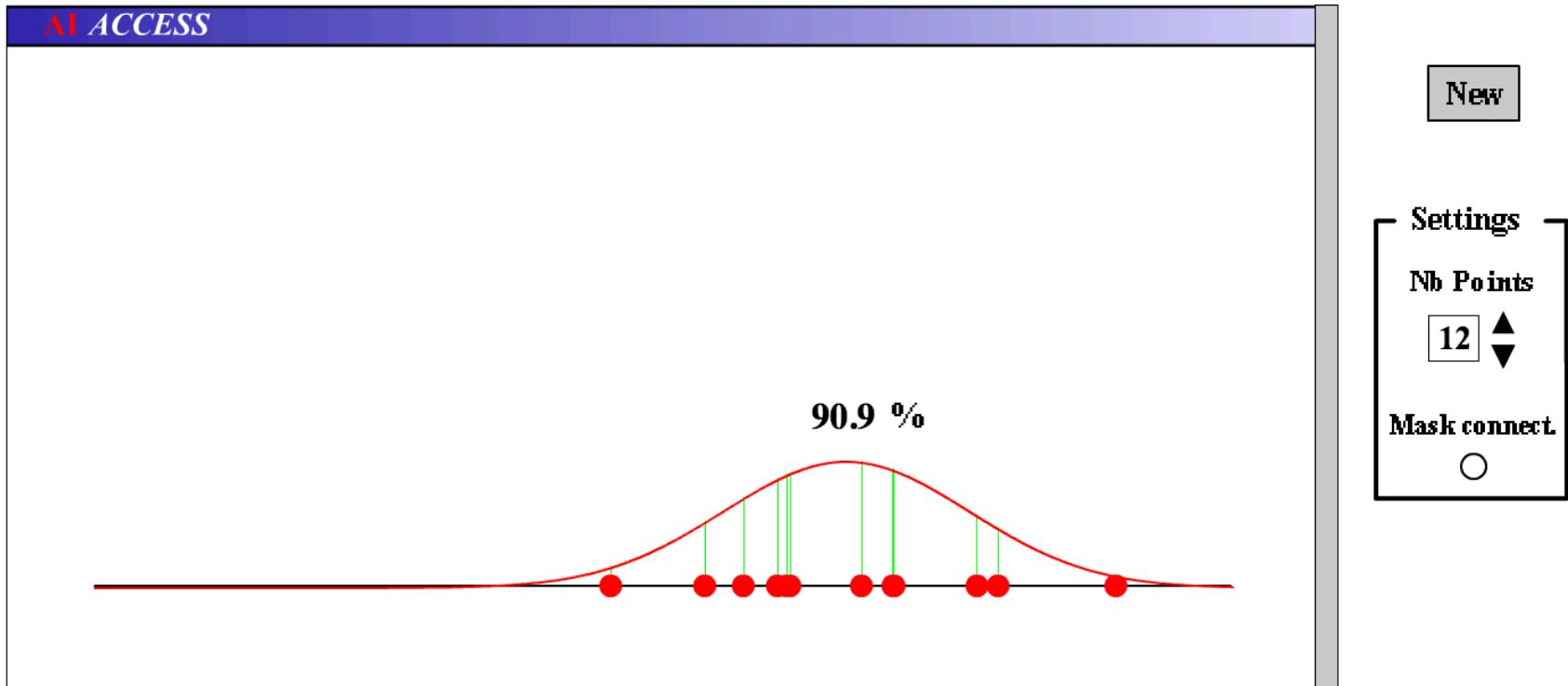


# Towards Machine Learning

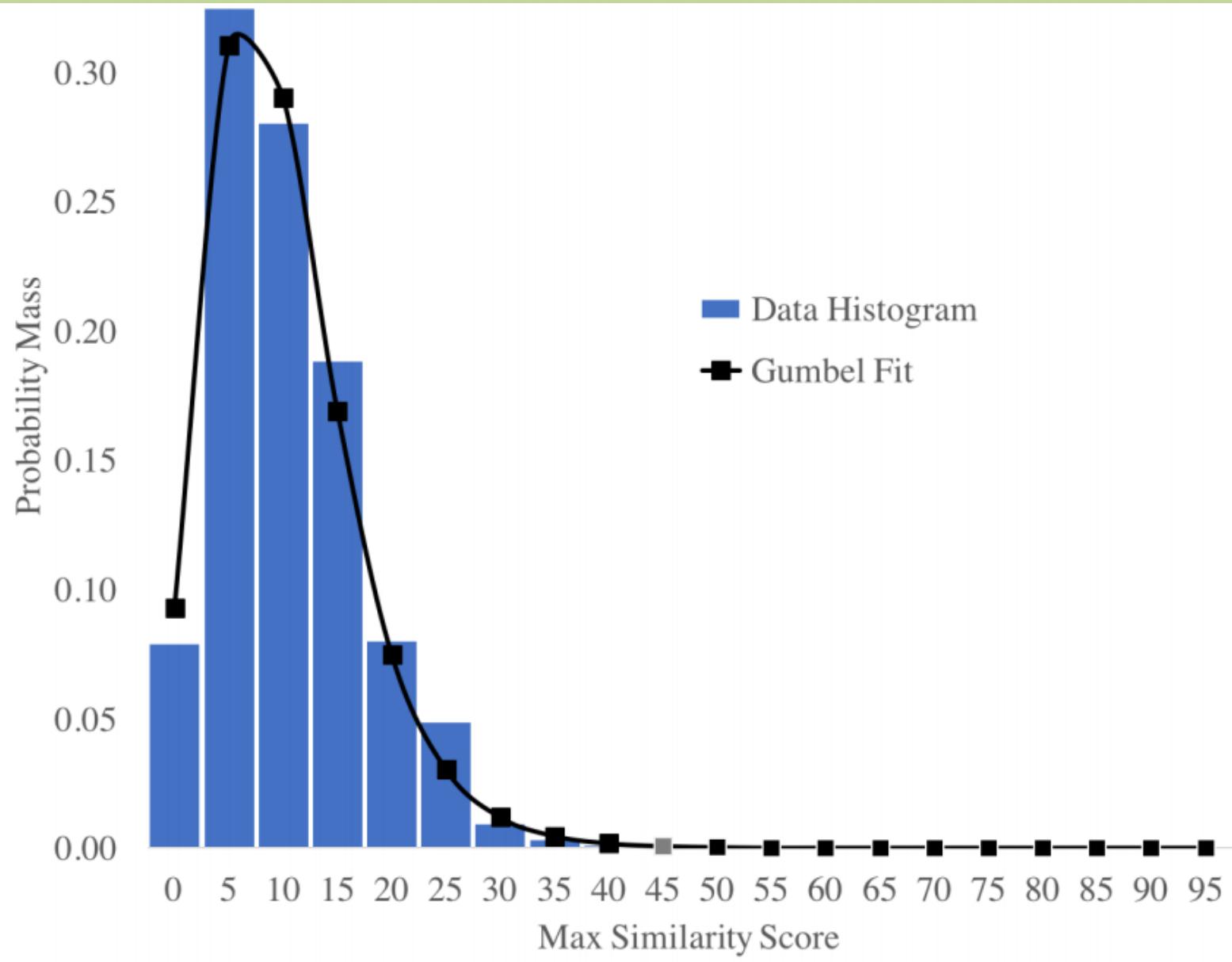


# MLE: Likelihood of Data

## Likelihood of Data from a Normal



# Gumbel Fit



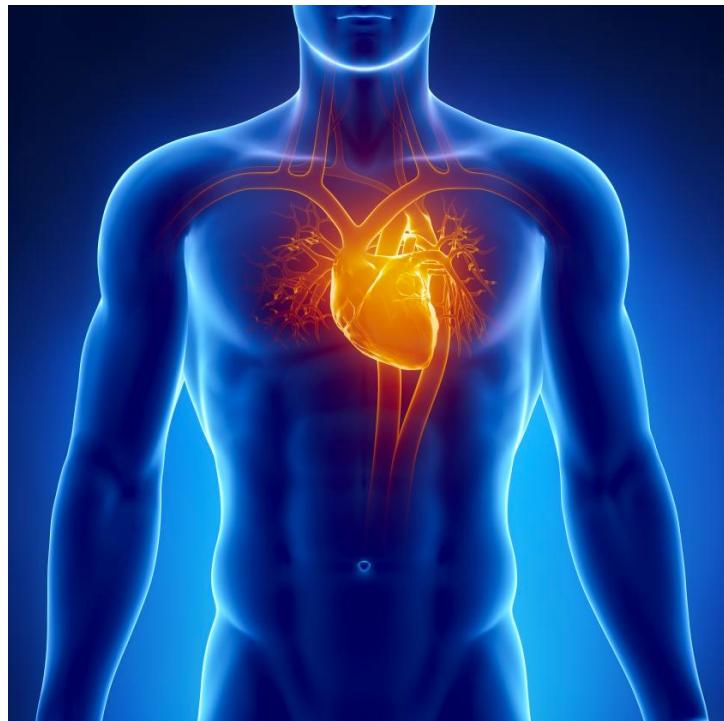
# MAP: Most Probable Parameter

So good to see  
you again!



# Machine Learning

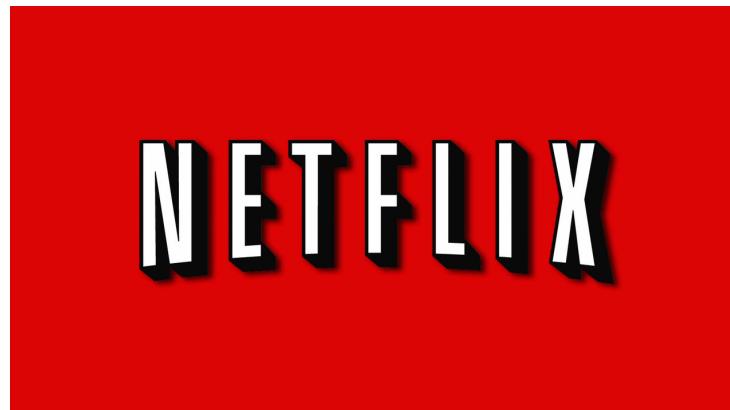
Heart



Ancestry



Netflix



# Naïve Bayes

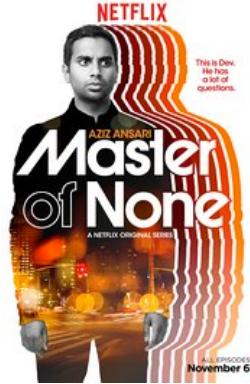
$x_1$



$x_2$



$x_3$



$y$



User 1

1

0

1

1

User 2

1

0

1

0

⋮

User  $n$

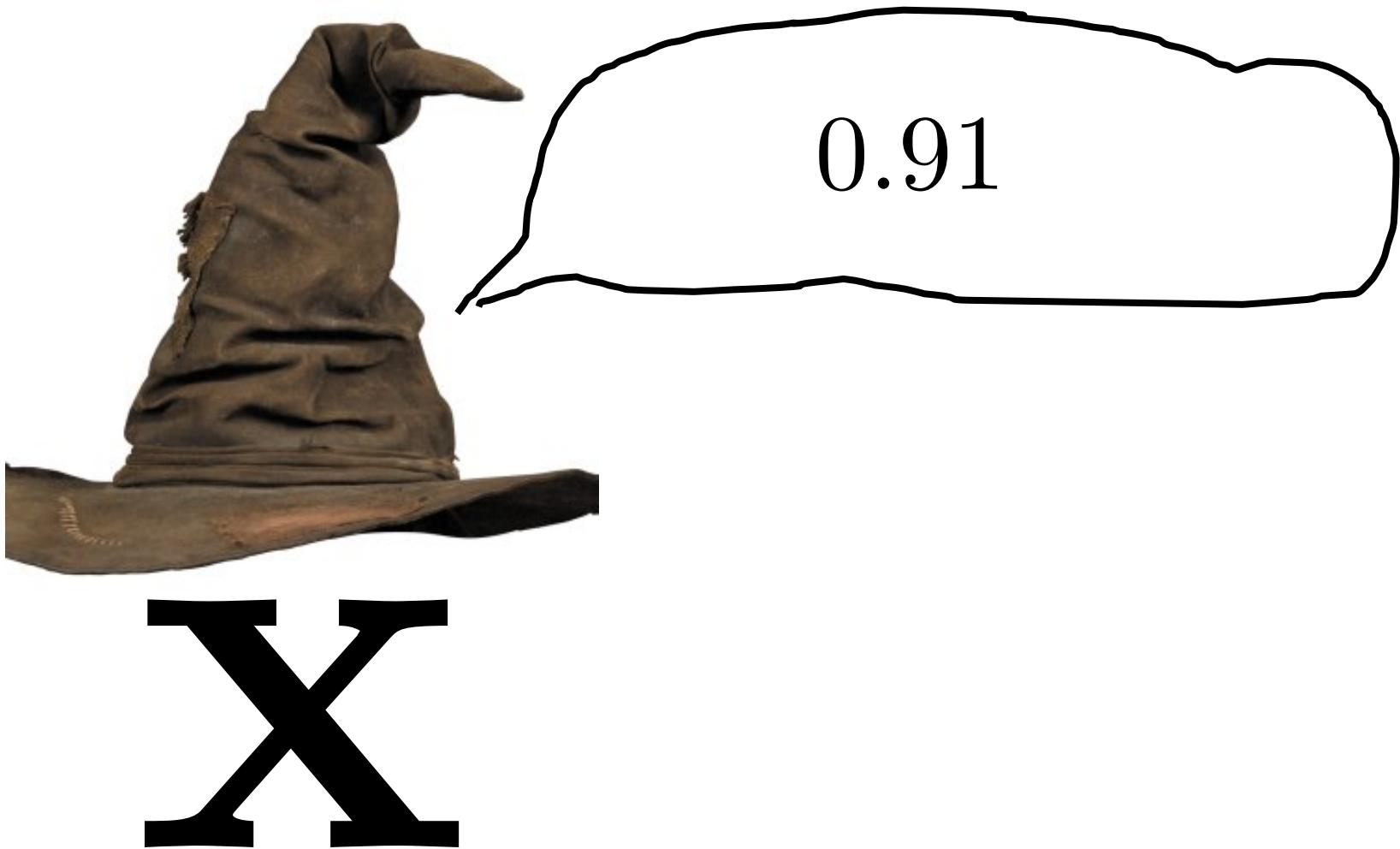
0

1

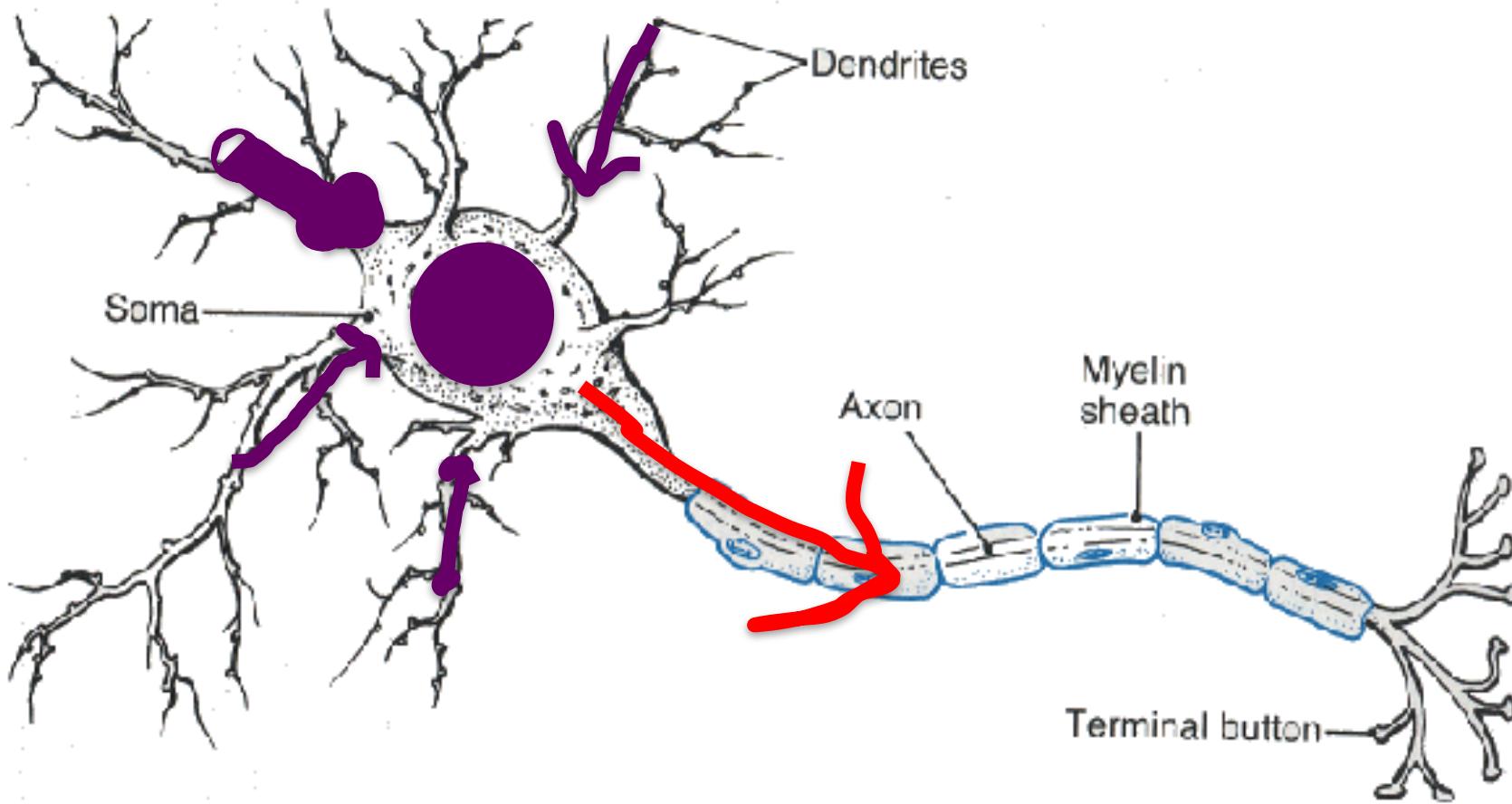
1

1

# Logistic Regression

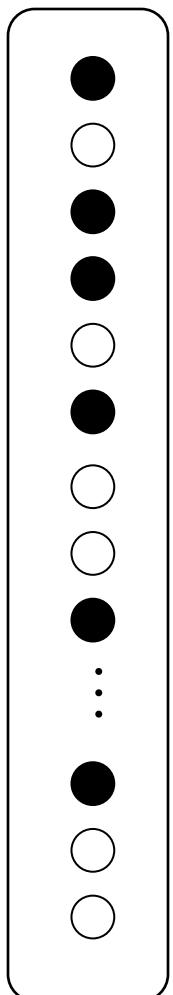


# Logistic Regression

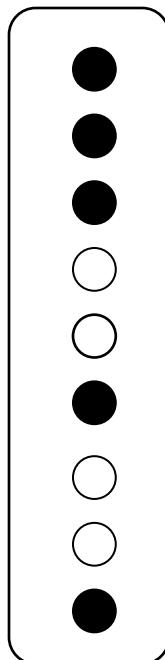


# Deep Learning

Layer  $\mathbf{x}$

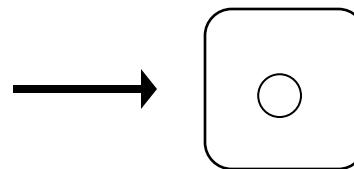


Layer  $\mathbf{h}$



Layer  $\hat{\mathbf{y}}$

$$\begin{aligned} LL(\theta) = & y \log \hat{y} \\ & + (1 - y) \log [1 - \hat{y}] \end{aligned}$$



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

$$\mathbf{h}_j = \sigma \left( \sum_{i=0}^{m_x} \mathbf{x}_i \theta_{i,j}^{(h)} \right)$$

# Deep Dream



By the numbers

# ~600 Fruit



# ~ 30 Major Keys



Naïve Bayes Assumption:

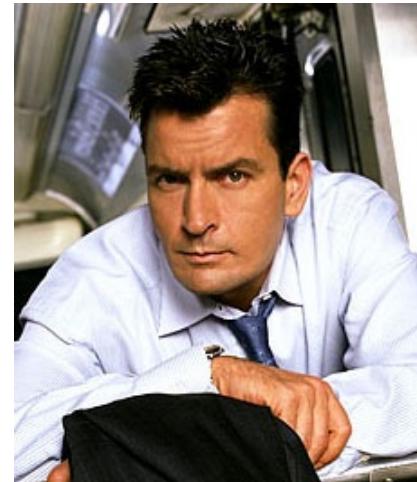
$$P(\mathbf{x}|y) = \prod_i P(x_i|y)$$

# 1 Contest



# Thomas Bayes

- Rev. Thomas Bayes (1702 –1761) was a British mathematician and Presbyterian minister



- He looked remarkably similar to Charlie Sheen
  - But that's not important right now...

# Jacob Bernoulli

- Jacob Bernoulli (1654-1705), also known as “James”, was a Swiss mathematician



- One of many mathematicians in Bernoulli family
- The Bernoulli Random Variable is named for him
- He is my *academic* great<sup>12</sup>-grandfather
- Same eyes as Ice Cube

# Simeon-Denis Poisson

- Simeon-Denis Poisson (1781-1840) was a prolific French mathematician



- Published his first paper at 18, became professor at 21, and published over 300 papers in his life
  - He reportedly said “*Life is good for only two things, discovering mathematics and teaching mathematics.*”
- I’m going with French Martin Freeman

# Carl Friedrich Gauss

- Carl Friedrich Gauss (1777-1855) was a remarkably influential German mathematician



- Started doing groundbreaking math as teenager
  - Did not invent Normal distribution, but popularized it
- He looked like Martin Sheen
  - Who is, of course, Charlie Sheen's father



# Open Problems

# One Shot Learning

Single training example:

କୁ

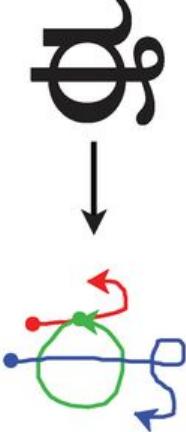
Test set:

a	ଶ	ଅ	ଶ
କୁ	ଅ	ପ୍ଲ	କୁ
ମ	କୁ	ଇ	ବ୍ର
ମ	ଅ	କୁ	ସ୍ତ୍ରୀ

# Bayesian Program Learning

i) 

අ	ඖ	භ	ය	සේ
කේ	ලා	නා	සු	රුමු
ඇ	තා	ඹ	තේ	ද්
නේ	යා	ලක්	හු	

iii) 

B Human drawings      Human parses      Machine parses

ග	ඇ	උ
ඇ	ඇ	ඇ
උ	ඇ	ඇ

ග	ඇ	ඇ
ඇ	ඇ	ඇ
ඇ	ග	ඇ

ග	ඇ	ඇ
ඇ	ඇ	ඇ
ඇ	ග	ඇ

ඇ	ඇ	ඇ
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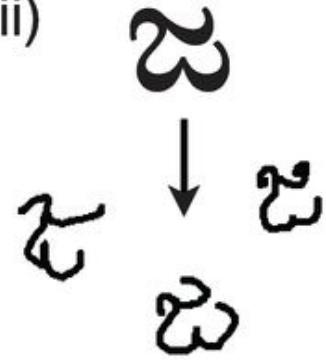
ඇ	ඇ	ඇ
ඇ	ඇ	ඇ
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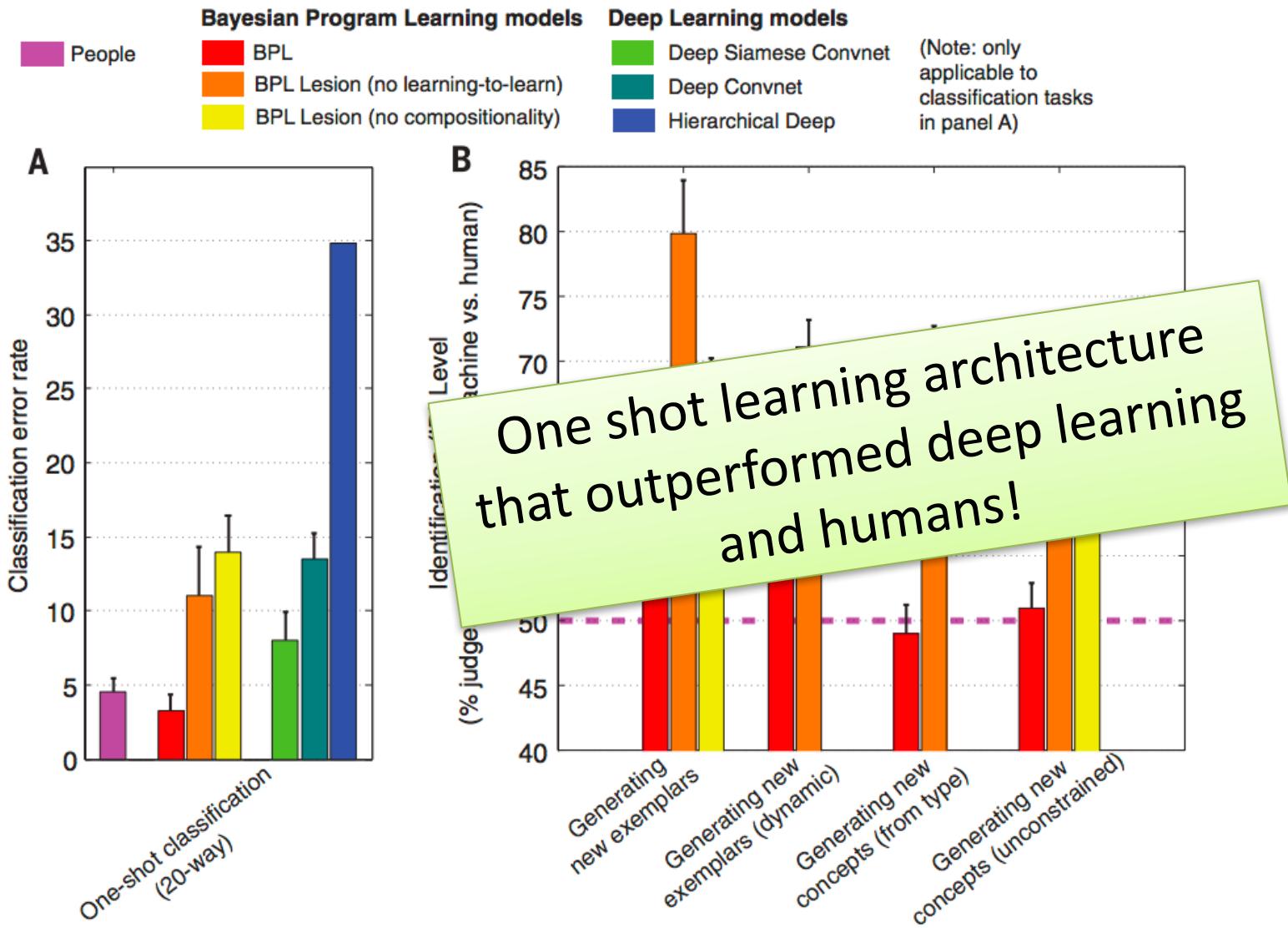
ඇ	ඇ	ඇ
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ඇ	ඇ	ඇ

ඇ	ඇ	ඇ
ඇ	ඇ	ඇ
ඇ	ඇ	ඇ

stroke order: — 1 — 2 — 3 — 4 — 5

ii) 

# Bayesian Program Learning



# Natural Language

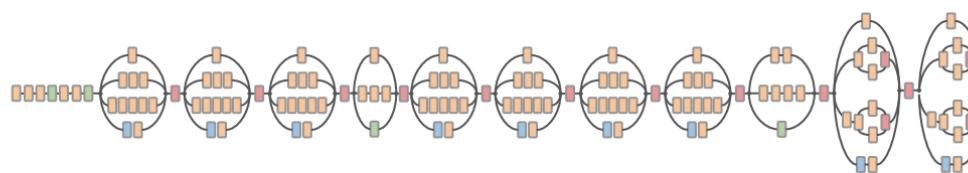


# AI for Medicine

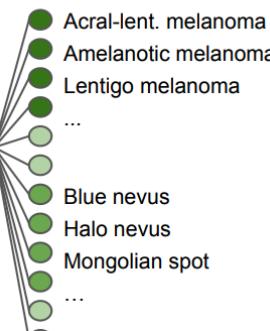
Skin Lesion Image



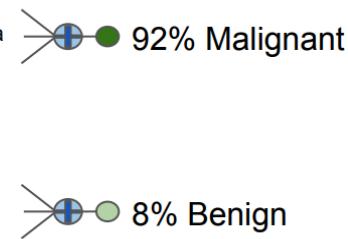
Deep Convolutional Neural Network (Inception-v3)



Training Classes (757)

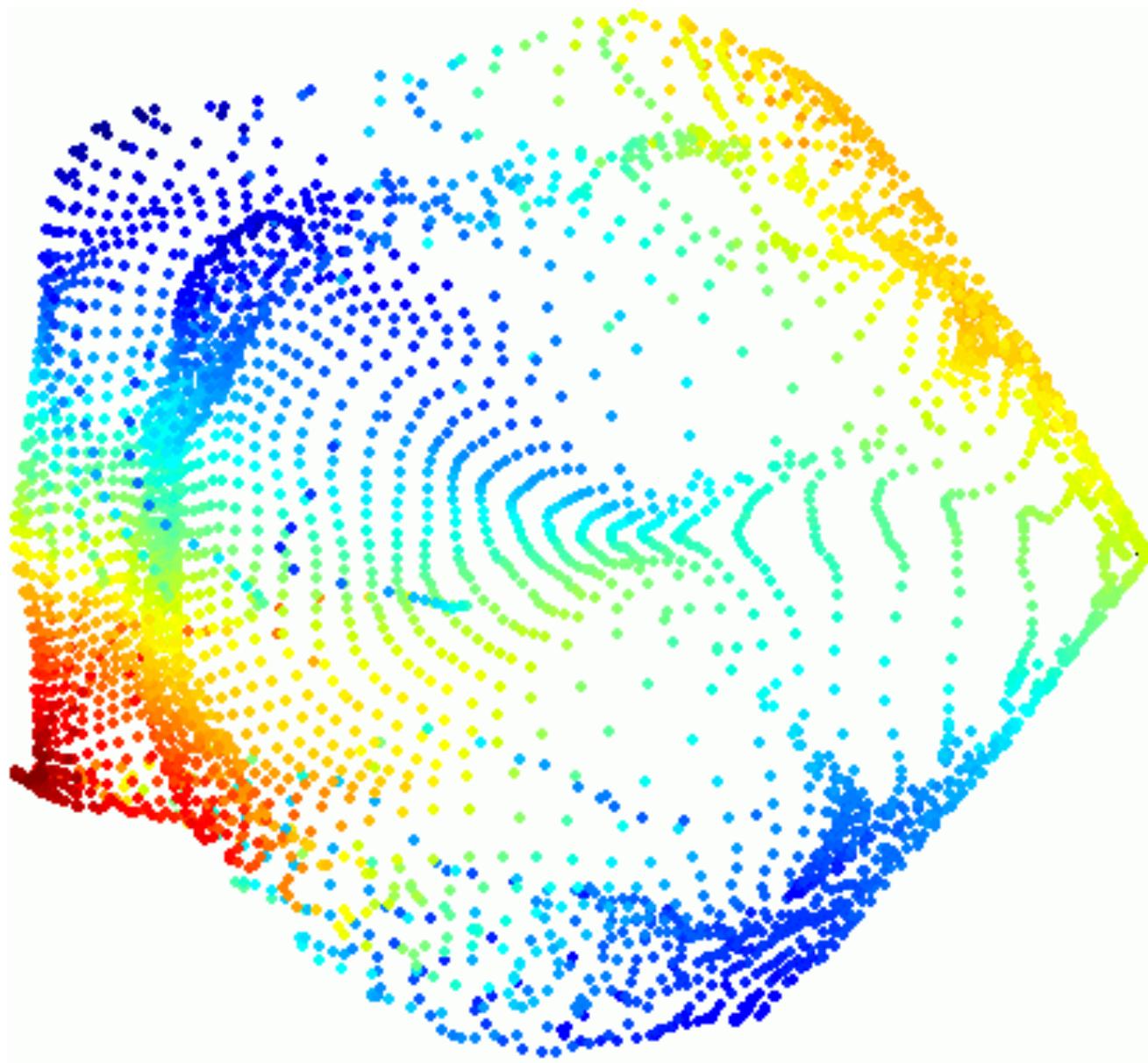


Inference Classes (varies by task)



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

# Theoretical Deep Learning



**Honorable Mentions:**  
Smart Grids  
Differential Privacy  
General AI  
Priors and Deep Learning

# Ethics and AI

# After CS109

## Theory

CS161 – Algorithmic analysis

Stats 217 – Stochastic Processes

CS 238 – Decision Making Under Uncertainty

CS 228 – Probabilistic Graphical Models

## AI

CS 221 – Intro to AI

CS 229 – Machine Learning

CS 230 – Deep Learning

CS 224N – Natural Language Processing

CS 234 – Reinforcement Learning

## Applications

CS 279 – Bio Computation

Literally any class with numbers in it



Technology magnifies.  
What do we want magnified?