### CS109 Summary Chris Piech CS109, Stanford University

FORD JUNION

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### **Counting Rules**



### What is a Probability?

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



### **Sources of Probability**



1. Experimentation



2. Dataset



3. Analytic Solution



4. Expert Opinion





### **Third Class with Coding!**



### **Target Revisited**



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 $\approx 0.1963$ 

### **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

### Everything in the world is either



a potato

### or not a potato.

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



### 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)



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

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



### 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(win) = 1/3 (random)



### **Second Ever Sections**

























### **Second Ever Sections**









### 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



### **Recall our Ebola Bats**



### **Discovered Pattern**



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These genes

don't impact T

G

 $G_3$ 

## **Random Variables**



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's

### 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<sup>n</sup>

### How much would you pay to play?

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



### **Binomial**



## 

### Poisson



### Geometric

### **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.



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



### **Bit Coin Mining**

You "mine a bitcoin" if, for given data D, you find a number N such that Hash(D, N) produces a string that starts with g zeroes.



### **Representative Juries**





# 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**



#### 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(wait < 5 min)?



#### Integrals



\*loving, not scary

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

# A probability!

#### **Climate Sensitivity**



#### **Probability Density Function**



#### **Simplicity is Humble**



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

#### PDF and CDF of a Normal



A CDF is the integral from –infinity 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.

- 3.5% 3.0% 2,5% 2.0% 1.5% 1.0% 0.5% 0.0% 22 24 26 28 30 32 34 36 38 40 42 44 46 48 50 52 54 56 58 60 62 64 66 68 70
- 2.1. Poziom podstawowy



http://freakonomics.com/2011/07/07/another-case-of-teacher-cheating-or-is-it-just-altruism/comment-page-2/

#### Will the Warriors Win?

What is the probability that the Warriors beat the Blazers? How do you model zero sum games?

OLDEN STA

#### **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 ~ N(S, 200<sup>2</sup>)
- The team with the higher sampled ability wins.



Arpad Elo



#### **Joint Distributions**

#### **Joint Distributions**



Go to this URL: https://goo.gl/Jh3Eu4

#### **Joint Probability Table**

	Walk	Bike	Scooter	Drive	Marginal Year
Freshman	0.04	0.04	0.01	0.03	0.12
Sophomore	0.03	0.34	0.03	0.00	0.40
Junior	0.04	0.21	0.01	0.00	0.25
Senior	0.07	0.08	0.01	0.00	0.16
5+	0.04	0.07	0.00	0.02	0.12
Marginal Mode	0.21	0.73	0.06	0.05	



#### **Joint Dart Distribution**



#### **Joint Dart Distribution**



Dart x location

#### 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





#### Midterm (part 1)



#### Midterm (part 2)



#### Midterm (part 3)



Midterm Score Bucket (m)



#### **Biometric Keystroke**





#### **Enchanted Die**



#### **Assignment Grades**



We have 2055 assignment distributions from grade scope

#### 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**



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Piech, CS106A, Stanford University

#### **C.L.T. Explains This**

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



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#### **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:



Happiness = {72, 85, 79, 91, 68, ..., 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



#### **Universal Sample**



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![](_page_74_Picture_3.jpeg)

#### **Peer Grading**

![](_page_75_Figure_1.jpeg)

Peer Grading on Coursera HCI.

31,067 peer grades for 3,607 students.

#### A/B Testing

![](_page_76_Picture_1.jpeg)

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![](_page_77_Picture_0.jpeg)

![](_page_78_Picture_0.jpeg)

# General "Inference"

#### tom Checker BETA

umo	ore symp	toms			AG	AE 30	GENDER Male	
Type your main symptom here						MY SYMPTOMS   cough × throat irritation ×		
					sneez	ing ×		
hoose	common syr	nptoms						
hoose	common syn	diarrhea	dizziness	fatigue				
hoose loating ver	common syn cough headache	nptoms diarrhea muscle cran	dizziness np nause	fatigue				

Continue

12)

#### Lots of Random Vars?

![](_page_80_Figure_1.jpeg)

#### **Bayes Nets!**

![](_page_81_Figure_1.jpeg)

	Alg #1: J	oi	nt	S	ar	npling
			N. werey	100.00		webMd — -bash — 38×22
		[0,	0,	0,	0]	
		[0,	1,	0,	1]	
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			1		11	one posterior sample:
5	" Joint Sa	10,	1,	0,	1 L L	ene perior estupie.
e	Here and the second	10,	1,	0,	01	
0	#	10,	0,	0,	0]	
7	# we can answer any	[0,	1,	1,	1]	5
à	# unith multium sints	[0,	1,	0,	0]	
8	# with multivariate	[0,	1,	0,	1]	
9	# where conditioned	[0]	1.	0.	01	
10	def and a ().	10	1	a	11	[Elu Harad Eaver Tired]
TO	der main():	10,	1	0	11	[i la, ogida, i ever, i i ea]
11	obs = getObserv	10,	1,	0,	101	
10		10,	0,	0,	10	
17	print Observat	11,	1,	1,	1]	
13		10,	0,	0,	0]	
14		[0,	0,	0,	0]	
14	samples = sampl	[1,	1,	1,	1]	
15	prob = probEluG	[0,	1,	0,	0]	
10		Obse	erva	atio	on =	[None, None, None, 1]
10	print Pr(Flu)	Pr	Flu	1 (	Obs)	= 0.140635888502
4.77			10		0007	012100000002
		-				

# Alg #2: MCMC

![](_page_83_Figure_1.jpeg)

#### MCMC is a way to sample with conditioned variables fixed

Each one of these is one joint sample:

[Flu, Undergrad, Fever, Tired]

#### **Towards Machine Learning**

![](_page_84_Picture_1.jpeg)

#### **MLE: Likelihood of Data**

#### Likelihood of Data from a Normal

![](_page_85_Figure_2.jpeg)

#### **Gumbel Fit**

![](_page_86_Figure_1.jpeg)

#### **MAP: Most Probable Parameter**

![](_page_87_Picture_1.jpeg)

#### **Machine Learning**

#### Heart

![](_page_88_Picture_2.jpeg)

![](_page_88_Picture_3.jpeg)

#### Netflix

# NETFLIX

#### **Logistic Regression**

![](_page_89_Figure_1.jpeg)

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![](_page_90_Figure_0.jpeg)

# 

![](_page_91_Picture_1.jpeg)

#### By the numbers

#### ~600 Fruit

![](_page_93_Picture_1.jpeg)

# ~ 30 Major Keys

![](_page_94_Figure_1.jpeg)

Naïve Bayes Assumption:  $P(\mathbf{x}|y) = \prod_{i} P(x_i|y)$ 

# 1 Contest

![](_page_95_Picture_1.jpeg)

#### **Thomas Bayes**

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

![](_page_96_Picture_2.jpeg)

- 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

![](_page_97_Picture_2.jpeg)

- 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

![](_page_98_Picture_2.jpeg)

- 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

![](_page_99_Picture_2.jpeg)

- 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

![](_page_100_Picture_0.jpeg)

![](_page_101_Picture_0.jpeg)

#### **Proximal Concepts**

#### **Bounds: Markov's Inequality**

• Say X is a **non-negative** random variable

$$P(X \ge a) \le \frac{E[X]}{a}$$
, for all  $a > 0$ 

 $\widehat{x}$ 

f(X)

 $\mathcal{X}$ 

- Proof:
  - I = 1 if  $X \ge a$ , 0 otherwise
  - Since  $X \ge 0$ ,  $I \le \frac{X}{a}$
  - Taking expectations:

$$E[I] = P(X \ge a) \le E\left[\frac{X}{a}\right] = \frac{E[X]}{a}$$

#### Markov and the Midterm

- Statistics from CS109 midterm
  - X = midterm score
  - Using sample mean  $\overline{X} = 102.0 \approx E[X]$
  - What is  $P(X \ge 110)$ ?

$$P(X \ge 110) \le \frac{E[X]}{110} = \frac{102}{110} = 0.93$$

- Markov bound:  $\leq 93\%$  of class scored 110 or greater
- In fact, 15.1% of class scored 110 or greater
  - Markov inequality can be a very loose bound
  - o But, it made <u>no</u> assumption at all about form of distribution!

#### Learn Bayes Nets Params?

![](_page_105_Figure_1.jpeg)

\* That is what we did with Naïve Bayes

#### Learn Bayes Nets Structure?

![](_page_106_Picture_1.jpeg)

\* That is what we did with Ebola Bats!

# Missing Data?

Missing Not at Random: You collect data on whether or not people intend to vote for Ayesha, a candidate in an upcoming election. You send an electronic poll to 100 randomly chosen people. You assume all 100 responses are IID.

User Response	Count
Responded that they will vote for Ayesha	40
Responded that they will not vote for Ayesha	45
Did not respond	15

#### \* Scratched the surface in section
#### **Temporal Patterns?**



\* Special type of Bayesian Network called a Markov Network

#### Ethics and Al

#### **Ethics and Datasets?**



Sometimes machine learning feels universally unbiased.

We can even call some estimators "unbiased"

Google/Nikon/HP had biased datasets



https://www.google.com/ingresTingurl=http%3A%2P%2Fww...

JANSPERSON,

#### Much more to Ethics + Al

#### **Open Problems**

#### **One Shot Learning**

Single training example:



Test set:

# とうりていていていていた。

#### **Bayesian Program Learning**





Lake et al. Human-level concept learning through probabilistic program induction

#### **Bayesian Program Learning**





Lake et al. Human-level concept learning through probabilistic program induction

#### **Transfer Learning**



#### **Neural Network Structure?**



#### **Neural Turing Machines**



Neural Turing Machine (NTM)



#### **Theoretical Deep Learning**



### Sampling + Deep Learning!





#### **Natural Language**



### Al for Medicine



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

#### **Climate Change?**



#### **Climate Change?**



**Honorable Mentions Differential Privacy** Fairness and AI General AI **Better Optimization** DeepLearning + X Self Driving Cars **Understanding Video** 

## 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 - Macine 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?

#### Why Study Probability + CS?

#### Interdisciplinary



#### **Closest Thing To Magic**



#### Now is the Time



#### Oh and Its Useful



Code.org

#### **Everyone is Welcome**



# I guarantee the techniques will change...

# You are close to the edge of human knowledge

# (all of you)

thank you!

## The End