

CS5841/EE5841 Machine Learning

Lecture 2: Probability Review

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Overview

- Basic probability
- Baye's Rule
- Random variables and distributions

Presentation contains material from Dr. Tim Havens, used with permission



Class updates

- Friday (January 12) class is cancelled in person
- We will have a video tutorial and extra credit assignment to introduce Julia
- Discord server link is on Canvas front page
- Two surveys due Friday!



Related reading

- Strongly suggested
 - Section 1.2 in Bishop
- Additional
 - Chapters 2-4 in Murphy
 - Chapter 3 in Goodfellow



Probability taxonomy

- **Experiment**
 - How is data being collected?
- **Sample Space**
 - What are the possible outcomes of the experiment?
- **Event**
 - A subset of the possible outcomes
- **Probability of an event**
 - The likelihood of an event occurring



Probability example

- **Experiment**
 - Flip 2 coins
- **Sample Space**
 - $S = \{HH, TT, HT, TH\}$
- **Event**
 - $A = \{HH\}; B = \{HT, TH\}$
- **Probability of an event**
 - Axiom 1: $\Pr(A) \geq 0$
 - Axiom 2: $\Pr(S) = 1$
 - Axiom 3: For every sequence of disjoint events

$$\Pr\left(\bigcup_i A_i\right) = \sum_i \Pr(A_i)$$



Frequentist statistics

- Count things!
- Example: $\Pr(A) = n(A)/N$ (frequentist statistics)



Joint Probability

- For events A and B , joint probability $\Pr(AB)$ [also shown as $\Pr(A,B)$] denotes the probability that both events happen
- Example: $A = \{HH\}$, $B = \{HT, TH\}$, what is the joint probability $\Pr(AB)$?



Independence

- Two events A and B are independent iff
 - $\Pr(AB) = \Pr(A)\Pr(B)$
- A set of events $\{A_i\}$ is independent iff

$$\Pr\left(\bigcap_i A_i\right) = \prod_i \Pr(A_i)$$



Independence, continued

Consider the experiment of tossing a coin twice

- Example 1:
 - $A = \{HT, HH\}$, $B = \{HT\}$
 - Is A independent from event B?
- Example 2:
 - $A = \{HT\}$, $B = \{TH\}$
 - Is A independent from event B?
- Disjoint \neq independence
- If A is independent from B, B is independent from C, is A independent from C?



Conditioning

- If A and B are events with $\Pr(A) > 0$, the **conditional probability of B given A** is

$$\Pr(B|A) = \frac{\Pr(AB)}{\Pr(A)}$$

- If A is independent from B, what is the relationship between $\Pr(A|B)$ and $\Pr(A)$?



Conditional Probability Example

$$Pr(B|A) = \frac{Pr(AB)}{Pr(A)}$$

$$Pr(A) = n(A)/N$$

	Female	Male
Success	200	1800
Failure	1800	200

A = {patient is female}

B = {drug fails}

$Pr(B|A) = ?$

$Pr(A|B) = ?$



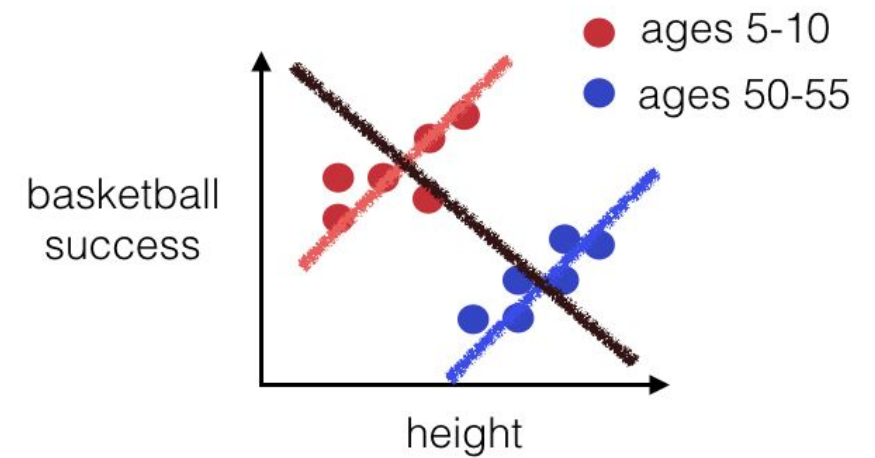
Conditional Probability Example, cont.

- $\Pr(A) = n(A)/N = 2000/4000 = 0.5$
- $\Pr(B) = n(B)/N = 2000/4000 = 0.5$
- $\Pr(AB) = n(AB)/N = 1800/4000 = 0.45$
- $\Pr(B|A) = \Pr(AB)/\Pr(A) = 0.45/0.5 = 0.9$
- $\Pr(A|B) = \Pr(AB)/\Pr(B) = 0.45/0.5 = 0.9$



Simpson's Paradox

A trend in data is reversed when data is grouped differently



Simpson's Paradox example

	Female		Male	
	Drug 1	Drug 2	Drug 1	Drug 2
Success	200	10	19	1000
Failure	1800	190	1	1000

Case 1: Grouped by drugs

Case 2: Grouped by gender



Simpson's Paradox example, Case 1

	Female		Male	
	Drug 1	Drug 2	Drug 1	Drug 2
Success	200	10	19	1000
Failure	1800	190	1	1000

Drug 2 is better than Drug 1

	Drug 1	Drug 2
Success	219	1010
Failure	1801	1190

$A = \{\text{Using Drug 1}\}$

$B = \{\text{Using Drug 2}\}$

$C = \{\text{Drug succeeds}\}$

$\Pr(C|A)$ is about 10%

$\Pr(C|B)$ is about 50%



Simpson's Paradox example, Case 2

	Female	
	Drug 1	Drug 2
Success	200	10
Failure	1800	190

A = {Using Drug 1}
B = {Using Drug 2}
C = {Drug succeeds}
 $\Pr(C|A)$ is about 20%
 $\Pr(C|B)$ is about 5%

	Male	
	Drug 1	Drug 2
Success	19	1000
Failure	1	1000

A = {Using Drug 1}
B = {Using Drug 2}
C = {Drug succeeds}
 $\Pr(C|A)$ is about 100%
 $\Pr(C|B)$ is about 50%

Drug 1 is better than Drug 2



Conditional Independence

- Event A and B are conditionally independent given C iff
 - $\Pr(AB|C) = \Pr(A|C)\Pr(B|C)$
- A set of events $\{A_i\}$ is conditionally independent given C iff

$$\Pr\left(\bigcup_i A_i | c\right) = \prod_i \Pr(A_i | C)$$



Conditional Independence, continued

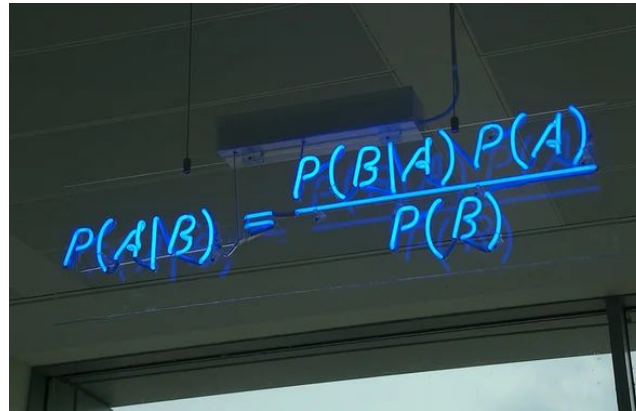
- Example: Consider three events: A, B, C
 - $\Pr(A) = \Pr(B) = \Pr(C) = 1/5$
 - $\Pr(A,C) = \Pr(B,C) = 1/25$, $\Pr(A,B) = 1/10$
 - $\Pr(A,B,C) = 1/125$
 - Are A and B independent?
 - Are A and B conditionally independent given C?
- A and B are independent \neq A and B are conditionally independent



Baye's Rule

- Given two events A and B, where $\Pr(A) > 0$, then:

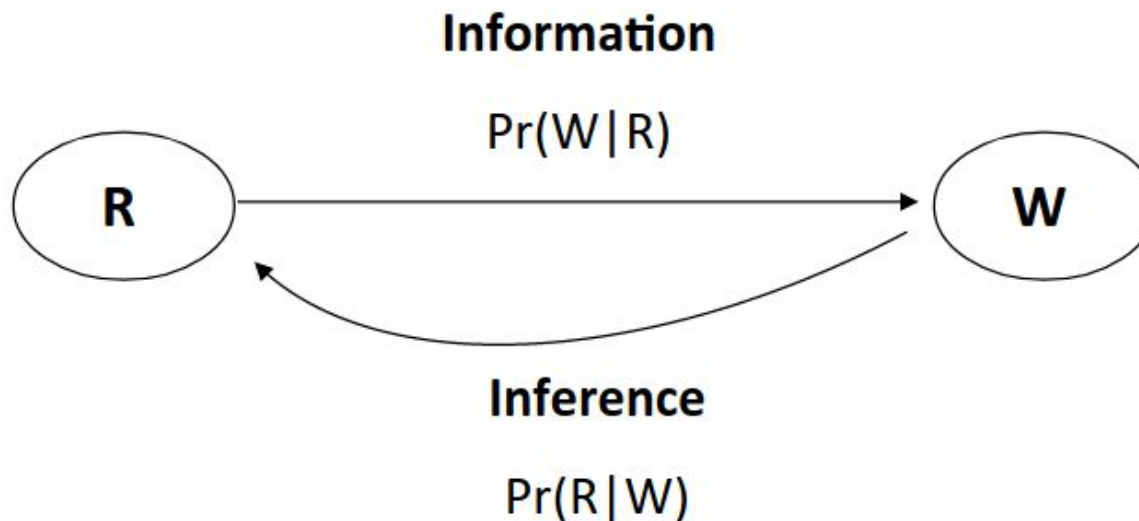
$$\Pr(B|A) = \frac{\Pr(AB)}{\Pr(A)} = \frac{\Pr(A|B)\Pr(B)}{\Pr(A)}$$



A photograph of a chalkboard with the formula for Bayes' Rule written in blue chalk. The formula is $P(A|B) = \frac{P(B|A)P(A)}{P(B)}$. The chalkboard is dark, and the chalk is bright blue. The formula is written in a clear, legible hand.



Baye's Rule, continued



$$\begin{array}{l} \text{Posterior} \\ \Pr(B|A) \end{array} = \frac{\Pr(AB)}{\Pr(A)} = \frac{\begin{array}{l} \text{Likelihood} \quad \text{Prior} \\ \Pr(A|B) \Pr(B) \end{array}}{\Pr(A)}$$



Baye's Rule example

$$Pr(B|A) = \frac{Pr(AB)}{Pr(A)} = \frac{Pr(A|B)Pr(B)}{Pr(A)}$$

Example:

- R: It's rainy
- W: Grass is wet
- $Pr(R) = 0.8$
- $Pr(R|W) = ?$

$Pr(W R)$	R	$\sim R$
W	0.7	0.4
$\sim W$	0.3	0.6



Interlude: How do we get $\Pr(A)$

- Law of Total Probability

$$P(B) = P(B | A_1)P(A_1) + \dots + P(B | A_n)P(A_n).$$



Baye's Rule example worked

$$Pr(B|A) = \frac{Pr(AB)}{Pr(A)} = \frac{Pr(A|B)Pr(B)}{Pr(A)}$$

Prior: $Pr(R) = 0.8$

Likelihood: $Pr(W|R) = 0.7$

Normalization: $Pr(W) = Pr(W|R)*P(R) + Pr(W|\sim R)*P(\sim R) =$

$$0.7*0.8 + 0.4*0.2 = 0.64$$

All together: $Pr(R|W) = (0.7*0.8)/0.64 = 0.875$



Random variables and distributions

- A random variable X is a numerical outcome of a random experiment
- The distribution of a random variable is the collection of possible outcomes along with their probabilities

- Discrete case: $Pr(X = x) = p_{\theta}(x)$

- Continuous case: $Pr(a \leq X \leq b) = \int_a^b p_{\theta}(x) dx$



PDFs, CDFs, etc.

Probability density function (PDF): a continuous function that describes the possible values and likelihoods of a random variable, represented as $f_x(x)$

Probability mass function (PMF): PDF specifically for discrete random variables, often represented as $p_x(x)$

Cumulative distribution function (CDF): a continuous function that describes the probability that a random variable is less than or equal to a given value, represented as $F_x(x)$



Random variable example

- Let S be the set of all sequences of two rolls of a fair 6-sided die. Let X be the sum of both rolls
 - What are all possible values for X ?
 - $\Pr(X=2) = ?$
 - Frequentist statistics!
 - 1 possible sequence sums to 2, 36 different sequences possible
 - $1/36$
 - $\Pr(X=7) = ?$



Expected value (mean)

- Given a random variable $X \sim \Pr(X=x)$
- Expectation is: $E[X] = \sum_x x \Pr(X = x)$

- For an empirical sample: x_1, x_2, \dots, x_n : $E[X] = \frac{1}{N} \sum_{i=1}^N x_i$
- Continuous case:

$$E[X] = \int_{-\infty}^{\infty} x p_{\theta}(x) dx$$

- Expectation of a sum of random variables:

$$E[X_1 + X_2] = E[X_1] + E[X_2]$$



Expectation examples

$$E[X] = \sum_x x \Pr(X = x)$$

- Let X be the sum of two rolls of a fair 6-sided die
 - What is $E[x]$?
 - $E[x] = 2 \cdot 1/36 + 3 \cdot 2/36 + \dots$
-



Variance

- The variance of a random variable X is the expectation of $(X - E[X])^2$

$$\begin{aligned} \text{Var}(X) &= E[(X - E[X])^2] \\ &= E[X^2 + E[X]^2 - 2XE[X]] \\ &= E[X^2 - E[X]^2] \\ &= E[X^2] - E[X]^2 \end{aligned}$$



Some common distributions

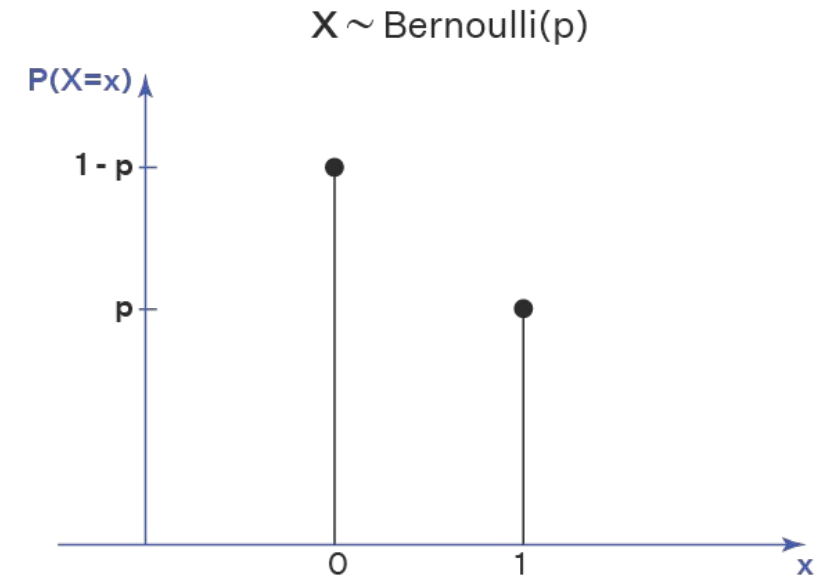
- Bernoulli
- Binomial
- Poisson
- Gaussian (aka. Normal)



Bernoulli Distribution

- The outcome of an experiment is a success or a failure
 - $\Pr(X=1) = p, \Pr(X=0) = 1-p$

Bernoulli Distribution Graph

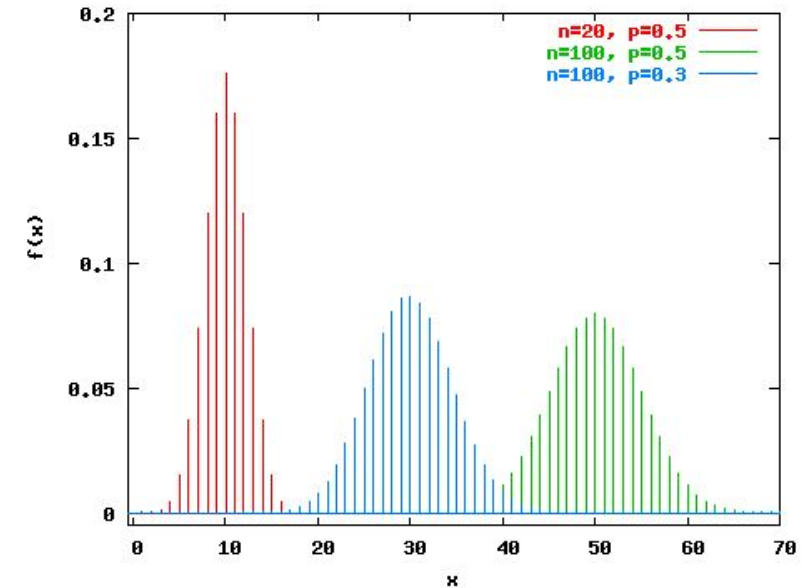


Binomial Distribution

- n draws of a Bernoulli distribution
 - $X_i \sim \text{Bernoulli}(p)$, $X \sim \text{Bin}(p, n)$
- Random experiment X stands for number of successes

$$\Pr(X = x) = p_{\theta}(x) = \begin{cases} \binom{n}{x} p^x (1-p)^{n-x} & x = 1, 2, \dots, n \\ 0 & \text{else} \end{cases}$$

- $E[X] = np$, $\text{Var}(X) = np(1-p)$



Poisson Distribution

- Comes from Binomial Distribution
 - Fix expectation $E[x] = \lambda$
 - Let number of trials $n \gg \lambda$
 - Binomial distribution becomes a Poisson distribution

$$Pr(X = x) = p_{\theta}(x) = \begin{cases} \frac{\lambda^x}{x!} e^{-\lambda} & x \geq 0 \\ 0 & \text{else} \end{cases}$$

- $E[x] = \text{Var}[x] = \lambda$



Gaussian (Normal) Distribution

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

$$p_{\theta}(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left\{ \frac{-(x - \mu)^2}{2\sigma^2} \right\}$$

$$Pr(a \leq X \leq b) = \int_a^b p_{\theta}(x)$$

- $E[x] = ?, \text{Var}(X) = ?$



Questions + Comments?

