

# CMPSCI 240: “Reasoning Under Uncertainty”

## Lecture 4

Prof. Hanna Wallach  
wallach@cs.umass.edu

February 2, 2012

## Reminders

- ▶ Pick up a copy of B&T
- ▶ Check the course website: `http://www.cs.umass.edu/~wallach/courses/s12/cmpsci240/`
- ▶ First homework will be assigned tomorrow

Recap

## Last Time: Total Probability Theorem

- ▶ If  $A_1, \dots, A_n$  partition  $\Omega$  then for any event  $B$

$$\begin{aligned} P(B) &= P(B \cap A_1) + \dots + P(B \cap A_n) \\ &= \sum_{i=1}^n P(A_i) P(B | A_i) \end{aligned}$$

## Last Time: Total Probability Theorem

- ▶ If  $A_1, \dots, A_n$  partition  $\Omega$  then for any event  $B$

$$\begin{aligned} P(B) &= P(B \cap A_1) + \dots + P(B \cap A_n) \\ &= \sum_{i=1}^n P(A_i) P(B | A_i) \end{aligned}$$

- ▶ “Divide-and-conquer” approach to finding  $P(B)$

## Last Time: Bayes' Rule

- ▶ If  $A_1, \dots, A_n$  partition  $\Omega$  then for any event  $B$

$$\underbrace{P(A_i | B)}_{\text{posterior}} = \frac{P(B \cap A_i)}{P(B)} = \frac{\overbrace{P(A_i)}^{\text{prior}} P(B | A_i)}{\sum_{i=1}^n P(A_i) P(B | A_i)}$$

## Last Time: Bayes' Rule

- ▶ If  $A_1, \dots, A_n$  partition  $\Omega$  then for any event  $B$

$$\underbrace{P(A_i | B)}_{\text{posterior}} = \frac{P(B \cap A_i)}{P(B)} = \frac{\overbrace{P(A_i)}^{\text{prior}} P(B | A_i)}{\sum_{i=1}^n P(A_i) P(B | A_i)}$$

- ▶ Useful for **inference**, i.e., where we know  $P(B | A_i)$  and  $P(A_i)$  for every  $i$  and want to find  $P(A_i | B)$  for some  $i$

Independence



# Independence

- ▶ Two events  $A$  and  $B$  are **independent** if and only if

$$P(A \cap B) = P(A) P(B)$$

# Independence

- ▶ Two events  $A$  and  $B$  are **independent** if and only if

$$P(A \cap B) = P(A)P(B)$$

- ▶ If  $P(B) > 0$  this is equivalent to

$$\frac{P(A \cap B)}{P(B)} = P(A|B) = P(A)$$

# Independence

- ▶ Two events  $A$  and  $B$  are **independent** if and only if

$$P(A \cap B) = P(A)P(B)$$

- ▶ If  $P(B) > 0$  this is equivalent to

$$\frac{P(A \cap B)}{P(B)} = P(A|B) = P(A)$$

- ▶ If  $P(A) > 0$  this is equivalent to  $P(B|A) = P(B)$

## Examples of Independence

- ▶ e.g., are  $A$  and  $A^c$  independent?

## Examples of Independence

- ▶ e.g., are  $A$  and  $A^c$  independent?
- ▶ e.g., if  $A$  and  $B$  are independent, are  $A$  and  $B^c$ ?

## Examples of Independence

- ▶ e.g., are  $A$  and  $A^c$  independent?
- ▶ e.g., if  $A$  and  $B$  are independent, are  $A$  and  $B^c$ ?
- ▶ e.g., flipping two coins,  $H_1 =$  “first flip is heads”  
 $= \{HH, HT\}$ ,  $H_2 =$  “second flip is heads”  $= \{HH, TH\}$ ,  
and  $D =$  “the flips are different”  $= \{HT, TH\}$ :  $H_1$  and  $H_2$   
are independent, what about  $H_1$  and  $D$ ?

## Independence of Multiple Events

- ▶ Events  $A$ ,  $B$ , and  $C$  are independent if and only if

$$P(A \cap B) = P(A) P(B)$$

$$P(A \cap C) = P(A) P(C)$$

$$P(B \cap C) = P(B) P(C)$$

$$P(A \cap B \cap C) = P(A) P(B) P(C)$$

## Independence of Multiple Events

- ▶ Events  $A$ ,  $B$ , and  $C$  are independent if and only if

$$P(A \cap B) = P(A) P(B)$$

$$P(A \cap C) = P(A) P(C)$$

$$P(B \cap C) = P(B) P(C)$$

$$P(A \cap B \cap C) = P(A) P(B) P(C)$$

- ▶ **Pairwise independence** does not imply independence



## Examples of Independence of Multiple Events

- ▶ e.g., flipping two coins,  $H_1 =$  “first flip is heads”  
 $= \{HH, HT\}$ ,  $H_2 =$  “second flip is heads”  $= \{HH, TH\}$ ,  
and  $D =$  “the flips are different”  $= \{HT, TH\}$ :  $H_1$  and  $H_2$   
are independent,  $H_1$  and  $D$  are independent,  $H_2$  and  $D$  are  
independent, what about  $H_1$ ,  $H_2$ , and  $D$ ?

## Conditional Independence

- ▶  $A$  and  $B$  are **conditionally independent** given  $C$  if and only if

$$P(A \cap B | C) = P(A | C) P(B | C)$$

## Conditional Independence

- ▶  $A$  and  $B$  are **conditionally independent** given  $C$  if and only if

$$P(A \cap B | C) = P(A | C) P(B | C)$$

- ▶ If  $P(B | C) > 0$  this is equivalent to

$$\frac{P(A \cap B | C)}{P(B | C)} = P(A | B \cap C) = P(A | C)$$

## Conditional Independence

- ▶  $A$  and  $B$  are **conditionally independent** given  $C$  if and only if

$$P(A \cap B | C) = P(A | C) P(B | C)$$

- ▶ If  $P(B | C) > 0$  this is equivalent to

$$\frac{P(A \cap B | C)}{P(B | C)} = P(A | B \cap C) = P(A | C)$$

- ▶ If  $P(A | C) > 0$  this is equivalent to  $P(B | A \cap C) = P(B | C)$

## Examples of Conditional Independence

- ▶ e.g., flipping two coins,  $H_1 =$  “first flip is heads”  
 $= \{HH, HT\}$ ,  $H_2 =$  “second flip is heads”  $= \{HH, TH\}$ ,  
and  $D =$  “the flips are different”  $= \{HT, TH\}$ : we already  
know that  $H_1$  and  $H_2$  are independent, but are they  
conditionally independent given  $D$ ?

# Counting

## Equally Likely Outcomes

- ▶ **Discrete probability law:** If  $\Omega$  is finite and  $A = \{x_1, \dots, x_n\} \subseteq \Omega$  then  $P(A) = P(x_1) + \dots + P(x_n)$

## Equally Likely Outcomes

- ▶ **Discrete probability law:** If  $\Omega$  is finite and  $A = \{x_1, \dots, x_n\} \subseteq \Omega$  then  $P(A) = P(x_1) + \dots + P(x_n)$
- ▶ If  $P(x_i) = p$  for all  $i = 1 \dots n$ , then  $P(A) = p|A|$



## Equally Likely Outcomes

- ▶ **Discrete probability law:** If  $\Omega$  is finite and  $A = \{x_1, \dots, x_n\} \subseteq \Omega$  then  $P(A) = P(x_1) + \dots + P(x_n)$
- ▶ If  $P(x_i) = p$  for all  $i = 1 \dots n$ , then  $P(A) = p|A|$
- ▶ **Discrete uniform probability law:** If  $\Omega$  is finite and all outcomes are equally likely, then  $P(A) = |A| / |\Omega|$

## Equally Likely Outcomes

- ▶ **Discrete probability law:** If  $\Omega$  is finite and  $A = \{x_1, \dots, x_n\} \subseteq \Omega$  then  $P(A) = P(x_1) + \dots + P(x_n)$
- ▶ If  $P(x_i) = p$  for all  $i=1 \dots n$ , then  $P(A) = p|A|$
- ▶ **Discrete uniform probability law:** If  $\Omega$  is finite and all outcomes are equally likely, then  $P(A) = |A| / |\Omega|$
- ▶ How can we count  $|A|$  and  $|\Omega|$ ?

# The Counting Principle: When Order Matters

- ▶ Consider a process with  $r$  stages, e.g., rolling  $r$  dice

# The Counting Principle: When Order Matters

- ▶ Consider a process with  $r$  stages, e.g., rolling  $r$  dice
- ▶ There are  $n_1$  possible choices at the first stage

# The Counting Principle: When Order Matters

- ▶ Consider a process with  $r$  stages, e.g., rolling  $r$  dice
- ▶ There are  $n_1$  possible choices at the first stage
- ▶ For each of these, there are  $n_2$  possible choices at stage 2

## The Counting Principle: When Order Matters

- ▶ Consider a process with  $r$  stages, e.g., rolling  $r$  dice
- ▶ There are  $n_1$  possible choices at the first stage
- ▶ For each of these, there are  $n_2$  possible choices at stage 2
- ▶ In general, for each possible choice at stage  $i - 1$ , there are  $n_i$  possible choices at stage  $i \implies$  the total number of choices (i.e., outcomes for the entire process) is  $n_1 n_2 n_3 \dots n_r$

## Examples of the Counting Principle

- ▶ e.g., a local phone number is a 7 digit sequence, but the first digit can't be a 0 or 1. How many local numbers are there?

## Examples of the Counting Principle

- ▶ e.g., a local phone number is a 7 digit sequence, but the first digit can't be a 0 or 1. How many local numbers are there?
- ▶ e.g., if  $A = \{x_1, \dots, x_n\}$  how many subsets does  $A$  have?



## Sampling with Replacement

- ▶ e.g., drawing  $r = 5$  cards from a deck of  $n = 52$  cards with replacement:  $n_1 = n_2 = \dots n_5 = n = 52$ , so there are  $n^r = 52^5$  ways of drawing 5 cards with replacement

# Permutations

# Permutations

- ▶ e.g., how many ways can we assign  $n$  threads to  $n$  processors, such that each thread is assigned to exactly one processor and each processor is assigned exactly one thread?

## $k$ -Permutations

- ▶ e.g., how many ways can we assign  $n$  threads to  $k \leq n$  processors such that no thread is assigned to multiple processors and each processor is assigned exactly one thread?

## Examples of Permutations

- ▶ e.g., suppose you have 4 books about competitive eating, 10 books about Linux, and 2 books about roller derby. How many ways can you arrange these books on a shelf such that all of the books on a given subject are grouped together?

# Combinations

## When Order Doesn't Matter

- ▶ **Combination:** order of the selected elements doesn't matter

## When Order Doesn't Matter

- ▶ **Combination:** order of the selected elements doesn't matter
- ▶ When order doesn't matter some permutations are indistinguishable from others, e.g., pizza toppings: bacon, ham, and sausage vs. sausage, bacon, and ham



# Combinations

- ▶ If we take a set of  $k$ -permutations and group “duplicates” then  $k!$  permutations will correspond to each combination

# Combinations

- ▶ If we take a set of  $k$ -permutations and group “duplicates” then  $k!$  permutations will correspond to each combination
- ▶ When order doesn't matter, the number of ways to choose  $k$  elements from a set of  $n$  elements (i.e., combinations) is

$$\frac{\# \text{ } k\text{-permutations}}{k!} = \frac{n!}{k!(n-k)!} = \binom{n}{k}$$

## Examples of Combinations

- ▶ e.g., Antonios offers fifteen different toppings. How many ways can you create a pizza with three distinct toppings?

# Real-World Examples of Combinations



- ▶ 4food claims “more than a million” burger combinations

## Real-World Examples of Combinations



- ▶ 4food claims “more than a million” burger combinations
- ▶ Last year, CNN Money decided to verify this claim...

## How Many Combinations?!

- ▶ 5 buns (bagel, brioche, multigrain, ...): can have 1

## How Many Combinations?!

- ▶ 5 buns (bagel, brioche, multigrain, ...): can have 1
- ▶ 4 add-ons (lettuce, pickle, tomato, onion): can have 0–4

## How Many Combinations?!

- ▶ 5 buns (bagel, brioche, multigrain, ...): can have 1
- ▶ 4 add-ons (lettuce, pickle, tomato, onion): can have 0–4
- ▶ 12 sauces (mustard, mayo, ketchup, ...): can have 0–3



## How Many Combinations?!

- ▶ 5 buns (bagel, brioche, multigrain, ...): can have 1
- ▶ 4 add-ons (lettuce, pickle, tomato, onion): can have 0–4
- ▶ 12 sauces (mustard, mayo, ketchup, ...): can have 0–3
- ▶ 7 cheeses (blue, goat, cheddar, ...): can have 0–2

## How Many Combinations?!

- ▶ 5 buns (bagel, brioche, multigrain, ...): can have 1
- ▶ 4 add-ons (lettuce, pickle, tomato, onion): can have 0–4
- ▶ 12 sauces (mustard, mayo, ketchup, ...): can have 0–3
- ▶ 7 cheeses (blue, goat, cheddar, ...): can have 0–2
- ▶ 8 patties (beef, pork, egg, lamb, ...): can have 1

## Less Exciting Examples of Combinations

- ▶ e.g., a system contains  $2x$  disks divided into  $x$  pairs, where each pair of disks contains the same data. If one of the disks in a pair fails, the data can be recovered, but if both disks fail, it cannot. Suppose two random disks fail. What is the probability that some data is inaccessible?

[Got to here in class...]

## Binomial Probabilities

## Sequences of Independent Trials

- ▶ Suppose we have a biased coin that lands heads with probability  $p$ . If we flip this coin 5 times, what is the probability that the outcome is *HHTTH*?

## Sequences of Independent Trials

- ▶ Suppose we have a biased coin that lands heads with probability  $p$ . If we flip this coin 5 times, what is the probability that the outcome is *HHTTH*?
- ▶ If we flip the coin 5 times, what is the probability that the outcome consists of 3 heads and 2 tails in any order?

# Binomial Probabilities

- ▶ Consider a sequence of  $n$  independent trials, each with a “success” probability  $p$ . The probability of any particular sequence with  $k$  successes is  $p^k (1 - p)^{n-k}$



# Binomial Probabilities

- ▶ Consider a sequence of  $n$  independent trials, each with a “success” probability  $p$ . The probability of any particular sequence with  $k$  successes is  $p^k (1 - p)^{n-k}$
- ▶ The probability of exactly  $k$  successful trials is

$$\binom{n}{k} p^k (1 - p)^{n-k}$$

## Examples of Sequences of Independent Trials

- ▶ A cell phone provider can handle up to  $r$  data requests at once. Assume that every minute, each of the provider's  $n$  customers makes a request with probability  $p$ , independent of the behavior of the other customers. What is the probability that exactly  $x$  customers will make a data request during a particular minute? What is the probability that  $> r$  customers will make a data request during a particular minute?

# Partitions

# Combinations and Partitions

- ▶ **Combination:**  $k$  elements from a  $n$  element set, ignoring order

# Combinations and Partitions

- ▶ **Combination:**  $k$  elements from a  $n$  element set, ignoring order
- ▶ A combination partitions the set in two: elements that belong to the  $k$ -element combination, and elements that don't

# Combinations and Partitions

- ▶ **Combination:**  $k$  elements from a  $n$  element set, ignoring order
- ▶ A combination partitions the set in two: elements that belong to the  $k$ -element combination, and elements that don't
- ▶ What about partitioning  $n$  elements into  $r$  disjoint subsets of sizes  $n_1, n_2, \dots, n_r$ ? How many ways can we do this?

# Partitions

- ▶ Form the subsets one at a time using a  $r$ -stage process

# Partitions

- ▶ Form the subsets one at a time using a  $r$ -stage process
- ▶ There are  $\binom{n}{n_1}$  ways to form the first subset



# Partitions

- ▶ Form the subsets one at a time using a  $r$ -stage process
- ▶ There are  $\binom{n}{n_1}$  ways to form the first subset
- ▶ For each of these, there are  $\binom{n-n_1}{n_2}$  ways to form subset 2

# Partitions

- ▶ Form the subsets one at a time using a  $r$ -stage process
- ▶ There are  $\binom{n}{n_1}$  ways to form the first subset
- ▶ For each of these, there are  $\binom{n-n_1}{n_2}$  ways to form subset 2
- ▶ In general, for each possible way to form subset  $i - 1$ , there are  $\binom{n-n_1-\dots-n_{i-1}}{n_i}$  ways to form subset  $i \implies$  there are  $\frac{n!}{n_1! n_2! \dots n_r!} = \binom{n}{n_1, n_2, \dots, n_r}$  ways to partition the set

## For Next Time

- ▶ Read B&T 2.1, 2.2, 2.3
- ▶ Check the course website: `http://www.cs.umass.edu/~wallach/courses/s12/cmpsci240/`
- ▶ First homework will be assigned tomorrow