

CMPSCI 240: “Reasoning Under Uncertainty”

Lecture 5

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Reminders

- ▶ Pick up a copy of B&T
- ▶ Check the course website: `http://www.cs.umass.edu/~wallach/courses/s12/cmpsci240/`
- ▶ First homework is due on Friday

Recap

Last Time: Counting

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- ▶ **Combinations** of k of n objects: $\binom{n}{k} = \frac{n!}{k!(n-k)!}$
- ▶ **Partitions** of n objects into r subsets: $\frac{n!}{n_1! n_2! \dots n_r!}$
- ▶ Challenge: try to show this yourself using a r -stage process!

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?

Binomial Probabilities

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- ▶ If we flip the coin 5 times, what is the probability that the outcome consists of 3 heads and 2 tails in any order?

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- ▶ 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?

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- ▶ A random variable can map **several outcomes** to **one value**

Examples of Random Variables

- ▶ e.g., suppose we have 4 disks, each of which fails with probability p , $\Omega = \{0000, 0001, 0010, \dots, 1111\}$: examples of random variables include the number of disks that failed, a boolean value indicating whether disks 1 or 2 failed, a boolean value indicating whether 3 or more disks failed, etc.

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- ▶ e.g., flipping two coins, $\Omega = \{HH, HT, TH, TT\}$: if X is the number of heads flipped, what is the event $\{X=1\}$?

Discrete vs. Continuous Random Variables

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- ▶ **Continuous random variable:** the set of values is uncountably infinite, e.g., choosing a in $[-1, 1]$: a^2 , $a/2$, $a + 3.1$, etc.

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- ▶ e.g., what is $\sum_x p_X(x)$ for any discrete random variable X ?

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- ▶ e.g., $Y = |X|$ and $p_X(x) = 1/9$ if x is an integer in the range $[-4, 4]$ and 0 otherwise: what do p_X and p_Y look like?

Common Discrete Random Variables

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- ▶ e.g., Used to model probabilistic situations with $b - a + 1$ equally likely outcomes $(a, a + 1, \dots, b)$, e.g., rolling a die

Bernoulli Random Variables

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- ▶ Used to model probabilistic situations with **two outcomes**, e.g., whether a server is online or an email is spam, etc.

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- ▶ X is the **sum** of n independent Bernoulli random variables, e.g., number of servers (out of n) that are online, etc.

Examples of Binomial Random Variables

- ▶ e.g., you go to a party with 500 guests. What is the probability that one other guest has the same birthday as you?

[Got to here in class...]

Geometric Random Variables

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- ▶ Used to model the number of repeated independent trials up to (and including) the **first “successful” trial**, e.g., spam

Examples of Geometric Random Variables

- ▶ e.g., you have 5 keys for your new apartment but you don't know which key opens the front door. What is the PMF of the number of trials needed to open the front door assuming that at each trial you are equally likely to choose any of the 5 keys?

Poisson Random Variables

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- ▶ e.g., the number of typos in a book with n words, number of cars (out of n) that crash in a city on a given day, etc.

Examples of Poisson Random Variables

- ▶ e.g., you go to a party with 500 guests. What is the probability that one other guest has the same birthday as you?

For Next Time

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- ▶ Check the course website: `http://www.cs.umass.edu/~wallach/courses/s12/cmpsci240/`
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