

CMPSCI 240: “Reasoning Under Uncertainty”

Lecture 6

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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 TOMORROW

Recap

Last Time: Random Variables

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- ▶ **Random variable:** a real-valued function of the outcome of an experiment... though not written as a function!
- ▶ A random variable X can be either **continuous** or **discrete**, i.e., the **range** of X can be either continuous or discrete
- ▶ Event $\{X=x\}$ is the event consisting of all outcomes in Ω that are **mapped to value x** by random variable X

Last Time: Probability Mass Functions

- ▶ Probability mass function (PMF) of X : denoted by p_X where

$$p_X(x) = P(X=x) = P(\{X=x\})$$

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$$p_X(x) = P(X=x) = P(\{X=x\})$$

- ▶ If X is a random variable and $g(\cdot)$ is some function, then $Y = g(X)$ is another random variable with PMF

$$p_Y(y) = \sum_{\{x \mid g(x)=y\}} p_X(x)$$

Last Time: Common Discrete Random Variables

- ▶ **Discrete uniform:** X has range $[a, b]$ and PMF

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- ▶ **Bernoulli:** X has range $\{0, 1\}$ and PMF

$$p_X(k) = p \text{ if } k=1 \text{ and } p_X(k) = (1 - p) \text{ if } k=0$$

Last Time: Common Discrete Random Variables

- ▶ **Binomial:** X is used to model the number of successes k in n independent trials and has range $[0, n]$ and PMF

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

Common Discrete Random Variables (Cont.)

Geometric Random Variables

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$$p_X(k) = (1 - p)^{k-1} p \text{ for } k = 1, 2, 3, \dots$$

- ▶ 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?

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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?

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- ▶ A weighted average of the possible values of X
- ▶ The value we'd "expect" to get for X "on average" if we repeated the same experiment (and calculated X) many times
- ▶ Useful if we want a single number that "summarizes" p_X

Examples of Expectation

- ▶ e.g., on any given day, with probability 0.2, I am not especially tired and will drink only 3 cups of coffee before work. With probability 0.8, however, I am VERY tired and will drink 6 cups of coffee before work. What is the expected number of cups that I will drink on any given day?

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- ▶ e.g., if X is a Poisson random variable, what is $\mathbb{E}[X]$?

[Got to here in class...]

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- ▶ e.g., if $Y = aX + b$ for any scalars a and b , what is $\mathbb{E}[Y]$?

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- ▶ **Variance** of X measures the **dispersion** of X around $\mathbb{E}[X]$:

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- ▶ **Standard deviation** of X is

$$\sigma_X = \sqrt{\text{var}(X)}$$

Examples of Variance

- ▶ e.g., on any given day, with probability 0.2, I am not especially tired and will drink only 3 cups of coffee before work. With probability 0.8, however, I am VERY tired and will drink 6 cups of coffee before work. Let X be the number of cups that I will drink. $\mathbb{E}[X] = 5.4$. What is $\text{var}(X)$?

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- ▶ e.g., $\text{var}(X)$ and σ_X for a Poisson random variable?

Variance of a Linear Function of a Random Variable

- ▶ Let $Y = aX + b$ for any random variable X and scalars a and b . $\mathbb{E}[Y] = a\mathbb{E}[X] + b$. What is $\text{var}(Y)$? What about σ_Y ?

Multiple Random Variables

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- ▶ e.g., choosing a random faculty member in the department: X is the person's height, Y is the person's weight

Tabular Representation

	y_1	y_2	y_3	y_4
x_1	0.1	0.1	0	0.2
x_2	0.05	0.05	0.1	0
x_3	0	0.1	0.2	0.1

- ▶ e.g., $p_{X,Y}(x_2, y_3) = 0.1$, $p_{X,Y}(x_3, y_1) = 0$, ...

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- ▶ e.g., $p_{X,Y}(x_2, y_3) = 0.1$, $p_{X,Y}(x_3, y_1) = 0$, ...
- ▶ Is this a valid joint PMF for X and Y ? How do we know?

Marginal PMFs

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x_2	0.05	0.05	0.1	0
x_3	0	0.1	0.2	0.1

- ▶ We can compute the PMFs of X and Y from $p_{X,Y}$

$$p_X(x) = \sum_y p_{X,Y}(x,y) \text{ and } p_Y(y) = \sum_x p_{X,Y}(x,y)$$

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- ▶ e.g., what is $\mathbb{E}[aX + bY + c]$ for any scalars a , b , and c ?

Three or More Random Variables

- ▶ The joint PMF of X , Y , and Z is denoted $p_{X,Y,Z}$ where

$$p_{X,Y,Z}(x, y, z) = P(\{X=x\} \cap \{Y=y\} \cap \{Z=z\})$$

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- ▶ $\mathbb{E}[aX + bY + cZ + d] = a\mathbb{E}[X] + b\mathbb{E}[Y] + c\mathbb{E}[Z] + d$

Examples of Three or More Random Variables

- ▶ e.g., an office party decides to do a gift exchange for their annual holiday party. Everyone in the office purchases a gift and places it in a box. Each person is then given a random gift from the box. What is the expected number of people who get back the gift that they purchased?

Expectations of Standard Random Variables

- ▶ e.g., if X is a binomial random variable, what is $\mathbb{E}[X]$?

For Next Time

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