# CMPSCI 240: "Reasoning Under Uncertainty" Lecture 7 

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## Reminders

- Check the course website: http://www.cs.umass.edu/ ~wallach/courses/s12/cmpsci240/
- Second homework is due on Friday

Recap

## Last Time: Common Discrete Random Variables

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- Geometric: $p_{X}(k)=(1-p)^{k-1} p$ for $k=1,2,3, \ldots$
- Poisson: $p_{X}(k)=e^{-\lambda} \lambda^{k} / k$ ! for $k=0,1,2, \ldots$


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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" $p_{X}$


## Last Time: Expectations of Standard Random Variables

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- Poisson random variable $X: \mathbb{E}[X]=\lambda$


## Expectation (Cont.)

## Expected Value Rule and Linearity of Expectation

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- Can compute $\mathbb{E}[Y]$ without computing $p_{Y}$ directly!
- e.g., if $Y=a X+b$ for any scalars $a$ and $b$, what is $\mathbb{E}[Y]$ ?


## Examples of Linearity of Expectation

- e.g., Amherst's temperature is modeled by a random variable $X$ with mean $\mathbb{E}[X]$ equal to 10 degrees Celsius. What is the mean temperature expressed in degrees Farenheit?

Variance

## Variance

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

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\operatorname{var}(X)=\underbrace{\mathbb{E}\left[(X-\mathbb{E}[X])^{2}\right]}_{\geq 0}=\mathbb{E}\left[X^{2}\right]-\mathbb{E}[X]^{2}
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- Standard deviation of $X$ is

$$
\sigma_{X}=\sqrt{\operatorname{var}(X)}
$$

## 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 $\operatorname{var}(X)$ ?


## Variance of Standard Random Variables

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- e.g., $\operatorname{var}(X)$ and $\sigma_{X}$ for a Bernoulli random variable?
- e.g., $\operatorname{var}(X)$ and $\sigma_{X}$ for a Poisson random variable?


## Variance of a Linear Function of a Random Variable

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


## Examples of Linearity of Variance

- e.g., Amherst's temperature is modeled by a random variable $X$ with mean $\mathbb{E}[X]$ and standard deviation $\sigma_{X}$ both equal to 10 degrees Celsius. A day is "typical" if it is within one standard deviation of the mean. What is the temperature range for a "typical" day expressed in degrees Farenheit?


## 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., $p_{X, Y}\left(x_{2}, y_{3}\right)=0.1, p_{X, Y}\left(x_{3}, y_{1}\right)=0, \ldots$


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- e.g., $p_{X, Y}\left(x_{2}, y_{3}\right)=0.1, p_{X, Y}\left(x_{3}, y_{1}\right)=0, \ldots$
- Is this a valid joint PMF for $X$ and $Y$ ? How do we know?


## Marginal PMFs

|  | $y_{1}$ | $y_{2}$ | $y_{3}$ | $y_{4}$ |
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| $x_{2}$ | 0.05 | 0.05 | 0.1 | 0 |
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- 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 $\mathbb{E}[a X+b Y+c]$ 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

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p_{X, Y, Z}(x, y, z)=P(\{X=x\} \cap\{Y=y\} \cap\{Z=z\})
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- $\mathbb{E}[a X+b Y+c Z+d]=a \mathbb{E}[X]+b \mathbb{E}[Y]+c \mathbb{E}[Z]+d$
[Got to here in class...]


## 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 drawn 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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