



Probability in Computing

CS
237

Reminders

- HW 7 is due Thursday

Reading

- LLM 19.4-19.5
P 3.2.2

LECTURE 16

Last time

- Expected value of a random variable

Today

- Expectation of continuous random variables
- Linearity of expectation

Reminder: expectation

Definition: Expectation

The **expectation** (also called the **expected value** or **mean**) of a discrete random variable X over a sample space Ω is

$$\mathbb{E}(X) = \sum_{\omega \in \Omega} X(\omega) \cdot \Pr(\omega).$$

$$\mathbb{E}(X) = \sum_{a \in \text{Range}(X)} a \cdot \Pr(X = a).$$

This version of the definition is more useful for computations.

Expectation for Continuous RVs

- The expectation of a continuous random variable is defined as in the discrete case, but we replace the sum with an integral
- Let X be a random variable with PDF f

discrete:

$$\mathbb{E}(X) = \sum_{x \in \mathbb{R}} x \cdot f(x)$$

continuous:

$$\mathbb{E}(X) = \int_{-\infty}^{\infty} x \cdot f(x) dx$$

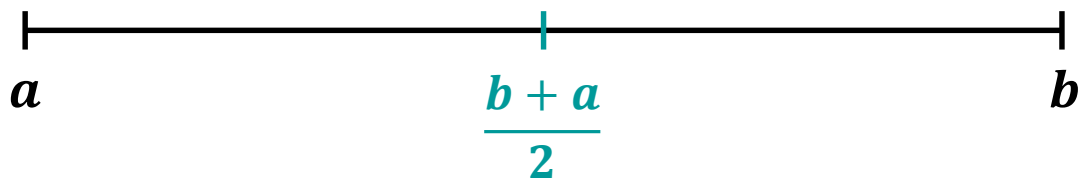
Example: Uniform Random Variable

- Suppose X is chosen uniformly at random from the interval $[a, b]$
- The PDF is $f(x) = \frac{1}{b-a}$ for $x \in [a, b]$

- $$\mathbb{E}(X) = \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_a^b x \cdot \frac{1}{b-a} dx = \frac{1}{b-a} \int_a^b x dx$$

$$= \frac{1}{b-a} \left. \frac{x^2}{2} \right|_a^b = \frac{1}{b-a} \frac{b^2 - a^2}{2} = \frac{b+a}{2}$$

- **Note:** $\mathbb{E}(X)$ is the midpoint of the interval $[a, b]$



Example: Darts

- Suppose we throw a dart at a circular board of radius r
- Let D be the distance from the center of the board to the hitting point of the dart
- The PDF is $f(x) = \frac{2x}{r^2}$ for $x \in [0, r]$

$$\begin{aligned} \mathbb{E}(D) &= \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_0^r x \cdot \frac{2x}{r^2} dx = \frac{2}{r^2} \int_0^r x^2 dx \\ &= \frac{2}{r^2} \left. \frac{x^3}{3} \right|_0^r = \frac{2}{r^2} \frac{r^3}{3} = \frac{2r}{3} \end{aligned}$$

Linearity of Expectation

Theorem. For any two random variables X and Y on the same probability space,

$$\mathbb{E}(X + Y) = \mathbb{E}(X) + \mathbb{E}(Y).$$

Also, for all $c \in \mathbb{R}$,

$$\mathbb{E}(cX) = c \cdot \mathbb{E}(X).$$

- More generally, for all $a, b \in \mathbb{R}$,

$$\mathbb{E}(aX + bY) = a \cdot \mathbb{E}(X) + b \cdot \mathbb{E}(Y).$$

- And for multiple random variables X_1, \dots, X_n and numbers $c_1, \dots, c_n \in \mathbb{R}$,

$$\begin{aligned} \mathbb{E}(c_1X_1 + c_2X_2 + \dots + c_nX_n) \\ = c_1\mathbb{E}(X_1) + c_2\mathbb{E}(X_2) + \dots + c_n\mathbb{E}(X_n) \end{aligned}$$

Linearity of expectation holds even for dependent random variables!

Indicator Random Variables

- An **indicator random variable** takes on two values: 0 and 1.
- **Lemma.** For an indicator random variable Y ,

$$\mathbb{E}(Y) = \Pr(Y = 1).$$

Proof:

An **indicator random variable** Y for an event E is

$$Y = \begin{cases} 1 & \text{if } E \text{ occurs} \\ 0 & \text{otherwise} \end{cases}$$

Method for computing expectation of a random variable X :
Represent X as a sum of indicators and use linearity of expectation.

You have a coin that shows HEADS with probability $3/4$.
Let X be the number of HEADS in 1000 tosses of your coin.
You represent X as the sum: $X = X_1 + X_2 + \cdots + X_{1000}$.

What is X_1 ?

- A. $3/4$.
- B. The number of HEADS.
- C. The probability of HEADS in toss 1.
- D. The number of heads in toss 1.
- E. None of the above.

You have a coin that shows HEADS with probability $3/4$.
Let X be the number of HEADS in 1000 tosses of your coin.

What is the expectation of X ?

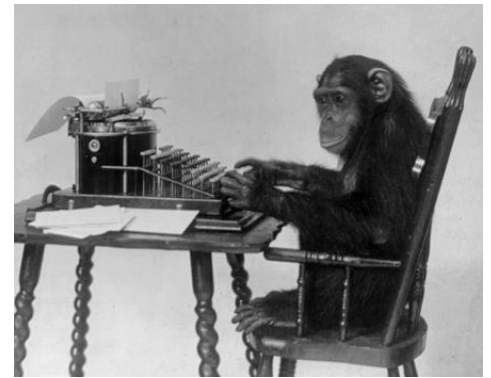
- A. $3/4$.
- B. $4/3$.
- C. 500.
- D. 750.
- E. None of the above.

Example: Random Proofs

- **Infinite Monkey Theorem.** Given enough time, a hypothetical chimpanzee typing at random would, as part of its output, almost surely produce all of Shakespeare's plays (or any other text.)

https://en.wikipedia.org/wiki/Infinite_monkey_theorem

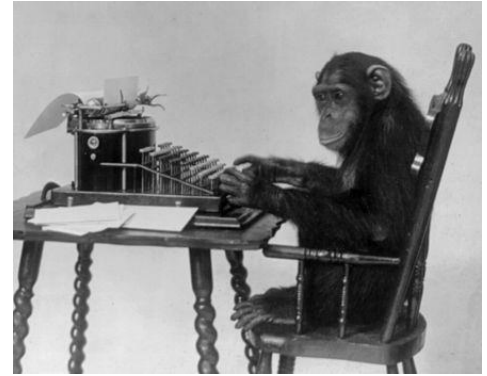
- Suppose that the monkey types on a 26-letter keyboard
- Each letter is chosen independently and uniformly at random from the 26 letters
- Suppose the monkey types $n = 1,000,000$ letters
- Let X = number of times the sequence "proof" appears
- What is $\mathbb{E}(X)$?



Example: Random Proofs

- Suppose the monkey types $n = 1,000,000$ letters
- Let $X =$ number of times the sequence “proof” appears
- What is $\mathbb{E}(X)$?

Solution: For each position i in the sequence, let



- **Example:** permutations
 - 3 students exchange their hats, so that everybody gets a random hat
 - Let X be the number of students that got their own hats.
 - E.g., if students 1,2,3 got hats 2,1,3 then $X=1$.

- PDF of X :

$$\Pr(X = 0) = \frac{1}{3}, \Pr(X = 1) = \frac{1}{2}, \Pr(X = 3) = \frac{1}{6}.$$

- **What's the expectation of X ?**

- A. 3/4
- B. 1
- C. 3/2
- D. None of the above.

Example: Random Hats

- **Example:** permutations
 - n students exchange their hats, so that everybody gets a random hat
 - Let X be the number of students that got their own hats.
- What's the expectation of X for general n ?

- A random variable can be a function of another random variable
- **Examples:** $Y = 3X + 1$; $Y = X^2$; $Y = |X|$
- If Y is a function of X then PDF of Y is related to the PDF of X
- **Law of the Unconscious Statistician (LOTUS):** If $Y = g(X)$ then

discrete:
$$\mathbb{E}(Y) = \mathbb{E}(g(X)) = \sum_{x \in \mathbb{R}} g(x) \cdot f_X(x)$$

continuous:
$$\mathbb{E}(Y) = \mathbb{E}(g(X)) = \int_{-\infty}^{\infty} g(x) \cdot f_X(x) dx$$

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- Example 1: Let X be uniform over $\{-4, \dots, 4\}$. Then PDF of X is

$$f_X(x) = \begin{cases} \frac{1}{9} & \text{if } x \in \{-4, \dots, 4\} \\ 0 & \text{otherwise} \end{cases}$$

$$\mathbb{E}(X^2) = \sum_{x \in \mathbb{R}} x^2 \cdot f_X(x) = \sum_{x \in \{-4, \dots, 4\}} x^2 \cdot \frac{1}{9} = \frac{2}{9} (1^2 + 2^2 + 3^2 + 4^2)$$

- **Law of the Unconscious Statistician (LOTUS):** If $Y = g(X)$ then

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- **Example 2:** Let X be uniform over $[0,1]$. Then PDF of X is

$$f_X(x) = \begin{cases} 1 & \text{if } x \in [0,1] \\ 0 & \text{otherwise} \end{cases}$$

$$\mathbb{E}(X^2) = \int_{-\infty}^{\infty} x^2 \cdot f_X(x) dx = \int_0^1 x^2 dx = \left. \frac{x^3}{3} \right|_0^1 = \frac{1}{3}$$