Linearity of Expectation CSE 312 Winter 25 Lecture 11

Outline

Linearity of expectation Statement Proof

A whole bunch of examples

Expectation

Expectation

The "expectation" (or "expected value") of a random variable X is:

$$\mathbb{E}[X] = \sum_{k \in \Omega_X} k \cdot \mathbb{P}(X = k)$$

$$\mathbb{E}[X] = \sum_{\omega \in \Omega} X(\omega) \cdot \mathbb{P}(\omega)$$

Intuition: The weighted average of values X could take on.

Weighted by the probability you actually see them.

Linearity of Expectation

Linearity of Expectation

For any two random variables *X* and *Y*:

$$\mathbb{E}[X+Y] = \mathbb{E}[X] + \mathbb{E}[Y]$$

Note: *X* and *Y* do not have to be independent

Linearity of Expectation

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Extending this to n random variables,
$$X_1, X_2, ..., X_n$$

$$\mathbb{E}[X_1 + X_2 + \cdots + X_n] = \mathbb{E}[X_1] + \mathbb{E}[X_2] + \cdots + \mathbb{E}[X_n]$$

This can be proven by induction.

Linearity of Expectation - Proof

Linearity of Expectation

For any two random variables X and Y: $\mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y]$

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Proof:

$$\begin{split} \underline{\mathbb{E}[X+Y]} &= \Sigma_{\omega \in \Omega} \mathbb{P}(\omega) \big(X(\omega) + Y(\omega) \big) \\ &= \Sigma_{\omega \in \Omega} \mathbb{P}(\omega) X(\omega) + \mathbb{P}(\omega) Y(\omega) \\ &= \Sigma_{\omega \in \Omega} \mathbb{P}(\omega) X(\omega) + \Sigma_{\omega \in \Omega} \mathbb{P}(\omega) Y(\omega) \\ &= \underline{\mathbb{E}[X]} + \underline{\mathbb{E}[Y]} \end{split}$$

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Note: X and Y do not have to be independent

Constants are also fine:

For real numbers a, b, c

$$\mathbb{E}[aX + bY + c] = \mathbb{E}[aX] + \mathbb{E}[bY + c]$$
$$= a\mathbb{E}[X] + b\mathbb{E}[Y] + c$$

Say you and your friend go fishing everyday.

- You catch X fish, with $\mathbb{E}[X] = 3$
- Your friend catches Y fish, with $\mathbb{E}[Y] = 7$

How many fish do both of you bring on an average day?

Z = X + Y

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You can sell each for \$10 per fish, but you need \$15 (total) for expenses.
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 What is your average profit?

$$\mathbb{E}[10Z - 15] = 10\mathbb{E}[Z] - 15 = 100 - 15 \neq 85$$

Coin Tosses

If we flip a coin twice, what is the expected number of heads that come up?

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Let Y be the r.v. representing the total number of heads

$$p_{Y}(y) = \begin{cases} \frac{1}{4} & \text{if } y = 0\\ \frac{1}{2} & \text{if } y = 1\\ \frac{1}{4} & \text{if } y = 2\\ 0 & \text{otherwise} \end{cases}$$

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$$\mathbb{E}[Y] = \sum_{k \in \Omega_{Y}} p_{Y}(k) \cdot k = \frac{1}{4} \cdot 0 + \frac{1}{2} \cdot 1 + \frac{1}{4} \cdot 2 = 1$$



Now what if the probability of flipping a head was p and that we wanted to find the total number of heads flipped when we flip the coin n times?

Let X be the r.v. representing the total number of heads.

Make a prediction --- what should $\mathbb{E}[X]$ be?

Now what if the probability of flipping a head was p and that we wanted to find the total number of heads flipped when we flip the coin n times?

Let X be the r.v. representing the total number of heads.

$$\mathbb{E}[X] = \sum_{k=0}^{n} k \cdot \mathbb{P}(Y = k) = \sum_{k=0}^{n} k \cdot \binom{n}{k} p^{k} (1-p)^{n-k}$$

Ok, but what actually is it? I don't have intuition for this formula.

Now what if the probability of flipping a head was p and that we wanted to find the total number of heads flipped when we flip the coin n times?

$$\mathbb{E}[X] = \sum_{k=0}^{n} k \cdot \mathbb{P}(Y = k) = \sum_{k=0}^{n} k \cdot \binom{n}{k} p^{k} (1 - p)^{n-k}$$

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

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

$$= np \sum_{i=0}^{n-1} \binom{n-1}{i} p^{i} (1 - p)^{n-1-i}$$

$$= np (p + (1 - p))^{n-1} = np$$
Tem!

Binomial Theorem!

We did it! And all it took was a clever application of the binomial theorem, setup by a very non-obvious application of an obscure combinatorial identity. Ezpz.

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To find the total number of heads flipped when we flip the com
$$n$$
 times?

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Linearity of Expectation

Linearity of Expectation

For any two random variables X and Y: $\mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y]$

Note: *X* and *Y* do not have to be independent

Extending this to n random variables,
$$X_1, X_2, ..., X_n$$

$$\mathbb{E}[X_1 + X_2 + \cdots + X_n] = \mathbb{E}[X_1] + \mathbb{E}[X_2] + \cdots + \mathbb{E}[X_n]$$

Indicator Random Variables

For any event A_{ι} we can define the indicator random variable $\mathbf{1}[A]$ for A

$$\mathbf{1}[A] = \begin{cases} 1 & \text{if event A occurs} \\ 0 & \text{otherwise} \end{cases}$$

$$\mathbb{P}(X = 1) = \mathbb{P}(A)$$
$$\mathbb{P}(X = 0) = 1 - \mathbb{P}(A)$$

$$p_X(x) = \begin{cases} \mathbb{P}(A) & \text{if } x = 1\\ 1 - \mathbb{P}(A) & \text{if } x = 0\\ 0 & \text{otherwise} \end{cases}$$



$$\mathbb{E}[X]$$

$$= 1 \cdot p_X(1) + 0 \cdot p_X(0)$$

$$= p_X(1) = \mathbb{P}(A)$$

Repeated Coin Tosses (Again)

The probability of flipping a head is p and we want to find the total number of heads flipped when we flip the coin n times?

Let *X* be the total number of heads

What indicators can we define? What 'Booleans' have enough information to combine (add) and solve the problem?

Repeated Coin Tosses (Again)

The probability of flipping a head is p and we want to find the total number of heads flipped when we flip the coin n times?

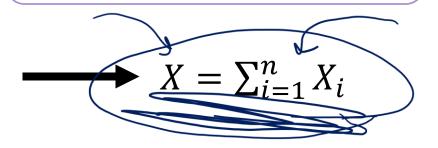
Let *X* be the total number of heads

Define X_i as follows:

$$X_i = \begin{cases} 1 \\ 0 \end{cases}$$

if the ith coin flip is heads otherwise

$$\mathbb{P}(X_i = 1) = p$$
$$\mathbb{P}(X_i = 0) = 1 - p$$



$$\mathbb{E}[X_i] = 1 \cdot p + 0 \cdot (1 - p) = p$$

Repeated Coin Tosses (Again)

The probability of flipping a head is p and we want to find the total number of heads flipped when we flip the coin n times?

Let *X* be the total number of heads

$$\mathbb{E}[X] = \mathbb{E}[\sum_{i=1}^{n} X_i]$$

$$= \mathbb{E}[X_1 + X_2 + \dots + X_n]$$

$$= \mathbb{E}[X_1] + \mathbb{E}[X_2] + \dots + \mathbb{E}[X_n]$$

$$= \sum_{i=1}^{n} \mathbb{E}[X_i]$$

$$= \sum_{i=1}^{n} p \neq np$$

$$X = \sum_{i=1}^{n} X_i$$
$$\mathbb{E}[X_i] = p$$



Computing complicated expectations

We often use these three steps to solve complicated expectations

1. <u>Decompose</u>: Finding the right way to decompose the random variable into sum of simple random variables

$$X = X_1 + X_2 + \dots + X_n$$

- 2. LOE: Apply Linearity of Expectation $\mathbb{E}[X] = \mathbb{E}[X_1] + \mathbb{E}[X_2] + \cdots + \mathbb{E}[X_n]$
- 3. Conquer: Compute the expectation of each X_i

Often X_i are indicator random variables

2 Pairs

In a class of m students, on average how many pairs of people have the same birthday?

Decompose:

LOE:

In a class of m students, on average how many pairs of people have the same birthday?

<u>Decompose</u>: Let *X* be the number of pairs with the same birthday

Define X_{ij} as follows:

$$X_{ij} = \begin{cases} 1 \\ 0 \end{cases}$$

if person i, j have the same bithday otherwise

$$X = \sum_{i,j} X_{ij}$$

<u>LOE:</u>

<u>Conquer:</u>

In a class of m students, on average how many pairs of people have the same birthday?

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Define X_{ij} as follows:

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$$X = \Sigma_{i,j} X_{ij}$$

LOE:

$$\mathbb{E}[X] = \mathbb{E}[\Sigma_{i,j}X_{ij}] = \Sigma_{i,j}\mathbb{E}[X_{ij}]$$

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LOE:

$$\mathbb{E}[X] = \mathbb{E}[\Sigma_{i,j}X_{ij}] = \Sigma_{i,j}\mathbb{E}[X_{ij}]$$

$$\mathbb{E}[X_{ij}] = \mathbb{P}(X_{ij} = 1) = \frac{365}{365 \cdot 365} = \frac{1}{365}$$

$$\mathbb{E}[X] = {m \choose 2} \cdot \mathbb{E}[X_{ij}] = {m \choose 2} \cdot \frac{1}{365}$$

Rotating the table

n people are sitting around a circular table. There is a name tag in each place. Nobody is sitting in front of their own name tag.

Rotate the table by a random number k of positions between 1 and n-1 (equally likely)

Let X be the number of people that end up in front of their own name tag. Find $\mathbb{E}[X]$.

Decompose:

What X_i can we define that have the needed information?

LOE:

Rotating the table

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Rotate the table by a random number k of positions between 1 and n-1 (equally likely)

X is the number of people that end up in front of their own name tag. Find $\mathbb{E}[X]$.

<u>Decompose:</u> Define X_i as follows:

$$X_i = \begin{cases} 1 \\ 0 \\ \text{Note: } X = \sum_{i=1}^n X_i \end{cases}$$

if person i sits infront of their own name tag otherwise

LOE:

$$\mathbb{E}[X] = \mathbb{E}[\Sigma_{i=1}^n X_i] = \Sigma_{i=1}^n \mathbb{E}[X_i]$$

Conquer:

These X_i are not independent! That's ok!!

Rotating the table

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<u>Decompose:</u> Define X_i as follows:

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LOE:

$$\mathbb{E}[X] = \mathbb{E}[\Sigma_{i=1}^{n} X_i] = \Sigma_{i=1}^{n} \mathbb{E}[X_i]$$

 $X = \sum_{i=1}^{n} X_i$

$$\mathbb{E}[X_i] = P(X_i = 1) = \frac{1}{n-1}$$

$$\mathbb{E}[X] = n \cdot \mathbb{E}[X_i] = \frac{n}{n-1}$$

Extra Practice

Frogger



A frog starts on a 1-dimensional number line at 0.

Each second, independently, the frog takes a unit step right with probability p_1 , to the left with probability p_2 , and doesn't move with probability p_3 , where $p_1 + p_2 + p_3 = 1$.

After 2 seconds, let X be the location of the frog. Find $\mathbb{E}[X]$.

Frogger – Brute Force



A frog starts on a 1-dimensional number line at 0. At each second, independently, the frog takes a unit step right with probability p_R , to the left with probability p_L , and doesn't move with probability p_S , where $p_L + p_R + p_S = 1$. After 2 seconds, let X be the location of the frog. Find $\mathbb{E}[X]$.

We could find the PMF by computing the probability for each value in the range of X, and then applying definition of expectation:

$$p_X(x) = egin{cases} p_L^2 \ 2p_L p_S \ 2p_L p_R + p_S^2 \ 2p_R p_S \ p_R^2 \ 0 \end{cases}$$

x = 2otherwise

We think about the outcomes that correspond to each value of X and compute the probability of that. For x = 0 example, X=0 happens when the frog is at the same x = 1 position after 2 sec – this means it either moved left and then right, or right and then left, or did not move both seconds.

$$\mathbb{E}[X] = \Sigma_{\omega} P(\omega) X(\omega) = (-2) p_L^2 + (-1) 2 p_L p_S + 0 \cdot (2 p_L p_R + p_S^2) + (1) 2 p_R p_S + (2) p_R^2 = 2 (p_R - p_L)$$

Frogger – LOE



Or we can apply LoE!

A frog starts on a 1-dimensional number line at 0. At each second, independently, the frog takes a unit step right with probability p_R , to the left with probability p_L , and doesn't move with probability p_S , where $p_L + p_R + p_S = 1$. After 2 seconds, let X be the location of the frog. **Find** $\mathbb{E}[X]$.

Define X_i as follows:

$$X_i = \begin{cases} -1 & \text{if the frog moved left on the } i\text{th step} \\ 0 & \text{otherwise} \\ 1 & \text{if the frog moved right on the } i\text{th step} \end{cases}$$

$$\mathbb{E}[X_i] = -1 \cdot p_L + 1 \cdot p_R + 0 \cdot p_S = (p_R - p_L)$$

By Linearity of Expectation,

$$\mathbb{E}[X] = \mathbb{E}[\sum_{i=1}^{2} X_i] = \sum_{i=1}^{2} \mathbb{E}[X_i] = 2(p_R - p_L)$$

Frogger – LOE

If we interested in a whole minute (60 sec), the first approach would be awful because we would need to compute many probabilities or deal with a gnarly summation! Instead, we can use LoE!

A frog starts on a 1-dimensional number line at 0. At each second, independently, the frog takes a unit step right with probability p_R , to the left with probability p_L , and doesn't move with probability p_S , where $p_L + p_R + p_S = 1$. After 60 seconds, let X be the location of the frog. Find $\mathbb{E}[X]$.

Define X_i as follows:

$$X_i = \begin{cases} -1 & \text{if the frog moved left on the } i\text{th step} \\ 0 & \text{otherwise} \\ 1 & \text{if the frog moved right on the } i\text{th step} \end{cases}$$

$$\mathbb{E}[X_i] = -1 \cdot p_L + 1 \cdot p_R + 0 \cdot p_S = (p_R - p_L)$$

By Linearity of Expectation,

$$\mathbb{E}[X] = \mathbb{E}\left[\sum_{i=1}^{60} X_i\right] = \sum_{i=1}^{60} \mathbb{E}[X_i] = 60(p_R - p_L)$$