

Linearity of Expectation: Examples

Question: Suppose we ~~flip~~^{have} a coin that has probability $0 \leq p \leq 1$ of landing H and prob. $1-p$ of landing tails. If we flip this coin n times, what is the expected number of heads?

Recall: $E[X_1 + X_2 + \dots + X_n] = E[X_1] + \dots + E[X_n]$.

Soln: Let X be the number of heads. How should we choose X_1, X_2, \dots, X_n so that

$$\forall \omega \in \Omega \quad X(\omega) = X_1(\omega) + \dots + X_n(\omega)?$$

Note

$$X_k = \begin{cases} 1 & \text{kth flip is H} \\ 0 & \text{otherwise} \end{cases}$$

does the trick.

Therefore

$$E[X] = E[X_1] + \dots + E[X_n]$$

$$= p + \dots + p$$

$$= np$$

because, by defn $E[X_k] = 0 \cdot \Pr(X_k=0) + 1 \cdot \Pr(X_k=1) = p$

Question: Using the same coin, what is the expected number of runs of heads? (A run of heads is a contiguous block of 1 or more heads; THTTHHTHHHTH has 4 runs of heads.) For which probability p do we maximize this ~~prob~~ expectation?

Soln: Let X be the number of runs of heads.

Note that for each $\omega \in \Omega$, $X(\omega)$ is the

number of times that we flip the sequence HT ,³
plus 1 if the last flip is H . For example

\underline{HTTHAT}
 $T\underline{HTTHHTHHHTTHH}$

Let For $1 \leq k \leq n-1$, let

$$X_k = \begin{cases} 1 & \text{kth flip is H and} \\ & \text{(k+1)th flip is T} \\ 0 & \text{otherwise} \end{cases}$$

$$X_n = \begin{cases} 1 & \text{nth flip is H} \\ 0 & \text{otherwise} \end{cases}$$

and note $\forall \omega \in \Omega$ $X(\omega) = X_1(\omega) + \dots + X_n(\omega)$.

Therefore

$$\begin{aligned} E[X] &= \sum_{k=1}^{n-1} E[X_k] + E[X_n] \\ &= (n-1) \Pr(X_k=1) + \Pr(X_n=1) \\ &= (n-1) p(1-p) + p \end{aligned}$$

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Exercise: Check that this answer ^{is correct} ~~not~~ ~~same~~ for small values of n and $p \in \{0, \frac{1}{2}, 1\}$.

Another form of checking our answer is to compute the probability that maximizes the runs of heads.

Our intuition says that p should be roughly $\frac{1}{2}$.

Using calculus, we know $E[X]$ is maximized when $p=0$, $p=1$, or $\frac{d}{dp} E[X] = 0$.

Note

$$\begin{aligned}\frac{d}{dp} E[X] &= \frac{d}{dp} [(n-1)p(1-p) + p] \\ &= (n-1)(1-p - p) + 1 \\ &= (n-1)(1-2p) + 1\end{aligned}$$

and then using algebra, we see that

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 $(n-1)(1-2p)+1=0$ requires

$$1-2p = \frac{-1}{n-1}$$

$$2p-1 = \frac{1}{n-1}$$

$$2p = \frac{n}{n-1}$$

$$p = \frac{n}{2(n-1)}$$

for $n > 1$. Clearly the expected number of runs is not maximized at $p=0$ or $p=1$ (for large n), so $p = \frac{n}{2(n-1)}$ maximizes $E[X]$.

Note $\lim_{n \rightarrow \infty} \frac{n}{2(n-1)} = \frac{1}{2}$ confirming our intuition.

Also note $\forall n > 1 \quad \frac{n}{2(n-1)} > \frac{1}{2}$, so we want a slight bias for the coin to land **H**. This also makes sense: we want the last flip to be **H**.

Conditional Expectation

def Let X be a r.v. and let A be an event. The expectation of X given A is

$$E[X|A] = \sum_a a \cdot \Pr(X=a | A).$$

• Similarly to the method of conditional probabilities, we have another tool for computing $E[X]$:

Prop Let X be a r.v. and suppose

A_1, A_2, \dots, A_n are events which partition

the probability space (i.e. $\forall i \neq j, A_i \cap A_j = \emptyset$

so the A_j are pairwise disjoint, and

$\bigcup_{j=1}^n A_j = \Omega$). Then

$$E[X] = \sum_{j=1}^n E[X|A_j] \cdot \Pr(A_j)$$

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Pr: $\sum_{j=1}^n E[X|A_j] \cdot \Pr(A_j) = \sum_{j=1}^n \left(\sum_a a \cdot \Pr(X=a|A_j) \right) \cdot \Pr(A_j)$

$$= \sum_{j=1}^n \left(\sum_a a \cdot \frac{\Pr(X=a \cap A_j)}{\Pr(A_j)} \right) \cdot \Pr(A_j)$$

$$= \sum_{j=1}^n \sum_a a \cdot \Pr(X=a \cap A_j)$$

$$= \sum_a a \left(\sum_{j=1}^n \Pr(X=a \cap A_j) \right)$$

$$= \sum_a a \cdot \Pr(X=a)$$

$$= E[X]$$



Facts about Conditional Expectation

~~pf~~

Prop If $\forall a (X=a)$ and A are independent events, then $E[X|A] = E[X]$.

• Also, Linearity of Expectation still holds:

Prop: $E[X+Y|A] = E[X|A] + E[Y|A]$

Cor: $E[X_1 + X_2 + \dots + X_n | A] = E[X_1 | A] + \dots + E[X_n | A]$

• The proofs are left as exercises.

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Application: Let us revisit our coin which has prob. p of landing H and $\overbrace{1-p}^{\text{prob}}$ of landing T .

What is the expected number of flips ~~before~~^{required} ~~we~~^{to} see the coin land H ?

- Direct computation is possible but involves a sum we ~~don't~~ have not seen before.
- Linearity of expectation will give us a geometric series, which we know how to solve.
(Exercise: ~~compute~~ solve this problem ~~with~~ using Linearity of Expectation.)
- The method of conditional expectations ~~is~~ makes the problem even easier.
- Our ~~sample~~ sample space Ω contains all infinite sequences of heads and tails.

- Let X be the the smallest number r so that the r th flip is H . Let A be the event that the first flip is H and let $B = \bar{A}$ = the event that first flip is T .
- By the method of conditional expectations,

$$E[X] = E[X|A] \cdot Pr(A) + E[X|B] \cdot Pr(B).$$

- Clearly, $Pr(A) = p$ and $Pr(B) = 1 - p$. Also, if we are told that the first flip is H , then the number of flips is 1 (i.e.

$$\omega \in A \implies X(\omega) = 1)$$

and hence $E[X|A] = 1$.

- What is $E[X|B]$? Let Y be the number of flips needed to see an H , starting from the second flip.

10.1

$$\Pr(X=a|A) = \begin{cases} 1 & a=1 \\ 0 & \text{otherwise} \end{cases}$$

$$E[X|A] = \sum_a a \cdot \Pr(X=a|A)$$

$$= \sum_a a \cdot \begin{cases} 1 & a=1 \\ 0 & \text{otherwise} \end{cases}$$

$$= 1$$

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Ex: $Y(\underline{HTTHTTTH}\dots) = 3$

$Y(\underline{THHT}\dots) = 1$

Note that

$$\omega \in B \implies X(\omega) = 1 + Y(\omega)$$

So Linearity of Expectation ~~gives~~ — applied
in the probability space $\Pr(\cdot | B)$ — gives
us

$$\begin{aligned} E[X|B] &= E[1|B] + E[Y|B] \\ &= 1 + E[Y] \\ &= 1 + E[X] \end{aligned}$$

because the event B that the first flip is T
is independent of the value of Y (i.e. $\forall a$,
 $(Y=a)$ and B are independent events) so that
 $E[Y|B] = E[Y]$. (Clearly $E[X] = E[Y]$.)

• Therefore

$$\begin{aligned}
 E[X] &= E[X|A] \cdot Pr(A) + E[X|B] \cdot Pr(B) \\
 &= 1 \cdot p + (1+E[X]) \cdot (1-p) \\
 &= 1 + (1-p)E[X]
 \end{aligned}$$

and solving for $E[X]$ gives $E[X] = \frac{1}{p}$.

Coupon Collection

- Each time you visit a store, the store gives you a coupon randomly chosen from n kinds of coupons.
- What is the expected number of visits to the store that are required to obtain each type of coupon?

Let X be the number of visits needed.

- Our sample space Ω contains all sequences of coupon types:

$$\Omega = \{ (a_1, a_2, a_3, \dots) \mid \forall j, 1 \leq a_j \leq n \}$$

- Let X be the number of visits needed.

Ex: $n = 4$

$$X(\underline{1332124132 \dots}) = 7$$

$$X(\underline{142121212121 \dots}) = \infty$$

• How do we compute $E[X]$? Use Linearity of Expectation!

• For $1 \leq k \leq n$, let X_k be the number of visits required to see the k th new type of coupon, after seeing the $(k-1)$ th new coupon type.

Ex: $n=4$ $\omega = \underline{1332124132 \dots}$

$$X_1(\omega) = 1, X_2(\omega) = 1, X_3(\omega) = 2, X_4(\omega) = 3$$

• Note $\forall \omega \in \Omega \quad X(\omega) = X_1(\omega) + X_2(\omega) + \dots + X_n(\omega)$

• Therefore

$$E[X] = E[X_1] + \dots + E[X_n]$$

• What is $E[X_k]$? Well, if we've seen $k-1$ of the n coupon types and we're waiting for the k th ^{new} coupon type, the probability that a particular visit results in a new coupon type is $p = \frac{n-(k-1)}{n}$.

• Therefore $E[X_k] = \frac{1}{p} = \frac{n}{n-(k-1)} = \frac{n}{n-k+1}$.

• Hence

$$\begin{aligned}
 E[X] &= \sum_{k=1}^n E[X_k] \\
 &= \sum_{k=1}^n \frac{n}{n-k+1} \\
 &= \frac{n}{n} + \frac{n}{n-1} + \frac{n}{n-2} + \dots + \frac{n}{1} \\
 &= n \left(\frac{1}{n} + \frac{1}{n-1} + \dots + 1 \right) \\
 &= n H_n \approx n \ln n \\
 &= \Theta(n \log n)
 \end{aligned}$$