

1 Such High Expectations

Note 19

Suppose X and Y are independently drawn from a Geometric distribution with parameter p .

(a) Compute $\mathbb{E}[\min(X, Y)]$.

(b) Compute $\mathbb{E}[\max(X, Y)]$.

Solution:

(a) By independence,

$$\mathbb{P}[\min(X, Y) \geq t] = \mathbb{P}[X \geq t]\mathbb{P}[Y \geq t] = (1 - p)^{2(t-1)}.$$

By Tail Sum,

$$\mathbb{E}[\min(X, Y)] = \sum_{t=1}^{\infty} \mathbb{P}[\min(X, Y) \geq t] = \sum_{t=1}^{\infty} (1 - p)^{2(t-1)} = \frac{1}{1 - (1 - p)^2}.$$

(b) We see that

$$\begin{aligned} \mathbb{P}[\max(X, Y) \geq t] &= 1 - \mathbb{P}[\max(X, Y) < t] = 1 - \mathbb{P}[X < t]\mathbb{P}[Y < t] \\ &= 1 - (1 - \mathbb{P}[X \geq t])(1 - \mathbb{P}[Y \geq t]) \\ &= 1 - (1 - (1 - p)^{t-1})(1 - (1 - p)^{t-1}) \\ &= 1 - (1 - 2(1 - p)^{t-1} + (1 - p)^{2(t-1)}) \\ &= 2(1 - p)^{t-1} - (1 - p)^{2(t-1)}. \end{aligned}$$

Using the result from part (a),

$$\begin{aligned} \mathbb{E}[\max(X, Y)] &= \sum_{t=1}^{\infty} \mathbb{P}[\max(X, Y) \geq t] \\ &= \sum_{t=1}^{\infty} 2(1 - p)^{t-1} - (1 - p)^{2(t-1)} \\ &= \sum_{t=1}^{\infty} 2(1 - p)^{t-1} - \sum_{t=1}^{\infty} (1 - p)^{2(t-1)} \\ &= \frac{2}{p} - \frac{1}{1 - (1 - p)^2}. \end{aligned}$$

Alternate Solution: An extremely elegant one-liner with linearity:

$$\mathbb{E}[\max(X, Y)] = \mathbb{E}[X + Y - \min(X, Y)] = \mathbb{E}[X] + \mathbb{E}[Y] - \mathbb{E}[\min(X, Y)] = \frac{2}{p} - \frac{1}{1 - (1 - p)^2}.$$

2 Number Game

Note 20

Sinho and Vrettos are playing a game where they each choose an integer uniformly at random from $[0, 100]$, then whoever has the larger number wins (in the event of a tie, they replay). However, Vrettos doesn't like losing, so he's rigged his random number generator such that it instead picks randomly from the integers between Sinho's number and 100. Let S be Sinho's number and V be Vrettos' number.

- (a) What is $\mathbb{E}[S]$?
- (b) What is $\mathbb{E}[V \mid S = s]$, where s is any constant such that $0 \leq s \leq 100$?
- (c) What is $\mathbb{E}[V]$?

Solution:

- (a) S is a (discrete) uniform random variable between 0 and 100, so its expectation is $\frac{0+100}{2} = 50$.
- (b) If $S = s$, we know that V will be uniformly distributed between s and 100. Similar to the previous part, this gives us that $\mathbb{E}[V \mid S = s] = \frac{s+100}{2}$.
- (c) With the law of total expectation, we have that

$$\begin{aligned}\mathbb{E}[V] &= \sum_{s=0}^{100} \mathbb{E}[V \mid S = s] \cdot \mathbb{P}[S = s] \\ &= \sum_{s=0}^{100} \frac{s+100}{2} \cdot \frac{1}{101} \\ &= \frac{1}{202} \left(\sum_{s=0}^{100} s + \sum_{s=0}^{100} 100 \right)\end{aligned}$$

The first summation comes out to $\frac{100(100+1)}{2} = 50 \cdot 101$; the second summation is just adding 100 to itself 101 times, so it comes out to $100 \cdot 101$. Plugging these values in, we get $\mathbb{E}[V] = 75$.

3 Number of Ones

Note 20

In this problem, we will revisit dice-rolling, except with conditional expectation. (*Hint:* for both of these subparts, the law of total expectation may be helpful.)

- (a) If we roll a die until we see a 6, how many ones should we expect to see?
- (b) If we roll a die until we see a number greater than 3, how many ones should we expect to see?

Solution:

- (a) Let Y be the number of ones we see. Let X be the number of rolls we take until we get a 6.

Let us first compute $\mathbb{E}[Y \mid X = k]$. We know that in each of our $k - 1$ rolls before the k th, we necessarily roll a number in $\{1, 2, 3, 4, 5\}$. Thus, we have a $1/5$ chance of getting a one in each of these $k - 1$ previous rolls, giving

$$\mathbb{E}[Y \mid X = k] = \frac{1}{5}(k - 1).$$

If this is confusing, we can write Y as a sum of indicator variables, $Y = Y_1 + Y_2 + \cdots + Y_k$, where Y_i is 1 if we see a one on the i th roll. This means that by linearity of expectation,

$$\mathbb{E}[Y \mid X = k] = \mathbb{E}[Y_1 \mid X = k] + \mathbb{E}[Y_2 \mid X = k] + \cdots + \mathbb{E}[Y_k \mid X = k].$$

We know that on the k th roll, we must roll a 6, so $\mathbb{E}[Y_k] = 0$. Further, by symmetry, each term in this summation has the same value; this means that we have

$$\begin{aligned} \mathbb{E}[Y_1 \mid X = k] + \mathbb{E}[Y_2 \mid X = k] + \cdots + \mathbb{E}[Y_{k-1} \mid X = k] &= (k - 1) \mathbb{E}[Y_1 \mid X = k] \\ &= (k - 1) \mathbb{P}[Y_1 = 1 \mid X = k] \\ &= (k - 1) \frac{1}{5}. \end{aligned}$$

Using the law of total expectation, we now have

$$\begin{aligned} \mathbb{E}[Y] &= \sum_{k=1}^{\infty} \mathbb{E}[Y \mid X = k] \mathbb{P}[X = k] && \text{(total expectation)} \\ &= \sum_{k=1}^{\infty} \frac{1}{5}(k - 1) \mathbb{P}[X = k] \end{aligned}$$

Here, we can see that this is an application of LOTUS for $f(X) = \frac{1}{5}(X - 1)$, so we can simplify this to

$$\begin{aligned} &= \mathbb{E}\left[\frac{1}{5}(X - 1)\right] && \text{(LOTUS)} \\ &= \frac{1}{5}(\mathbb{E}[X] - 1) && \text{(linearity)} \end{aligned}$$

Since $X \sim \text{Geometric}(\frac{1}{6})$, the expected number of rolls until we roll a 6 is $\mathbb{E}[X] = 6$:

$$= \frac{1}{5}(6 - 1) = 1$$

Alternatively, we can use iterated expectation, along with the fact that $\mathbb{E}[Y \mid X] = \frac{1}{5}(X - 1)$, to give

$$\begin{aligned}\mathbb{E}[Y] &= \mathbb{E}[\mathbb{E}[Y \mid X]] \\ &= \mathbb{E}\left[\frac{1}{5}(X - 1)\right] \\ &= \frac{1}{5}(\mathbb{E}[X] - 1) \\ &= \frac{1}{5}(6 - 1) = 1\end{aligned}$$

- (b) We use the same logic as the first part, except now each of the first $k - 1$ rolls can only be 1, 2, or 3, so

$$\mathbb{E}[Y \mid X = k] = \frac{1}{3}(k - 1).$$

Using the law of total expectation, we have

$$\begin{aligned}\mathbb{E}[Y] &= \sum_{k=1}^{\infty} \mathbb{E}[Y \mid X = k] \mathbb{P}[X = k] && \text{(total expectation)} \\ &= \sum_{k=1}^{\infty} \frac{1}{3}(k - 1) \mathbb{P}[X = k] \\ &= \mathbb{E}\left[\frac{1}{3}(X - 1)\right] && \text{(LOTUS)} \\ &= \frac{1}{3}(\mathbb{E}[X] - 1) && \text{(linearity)}\end{aligned}$$

Since now $X \sim \text{Geometric}(\frac{1}{2})$, the expected number of rolls until we roll a number greater than 3 is $\mathbb{E}[X] = 2$:

$$= \frac{1}{3}(2 - 1) = \frac{1}{3}$$

Alternatively, we can use iterated expectation, along with the fact that $\mathbb{E}[Y \mid X] = \frac{1}{3}(X - 1)$, to give

$$\begin{aligned}\mathbb{E}[Y] &= \mathbb{E}[\mathbb{E}[Y \mid X]] \\ &= \mathbb{E}\left[\frac{1}{3}(X - 1)\right] \\ &= \frac{1}{3}(\mathbb{E}[X] - 1) \\ &= \frac{1}{3}(2 - 1) = \frac{1}{3}\end{aligned}$$