# **CSE 312: Foundations of Computing II**

# Section 4: Random Variables, Linearity of Expectation Solutions

### 0. Balls in Bins

Let X be the number of bins that remain empty when m balls are distributed into n bins randomly and independently. For each ball, each bin has an equal probability of being chosen. (Notice that two bins being empty are not independent events: if one bin is empty, that decreases the probability that the second bin will also be empty. This is particularly obvious when n = 2 and m > 0.) Find  $\mathbb{E}[X]$ .

#### Solution:

For  $i \in [n]$ , let  $X_i$  be 1 if bin i is empty, and 0 otherwise. Then,  $X = \sum_{i=1}^n X_i$ . We first compute  $\mathbb{E}[X_i] = 1 \cdot \Pr(X_i = 1) + 0 \cdot \Pr(X_i = 0) = \Pr(X_i = 1) = (\frac{n-1}{n})^m$ . Hence,

$$\mathbb{E}[X] = \mathbb{E}\left[\sum_{i=1}^{n} X_i\right] = \sum_{i=1}^{n} \mathbb{E}[X_i] = n \cdot \left(\frac{n-1}{n}\right)^m$$

### 1. Fair Game?

You flip a fair coin independently and count the number of flips until the first tail, including that tail flip in the count. If the count is n, you receive  $2^n$  dollars. What is the expected amount you will receive? How much would you be willing to pay at the start to play this game?

#### Solution:

The expected amount is  $\infty$ . Let N be the number of flips until the first tail, so  $p_N(n) = \frac{1}{2^n}$  for  $n \in \mathbb{N}$ . Hence  $\mathbb{E}[2^N] = \sum_{n=1}^{\infty} 2^n \frac{1}{2^n} = \sum_{n=1}^{\infty} 1 = \infty$ . In theory, you should be willing to pay any finite amount of money to play this game, but I admit I would be nervous to pay a lot. For instance, if you pay \$1000, you will lose money unless the first 9 flips are all heads. With high probability you will lose money, and with low probability you will win a lot of money.

### 2. Symmetric Difference

Suppose A and B are random, independent (possibly empty) subsets of  $\{1, 2, ..., n\}$ , where each subset is equally likely to be chosen as A or B. Consider  $A\Delta B = (A \cap B^C) \cup (B \cap A^C) = (A \cup B) \cap (A^C \cup B^C)$ , i.e., the set containing elements that are in exactly one of A and B. Let X be the random variable that is the size of  $A\Delta B$ . What is  $\mathbb{E}[X]$ ?

#### Solution:

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For i = 1, 2, ..., n, let  $X_i$  be the indicator of whether  $i \in A\Delta B$ . Then  $\mathbb{E}[X_i] = \Pr(X_i = 1) = \frac{1}{2}$ , and  $X = \sum_{i=1}^n X_i$ , so

$$\mathbb{E}[X] = \mathbb{E}\left[\sum_{i=1}^{n} X_i\right] = \frac{n}{2}$$

### 3. Negative Binomial Random Variable

Recall that  $W \sim Geo(p)$  (W has a geometric distribution with success parameter p) if it is the number of independent coin flips up to and including the first head, where Pr(HEAD) = p. The probability mass function is  $p_W(k) = (1-p)^{k-1}p$  and  $\mathbb{E}[W] = \frac{1}{p}$ . What if we wanted to flip until the  $r^{th}$  head, and not just the first? We say X is a **negative binomial** random variable with parameters r a positive integer and p = Pr(HEAD) (written  $X \sim NegBin(r, p)$ ) if X is the number of independent coin flips up to and including the  $r^{th}$  head.

(a) What is the codomain  $\Omega_X$ , and the probability mass function  $p_X(k)$ , if  $X \sim NegBin(r, p)$ ?

### Solution:

We must flip at least r times, and can flip any number of times, so  $\Omega_X = \{r, r+1, ...\}$ . To get the  $r^{th}$  head on the  $k^{th}$  flip, the first k-1 must have exactly r-1 heads and k-r tails, followed by a head; there are  $\binom{k-1}{r-1}$  ways to choose positions of these heads and tails. There are r heads and k-r tails total, with probability  $p^r(1-p)^{k-r}$  for any particular sequence. Hence,

$$p_X(k) = {\binom{k-1}{r-1}} p^r (1-p)^{k-r}$$

(b) Find  $\mathbb{E}[X]$  (hint: use linearity of expectation with r appropriate random variables, which are not necessarily indicator variables).

### Solution:

Let  $X_1, ..., X_r$  be independent Geo(p) random variables. Then,  $X = \sum_{i=1}^r X_i$ . Hence, by linearity of expectation,

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

## 4. Hypergeometric Random Variable

Recall the trick or treating scenario: Suppose on Halloween, someone is too lazy to keep answering the door, and leaves a jar of exactly N total candies. You count that there are exactly K of them which are kit kats (and the rest are not). The sign says to please take exactly n candies. Each item is equally likely to be drawn. Let X be the number of kit kats we draw (out of n). We say X is a **hypergeometric** random variable, and write  $X \sim HypGeo(N, K, n)$ .

(a) Find  $p_X(k) = \Pr(X = k)$ .

### **Solution:**

$$p_X(k) = \Pr(X = k) = \frac{\binom{K}{k}\binom{N-K}{n-k}}{\binom{N}{n}}$$

We choose k out of the K kit kats, and n - k out of the N - K other candies. The denominator is the total number of ways to choose n candies out of N total.

(b) Compute  $\mathbb{E}[X]$  (hint: define appropriate indicator variables and use linearity of expectation).

### Solution:

For i = 1, ..., n, let  $X_i$  be 1 if candy i was a kit kat, and 0 otherwise. Then,  $\mathbb{E}[X_i] = 1 \cdot \Pr(X_i = 1) + 0 \cdot \Pr(X_i = 0) = \Pr(X_i = 1) = \frac{K}{N}$ . So,

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

(c) Suppose we have the same setup: N candies total, K of which are kit kats, and we plan to draw n of them. This time, however, we just want to sniff the candies. We will draw a candy, sniff the candy, **put it back**, and draw another,.... We do this n times total. Let Y be the number of kit kats sniffed. What distribution does Y have, and what is  $\mathbb{E}[Y]$ ? Compare it to the expectation from the previous part from when we didn't return the candies.

## Solution:

 $Y \sim Bin(n, \frac{K}{N})$  , and we know  $\mathbb{E}[Y] = n \frac{K}{N}.$  It is the same!