Yale University S&DS 551, Spring 2023 Homework 1

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Problem 1.

(1) Mean

Let X_1, \ldots, X_{n-k} be random variables denoting the value of each of the n-k remaining balls and $X = \sum_{i=1}^{n-k} X_i$ be their sum. By the linearity of expectation,

$$\mathbb{E}[X] = \sum_{i=1}^{n-k} \mathbb{E}[X_i]$$
$$= \left[(n-k) \frac{n+1}{2} \right].$$

(2) Variance

We compute the second moment in similar fashion. Indeed,

$$\begin{split} \mathbb{E}[X^2] &= \sum_{i,j=1}^{n-k} \mathbb{E}[X_i X_j] \\ &= \sum_{\ell=1}^{n-k} \mathbb{E}[X_\ell^2] + 2 \sum_{1 \leq i < j \leq n-k} \mathbb{E}[X_i X_j]. \end{split}$$

For any $\ell \in [n-k]$,

$$\mathbb{E}[X_{\ell}^2] = \frac{1}{n} \sum_{i=1}^n i^2$$

$$= \frac{(n+1)(2n+1)}{6}$$

$$\sum_{\ell=1}^{n-k} \mathbb{E}[X_{\ell}^2] = \frac{(n-k)(n+1)(2n+1)}{6}.$$

Now, there are $\binom{n}{2}$ possible pairs $1 \leq i < j \leq n$ and $\binom{n-k}{2}$ pairs of indices from [n-k]. It follows that

$$2\sum_{1 \le i < j \le n-k} \mathbb{E}[X_i X_j] = \frac{\binom{n-k}{2}}{\binom{n}{2}} \sum_{1 \le i, j \le n: i \ne j} ij$$

$$= \frac{(n-k)(n-k-1)}{n(n-1)} \left(\sum_{1 \le i, j \le n} ij - \sum_{\ell=1}^n \ell^2 \right)$$

$$= \frac{(n-k)(n-k-1)}{n(n-1)} \left(\left[\sum_{i=1} i \right]^2 - \sum_{j=1}^n j^2 \right)$$

$$= \frac{(n-k)(n-k-1)}{n(n-1)} \left(\frac{n^2(n+1)^2}{4} - \frac{n(n+1)(2n+1)}{6} \right)$$

$$= \frac{(n-k)(n-k-1)}{n(n-1)} \left(\frac{n(n+1)[3n(n+1) - 2(2n+1)]}{12} \right)$$

$$= \frac{(n-k)(n-k-1)}{n(n-1)} \left(\frac{n(n+1)(3n^2 - n - 2)}{12} \right).$$

Finally,

$$Var[X] = \mathbb{E}[X^2] - \mathbb{E}[X]^2$$

$$= \frac{(n-k)(n+1)(2n+1)}{6} + \frac{(n-k)(n-k-1)}{n(n-1)} \left(\frac{n(n+1)(3n^2-n-2)}{12}\right) - (n-k)^2 \frac{(n+1)^2}{4}$$

$$= \frac{k(n+1)(n-k)}{12}.$$

Problem 2.

First consider the special case of m = n. Clearly, we must have

$$\mathbb{E}[S_m/S_n] = 1.$$

Note we use the assumption here that $\mathbb{E}[X_1^{-1}]$ exists.

By the linearity of expectation,

$$\mathbb{E}[S_m/S_n] = \sum_{i=1}^m \mathbb{E}[X_i/S_n].$$

However, since each X_i is iid, we conclude that

$$\mathbb{E}[X_i/S_n] = \frac{1}{n}$$

so that

$$\mathbb{E}[S_m/S_n] = \frac{m}{n}$$

as desired.

Problem 3.

We remark that

$$\mathbb{P}\{N > n\} = \mathbb{P}\left\{\sum_{i=1}^{n} X_i \le x\right\}$$
$$=: p_{n,x}.$$

We argue by induction that $p_{n,x} = \frac{x^n}{n!}$. It is clear that $p_{0,x} = 1$ since x > 0. Now suppose the induction hypothesis holds up to n - 1. We have

$$p_n = \int_0^x \mathbb{P}\left\{\sum_{i=1}^{n-1} X_i \le x - z\right\} p(z) dz$$

$$= \int_0^x p_{n,x-z} dz$$

$$= \int_0^x \frac{(x-z)^{n-1}}{(n-1)!}$$

$$= \left[-\frac{(x-z)^n}{n!}\right]_0^x$$

$$= \frac{x^n}{n!}.$$

By induction, we conclude the proof.

In order to compute the expectation, recall the identity

$$\mathbb{E}[N] = \sum_{n=1}^{\infty} \mathbb{P}\{N = n\} \cdot n$$

$$= \sum_{n=1}^{\infty} \mathbb{P}\{N \ge n\}$$

$$= \sum_{n=0}^{\infty} \mathbb{P}\{N > n\}$$

$$= \sum_{n=0}^{\infty} \frac{x^n}{n!}$$

$$= \boxed{e^x}.$$

We can similarly compute the variance as

$$\begin{split} \mathbb{E}[N^2] &= \sum_{n=1}^{\infty} \mathbb{P}\{N=n\} \cdot n^2 \\ &= \sum_{n=1}^{\infty} \mathbb{P}\{N \geq n\} [n^2 - (n-1)^2] \\ &= \sum_{n=0}^{\infty} \mathbb{P}\{N > n\} [2n-1] \\ &= 2 \sum_{n=1}^{\infty} \mathbb{P}\{N > n\} \cdot n - \sum_{n=0}^{\infty} \mathbb{P}\{N > n\} \\ &= 2x \sum_{n=0}^{\infty} \frac{x^n}{n!} - e^x \\ &= e^x (2x-1). \end{split}$$

It follows that the variance is

$$Var[N] = e^x(2x-1) - e^{2x}$$
.

Problem 4.

let us determine the density of XY, Z^2 separately and recall that the joint density is just the product of densities since XY, Z^2 are independent.

Let F_{ξ}, p_{ξ} denote the distribution function and density of a random variable ξ , respectively.

We have

$$F_{XY}(t) = \mathbb{P}\{X \le t\} + \int_{t}^{1} \mathbb{P}\{Y \le t/x\} p_{X}(x) dx$$

$$= t + \int_{t}^{1} \frac{t}{x} dx$$

$$= t + [t \log x]_{t}^{1}$$

$$= t - t \log t.$$

$$t \in [0, 1]$$

It follows that

$$p_{XY}(t) = \frac{d}{dt} F_{XY}(t)$$

$$= 1 - \log t - 1$$

$$= -\log t.$$

$$t \in [0, 1]$$

Similarly,

$$F_{Z^2}(t) = \sqrt{t}$$

$$t \in [0, 1]$$

$$p_{Z^2}(t) = \frac{1}{2\sqrt{t}}.$$

$$t \in (0, 1]$$

The joint density for $t_1, t_2 \in [0, 1], t_2 \neq 0$ is thus

$$p_{XY,Z^2}(t_1,t_2) = -\frac{\log t_1}{2\sqrt{t_2}}.$$

We now compute the desired probability

$$\mathbb{P}\{XY < Z^2\} = \int_0^1 \mathbb{P}\{XY \le t\} p_{Z^2}(t) dt$$

$$= \int_0^1 (t - t \log t) \frac{1}{2\sqrt{t}} dt$$

$$= \frac{1}{2} \int_0^1 \sqrt{t} - \sqrt{t} \log t dt$$

$$= \left[-\frac{t^{3/2} (3 \log x - 5)}{9} \right]_0^1$$

$$= \frac{5}{9}$$

as desired.

Problem 5.

(1)

Write N to be the random variable indicating the number of draws until we stop. Clearly for $n \leq 3$,

$$\mathbb{P}\{N=n\}=0.$$

Let us compute the distribution function $F(n) := \mathbb{P}\{N \leq n\}$. This the probability that after n draws, we have at least one card from each shade. This is equal to 1 minus the probability that we see at most 3 shades. Let p_1, p_2, p_3 denote the probability we do not see 1, 2, 3 shades for some FIXED shades (it is symmetric so it does not matter the particular choice of shades), respectively. By the inclusion-exclusion principle, the desired probability is thus

$$F(n) = 1 - \binom{4}{1}p_1 + \binom{4}{2}p_2 - \binom{4}{3}p_3$$

= 1 - 4\left(\frac{3}{4}\right)^n + 6\left(\frac{2}{4}\right)^n - 4\left(\frac{1}{4}\right)^n. \qquad n \geq 4

This is because we need to count for the $\binom{4}{1}$ ways we avoid a particular shade, but avoid double counting the $\binom{4}{2}$ ways we avoid a particular pair of shades, and adjust for the overcorrection of the $\binom{4}{3}$ ways we avoid any triple of shades.

Then for $n \geq 4$,

$$\mathbb{P}\{N=n\} = F(n) - F(n-1)
= -4\left(\frac{3}{4}\right)^n + 6\left(\frac{2}{4}\right)^n - 4\left(\frac{1}{4}\right)^n + 4\left(\frac{3}{4}\right)^{n-1} - 6\left(\frac{2}{4}\right)^{n-1} + 4\left(\frac{1}{4}\right)^{n-1}
= \left[\left(\frac{3}{4}\right)^{n-1} - 3\left(\frac{2}{4}\right)^{n-1} + 3\left(\frac{1}{4}\right)^{n-1}\right].$$

(2)

It is easy to see the that the sum of probabilities tend to 1 since we computed the individual probabilities as the difference between consecutive values of the distribution function. Thus the partial sum up to n is just F(n) and $F(n) \to 1$ as $n \to \infty$ by inspection.