

Stochastic Models: Exercise 1

1. Let X_k denote the number of meals ordered to collect the k th different type of mini toy given that $k-1$ different toys have been collected. Then each X_k is a geometric distribution with $p = \frac{N-(k-1)}{N}$ and

$$E[X_k] = \frac{N}{N-k+1}$$

The expected number of meals ones need to order before collecting a complete set of at least one toy is

$$\begin{aligned} E\left[\sum_{k=1}^N X_k\right] &= \sum_{i=1}^N \frac{N}{N-k+1} \\ &= N \sum_{k=1}^N \frac{1}{k} \end{aligned}$$

2. First observe that

$$\sum_{k=1}^{\infty} \mathbb{P}(X \geq k) = \lim_{n \rightarrow \infty} \sum_{k=1}^n \mathbb{P}(X \geq k) = \lim_{n \rightarrow \infty} \sum_{k=0}^n \mathbb{P}(X > k) = \sum_{k=0}^{\infty} \mathbb{P}(X > k)$$

We shall now show that it equals $E[X]$ by first defining

$$I_n := \begin{cases} 1 & \text{if } X \geq n \text{ for } n \geq 1 \\ 0 & \text{otherwise} \end{cases}$$

then $\mathbb{P}(X \geq n) = E[I_n]$.

$$\begin{aligned} \sum_{k=1}^{\infty} \mathbb{P}(X \geq k) &= \sum_{k=1}^{\infty} E[I_k] \\ &= E\left[\sum_{k=1}^{\infty} I_k\right], \text{ by Fubini's Theorem} \\ &= E[X] \end{aligned}$$

since the infinite sum of the indicator random variables is equals to X .

$$\begin{aligned} \int_0^{\infty} \mathbb{P}(X > x) dx &= \int_0^{\infty} \int_x^{\infty} f_X(t) dt dx \\ &= \int_0^{\infty} \int_0^t f_X(t) dx dt \\ &= \int_0^{\infty} t f_X(t) dt \\ &= \int_{-\infty}^{\infty} t f_X(t) dt \\ &= E[X] \end{aligned}$$

Using the earlier result,

$$\begin{aligned} E[X^n] &= \int_0^{\infty} \mathbb{P}(X^n > x) dx \\ &= \int_0^{\infty} \mathbb{P}(X > x^{1/n}) dx, \text{ only consider positive root of } x^{1/n} \text{ since } X \text{ is non-negative} \end{aligned}$$

we can do a change of variables, by letting $y = x^n$

$$\begin{aligned} dx &= ny^{n-1} dy \\ \int_0^{\infty} \mathbb{P}(X > x^{1/n}) dx &= \int_0^{\infty} ny^{n-1} \mathbb{P}(X > y) dy \end{aligned}$$

and we are done.

3.

(a) Let $Y = F(X)$ and $0 \leq Y \leq 1$ since F_X is a cdf.

$$\begin{aligned} F_Y(y) &= \mathbb{P}(Y \leq y) = \mathbb{P}(F(X) \leq y) \\ &= \mathbb{P}(X \leq F^{-1}(y)), \text{ since } F \text{ is non-decreasing} \\ &= F(F^{-1}(y)) = y \end{aligned}$$

since $F_Y(y) = y$, Y is the uniform distribution as the cdf of U over $(0, 1)$, $F_U(x) = x$.

(b) Let $W = F^{-1}(U)$, then

$$\begin{aligned} F_W(w) &= \mathbb{P}(W \leq w) = \mathbb{P}(F^{-1}(U) \leq w) \\ &= \mathbb{P}(U \leq F(w)), \text{ since } F_X \text{ is non-decreasing} \\ &= F(w) \end{aligned}$$

since $F_W(w) = F(w)$ for all w , $F^{-1}(U)$ has the same distribution as F .

4.

$$E[X^2] = \int_0^a x^2 dF(x) dx \leq \int_0^a ax dF(x) = aE[X], \text{ since } 0 \leq x \leq a.$$

then

$$\begin{aligned} \text{Var}[X] &= E[X^2] - E[X]^2 \\ &\leq aE[X] - E[X]^2 \\ &= a^2 \left(\frac{E[X]}{a} - \left(\frac{E[X]}{a} \right)^2 \right) \\ &= a^2 [\alpha(1 - \alpha)], \text{ where } \alpha = \frac{E[X]}{a} \end{aligned}$$

it suffices to show that $\alpha(1 - \alpha) \leq 1/4$ and we are done which is true since we know that the function $y = x - x^2$ attains a maximum value of $y = 1/4$ at $x = 1/2$. Therefore,

$$\text{Var}[X] \leq \frac{a}{4}$$