

Problem 1. Special Random Variables

(a) $P(X=n) = (1-p)^{n-1} \cdot p$; X can be $p, (1-p)p, (1-p)^2p, \dots$

$$P(X < p) = \left(\sum_{k=2}^{\infty} (1-p)^{k-1} \cdot p \right)^n \Rightarrow P(X=p) = 1 - \left(\sum_{k=2}^{\infty} (1-p)^{k-1} \cdot p \right)^n = \left(\sum_{k=1}^{\infty} (1-p)^{k-1} \cdot p \right)^n - \left(\sum_{k=2}^{\infty} (1-p)^{k-1} \cdot p \right)^n$$

$$P(X=(1-p)p) = 1 - \left[1 - \left(\sum_{k=2}^{\infty} (1-p)^{k-1} \cdot p \right)^n \right] - \left(\sum_{k=3}^{\infty} (1-p)^{k-1} \cdot p \right)^n = \left(\sum_{k=2}^{\infty} (1-p)^{k-1} \cdot p \right)^n - \left(\sum_{k=3}^{\infty} (1-p)^{k-1} \cdot p \right)^n$$

$$P(X=(1-p)^2p) = 1 - \left[1 - \left(\sum_{k=2}^{\infty} (1-p)^{k-1} \cdot p \right)^n \right] - \left[\left(\sum_{k=2}^{\infty} (1-p)^{k-1} \cdot p \right)^n - \left(\sum_{k=3}^{\infty} (1-p)^{k-1} \cdot p \right)^n \right] - \left(\sum_{k=4}^{\infty} (1-p)^{k-1} \cdot p \right)^n = \left(\sum_{k=2}^{\infty} (1-p)^{k-1} \cdot p \right)^n - \left(\sum_{k=3}^{\infty} (1-p)^{k-1} \cdot p \right)^n$$

\Rightarrow Thus we know that PMF of X is $\left(\sum_{k=1}^{\infty} (1-p)^{k-1} \cdot p \right)^n - \left(\sum_{k=2}^{\infty} (1-p)^{k-1} \cdot p \right)^n$, where X is a positive integer, otherwise it equals to 0.

For Y , it can also be $p, (1-p)p, (1-p)^2p, \dots$

$$P(Y=p) = p^n, P(Y=(1-p)p) = (p + (1-p)p)^n - p^n = \left(\sum_{k=1}^2 (1-p)^{k-1} \cdot p \right)^n - \left(\sum_{k=1}^1 (1-p)^{k-1} \cdot p \right)^n$$

$$P(Y=(1-p)^2p) = (p + (1-p)p + (1-p)^2p)^n - (p + (1-p)p)^n = \left(\sum_{k=1}^3 (1-p)^{k-1} \cdot p \right)^n - \left(\sum_{k=1}^2 (1-p)^{k-1} \cdot p \right)^n$$

\Rightarrow Thus we know that PMF of Y is $\left(\sum_{k=1}^X (1-p)^{k-1} \cdot p \right)^n - \left(\sum_{k=1}^{X-1} (1-p)^{k-1} \cdot p \right)^n$, $X \in \{2, 3, 4, \dots\}$
 $p^n, X=1$
 $0, \text{ otherwise.}$

By 等比级数, it's $\frac{p \cdot [1 - (1-p)^X]}{(1-p)} - \frac{p \cdot [1 - (1-p)^{X-1}]}{(1-p)} = 1 - (1-p)^X - (1 - (1-p)^{X-1}) = (1-p)^{X-1} - (1-p)^X = (1-p)^{X-1} (1 - (1-p)) = (1-p)^{X-1} \cdot p$

\Rightarrow So it's Geometric Random Variable.

(b) For $n=1$, $p(1) = p(2) = \frac{1}{2} = \frac{1}{(1+1)-1}$, it agrees.

Assume that for $n=k$, it agrees, i.e. $p(1) = p(2) = \dots = p(k+1) = \frac{1}{(k+1)-1} = \frac{1}{k}$, and there are $k+1$ balls

Then for $n=k+1$, $p(1) = \frac{1}{k} \times \frac{k}{k+1} = \frac{1}{k+1}$, $p(2) = \frac{1}{k} \times \frac{1}{k+1} + \frac{1}{k} \times \frac{k-1}{k+1} = \frac{k}{k(k+1)} = \frac{1}{k+1}$, $p(n) = \frac{1}{k} \times \frac{(n-1)}{k+1} + \frac{1}{k} \times \frac{k+1-n}{k+1} = \frac{k}{k(k+1)} = \frac{1}{k+1}$

\Rightarrow By the induction steps, it agrees, \star

(2) X has PDF like $\frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$, Y is $\begin{cases} ax+b, & \text{if } x \geq 0 \\ -ax+b, & \text{if } x < 0. \end{cases}$

and X 's CDF like $\Phi\left(\frac{t-\mu}{\sigma}\right)$.

Case 1: $a > 0$

$$\begin{aligned} P(Y \leq t) &= P(Y \leq t, X \geq 0) + P(Y \leq t, X < 0) \\ &= P(ax+b \leq t) + P(-ax+b \leq t) \\ &= P\left(X \leq \frac{t-b}{a}\right) + P\left(X \leq \frac{t+b}{-a}\right) \\ &= \Phi\left(\frac{t-b}{a}\right) + \Phi\left(\frac{t+b}{-a}\right) \end{aligned}$$

Case 2: $a = 0$.

$$\begin{aligned} \Rightarrow P(Y \leq t) &= \begin{cases} 0 & \text{for } Y < b \\ 1 & \text{for } Y \geq b \end{cases} \end{aligned}$$

Case 3: $a < 0$.

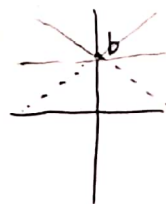
$$\begin{aligned} \Rightarrow P(Y \leq t) &= P(Y \leq t, X \geq 0) + P(Y \leq t, X < 0) \\ &= P(ax+b \leq t) + P(-ax+b \leq t) \\ &= P\left(X \leq \frac{t-b}{a}\right) + P\left(X \leq \frac{t+b}{-a}\right) \\ &= \Phi\left(\frac{t-b}{a}\right) + \Phi\left(\frac{t+b}{-a}\right) \end{aligned}$$

\Rightarrow And Y 's PDF is like:

If $a \neq 0$, $\Phi'\left(\frac{t-b}{a}\right) \cdot \frac{1}{a} + \Phi'\left(\frac{t+b}{-a}\right) \cdot \frac{1}{a}$

If $a = 0$, $\begin{cases} 1, & \text{for } Y=b \\ 0, & \text{for elsewhere.} \end{cases}$

$\Rightarrow Y$ cannot be a normal random variable, since normal random variables has PDF symmetric about μ , while taking absolute breaks the characteristic.



Problem 2 (PMF and Entropy)

(a)

$$H(X) = - \sum_{i=1}^n P_i \ln P_i = \sum_{i=1}^n P_i \cdot (-\ln P_i)$$

By the weighted inequality of arithmetic and geometric means, $\frac{P_1(-\ln P_1) + P_2(-\ln P_2) + \dots + P_n(-\ln P_n)}{P_1 + \dots + P_n} \geq \frac{P_1}{P_1 + \dots + P_n} \cdot \frac{P_2}{P_1 + \dots + P_n} \cdot \dots \cdot \frac{P_n}{P_1 + \dots + P_n} = \left(\frac{P_1 P_2 \dots P_n}{(P_1 + \dots + P_n)^n} \right)^{\frac{1}{n}}$

Since $P_1 + \dots + P_n = 1$, $-P_1 \ln P_1 - P_2 \ln P_2 - \dots - P_n \ln P_n \geq \left(\frac{P_1 P_2 \dots P_n}{1^n} \right)$

The "=" holds when $P_1 \ln P_1 = P_2 \ln P_2 = \dots = P_n \ln P_n$, then $P_1 = P_2 = \dots = P_n = \frac{1}{n}$

$\Rightarrow \sum_{i=1}^n \frac{1}{n} \ln \left(\frac{1}{n} \right) = \sum_{i=1}^n \frac{1}{n} \cdot \ln \left(\frac{1}{n} \right) = \ln n$, $\ln n$ is the maximum value of entropy.

The PMF is $\begin{cases} \frac{1}{n}, & \text{if } i=1, 2, \dots, n \\ 0, & \text{otherwise.} \end{cases}$

(b)

The minimum value happens when only for one i , $P_i = 1$, otherwise $P_i = 0$.

then the value is $-1 \ln 1 = 0$, since $P_i(-\ln P_i)$ always larger than or equals to 0, 0 is the minimum value for $H(X)$.

\Rightarrow The PMFs are $\begin{cases} 1, & \text{if } i=k, \\ 0, & \text{otherwise} \end{cases}, k \in \{1, 2, \dots, n\}$.

Problem 3 (Expectation and Moments)

(a) (i) $E[X] = 1 \cdot p + 2 \cdot (1-p) \cdot p + \dots + n \cdot (1-p)^{n-1} \cdot p + \dots = \sum_{n=1}^{\infty} n(1-p)^{n-1} \cdot p = \frac{p}{1-p} \sum_{n=1}^{\infty} n \cdot (1-p)^n$

Let $A = \sum_{n=1}^{\infty} n(1-p)^n = (1-p) + 2(1-p)^2 + 3(1-p)^3 + \dots$

$\Rightarrow \frac{(1-p)A}{pA} = \frac{(1-p) + 2(1-p)^2 + 3(1-p)^3 + \dots}{(1-p) + (1-p)^2 + (1-p)^3 + \dots} \Rightarrow pA = \frac{(1-p)}{1-(1-p)} \Rightarrow A = \frac{1-p}{p} \Rightarrow E[X] = \frac{p}{1-p} \cdot \frac{1-p}{p} = \frac{1}{p}$

(ii) $E[e^{tx}] = e^t \cdot p + e^{2t} \cdot (1-p) \cdot p + e^{3t} \cdot (1-p)^2 \cdot p + \dots = \sum_{n=1}^{\infty} e^{nt} \cdot (1-p)^{n-1} \cdot p = \frac{p}{1-p} \sum_{n=1}^{\infty} e^{nt} \cdot (1-p)^n$

Let $B = \sum_{n=1}^{\infty} e^{nt} \cdot (1-p)^n = e^t \cdot (1-p) + e^{2t} \cdot (1-p)^2 + e^{3t} \cdot (1-p)^3 + \dots$

$\Rightarrow \frac{e^t(1-p)B}{(1-e^t(1-p))B} = \frac{e^t(1-p) + e^{2t}(1-p)^2 + e^{3t}(1-p)^3 + \dots}{(1-e^t(1-p)) + (1-e^t(1-p))^2 + (1-e^t(1-p))^3 + \dots} \Rightarrow B = \frac{e^t(1-p)}{(1-e^t(1-p))} \Rightarrow E[e^{tx}] = \frac{p}{1-p} \cdot \frac{e^t(1-p)}{1-e^t(1-p)} = \frac{pe^t}{1-(1-p)e^t}$

(b) Prove:

$E[X^m] = E[Y^m], \forall m \in \{1, 2, \dots, n-1\} \Leftrightarrow P(X=t) = P(Y=t), \forall t \in \{a_1, \dots, a_n\}$.

" \Leftarrow ": It is trivial since $X=Y$.

" \Rightarrow ": $E[X^m] = 1^m p_1 + 2^m p_2 + 3^m p_3 + \dots + n^m p_n$

We could write all possibilities into matrices multiplying:

$\begin{bmatrix} p_1 & p_2 & p_3 & \dots & p_n \end{bmatrix} \begin{bmatrix} 1 & 1^2 & 1^3 & \dots & 1^n \\ 2 & 2^2 & 2^3 & \dots & 2^n \\ 3 & 3^2 & 3^3 & \dots & 3^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ n & n^2 & n^3 & \dots & n^n \end{bmatrix} = \begin{bmatrix} E[X] & E[X^2] & E[X^3] & \dots & E[X^n] \end{bmatrix}$

(notice that the right matrix is the Vandermonde matrix, it has property that $\det[V] \neq 0$ when $\alpha_1, \alpha_2, \dots, \alpha_n (1, 2, 3, \dots, n)$ in this question), it implies that the result of $E[X], E[X^2], \dots, E[X^n]$ is unique, i.e. $E[X^m] = E[Y^m]$ only when X and Y are identically distributed, so it has been proved.

By " \Rightarrow " and " \Leftarrow ", $E[X^m] = E[Y^m], \forall m \in \{1, 2, \dots, n-1\} \Leftrightarrow P(X=t) = P(Y=t), \forall t \in \{a_1, \dots, a_n\}$. #

(c)

$$\text{Var}[Z] = E[Z^2] - E[Z]^2$$

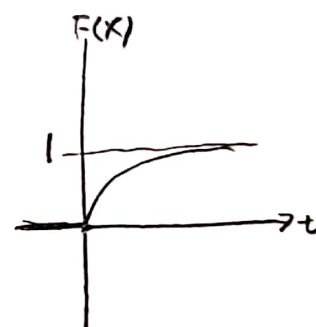
$$E[Z^2] = \sum_{n=1}^{\infty} ((-1)^n \sqrt{n})^2 \cdot \frac{6}{(\pi n)^2} = \sum_{n=1}^{\infty} n \cdot \frac{6}{\pi^2 n^2} = \frac{6}{\pi^2} \sum_{n=1}^{\infty} \frac{1}{n} = \infty. \Rightarrow \text{Var}[Z] \text{ doesn't exist.}$$

$$\sum_{n=1}^{\infty} Z_n^3 \cdot p_Z(Z_n) = \sum_{n=1}^{\infty} (-1)^{3n} \cdot n^{\frac{3}{2}} \cdot \frac{6}{\pi n^2} = \sum_{n=1}^{\infty} (-1)^n \cdot \frac{6}{\pi^2} \cdot \frac{1}{n^{\frac{1}{2}}} = \frac{6}{\pi^2} \sum_{n=1}^{\infty} \frac{1}{n^{\frac{1}{2}}} \cdot (-1)^n \doteq \frac{6}{\pi^2} \times -0.605 \doteq -0.3678.$$

$E[Z^3]$ exists, and above is its approximated value

Problem 4 (Inverse Transform Sampling)

(a) Its CDF is $\int_0^t \lambda e^{-\lambda x} dx = -e^{-\lambda x} \Big|_0^t = -e^{-\lambda t} + 1$, for $t \geq 0$
 $F(x) = 0$, otherwise.



$$y = 1 - e^{-\lambda t}$$

$$e^{-\lambda t} = 1 - y$$

$$-\lambda t = \ln(1 - y)$$

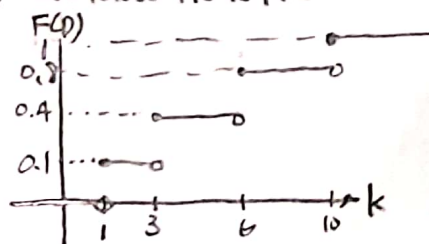
$$t = -\frac{1}{\lambda} \ln(1 - y)$$

$$F^{-1}(x) = \inf\{z: F(z) \geq x\}$$

$$= \begin{cases} -\infty, & x \leq 0 \\ -\frac{1}{\lambda} \ln(1 - y), & 0 < x < 1 \\ \infty, & x \geq 1. \end{cases}$$

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(b) We know its is like



, then we could write $F^{-1}(p) =$

$$\begin{cases} -\infty, & \text{for } p \leq 0 \\ 1, & \text{for } 0 < p \leq 0.1 \\ 3, & \text{for } 0.1 < p \leq 0.4 \\ 6, & \text{for } 0.4 < p \leq 0.8 \\ 10, & \text{for } 0.8 < p \leq 1 \\ \infty, & \text{for } p > 1. \end{cases}$$