

1. Solution:

According to the definition of expectation

$$E(X) = \int_0^{+\infty} xf(x)dx$$

where X is non-negative, and $f(x)$ is the density of X .

$$E(X) = \int_0^{+\infty} xf(x)dx$$

$$\geq \int_a^{+\infty} af(x)dx \text{ if } X \geq a$$

$$= a \int_a^{+\infty} f(x)dx$$

$$= aP(X \geq a)$$

$$\Rightarrow P(X \geq a) \leq \frac{E(X)}{a}.$$

2. Solution:**(a)**

As we know:

$$\phi_X(s) = E(e^{sX}) = \sum_{k \geq 0} \frac{s^k}{k!} E(X^k)$$

Differentiating $\phi_X(s)$ with respect to s will result in

$$\phi'_X(s) = 0 + E(X) + sE(X^2) + \dots$$

So: $\phi'_X(0) = E(X)$.

(b)

Similarly as question (a), we can get:

$$\begin{aligned} \frac{d^k \phi_X(s)}{ds^k} &= \frac{k!}{k!} E(X^k) + s \frac{(k+1)!}{(k+1)!} E(X^{k+1}) + \dots \\ \Rightarrow \left. \frac{d^k \phi_X(s)}{ds^k} \right|_{s=0} &= E(X^k) \end{aligned}$$

(c) First method

$$\begin{aligned}
\phi_X(s) &= E(e^{sX}) \\
&= \sum_{X=0}^n \binom{n}{X} p^X (1-p)^{n-X} e^{sX} \quad (\text{The definition of expectation}) \\
&= \sum_{X=0}^n \binom{n}{X} (pe^s)^X (1-p)^{n-X} \\
&= (pe^s + (1-p))^n
\end{aligned}$$

(c) Second method

If we note the fact that Binomial is the sum of Bernoulli trials, i.e.

$$X = \sum_{i=1}^n X_i, \quad X_i = \begin{cases} 1, & \text{with prob. } p \\ 0, & \text{with prob. } 1-p \end{cases} \quad i=1, \dots, n$$

where X is Binomial and X_i is Bernoulli, then:

$$\begin{aligned}
E(e^{sX}) &= E\left(e^{s \sum_{i=1}^n X_i}\right) \\
&= \left(E(e^{sX_1})\right)^n \quad (X_i, i=1, \dots, n \text{ are independent and identically distributed}) \\
&= (pe^s + (1-p))^n
\end{aligned}$$

For exponential distribution

$$E(e^{sX}) = \int_0^{\infty} \lambda e^{-\lambda u} e^{su} du = \lambda \frac{1}{s-\lambda} e^{(s-\lambda)u} \Big|_0^{\infty} = \frac{\lambda}{\lambda-s}$$

3. Solution:

(a)

$$P(|X - E(X)| \geq a) = P(|X - E(X)|^2 \geq a^2)$$

Since $|X - E(X)|^2$ is non-negative, we can apply Markov inequality

$$P(|X - E(X)|^2 \geq a^2) \leq \frac{E(|X - E(X)|^2)}{a^2} = \frac{\text{Var}(X)}{a^2}$$

(b)

For any $t > 0$

$$\begin{aligned} P(X \geq a) &= P(e^{tX} \geq e^{ta}) \\ &\leq \frac{E[e^{tX}]}{e^{ta}} \quad (\text{Markov inequality}) \end{aligned}$$

$$\text{Particular, } P(X \geq a) \leq \min_{t>0} \frac{E[e^{tX}]}{e^{ta}}$$

4. Solution:

(a)

Since $X_i, i=1, \dots, n$ are independent Bernoulli variables, $X = \sum_{i=1}^n X_i$ is Binomial with

$$X \sim B\left(n, \frac{1}{2}\right)$$

$$E(X) = E\left(\sum_{i=1}^n X_i\right) = nE(X_1) = np \quad (p=1/2)$$

$$\begin{aligned} \text{Var}(X) &= \text{Var}\left(\sum_{i=1}^n X_i\right) \\ &= n\text{Var}(X_1) \quad (\text{Independence}) \\ &= np(1-p) \quad (p=1/2) \end{aligned}$$

(b)

i. If we note that

$$\left\{ \left| X - \frac{n}{2} \right| \geq \frac{n}{4} \right\} = \left\{ X \geq \frac{3n}{4} \right\} \cup \left\{ X \leq \frac{n}{4} \right\}$$

then

$$P\left(\left| X - \frac{n}{2} \right| \geq \frac{n}{4}\right) \geq P\left(X \geq \frac{3n}{4}\right)$$

ii.

$$P\left(X \geq \frac{3n}{4}\right) \leq P\left(\left| X - \frac{n}{2} \right| \geq \frac{n}{4}\right) \leq \frac{\text{var}(X)}{(n/4)^2} = \frac{4}{n}$$

$$\text{iii. } \lim_{n \rightarrow \infty} P\left(X \geq \frac{3n}{4}\right) = 0$$

(c)

$$\text{i. } P(X \geq x) = P(e^{\theta X} \geq e^{\theta x}) \leq \frac{E(e^{\theta X})}{e^{\theta x}} \quad (\text{Markov's inequality})$$

$$\text{where } E(e^{\theta X}) = \left(\frac{1}{2} + \frac{1}{2}e^{\theta}\right)^n \quad (\text{see solutions of 2.(c)})$$

$$= \left(1 + \frac{1}{2}(e^{\theta} - 1)\right)^n \leq \exp\left(\frac{1}{2}(e^{\theta} - 1)\right)^n \quad (\text{hint: } 1 + \alpha \leq e^{\alpha}, \text{ where } \alpha = \frac{1}{2}(e^{\theta} - 1))$$

$$= \exp\left(\frac{n}{2}(e^{\theta} - 1)\right)$$

$$\text{Therefore } P(X \geq x) = \exp\left(\frac{n}{2}(e^{\theta} - 1) - \theta x\right) \quad (*)$$

$$\text{ii. set } \theta = \log 2, \text{ then } \frac{1}{2}(e^{\theta} - 1) - \frac{3}{4}\theta \approx -0.0199 \leq -0.01 \quad (**)$$

$$\text{iii. let } x = \frac{3}{4}n \text{ and using } (*), \text{ we have}$$

$$P\left(X \geq \frac{3}{4}n\right) = \exp\left(\frac{n}{2}(e^\theta - 1) - \theta \frac{3}{4}n\right) \leq \exp(-0.01n) \text{ (by using (**))}$$

5. Solution:

We denote

$$I_k = \begin{cases} 1, & \text{the } k\text{-th post is rotten} \\ 0, & \text{otherwise} \end{cases}$$

$k+1, k+2, \dots, k+7$ are 7 consecutive posts (if $k+i > 17$, then $k+i \Leftarrow k+i-17$)

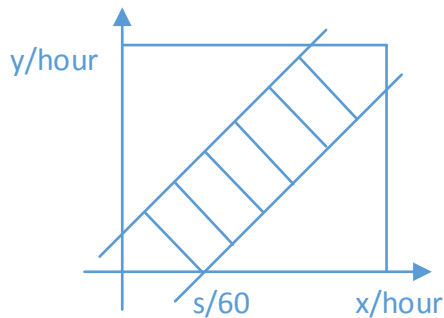
$R_k = \sum_{i=1}^7 I_{k+i}$ is the number of rotten posts.

$$\text{So, } E(R_k) = E\left(\sum_{i=1}^7 I_{k+i}\right) = \sum_{i=1}^7 E(I_{k+i}) = 7E(I_k) = \frac{35}{17} > 2$$

where $E(I_k) = \frac{5}{17}$ (5 of 17 posts are rotten)

Since R_k should be integer, it must be the case that $P(R_k \geq 3) > 0$, for some k , i.e. there necessarily exists a run of 7 consecutive posts at least 3 of which are rotten.

6. Solution:



See the unit square in the figure, the origin is 12:00, x axis (unit: hour) represents the time Alice arrive and y axis represents the time Bob arrives.

So, “each is prepared to wait $s/60$ hours before leaving and they meet each other” can be mathematically described as

$$\left\{ |x - y| \leq \frac{s}{60} \right\} \text{ (see the shadow area in the figure)}$$

$$P\left(|x - y| \leq \frac{s}{60}\right) = \iint_{|x-y| \leq \frac{s}{60}} f(x, y) dx dy = 1 - \left(1 - \frac{s}{60}\right)^2$$

where (x, y) is a two-dimensional variable stands for the time Alice and Bob arrive, and its

$$\text{density function is } f(x, y) = \begin{cases} 1, & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

$$\text{Finally, } 1 - \left(1 - \frac{s}{60}\right)^2 \geq \frac{1}{2} \Rightarrow s \geq 18 \text{ min}$$

7. Solution:

See

[https://en.wikipedia.org/wiki/Bertrand_paradox_\(probability\)](https://en.wikipedia.org/wiki/Bertrand_paradox_(probability))

<http://www.cut-the-knot.org/bertrand.shtml>

<http://web.mit.edu/tee/www/bertrand/problem.html>

for more details.