

1.

$$P(|X - (h+1)| > (h+1)) < E \frac{[X - (h+1)]^2}{(h+1)^2}$$

$$P[|X - (h+1)| < (h+1)] \gg 1 - E \frac{[X - (h+1)]^2}{(h+1)^2}$$

$$P[-(h+1) < X - (h+1) < (h+1)] \gg 1 - E \frac{[X - (h+1)]^2}{(h+1)^2}$$

$$P\{0 < X < 2(h+1)\} \gg 1 - E \frac{[X - (h+1)]^2}{(h+1)^2} \quad \text{--- (1)}$$

$$\begin{aligned} E[X - (h+1)]^2 &= E\{X^2 + (h+1)^2 - 2X(h+1)\} \\ &= E X^2 + (h+1)^2 - 2(h+1)E(X) \\ &= E X^2 + (h+1)^2 - \frac{2(h+1)}{h} h \\ &= h^2 + h + (h+1)^2 - 2h(h+1) \\ &= h^2 + h + h^2 + 2h + 1 - 2h^2 - 2h \\ &= 1 + h \end{aligned}$$

$$P\{0 < X < 2(h+1)\} \gg 1 - \frac{1+h}{(1+h)^2} = 1 - \frac{1}{1+h} = \frac{h}{h+1}$$

Thus, $P\{0 < X < 2(h+1)\} \gg \frac{h}{h+1}$

2. $P\{\mu - 2\sigma \leq X \leq \mu + 2\sigma\} = 0.6$

$$P\{|X - \mu| \leq 2\sigma\} = 0.6$$

$$P\{|X - \mu| > 2\sigma\} = 0.4 \quad \text{--- (1)}$$

$$P\{|X - \mu| > 2\sigma\} < E \frac{(X - \mu)^2}{(2\sigma)^2} = E \frac{(X - \mu)^2}{4\sigma^2} = \frac{5-2}{4 \cdot 5^2} = 0.25 \quad \text{--- (2)}$$

Using (1) and (2) gives a contradiction

3. $N \sim \text{geometric} (1/4) \rightarrow p_1$
 $K \sim \text{geometric} (1/10^5) \rightarrow p_2$
 # bits in a page

Let N be the number of pages in the transmission
 Thus the total number of bits K can be written as

$$K|_N = K_1 + K_2 + \dots + K_N$$

where $K_i \sim \text{geom}(p_2)$

$$\begin{aligned} MGF(K) &= E[e^{tK}] = E\{E[e^{tK}|N]\} \\ &= E\{E[e^{t(K_1+K_2+\dots+K_N)}]\} \\ &= E\{E[e^{tK_1} \cdot e^{tK_2} \cdot \dots \cdot e^{tK_N}]\} \end{aligned}$$

but the # bits in a page is independent of other
 thus, by using independence, we have

$$= E\{[E(e^{tK_1})]^N\} = E\{(m(t))^N\}$$

where $m(t) = MGF(K_1) = \frac{p_2 e^t}{1 - (1-p_2)e^t}$

$$\begin{aligned} &= \frac{p_1 m(t)}{1 - (1-p_1)m(t)} = \frac{p_1 \left\{ \frac{p_2 e^t}{1 - (1-p_2)e^t} \right\}}{1 - \frac{(1-p_1)p_2 e^t}{1 - (1-p_2)e^t}} \\ &= \frac{p_1 p_2 e^t}{1 - (1-p_2)e^t - (1-p_1)p_2 e^t} \\ &= \frac{p_1 p_2 e^t}{1 - e^t + p_2 e^t - p_2 e^t + p_1 p_2 e^t} \\ &= \frac{p_1 p_2 e^t}{1 - (1-p_1 p_2)e^t} \end{aligned}$$

thus K is geometric with parameter $p_1 p_2$

4. Let $N(t)$ denote the # bacteria at time t

$$\text{Let } X = \begin{cases} 0 & , 0.25 \\ 1 & , 0.25 \\ 2 & , 0.5 \end{cases}$$

be a random variable

$$\text{then } N(t+1) | N(t) = X_1 + X_2 + \dots + X_{N(t)}$$

$$P_{GF}(N(t+1))_{(s)} = E(s^{N(t+1)}) = E(s^{N(t+1)} | N(t))$$

$$= E \left\{ E(s^{X_1 + X_2 + \dots + X_{N(t)}}) \right\} \text{ but each } X_i \text{ is independent of other since bacteria act independently}$$

$$= E \left\{ [E(s^{X_1})]^{N(t)} \right\}$$

$$= E \left\{ (m(s))^{N(t)} \right\} \text{ where } m(s) = P_{GF}(X_1) = \frac{1}{4} + \frac{1}{4}s + \frac{1}{2}s^2$$

$$= P_{GF}(N(t)) [m(s)]$$

it is the same as P_{GF} of $N(t)$ evaluated at $m(s)$

$$= P_{GF}(N(t-1)) [m(s)]$$

which is the same as P_{GF} of $N(t-1)$ evaluated at $m(s)$

$$\text{Thus, } P_{GF}[N(t+1)](s) = P_{GF}(N(t)) \left[\underbrace{m(m(s))}_{t+1 \text{ times}} \right]$$

Assuming that at $t=0$, # bacteria at

$$\begin{aligned} E[N(t+1) | N(t)] &= E(X_1 + X_2 + \dots + X_{N(t)}) \\ &= E(X_1) + E(X_2) + \dots + E(X_{N(t)}) \\ &= N(t) E(X_1) \\ &= 1000 \left\{ \frac{1}{4}(0) + \frac{1}{4}(1) + \frac{1}{2}(2) \right\} \\ &= 1000 \left(\frac{5}{4} \right) = 1250 \end{aligned}$$

$$\underline{5.} \quad MGF(S_N) = \text{MPE} \quad E(s^{S_N})$$

$$\begin{aligned}
 &= E[E(s^{S_N} | N)] = E\left\{ E(s^{X_1 + X_2 + \dots + X_N}) \right\} \\
 &= E\left\{ E(s^{X_1} s^{X_2} \dots s^{X_N}) \right\} \\
 &\quad \text{now } X_1, X_2, \dots, X_N \text{ are independent} \\
 &\quad \text{of each other given } N \\
 &= E\left\{ [E(s^{X_1})]^N \right\} \\
 &= E\left\{ (m(s))^N \right\} \\
 &\quad \text{where } m(s) \text{ is the PGF of } X_1
 \end{aligned}$$

$$\begin{aligned}
 \text{and now} \quad &= PGF(N)[m(s)] \\
 &\text{which is same as PGF of } N \text{ evaluated at } m(s)
 \end{aligned}$$

$$\begin{aligned}
 a) \quad E(S_N) &= E[E(S_N | N)] = E\left\{ E(X_1 + X_2 + \dots + X_N | N) \right\} \\
 &= E[E(X_1) + E(X_2) + \dots + E(X_N) | N] \\
 &= E(N) E(X)
 \end{aligned}$$

$$\begin{aligned}
 b) \quad Var(S_N) &= E(S_N^2) - [E(S_N)]^2 \\
 &= E[E(S_N^2 | N)] - E(N)^2 E(X)^2 \\
 &= E\left\{ E[(X_1 + X_2 + \dots + X_N)^2 | N] \right\} - E(N)^2 E(X)^2 \\
 &= E\left\{ E[X_1^2 + X_2^2 + \dots + X_N^2 + 2X_1X_2 + \dots + 2X_NX_{N-1}] | N \right\} \\
 &\quad - E(N)^2 E(X)^2 \\
 &= E\left\{ E(X_1^2) + E(X_2^2) + \dots + E(X_N^2) + 2E(X_1X_2) \right. \\
 &\quad \left. + \dots + 2E(X_NX_{N-1}) \right\} - E(N)^2 E(X)^2 \\
 &= E\left\{ N E(X_1^2) + 2E(X_1)E(X_2) + \dots + 2E(X_N)E(X_{N-1}) \right\} \\
 &\quad - E(N)^2 E(X)^2 \\
 &= E\left\{ N E(X_1^2) + N(N-1) E(X_1)^2 \right\} - E(N)^2 E(X)^2 \\
 &= E(N) E(X_1^2) + E(N^2) E(X_1)^2 - E(N) E(X_1)^2 \\
 &\quad - E(N)^2 E(X)^2
 \end{aligned}$$

$$\begin{aligned}
&= E(N) E(X_1^2) + \cancel{E(N)^2 E(X_1^2)} \{ \text{Var}(N) + E(N)^2 \} \\
&\quad - E(N) E(X_1)^2 - \cancel{E(N)^2 E(X_1)^2} \\
&= E(N) E(X_1^2) + \text{Var}(N) E(X_1)^2 - E(N) E(X_1)^2 \\
&= E(N) \text{Var}(X) + \text{Var}(N) E(X)^2
\end{aligned}$$

6. $E(Y/X) = 1$

$$\begin{aligned}
\text{Var}(XY) &= E(X^2 Y^2) - [E(XY)]^2 \\
&= E[E(X^2 Y^2 | X)] - \{E[E(XY | X)]\}^2 \\
&= E[X^2 E(Y^2 | X)] - E(X)^2
\end{aligned}$$

$$\text{Var}(Y/X) = E(Y^2/X) - [E(Y/X)]^2 \geq 0$$

$$E(Y^2/X) \geq [E(Y/X)]^2 = 1$$

$$E(Y^2/X) \geq 1$$

thus $E[X^2 E(Y^2/X)] - E(X)^2 \geq E(X^2) - E(X)^2$

thus $\text{Var}(XY) \geq \text{Var}(X)$

7. # games played forms a geometric distribution
 Let the # games played = $N \sim \text{geom}(p)$ where $p = 1/3$
 let $X_i = \begin{cases} 2 & \text{with prob } 1/3 \\ 1 & \text{with prob } 1/3 \\ 0 & \text{with prob } 1/3 \end{cases}$ denote the points outcome of the i^{th} game

Then, total number of points earned Y can be written as

$$Y | N = X_1 + X_2 + \dots + X_N$$

$$\begin{aligned}
M_G F(Y) &= E(e^{tY}) = E\{E(e^{tY} | N)\} \\
&= E\{E(e^{t(X_1 + X_2 + \dots + X_N)} | N)\} \\
&= E\{E(e^{tX_1} e^{tX_2} \dots e^{tX_N} | N)\}
\end{aligned}$$

Now X_i 's are all independent given N

$$\begin{aligned}
&= E[(E e^{tX_1})^N] \\
&= E[(m(t))^N]
\end{aligned}$$

now $m(t) = M_G F(X)$

$$= \frac{p e^t}{1 - (1-p)e^t} = \frac{e^t}{3 - 2e^t}$$

8: $X \sim \text{Bin}(n, p)$

$$X = X_1 + X_2 + \dots + X_n$$

where each X_i is Bernoulli distributed with parameter p
and all are independent of each other

$$E(X) = np$$

$$\text{Var}(X) = np(1-p)$$

thus $Z = \frac{X - np}{\sqrt{np(1-p)}}$ can be considered standard normal using CLT

$$\begin{aligned} \text{thus } P(X \geq n/2) &= P\left(\frac{X - np}{\sqrt{np(1-p)}} \geq \frac{n(1/2 - p)}{\sqrt{np(1-p)}}\right) \\ &= P\left(Z \geq \frac{n(1/2 - p)}{\sqrt{np(1-p)}}\right) \end{aligned}$$

thus the value for which $P(X \geq n/2) \leq 1 - \alpha$
is same as that of for which $P\left(Z \geq \frac{n(1/2 - p)}{\sqrt{np(1-p)}}\right) \leq 1 - \alpha$

9: $T =$ total time to failure

$$T = T_1 + T_2 + \dots + T_{30}$$

where T_i denotes the failure time of i^{th} device

$$T_i \sim \exp(0.1)$$

$$E(T) = NE(T_i) = 30 \times (0.1)^{-1} = 300$$

$$\text{Var}(T) = N \text{Var}(T_i) = 30 \times (0.1)^{-2} = 3000$$

$$\begin{aligned} P(T \geq 350) &= P\left(\frac{T - 300}{\sqrt{3000}} \geq \frac{350 - 300}{\sqrt{3000}}\right) \\ &= P\left(Z \geq \frac{50}{\sqrt{3000}}\right) \text{ using CLT} \end{aligned}$$

11.

$$\xi = X_1 + X_2 + \dots + X_{30}$$

$$X_i \sim \text{Poisson}(\lambda = 0.01)$$

$$E(S) = 30 E(X_i) = 30(0.01) = 0.3$$

$$\text{Var}(S) = 30 \text{Var}(X_i) = 30(0.01) = 0.3$$

$$\begin{aligned} a) P(S > 3) &= P\left(\frac{S - 0.3}{\sqrt{0.3}} > \frac{3 - 0.3}{\sqrt{0.3}}\right) \\ &= P\left(Z > \frac{2.7}{\sqrt{0.3}}\right) \approx 0 \end{aligned}$$

b) Now sum of Poissons is also Poisson

$$\xi \sim \text{Poisson}(30\lambda) = \text{Poisson}(0.3)$$

$$\begin{aligned} P(\xi > 3) &= 1 - P(S=0) - P(S=1) - P(S=2) \\ &= 1 - \frac{(0.3)^0 e^{-0.3}}{0!} - \frac{e^{-0.3} (0.3)^1}{1!} - \frac{e^{-0.3} (0.3)^2}{2!} \end{aligned}$$

12.

$$P(|S_n - p| \leq 0.01) \approx 1 - \frac{E(S_n - p)^2}{(0.01)^2}$$

So we have

$$\frac{1 - E(S_n - p)^2}{(0.01)^2} = 0.95$$

$$\frac{E(S_n - p)^2}{(0.01)^2} = 0.05$$

$$\bar{X}_n = \frac{X_1 + X_2 + \dots + X_n}{n}$$

$$S_n - p = \frac{X_1 + X_2 + \dots + X_n}{n} - p = \frac{(X_1 - p) + (X_2 - p) + \dots + (X_n - p)}{n}$$

$$E(S_n - p)^2 = \frac{1}{n^2} E((X_1 - p) + (X_2 - p) + \dots + (X_n - p))^2$$

$$= \frac{1}{n^2} \left\{ n E(X_1 - p)^2 + n(n-1) \overset{0}{E(X_1 - p)} \overset{0}{E(X_2 - p)} \right\}$$

$$= \frac{n}{n^2} E(X_1 - p)^2 = \frac{1}{n} p(1-p)$$

$$\frac{p(1-p)}{n(0.01)^2} = 0.05$$

solve to get the # of samples that we need to take

13.

$$Y = \sum_{k=1}^{100} X_k$$

$$a) P(Y \gg 900) < \frac{E(Y)}{900} = \frac{100}{900} = 1/9$$

$$b) E(Y) = 100 E(X_k) = 100$$
$$Var(Y) = 100 Var(X_k) = 16 \times 10^4$$

$$P(Y \gg 900) = P\left(\frac{Y-100}{\sqrt{16 \times 10^4}} \gg \frac{900-100}{\sqrt{16 \times 10^4}}\right) = \cancel{\phi(2)} \quad \cancel{\phi(2)}$$

$$= P\left(Z \gg \frac{800}{400}\right)$$

$$= P(Z \gg 2) = 1 - \phi(2)$$