

HW 6, Harry Luo

ex4.6

recall from lec the normal approximation formula, where

$$P(|\hat{p} - p| < \varepsilon) \geq 2\Phi(2\varepsilon\sqrt{n}) - 1 \quad (1)$$

For this problem, we have $\varepsilon = 0.02$, $2\Phi(2\varepsilon\sqrt{n}) - 1 \geq 0.95$. We can solve for n when $n = n_{\min}$ with the following:

$$\begin{aligned} 2\Phi(2\varepsilon\sqrt{n}) - 1 &= 0.95 \\ \Phi(2 * 0.02\sqrt{n}) &= \frac{1.95}{2} \end{aligned} \quad (2)$$

according to the table of Phi values, we have

$$\begin{aligned} 0.04\sqrt{n} &= 1.96 \\ \Rightarrow \boxed{n = 2401} \end{aligned} \quad (3)$$

therefore the smallest size should be 2401

ex4.8

Rolling a biased die can be modeled as a binomial distribution as either “rolling the number 6” or not. We denote an unknown probability of rolling a 6 as p , and denote the number of getting 6 as X . We write $X \sim \text{Bin}(1000000, p)$. We want to find a confidence interval for p with 0.999 confidence. Using Equation 1, we have $n = 1000000$, $P(|\hat{p} - p| < \varepsilon) = 0.99$. We need to solve for ε at the lower bound, where:

$$\begin{aligned} 2\Phi(2\varepsilon\sqrt{n}) - 1 &= 0.999 \\ \Rightarrow \Phi(2 * 1000\varepsilon) &= 0.9995 \\ \Rightarrow 2000\varepsilon &\approx 3.32 \\ \varepsilon &= 0.00166 \end{aligned} \quad (4)$$

Since the number 6 shows up 180000 times when rolling 1000000 times, $\hat{p} = \frac{180000}{1000000} = 0.18$.

Therefore, the confidence interval is $[\hat{p} - \varepsilon, \hat{p} + \varepsilon] = [0.1783, 0.1817]$

ex4.10

We assume that scoring a goal in a certain game is a rare event for the player, we can approximate the r.v. X corresponding to the number of goals scored by the player as a Poisson distribution

$$P(X = k) = \frac{e^{-\lambda} \lambda^k}{k!} \quad (5)$$

probability of player scoring 0 goals is $P(X = 0) = e^{-\lambda} \frac{\lambda^0}{0!} = e^{-\lambda}$

Thus the probability of scoring at least 1 goal is $1 - e^{-\lambda} = 0.5 \Rightarrow \lambda = \ln(2) \approx 0.693$

We can now calculate the approximation for scoring 3 goals as

$$P(X = 3) = \frac{e^{-\lambda} \lambda^k}{k!} = \frac{e^{-0.693} 0.693^3}{3 * 2 * 1} = 0.028$$

ex4.34

Assume that accidents happen rarely and independently. We can model the number of accidents happen in a week with a Poisson distribution. We denote the r.v. X as the number of accidents in a week, and we have $X \sim \text{Poisson}(\lambda)$, where λ is the average number of accidents in a week, given as $\lambda = 3$. Therefore, the probability of **at most** 2 accidents happening next week can be calculated as

$$P(X = 1) + P(X = 2) + P(X = 0) = (e^{-3}) \left(\frac{3^1}{1} + \frac{3^2}{2 * 1} + \frac{3^3}{3 * 2 * 1} \right) = \boxed{0.59744} \quad (6)$$

ex4.46

We can consider the series of trials of “flipping a coin 5 times each day for 30 days” as a binomial distribution, where we either get 5 tails each day or not. We denote the r.v. X as the number of days that we get 5 tails. The probability of having 5 tails in a day is $p = \frac{1}{2^5} = \frac{1}{32}$. Therefore, $X \sim \text{Bin}(30, \frac{1}{32})$

Since $np(1 - p) = \frac{465}{512}$, the normal approximation is not valid.

Poisson approximation is a better choice, especially when our $np = 15/512$ is small.

We approximate the distribution of X with r.v. $Y \sim \text{Poisson}(\lambda)$ where $\lambda = E(X) = np = \frac{30}{32} = 0.9375$. Thus,

$$P(X = 2) \approx P(Y = 2) = \frac{e^{-0.9375} 0.9375^2}{2} \approx \boxed{0.1721} \quad (7)$$

ex5.2

- (a)

Given the MGF, we can calculate its derivatives as

$$M'(t) = -\frac{4}{3}e^{-4t} + \frac{5}{6}, M''(t) = \frac{16}{3}e^{-4t} - \frac{25}{6}e^{5t} \quad (8)$$

We can get

$$\begin{aligned} E(X) &= M'(0) = \frac{1}{2}, E(X^2) = M''(0) = \frac{19}{2} \\ \Rightarrow \text{Var}(X) &= E(X^2) - E(X)^2 = \frac{37}{4} \end{aligned} \quad (9)$$

- (b)

Given the MGF, we observe that the possible values for r.v. are 0, -4, 5; and the corresponding probabilities are 1/2, 1/3, 1/6. Thus the discrete probability mass function is $P(X = 0) = \frac{1}{2}, P(X = -4) = \frac{1}{3}, P(X = 5) = \frac{1}{6}$. From which we can calculate We can calculate

$$\begin{aligned} E(X) &= -4 * \frac{1}{3} + 5 * \frac{1}{6} = \frac{1}{2}; E(X^2) = \frac{1}{3} * 16 + \frac{1}{6} * 25 = \frac{19}{2} \\ \text{Var}(X) &= E(X^2) - E(X)^2 = \frac{37}{4} \end{aligned} \quad (10)$$

As calculated in (a).

ex5.18

- (a)

Given $X \sim \text{Geom}(p)$, the probability mass function is $P(X = k) = p(1 - p)^{k-1}$, where $k=1,2,3,\dots$

$$\begin{aligned} M_X(t) &= E(e^{tX}) = \sum_{k=1}^{\infty} e^{tk} P(X = k) = \sum_{k=1}^{\infty} e^{tk} p(1 - p)^{k-1} = pe^t \sum_{k=1}^{\infty} (e^t(1 - p))^{k-1} \\ &= pe^t \sum_{k=0}^{\infty} (e^t(1 - p))^k \end{aligned} \quad (11)$$

when $e^t(1 - p) < 1$, i.e. $t < \ln\left(\frac{1}{1-p}\right)$, the series converges, and

$$\boxed{M_X(t) = \frac{pe^t}{1 - e^t(1 - p)}} \quad (12)$$

while $t \geq \ln\left(\frac{1}{1-p}\right)$, the series diverges, and

$$M_X(t) = +\infty \quad (13)$$

- (b)

$$\begin{aligned} E(X) &= M'_{X(0)} = \frac{pe^t}{(1 - e^t(1 - p))^2} \Big|_{t=0} = \frac{1}{p}. \\ E(X^2) &= M''_X(0) = \frac{pe^t}{(1 - e^t(1 - p))^2} \Big|_{t=0} = \frac{2}{p^2} - \frac{1}{p} \\ \text{Var}(X) &= E(X^2) - E(X)^2 = \frac{1}{p^2} - \frac{1}{p} \end{aligned} \quad (14)$$

ex 5.20

- (a) by def, we know

$$\begin{aligned} M_X(t) &= \int_{-\infty}^{\infty} e^{tx} * \frac{1}{2} e^{-|x|} dx = \frac{1}{2} \int_0^{\infty} e^{(-1-t)x} dx + \frac{1}{2} \int_{-\infty}^0 e^{(t+1)x} dx \\ &= \frac{1}{2} \int_0^{\infty} e^{(-1-t)x} dx + \frac{1}{2} \int_0^{\infty} e^{-(t+1)x} dx \end{aligned} \quad (15)$$

Noticing that $\int_0^{\infty} e^{-cx} dx$ converges to $\frac{1}{c}$ iff $c > 0$, we can get

$$M_X(t) = \begin{cases} \frac{1}{2} \left(\frac{1}{1-t} \right) + \frac{1}{2} \left(\frac{1}{1+t} \right) = \frac{1}{2(1-t^2)}, & \text{when } -1 < t < 1 \\ \infty & \text{O.W.} \end{cases} \quad (16)$$

- (b) Taylor expanding $M_X(t)$ at $t=0$ when $-1 < t < 1$, we have

$$M_X(t) = \frac{1}{2(1-t^2)} = \frac{1}{2} + \frac{t^2}{2} + \frac{t^4}{2} + \frac{t^6}{2} + \dots = \sum_{k=0}^{\infty} \frac{1}{2} t^{2k} \quad (17)$$

Therefore,

odd-numbered moments are all 0, and the $2k$ -th moment is $\frac{1}{2} t^{2k}$