

HW #7

$$\begin{aligned} 1.) \quad M_X(t) &= 1 + \sum_{n=1}^{\infty} \frac{t^n}{n!} c \\ &= 1 + c \left[\sum_{n=0}^{\infty} \frac{t^n}{n!} - 1 \right] \\ &= 1 + c(e^t - 1) \end{aligned}$$

$$M_X(t) = ce^t + 1 - c$$

Which is the MGF for Bernoulli(c)

$$2.) \quad E[X^n] = \frac{2^n}{n+1} \quad \forall n \geq 1$$

$$\begin{aligned} M_X(t) &= \sum_{n=0}^{\infty} \frac{t^n (2^n)}{n! (n+1)} \\ &= \sum_{n=0}^{\infty} \frac{(2t)^n}{(n+1)!} \end{aligned}$$

$$= \frac{1}{2t} \sum_{n=0}^{\infty} \frac{(2t)^{n+1}}{(n+1)!} \quad \text{Let } i = n+1$$

$$= \frac{1}{2t} \sum_{i=1}^{\infty} \frac{(2t)^i}{i!}$$

$$= \frac{1}{2t} \left[\sum_{i=0}^{\infty} \frac{(2t)^i}{i!} - 1 \right]$$

$$= \frac{1}{2t} [e^{2t} - 1]$$

$$M_X(t) = \begin{cases} \frac{e^{2t} - 1}{2t}, & t \neq 0 \\ 1, & t = 0 \end{cases}$$

$M_X(0) = 1$ due to MGF restriction

$X \sim \text{Unif}(0, 2)$

3) a.) Let $Y = -X$ $g^{-1}(y) = -y$

$$\left| \frac{dg^{-1}(y)}{dy} \right| = |-1| = 1$$

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{dg^{-1}(y)}{dy} \right|$$

$$= f_X(-y)$$

$$= f_X(y)$$

Therefore X & Y have the same distribution.

b.) $M_X(0+\epsilon) = M_X(\epsilon) = \int_{-\infty}^{\infty} e^{\epsilon x} f_X(x) dx$

$$= \int_0^{\infty} e^{\epsilon x} f_X(x) dx + \int_{-\infty}^0 e^{\epsilon x} f_X(x) dx$$

Changing integration

$$= \int_{-\infty}^0 e^{-\epsilon x} f_X(-x) dx + \int_0^{\infty} e^{-\epsilon x} f_X(x) dx$$

$$= \int_{-\infty}^{\infty} e^{-\epsilon x} f_X(-x) dx$$

$$= \int_{-\infty}^{\infty} e^{-\epsilon x} f_X(x) dx \quad f_X \text{ Symmetric}$$

$$= M_X(-\epsilon)$$

$= M_X(0-\epsilon)$ Therefore $M_X(t)$ is symmetric about 0.

Question 4

There does not exist a distribution in which $M_X(t) = t/(1-t)$, $|t| < 1$

Since $M_X(t) = E(e^{tx})$,

then $M_X(0) = E(e^{0x})$

$$= E(1)$$

$$= 1$$

However, when $t=0$, $t/(1-t) = 0/(1-0) = 0 \neq 1$

Therefore, there does not exist such an mgf. ✓

Question 5

$$\frac{d}{dt} S(t) \Big|_{t=0} = \frac{d}{dt} (\log_e M_X(t)) \Big|_{t=0}$$

$$= \frac{M'_X(t)}{M_X(t)} \Big|_{t=0}$$

$$= \left(\frac{d}{dt} M_X(t) \Big|_{t=0} \right) / (M_X(t) \Big|_{t=0})$$

$$= (E(X)) / E(e^0)$$

$$= E(X) / 1$$

$$= E(X)$$

$$\frac{d^2}{dt^2} S(t) \Big|_{t=0} = \frac{d}{dt} \left(\frac{M'_X(t)}{M_X(t)} \Big|_{t=0} \right)$$

$$= \left(\frac{M''_X(t)}{M_X(t)} - \frac{(M'_X(t))^2}{(M_X(t))^2} \right) \Big|_{t=0}$$

$$= E(X^2) / 1 - (E(X) / 1)^2$$

$$= E(X^2) - (E(X))^2$$

$$= \text{Var}(X). \quad \checkmark$$

Question 6

(a) $M_X(t) = E(e^{tx})$

$$= \sum_{x=0}^{\infty} \frac{e^{-\lambda} \lambda^x}{x!} \cdot e^{tx}$$

$$= e^{-\lambda} \sum_{x=0}^{\infty} \frac{(\lambda e^t)^x}{x!}$$

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

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

$$= e^{\lambda(e^t - 1)}$$

$$E(X) = \frac{d}{dt} e^{\lambda(e^t - 1)} \Big|_{t=0} = \lambda e^t e^{\lambda(e^t - 1)} \Big|_{t=0} = \lambda(1)(1) = \lambda$$

$$\begin{aligned}
 E(X^2) &= \frac{d}{dt} \lambda e^t e^{\lambda(e^t-1)} \Big|_{t=0} \\
 &= \lambda e^t e^{\lambda(e^t-1)} + (\lambda e^t)^2 e^{\lambda(e^t-1)} \Big|_{t=0} \\
 &= \lambda(1)(1) + (\lambda(1))^2(1) \\
 &= \lambda + \lambda^2
 \end{aligned}$$

$$\text{Therefore, } \text{Var}(X) = E(X^2) - (E(X))^2 = \lambda + \lambda^2 - \lambda^2 = \lambda$$

$$\begin{aligned}
 \text{(b) } M_X(t) &= E(e^{tx}) \\
 &= \sum_{x=0}^{\infty} e^{tx} P(1-p)^x \\
 &= P \sum_{x=0}^{\infty} (e^t(1-p))^x \\
 &= (P) \left(\frac{1}{1 - e^t(1-p)} \right) \\
 &= \frac{P}{1 - e^t(1-p)}
 \end{aligned}$$

$$\begin{aligned}
 E(X) &= \frac{d}{dt} \frac{P}{1 - e^t(1-p)} \Big|_{t=0} = \frac{-P}{(1 - (1-p)e^t)^2} (- (1-p)e^t) \Big|_{t=0} \\
 &= \frac{-P}{(1 - 1 + p)^2} (-1 + p) = -\frac{1}{p} (- (1-p)) = \frac{1-p}{p}
 \end{aligned}$$

$$\begin{aligned}
 E(X^2) &= \frac{d}{dt} \left(\frac{-P}{(1 - (1-p)e^t)^2} (- (1-p)e^t) \right) \Big|_{t=0} \\
 &= \frac{(1 - (1-p)e^t)^2 (P(1-p)e^t) + P(1-p)e^t \cdot 2(1 - (1-p)e^t)(1-p)e^t}{(1 - (1-p)e^t)^4} \Big|_{t=0} \\
 &= \frac{P(1-p) + 2(1-p)^2}{p^2}
 \end{aligned}$$

$$\text{therefore, } \text{Var}(X) = \frac{P(1-p) + 2(1-p)^2}{p^2} - \frac{(1-p)^2}{p^2} = \frac{P(1-p) + (1-p)^2}{p^2} = \frac{1-p}{p^2}$$

$$\begin{aligned}
 \text{(c) } M_X(t) &= E(e^{tx}) \\
 &= \int_{-\infty}^{\infty} \frac{e^{-(x-\mu)^2/(2\sigma^2)}}{\sqrt{2\pi}\sigma} e^{tx} dx \\
 &= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{-(x-\mu)^2/(2\sigma^2)} e^{tx} dx \\
 &= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{(x - (\mu + \sigma^2 t))^2 - (2\mu\sigma^2 t + (\sigma^2 t)^2)} dx \\
 &= e^{\mu t + \frac{\sigma^2 t^2}{2}}
 \end{aligned}$$

$$\begin{aligned}
 E(X) &= \frac{d}{dt} e^{\mu t + \frac{\sigma^2 t^2}{2}} \Big|_{t=0} = (\mu + \sigma^2 t) e^{\mu t + \sigma^2 t^2/2} \Big|_{t=0} \\
 &= (\mu + 0) e^0 = \mu
 \end{aligned}$$

$$\begin{aligned}
 E(X^2) &= \frac{d}{dt} (\mu + \sigma^2 t) e^{\mu t + \sigma^2 t^2/2} \Big|_{t=0} \\
 &= (\mu + \sigma^2 t)^2 e^{\mu t + \sigma^2 t^2/2} + e^{\mu t + \sigma^2 t^2/2} \cdot \sigma^2 \Big|_{t=0} \\
 &= (\mu + 0)^2 e^0 + e^0 \sigma^2 = \mu^2 + \sigma^2
 \end{aligned}$$

$$\text{therefore, } \text{Var}(X) = E(X^2) - (E(X))^2 = \mu^2 + \sigma^2 - \mu^2 = \sigma^2$$

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7. C&B 2.36

$$\therefore f(x) = \frac{1}{\sqrt{2\pi}x} \cdot e^{-\frac{1}{2}(\ln x)^2}, \quad 0 \leq x < \infty$$

$$\therefore \text{MGF: } E(e^{tx}) = \int_0^{+\infty} f(x) \cdot e^{tx} dx = \int_0^{+\infty} \frac{1}{\sqrt{2\pi}x} \cdot e^{tx - \frac{1}{2}(\ln x)^2} dx$$

$$\therefore \lim_{x \rightarrow \infty} \frac{tx - \frac{1}{2}(\ln x)^2}{tx} \text{ is in } \frac{\infty}{\infty} \text{ form}$$

$$\therefore \lim_{x \rightarrow \infty} \frac{tx - \frac{1}{2}(\ln x)^2}{tx} = \lim_{x \rightarrow \infty} \frac{t - (\ln x) \cdot \frac{1}{x}}{t} \text{ by L'Hospital Rules.}$$

$$\therefore \lim_{x \rightarrow \infty} (\ln x) \cdot \frac{1}{x} \text{ is in } \frac{\infty}{\infty} \text{ form, } \therefore \lim_{x \rightarrow \infty} \frac{\ln x}{x} = \lim_{x \rightarrow \infty} \frac{\frac{1}{x}}{1} = 0$$

$$\therefore \lim_{x \rightarrow \infty} \frac{tx - \frac{1}{2}(\ln x)^2}{tx} = \lim_{x \rightarrow \infty} \frac{t - (\frac{\ln x}{x})}{t} = \frac{t}{t} = 1$$

$$\therefore \lim_{x \rightarrow \infty} tx - \frac{1}{2}(\ln x)^2 = \lim_{x \rightarrow \infty} tx = \infty$$

$$\therefore \forall k > 0, \exists c, c \text{ is constant. } \int_k^{+\infty} \frac{1}{\sqrt{2\pi}x} \cdot e^{tx - \frac{1}{2}(\ln x)^2} dx \geq c \cdot \int_k^{\infty} \frac{1}{\sqrt{2\pi}x} = \frac{1}{\sqrt{2\pi}} \ln x \Big|_k^{\infty} = \infty$$

$$\therefore \int_k^{+\infty} \frac{1}{\sqrt{2\pi}x} \cdot e^{-\frac{1}{2}(\ln x)^2} \cdot de^{tx} dx = \infty, \text{ does not exist.}$$

8. C&B 2.38

$$(a) \therefore f(x) = \binom{r+x-1}{x} p^r (1-p)^x, \quad x=0, 1, 2, \dots$$

$$\therefore \text{MGF: } E(e^{tx}) = \sum_{x=0}^{\infty} f(x) \cdot e^{tx} = \sum_{x=0}^{\infty} \binom{r+x-1}{x} p^r (1-p)^x \cdot e^{tx}$$

$$= \sum_{x=0}^{\infty} \binom{r+x-1}{x} p^r [(1-p) \cdot e^t]^x$$

$$= \sum_{x=0}^{\infty} \binom{r+x-1}{x} \cdot (1 - (1-p)e^t)^r (1-p)e^t \cdot \left(\frac{p}{1 - (1-p)e^t}\right)^r$$

$$= \left(\frac{p}{1 - (1-p)e^t}\right)^r \cdot \left[\sum_{x=0}^{\infty} \binom{r+x-1}{x} \cdot (1 - (1-p)e^t)^r (1-p)e^t \cdot \left(\frac{p}{1 - (1-p)e^t}\right)^r\right]$$

$$= \left(\frac{p}{1 - (1-p)e^t}\right)^r \cdot 1 = \left(\frac{p}{1 - (1-p)e^t}\right)^r, \quad \frac{p}{1 - (1-p)e^t} < 1 \Rightarrow t > 0$$

$$\therefore \text{MGF of } X \text{ is } M_X(t) = \left(\frac{p}{1 - (1-p)e^t}\right)^r, \quad t > 0$$

$$(b) \therefore Y = 2pX \quad \therefore \text{MGF of } Y \text{ is } M_Y(t) = E(e^{tY}) = E(e^{2ptX})$$

$$\therefore M_Y(t) = E(e^{2ptX}) = \sum_{x=0}^{\infty} f(x) \cdot e^{2ptx} = \sum_{x=0}^{\infty} \binom{r+x-1}{x} p^r (1-p)^x \cdot e^{2ptx}$$

$$= \sum_{x=0}^{\infty} \binom{r+x-1}{x} p^r \cdot (1-p) \cdot e^{2pt} \cdot \left(\frac{p}{1 - (1-p)e^{2pt}}\right)^r$$

$$= \left(\frac{p}{1 - (1-p)e^{2pt}}\right)^r \cdot \left[\sum_{x=0}^{\infty} \binom{r+x-1}{x} \cdot (1 - (1-p)e^{2pt})^r (1-p) \cdot e^{2pt} \cdot \left(\frac{p}{1 - (1-p)e^{2pt}}\right)^r\right]$$

$$= \left(\frac{p}{1 - (1-p)e^{2pt}}\right)^r \cdot 1 = \left(\frac{p}{1 - (1-p)e^{2pt}}\right)^r$$

8. (b) continued

$$\therefore \text{MGF of } Y \text{ is } M_Y(t) = \left(\frac{p}{1-(1-p)e^{2pt}} \right)^Y$$

$$\therefore \lim_{p \rightarrow 0} M_Y(t) = \lim_{p \rightarrow 0} \left(\frac{p}{1-(1-p)e^{2pt}} \right)^Y$$

$\therefore \lim_{p \rightarrow 0} \frac{p}{1-(1-p)e^{2pt}}$ is in $\frac{0}{0}$ form. by L'Hospital Rules,

$$\lim_{p \rightarrow 0} \frac{p}{1-(1-p)e^{2pt}} = \lim_{p \rightarrow 0} \frac{1}{e^{2pt} + (p-1) \cdot 2t \cdot e^{2pt}} = \frac{1}{1-2t}$$

$$\therefore \lim_{p \rightarrow 0} M_Y(t) = \left(\frac{1}{1-2t} \right)^Y,$$

$$\therefore M_Y(t) = E(e^{ty}) > 0, \quad \therefore \frac{1}{1-2t} > 0, \quad 0 < t < \frac{1}{2}$$

$$\therefore \lim_{p \rightarrow 0} M_Y(t) = \left(\frac{1}{1-2t} \right)^Y, \quad 0 < t < \frac{1}{2}$$

9. (a) $\therefore X \sim \text{ber}(p) \quad \therefore P(X=x) = p^x \cdot (1-p)^{1-x}, \quad x \in \{0, 1\}, \quad 0 \leq p \leq 1$

$$\therefore \phi_X(t) = E(e^{itX}) = \sum_{x=0}^1 P(X=x) \cdot e^{itx} \\ = p \cdot e^{it \cdot 1} + (1-p) \cdot e^{it \cdot 0} = p \cdot e^{it} + 1-p$$

$$\therefore \text{Let } q = 1-p, \quad \phi_X(t) = q + p \cdot e^{it}, \quad 0 \leq p \leq 1, q = 1-p.$$

(b) $\therefore X \sim \text{bin}(n, p) \quad \therefore P(X=x) = \binom{n}{x} \cdot p^x \cdot q^{n-x}, \quad x=0, 1, 2, \dots, n, \quad q=1-p, \quad 0 \leq p \leq 1$

$$\therefore \phi_X(t) = E(e^{itX}) = \sum_{x=0}^n P(X=x) \cdot e^{itx} \\ = \sum_{x=0}^n \binom{n}{x} \cdot p^x \cdot q^{n-x} \cdot e^{itx} = \sum_{x=0}^n \binom{n}{x} \cdot (p \cdot e^{it})^x \cdot q^{n-x} = (p \cdot e^{it} + q)^n$$

$$\therefore \phi_X(t) = (q + p e^{it})^n$$

(c) $\therefore X \sim \text{Geo}(p) \quad \therefore P(X=x) = (1-p)^x \cdot p, \quad x=0, 1, 2, 3, \dots$

$$\therefore \phi_X(t) = \sum_{x=0}^{\infty} P(X=x) \cdot e^{itx} = \sum_{x=0}^{\infty} (1-p)^x \cdot p \cdot e^{itx} \\ = \sum_{x=0}^{\infty} ((1-p) \cdot e^{it})^x \cdot p = \sum_{x=0}^{\infty} ((1-p)e^{it})^x \cdot (1-(1-p)e^{it}) \cdot \frac{p}{1-(1-p)e^{it}}$$

$$= \frac{p}{1-(1-p)e^{it}} \cdot \left[\sum_{x=0}^{\infty} ((1-p)e^{it})^x \cdot (1-(1-p)e^{it}) \right]$$

$$= \frac{p}{1-q \cdot e^{it}} \cdot 1, \quad \text{where } q = 1-p$$

$$\therefore \phi_X(t) = \frac{p}{1-q e^{it}}$$

$$9. (d) \because X \sim \text{Pois}(m), \quad \therefore P(X=x) = \frac{e^{-m} \cdot m^x}{x!}, \quad x=0, 1, 2, \dots$$

$$\therefore \phi_X(t) = E(e^{itX}) = \sum_{x=0}^{\infty} P(X=x) \cdot e^{itx}$$

$$= \sum_{x=0}^{\infty} \frac{e^{-m} \cdot m^x}{x!} \cdot e^{itx} = \sum_{x=0}^{\infty} \frac{e^{-m} \cdot (m \cdot e^{it})^x}{x!}$$

$$= \sum_{x=0}^{\infty} \frac{e^{-(m \cdot e^{it})} \cdot (m \cdot e^{it})^x}{x!} \cdot e^{m(e^{it}-1)}$$

$$= e^{m(e^{it}-1)} \left[\sum_{x=0}^{\infty} \frac{e^{-(m \cdot e^{it})} \cdot (m \cdot e^{it})^x}{x!} \right] = e^{m(e^{it}-1)} \cdot 1$$

$$= e^{m(e^{it}-1)}$$

$$\therefore \phi_X(t) = e^{m(e^{it}-1)}$$