

Total Points: 16

1. (4pts) Find $\text{Var}(aX + bY)$ and express it as simply as possible (don't make any assumptions about X and Y).

ANSWER:

$$\begin{aligned} \text{Var}(aX + bY) &= \text{Var}(aX) + \text{Var}(bY) + 2\text{Cov}(aX, bY) \\ &= a^2\text{Var}(X) + b^2\text{Var}(Y) + 2\text{Cov}(aX, bY) \\ &= a^2\text{Var}(X) + b^2\text{Var}(Y) + 2ab\text{Cov}(X, Y) \end{aligned}$$

(1pt) $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y)$

(1pt) $\text{Var}(aX) = a^2\text{Var}(X)$

(1pt) $\text{Cov}(aX, bY) = a\text{Cov}(X, Y)$

(1pt) for right solution

2. (7pts) Let X_1, X_2, \dots, X_n be iid Poisson(λ).

a. (3pts) What is the exact mgf of $S_n = \sum_{i=1}^n X_i$. Does this represent a named distribution? What is the mgf of $\bar{X}_n = \frac{S_n}{n}$?

ANSWER:

$$\begin{aligned} M_{S_n}(t) &= E[e^{tS_n}] \quad (\text{def. of mgf}) \\ &= E[e^{t(\sum_{i=1}^n X_i)}] \quad (\text{def. of } S_n) \\ &= E[e^{(tx_1 + tx_2 + \dots + tx_n)}] \\ &= E[e^{tx_1} \cdot e^{tx_2} \cdot \dots \cdot e^{tx_n}] \quad (\text{property of exponents}) \\ &= E[e^{tx_1}] E[e^{tx_2}] \dots E[e^{tx_n}] \quad (E[XY] = E[X]E[Y] \text{ if } X \text{ and } Y \text{ are independent}) \\ &= M_{X_1}(t) M_{X_2}(t) \dots M_{X_n}(t) \quad (\text{def. of mgf}) \\ &= (M_{X_1}(t))^n \quad (X_i \text{ are identically distributed same distribution} = \text{same mgf}) \\ &= (e^{\lambda(e^t - 1)})^n \quad (\text{mgf of poisson } (\lambda) \text{ i.v.}) \\ &= e^{n\lambda(e^t - 1)} \quad (\text{property of exponents}) \quad \square \end{aligned}$$

(1pt) for correctly calculating mgf

$$\text{mgf of } \bar{X}_n = E[e^{t\bar{X}_n}]$$

$$= E[e^{t\frac{S_n}{n}}]$$

$$= E[e^{\frac{t}{n}S_n}]$$

$$= E[e^{\frac{t}{n}X_1 + \frac{t}{n}X_2 + \dots + \frac{t}{n}X_n}]$$

$$= E[e^{\frac{t}{n}X_1}]E[e^{\frac{t}{n}X_2}] \dots E[e^{\frac{t}{n}X_n}]$$

$$= M_{X_1}(\frac{t}{n})M_{X_2}(\frac{t}{n}) \dots M_{X_n}(\frac{t}{n})$$

$$= (M_{X_1}(\frac{t}{n}))^n$$

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

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

(cont'd: Is this a known distribution?)

Let $\lambda' = n\lambda$, then

Mgf $S_n(t) = e^{\lambda'(e^t - 1)}$ which is the mgf of a poisson(λ') r.v.

$\therefore S_n \sim \text{poisson}(n\lambda)$

1pt for recognizing

1pt

b. (2pts) What is another mgf that \bar{X}_n should be close to? Why?

The mgf of \bar{X}_n should be close to the mgf for a normal

a normal $(\lambda, \frac{\lambda}{n})$ random variable

$\frac{1}{2}$ pt normal
 $\frac{1}{2}$ pt $\mu = \lambda$ $\sigma^2 = \frac{\lambda}{n}$

... this is because for large n the mean of an iid random sample \rightarrow normal $(\mu, \frac{\sigma^2}{n})$ by the central limit theorem

1pt reason is the CLT

c. (2pts) Compare the mgf of exact distribution (from a) to the approximate distribution (from b). What happens as $n \rightarrow \infty$?

ANSWER: $\lim_{n \rightarrow \infty} M_{\bar{X}_n}(t) = \lim_{n \rightarrow \infty} e^{n\lambda(e^{\frac{t}{n}} - 1)}$

$= e^{t\lambda}$ W.A.

$\lim_{n \rightarrow \infty} M_{N(\lambda, \frac{1}{n})}(t) = \lim_{n \rightarrow \infty} e^{(\lambda t + \frac{1}{2} \frac{\lambda}{n} t^2)}$

$= e^{t\lambda}$ W.A.

(1pt) mgf of a normal $(\lambda, \frac{1}{n})$ r.v

(1pt) for showing that the limits both go to $e^{t\lambda}$ as $n \rightarrow \infty$

Although the mgf look different, they converge to the same value in the limit

3. (5pts) Let X be a random variable with pdf $f(x) = 630x^4(1-x)^4, x \in (0, 1)$

a. (2pts) Find $E(X)$ and $Var(X)$

ANSWER: $E(X) = \int_0^1 x f(x) dx$
 $= \int_0^1 x 630x^4(1-x)^4 dx$
 $= \int_0^1 630x^5(1-x)^4 dx$
 $= \frac{1}{2}$

(1pt)

$$\begin{aligned}
\text{Var}(x) &= E(x^2) - E(x)^2 \\
&= E(x^2) - \frac{1}{4} \\
&= \int_0^1 x^2 (630x^4(1-x)^4) dx - \frac{1}{4} \\
&= \int_0^1 630x^6(1-x)^4 dx - \frac{1}{4} \\
&= \frac{3}{11} - \frac{1}{4} \\
&= \frac{1}{44} \approx 0.0227
\end{aligned}$$

1pt

b. (2pts) Obtain the lower bound given by Chebyshev's inequality for $P(0.2 < X < 0.8)$

ANSWER:

$$\begin{aligned}
P(0.2 < X < 0.8) &= P\left(|X - \frac{1}{2}| < 0.3\right) \\
&= P(|X - E(X)| < 0.3) \\
&= P\left(|X - E(X)| < \frac{0.3}{\sqrt{\frac{1}{44}}} \cdot \sqrt{\frac{1}{44}}\right) \\
&= P(|X - E(X)| < 1.99 \sqrt{\text{Var}(X)})
\end{aligned}$$

1pt for $k = 1.99$ or $\frac{0.3}{\sqrt{\text{Var}(X)}}$

$$= 1 - P(|X - E(X)| \geq 1.99 \sqrt{\text{Var}(X)})$$

1pt for taking probab of the complement to use chebyshev

$$\geq 1 - \frac{1}{(1.99)^2} \quad (\text{by Chebyshev's } \neq)$$

$$\geq 0.747 \quad (\text{approximately})$$

c. (1pt) Compute the exact $P(0.2 < X < 0.8)$

ANSWER:

$$\begin{aligned}
P(0.2 < X < 0.8) &= \int_{0.2}^{0.8} f(x) dx = \int_{0.2}^{0.8} 630x^4(1-x)^4 dx \\
&= 0.961 \quad \text{W.A.}
\end{aligned}$$

1pt for using correct integral w/ correct bounds