

University of Michigan, Dept of Statistics

Stats 510, Instructor: Long Nguyen

Homework 3

Oct 7, 2024, due by 11:59pm Oct 16, 2025

Notes: You may select to solve both problem 4 and problem 5 given below, or only one of them. If you decide the latter, then please toss a coin. If you get a head, do problem 4, otherwise, do problem 5.

1. (Expectations) Do problems 2.14, 2.17, 2.22.
2. (Moments) Do problems 2.30, 2.38, 2.39.
3. Let X be a standard normal variable, i.e., X has pdf $f_X(x) = \frac{1}{\sqrt{2\pi}}e^{-x^2/2}$ for all $x \in (-\infty, +\infty)$. Define a discrete random variable Y by $P(Y = \sqrt{3}) = P(Y = -\sqrt{3}) = 1/6$ and $P(Y = 0) = 2/3$.
 - (i) Show that X and Y have the same r -th moment, for each $r = 1, 2, 3, 4, 5$.
 - (ii) Find another random variable Z which has the same r -th moments as X and Y for all $r = 1, \dots, 5$.

4. Let $n \in \mathbb{N}$, define function f_{X_n} as follows: For $x \in \{0, 1, \dots, n\}$,

$$f_{X_n}(x) = (1/2) \binom{n}{x} (3n)^{-x} (1 - 1/(3n))^{n-x} + (1/2) \binom{n}{x} (6n)^{-x} (1 - 1/(6n))^{n-x}.$$

and $f_{X_n}(x) = 0$ otherwise.

- (i) Verify that f_{X_n} is a valid pmf for a discrete random variable, which we denote by X_n . Derive the moment generating function for X_n .
 - (ii) Show that the sequence of random variables $\{X_n\}$ converges in distribution to a random variable Y . What is the distribution of Y ?
5. Let $n \in \mathbb{N}$, define function f_{X_n} as follows: For $x \in \mathbb{R}$,

$$f_{X_n}(x) = (1/2) \frac{1}{\sqrt{2\pi/n}} e^{-\frac{n}{2}x^2} + (1/2) \frac{1}{\sqrt{2\pi/n}} e^{-\frac{n}{2}(x-1)^2}.$$

- (i) Verify that f_{X_n} is a valid pdf for a continuous random variable, which we denote by X_n .
- (ii) Show that the sequence of random variables $\{X_n\}$ converges in distribution to a random variable Y . What is the distribution of Y ?

1. (Expectations) Do problems 2.14, 2.17, 2.22.

- 2.14 (a) Let X be a continuous, nonnegative random variable [$f(x) = 0$ for $x < 0$]. Show that

$$E(X) = \int_0^\infty [1 - F_X(x)] dx,$$

where $F_X(x)$ is the cdf of X .

- (b) Let X be a discrete random variable whose range is the nonnegative integers. Show that

$$E(X) = \sum_{k=0}^\infty (1 - F_X(k)),$$

where $F_X(k) = P(X \leq k)$. Compare this with part (a).

$$\begin{aligned} \text{(a)} \int_0^\infty (1 - F_X(x)) dx &= \int_0^\infty P(X > x) dx \\ &= \int_0^\infty \int_x^\infty f_X(y) dy dx \\ &= \int_0^\infty \left(\int_x^\infty 1(x > y) f_X(y) dy \right) dx \\ &= \int_0^\infty \left(\int_0^\infty 1(x > y) f_X(y) dy \right) dx \\ &= \int_0^\infty y f_X(y) dy = E(X). \end{aligned} \quad \begin{aligned} \text{(b)} E(X) &= \sum_{k=0}^\infty k P(X=k) \\ &= \sum_{k=0}^\infty \frac{k}{k+1} P(X=k) \quad , \quad k = \sum_{j=1}^k 1 \\ &= \sum_{k=0}^\infty \frac{1}{k+1} (1 - P(X \leq k)) \\ &= \sum_{k=1}^\infty (1 - P(X \leq k-1)) \\ &= \sum_{k=1}^\infty 1 - P(X \leq k-1) \\ &= \sum_{k=0}^\infty (1 - F_X(k)) \quad , \quad k = j-1. \end{aligned}$$

- 2.17 A median of a distribution is a value m such that $P(X \leq m) \geq \frac{1}{2}$ and $P(X \geq m) \geq \frac{1}{2}$. (If X is continuous, m satisfies $\int_{-\infty}^m f(x) dx = \int_m^\infty f(x) dx = \frac{1}{2}$.) Find the median of the following distributions.

(a) $f(x) = 3x^2, \quad 0 < x < 1$

(b) $f(x) = \frac{1}{\pi(1+x^2)}, \quad -\infty < x < \infty$

(a) $\int_0^m 3x^2 dx = x^3 \Big|_0^m = m^3 = \frac{3}{2} \Rightarrow m = \left(\frac{3}{2}\right)^{\frac{1}{3}} = 0.794$

(b) $f(x)$ is the Cauchy pdf and therefore symmetric about zero, i.e.

$$\int_{-\infty}^m f(x) dx = \int_m^\infty f(x) dx \Rightarrow m = 0$$

to show formally:

$$\begin{aligned} \int_{-\infty}^m f(x) dx &= \frac{1}{\pi} \int_{-\infty}^m \frac{1}{1+x^2} dx = \frac{1}{\pi} \tan^{-1}(x) \Big|_{-\infty}^m = \frac{1}{\pi} (\tan^{-1}(m) + \frac{\pi}{2}) \stackrel{m \rightarrow 0}{\rightarrow} \frac{\pi}{2} \\ &\Rightarrow \frac{\pi}{2} = \tan^{-1}(m) + \frac{\pi}{2} \Rightarrow \tan^{-1}(m) = 0 \Rightarrow m = 0. \end{aligned}$$

- 2.22 Let X have the pdf

$$f(x) = \frac{4}{\beta^3 \sqrt{\pi}} x^2 e^{-x^2/\beta^2}, \quad 0 < x < \infty, \quad \beta > 0.$$

- (a) Verify that $f(x)$ is a pdf.

- (b) Find $E(X)$ and $\text{Var}(X)$.

- (a) I will show (i) nonnegativity and (ii) normalization, i.e.

i) For $x > 0$ and $\beta > 0$, every factor is nonnegative $\Rightarrow f(x) > 0$

ii) Now, I must show $\int_0^\infty f(x) dx = \frac{4}{\beta^3 \sqrt{\pi}} \int_0^\infty x^2 e^{-x^2/\beta^2} dx = 1$:

First, compute $A := \int_0^\infty x^2 e^{-x^2/\beta^2} dx$ via IBP

$$\begin{aligned} \text{let } u = x &\Rightarrow du = dx \\ \text{and } v = \int x e^{-x^2/\beta^2} dx & \end{aligned}$$

$$w = -x^2/\beta^2$$

$$dw = -2x/\beta^2 dx \Rightarrow x dx = -\frac{1}{2} dw$$

$$= \int e^w \left(-\frac{1}{2} dw\right) = -\frac{1}{2} \int e^w dw = -\frac{1}{2} e^{-x^2/\beta^2}$$

$$= u w \Big|_0^\infty - \int_0^\infty v dw$$

$$\text{where } \lim_{w \rightarrow \infty} w(-\frac{1}{2} e^{-x^2/\beta^2}) = 0 \text{ and } uw|_{w=0} = 0 \Rightarrow uw|_{w=0} = 0$$

$$= - \int_0^\infty \left(-\frac{1}{2} e^{-x^2/\beta^2}\right) dx$$

$$= \frac{1}{2} \int_0^\infty e^{-x^2/\beta^2} dx$$

$$\text{let } u = x/\beta \Rightarrow x = \beta u \Rightarrow dx = \beta du$$

$$= \frac{1}{2} \int_0^\infty e^{-(\beta u)^2/\beta^2} \beta du = \frac{1}{2} \int_0^\infty e^{-u^2} \beta du$$

$$= \beta \cdot \frac{\sqrt{\pi}}{2} \text{ since } \int_{-\infty}^\infty e^{-u^2} du = \sqrt{\pi}$$

Hence,

$$A = \frac{\beta^2}{2} \cdot \beta \frac{\sqrt{\pi}}{2} = \frac{\beta^3 \sqrt{\pi}}{4}.$$

Therefore,

$$\int_0^\infty f(x) dx = \frac{4}{\beta^3 \sqrt{\pi}} A = \frac{4}{\beta^3 \sqrt{\pi}} \cdot \frac{\beta^3 \sqrt{\pi}}{4} = 1.$$

- (b) Find $E(X) = \int_0^\infty x f(x) dx$ and $\text{Var}(X) = E(X^2) - [E(X)]^2$ where $E(X^2) = \int_0^\infty x^2 f(x) dx$

$$\begin{aligned} \text{compute } B := \int_0^\infty x^2 e^{-x^2/\beta^2} dx \text{ via IBP} \\ u = x^2 \quad dv = x e^{-x^2/\beta^2} dx \\ du = 2x dx \quad v = \int x e^{-x^2/\beta^2} dx \end{aligned}$$

$$= \frac{-x^3}{3} e^{-x^2/\beta^2} \text{ as seen in (a)}$$

$$= uv \Big|_0^\infty - \int_0^\infty v du$$

$$= 0 - \int_0^\infty \left(-\frac{x^3}{3} e^{-x^2/\beta^2}\right) x \beta^2 dx$$

$$= \beta^2 \int_0^\infty x^2 e^{-x^2/\beta^2} dx \text{ via substitution}$$

$$u = x^2/\beta^2$$

$$du = 2x/\beta^2 dx \Rightarrow x dx = \frac{\beta^2}{2} dw$$

$$\text{where } \int_0^\infty x^2 e^{-x^2/\beta^2} dx = \frac{\beta^3}{2} \int_0^\infty e^{-w} dw = \frac{\beta^3}{2} \text{ since } \left[-e^{-w}\right]_0^\infty = 1$$

$$= \beta^2 \int_0^\infty e^{-w} dw$$

$$= \beta^2 \cdot \frac{\sqrt{\pi}}{2} = \frac{\beta^3 \sqrt{\pi}}{4}$$

$$\text{so, } E(X) = \frac{4}{\beta^3 \sqrt{\pi}} B = \frac{4}{\beta^3 \sqrt{\pi}} \cdot \frac{\beta^3 \sqrt{\pi}}{4} = \frac{\beta}{\sqrt{\pi}}$$

$$\begin{aligned} \text{now compute } C := \int_0^\infty x^4 e^{-x^2/\beta^2} dx \text{ via IBP} \\ \text{let } u = x^4 \Rightarrow du = 4x^3 dx \\ v = x^2 e^{-x^2/\beta^2} \Rightarrow v = \int x^2 e^{-x^2/\beta^2} dx \end{aligned}$$

$$= \frac{-x^5}{5} e^{-x^2/\beta^2} \text{ as seen previously}$$

$$= uv \Big|_0^\infty - \int_0^\infty v du$$

$$= \int_0^\infty \left(\frac{x^5}{5} e^{-x^2/\beta^2}\right) 4x^3 dx$$

$$= \frac{3\beta^5}{2} \int_0^\infty x^2 e^{-x^2/\beta^2} dx$$

$$= \frac{3\beta^5}{2} A$$

$$= \frac{3\beta^5}{2} \cdot \frac{\beta^3 \sqrt{\pi}}{4} = \frac{3\beta^8 \sqrt{\pi}}{8}.$$

$$\text{Hence, } E(X^2) = \frac{4}{\beta^3 \sqrt{\pi}} C = \frac{4}{\beta^3 \sqrt{\pi}} \cdot \frac{3\beta^8 \sqrt{\pi}}{8} = \frac{3}{2} \beta^5.$$

$$\begin{aligned} \text{therefore, } \text{Var}(X) &= E(X^2) - [E(X)]^2 = \frac{3}{2} \beta^5 - \left(\frac{\beta}{\sqrt{\pi}}\right)^2 = \beta^2 \left(\frac{3}{2} - \frac{1}{\pi}\right) \end{aligned}$$

2.39 In each of the following cases calculate the indicated derivatives, justifying all operations.

$$(a) \frac{d}{dx} \int_0^x e^{-\lambda t} dt$$

$$(b) \frac{d}{d\lambda} \int_0^\infty e^{-\lambda t} dt$$

$$(c) \frac{d}{dt} \int_t^1 \frac{1}{x^2} dx$$

$$(d) \frac{d}{dt} \int_1^\infty \frac{1}{(x-t)^2} dx$$

$$\text{a) verify } \frac{d}{dx} \left[\int_0^x e^{-\lambda t} dt \right] = \frac{d}{dx} \int_0^x \frac{1}{\lambda} e^{-\lambda t} \left[\frac{t}{\lambda} \right] = \frac{d}{dx} \left[\left(\frac{-1}{\lambda} e^{-\lambda t} + \frac{1}{\lambda} \right) \right] = \frac{-1}{\lambda} e^{-\lambda x} + e^{-\lambda x}.$$

$$\text{b) verify } \frac{d}{d\lambda} \int_0^\infty e^{-\lambda t} dt = \frac{d}{d\lambda} \left[\frac{-1}{\lambda} e^{-\lambda t} \left[\frac{t}{\lambda} \right] \right] = \frac{d}{d\lambda} \left[0 + \frac{1}{\lambda} \right] = \frac{d}{d\lambda} \left(\lambda^{-1} \right) = -\frac{1}{\lambda^2}.$$

$$\text{c) verify } \frac{d}{dt} \left[\int_t^1 \frac{1}{x^2} dx \right] = \frac{d}{dt} \left[-\frac{1}{x} \right] = \frac{d}{dt} \left[-1 \cdot \frac{1}{t} \right] = -\frac{1}{t^2}.$$

$$\text{d) verify } \frac{d}{dt} \int_1^\infty \frac{1}{(x-t)^2} dx = \frac{d}{dt} \left[-\frac{1}{t-1} \right] = \frac{1}{(t-1)^2}.$$

3. Let X be a standard normal variable, i.e., X has pdf $f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$ for all $x \in (-\infty, +\infty)$. Define a discrete random variable Y by $P(Y = \sqrt{3}) = P(Y = -\sqrt{3}) = 1/6$ and $P(Y = 0) = 2/3$.

- (i) Show that X and Y have the same r -th moment, for each $r = 1, 2, 3, 4, 5$.
(ii) Find another random variable Z which has the same r -th moments as X and Y for all $r = 1, \dots, 5$.

$$\text{RECALL: } \forall n \in \mathbb{N}, \quad E(X^n) = \frac{d^n}{dt^n} M_X(t) \Big|_{t=0}$$

For $X \sim \text{N}(0, 1)$,

$$M_X(t) = E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = \int_{-\infty}^{\infty} \exp(tx - \frac{x^2}{2}) dx.$$

Now complete the square in the exponent:

$$tx - \frac{x^2}{2} = \frac{1}{2} (x^2 - 2tx) = \frac{1}{2} (x - t)^2 + \frac{t^2}{2}.$$

$$\text{So, } M_X(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp\left(-\frac{(x-t)^2}{2} + \frac{t^2}{2}\right) dx = e^{t^2/2} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-(x-t)^2/2} dx$$

$$= e^{t^2/2} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-(u-t)^2/2} du \quad \text{(*standard normal*)}$$

Thus, $M_X(t) = e^{t^2/2}$. Now evaluate $M_X^{(1)}(0)$, $M_X^{(2)}(0)$, ..., $M_X^{(5)}(0)$:

$$E(X) := M_X^{(1)}(0) = \frac{d}{dt} e^{t^2/2} \Big|_{t=0} = t e^{t^2/2} \Big|_{t=0} = 0$$

$$E(X^2) := M_X^{(2)}(0) = \frac{d}{dt} (t e^{t^2/2}) \Big|_{t=0} = e^{t^2/2} + t(t e^{t^2/2}) \Big|_{t=0} = (1+t^2) e^{t^2/2} \Big|_{t=0} = 1.$$

$$E(X^3) := M_X^{(3)}(0) = \frac{d}{dt} ((1+t^2) e^{t^2/2}) \Big|_{t=0} = 2t e^{t^2/2} + (t+t^3) \cdot (t e^{t^2/2}) \Big|_{t=0} = (2t+t^3) e^{t^2/2} \Big|_{t=0} = 0$$

$$E(X^4) := M_X^{(4)}(0) = \frac{d}{dt} ((2t+t^3) e^{t^2/2}) = (5+3t^2) e^{t^2/2} + (3t+t^3) \cdot (t e^{t^2/2}) = (5+6t^2+t^4) e^{t^2/2} \Rightarrow M_X^{(4)}(0) = 3$$

$$E(X^5) := M_X^{(5)}(0) = 0 \text{ since odd moments vanish by symmetry (as seen above).}$$

For discrete Y given by $P(Y = \sqrt{3}) = P(Y = -\sqrt{3}) = \frac{1}{6}$ and $P(Y = 0) = \frac{2}{3}$,

$$\begin{aligned} M_Y(t) &= \sum_y t^y P(Y=y) = \frac{1}{6} t^{\sqrt{3}} + \frac{1}{6} t^{-\sqrt{3}} + \frac{2}{3} t^0 \\ &= \frac{1}{6} (t^{\sqrt{3}} + t^{-\sqrt{3}}) + \frac{1}{3}, \\ \text{where } \cosh(u) &= \frac{e^u + e^{-u}}{2} \\ &= \frac{1}{3} \cosh(\sqrt{3}t) + \frac{1}{3} \end{aligned}$$

of interest,

$$\frac{d}{dt} \cosh(u) = \sinh(u), \quad \frac{d}{dt} \sinh(u) = \cosh(u), \quad \cosh(0) = 1, \quad \sinh(0) = 0$$

Given $M_Y(t) = \frac{1}{3} \cosh(\sqrt{3}t) + \frac{1}{3}$, find $E(Y)$, $E(Y^2)$, ..., $E(Y^5)$:

$$E(Y) := M_Y^{(1)}(0) = \frac{1}{3} (t^{\sqrt{3}}) \sinh(\sqrt{3}t) \Big|_{t=0} = \frac{1}{3} (t^0) \cdot 0 = 0$$

$$E(Y^2) := M_Y^{(2)}(0) = \frac{1}{3} (t^{\sqrt{3}})^2 \cosh(\sqrt{3}t) \Big|_{t=0} = \frac{1}{3} (t^0)^2 = 1$$

$$E(Y^3) := M_Y^{(3)}(0) = \frac{1}{3} (t^{\sqrt{3}})^3 \sinh(\sqrt{3}t) \Big|_{t=0} = \sqrt{3} \sinh(\sqrt{3}t) \Big|_{t=0} = 0$$

$$E(Y^4) := M_Y^{(4)}(0) = \frac{1}{3} (t^{\sqrt{3}})^4 \cosh(\sqrt{3}t) \Big|_{t=0} = 3 \cosh(\sqrt{3}t) \Big|_{t=0} = 3$$

$$E(Y^5) := M_Y^{(5)}(0) = 0 \text{ since odd moments vanish by symmetry}$$

Define $Z := P(Z=1) = P(Z=-1) = \frac{1}{6}$, $P(Z=2) = P(Z=-2) = \frac{1}{12}$, and $P(Z=0) = \frac{1}{2}$.

$$\text{then, } M_Z(t) = E(t^Z) = \frac{1}{6}(t^2 + e^{-t^2}) + \frac{1}{12}(e^{t^2} + e^{-t^2}) + \frac{1}{2} = \frac{1}{3} \cosh(t) + \frac{1}{6} \cosh(2t) + \frac{1}{2}$$

Next, we can compute $E(Z)$, $E(Z^2)$, ..., $E(Z^5)$.

As discussed in (i), the odd moments vanish by symmetry, i.e.

$$E(Z) = E(Z^3) = E(Z^5) = 0, \text{ and}$$

$$E(Z^2) := M_Z^{(2)}(0) = \frac{1}{3} \cosh(t) + \frac{1}{6} \cosh(2t) \Big|_{t=0} = 1$$

$$E(Z^4) := M_Z^{(4)}(0) = \frac{1}{3} \cosh(2t) + \frac{1}{6} \cosh(4t) \Big|_{t=0} = \frac{1}{3} + \frac{1}{3} = 3.$$

Therefore, Z contains the same r -th moments as X and Y for $r = 1, \dots, 5$.

4. Let $n \in \mathbb{N}$, define function f_{X_n} as follows: For $x \in \{0, 1, \dots, n\}$,

$$f_{X_n}(x) = (1/2)\binom{n}{x}(3n)^{-x}(1 - 1/(3n))^{n-x} + (1/2)\binom{n}{x}(6n)^{-x}(1 - 1/(6n))^{n-x}.$$

and $f_{X_n}(x) = 0$ otherwise.

- (i) Verify that f_{X_n} is a valid pmf for a discrete random variable, which we denote by X_n . Derive the moment generating function for X_n .
- (ii) Show that the sequence of random variables $\{X_n\}$ converges in distribution to a random variable Y . What is the distribution of Y ?

(i) Each summand is a binomial pmf: $\text{Bin}(n, \frac{1}{3n})$ and $\text{Bin}(n, \frac{1}{6n})$. Their avg. is a valid pmf.

Recall: For $X \sim \text{Bin}(n, p)$, $M_X(s) = [1 - p + ps]^n$.

Here, the avg. is a mixture of mgfs, i.e.

$$M_{X_n}(s) = \frac{1}{2}(1 - \frac{1}{3n} + \frac{s}{3n})^n + \frac{1}{2}(1 - \frac{1}{6n} + \frac{s}{6n})^n = \frac{1}{2}(1 + \frac{3s-1}{3n})^n + \frac{1}{2}(1 + \frac{6s-1}{6n})^n.$$

(ii) As $n \rightarrow \infty$, $(1 + \frac{s}{n})^n \rightarrow e^s$. Hence,

$$M_{X_n}(s) \rightarrow \frac{1}{2} \exp\left(\frac{3s-1}{3}\right) + \frac{1}{2} \exp\left(\frac{6s-1}{6}\right),$$

the mgf of the mixture $\frac{1}{2}\text{Poi}(\frac{1}{3}) + \frac{1}{2}\text{Poi}(\frac{1}{6})$.

Therefore, $X_n \xrightarrow{d} Y$ where Y is the Poisson mixture.

5. Let $n \in \mathbb{N}$, define function f_{X_n} as follows: For $x \in \mathbb{R}$,

$$f_{X_n}(x) = (1/2)\frac{1}{\sqrt{2\pi/n}}e^{-\frac{n}{2}x^2} + (1/2)\frac{1}{\sqrt{2\pi/n}}e^{-\frac{n}{2}(x-1)^2}.$$

- (i) Verify that f_{X_n} is a valid pdf for a continuous random variable, which we denote by X_n .

- (ii) Show that the sequence of random variables $\{X_n\}$ converges in distribution to a random variable Y . What is the distribution of Y ?

For $n \in \mathbb{N}$, define

$$f_{X_n}(x) = \frac{1}{\sqrt{2\pi/n}}e^{-\frac{n}{2}x^2} + \frac{1}{\sqrt{2\pi/n}}e^{-\frac{n}{2}(x-1)^2}, \quad x \in \mathbb{R} \quad (1)$$

where each (equally weighted) term is the density of a normal curve, i.e.

$$\frac{1}{\sqrt{2\pi/n}}e^{-\frac{n}{2}x^2} \sim N(0, \frac{1}{n}) \quad \text{and} \quad (2)$$

$$\frac{1}{\sqrt{2\pi/n}}e^{-\frac{n}{2}(x-1)^2} \sim N(1, \frac{1}{n}) \quad (3)$$

(i) Verify that f_{X_n} is a valid pdf, i.e.

$$\Leftrightarrow f_{X_n}(x) \geq 0 \quad \text{for all } x \in \mathbb{R} \quad \text{and} \quad \Leftrightarrow \int_{-\infty}^{\infty} f_{X_n}(x) dx = 1.$$

Nonnegativity (1) holds since (1) is comprised of two nonnegative Gaussian densities, i.e.

$$(2) \geq 0 \quad \forall x \in \mathbb{R} \quad \text{and} \quad (3) \geq 0 \quad \forall x \in \mathbb{R}. \quad \text{Thus, (1)} \geq f_{X_n} \geq 0 \quad \forall x \in \mathbb{R}.$$

We can also verify (1), i.e.

$$\int_{-\infty}^{\infty} f_{X_n}(x) dx = \frac{1}{2} \int_{-\infty}^0 \frac{1}{\sqrt{2\pi/n}}e^{-\frac{n}{2}x^2} dx + \frac{1}{2} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi/n}}e^{-\frac{n}{2}(x-1)^2} dx = \frac{1}{2} + \frac{1}{2} = 1. \quad \square$$

(ii) By definition, the MGF of X_n is

$$\begin{aligned} M_{X_n}(t) &= E(e^{tX_n}) = \int_{-\infty}^{\infty} e^{tx_n} f_{X_n}(x) dx \\ &= \frac{1}{2} \sqrt{\frac{n}{\pi}} \int_{-\infty}^0 e^{tx_n} e^{-\frac{n}{2}x^2} dx + \frac{1}{2} \sqrt{\frac{n}{\pi}} \int_{-\infty}^{\infty} e^{tx_n} e^{-\frac{n}{2}(x-1)^2} dx. \end{aligned}$$

To compute the first integral

$$I_1 := \sqrt{\frac{n}{\pi}} \int_{-\infty}^0 e^{tx_n} e^{-\frac{n}{2}x^2} dx,$$

compute the square in the exponent, i.e.

$$tx_n - \frac{n}{2}x^2 = -\frac{n}{2}(x^2 - \frac{2t}{n}x) = -\frac{n}{2}((x - \frac{t}{n})^2 - \frac{t^2}{n^2}) = -\frac{n}{2}(x - \frac{t}{n})^2 + \frac{t^2}{2n}.$$

Therefore,

$$I_1 = \sqrt{\frac{n}{\pi}} \int_{-\infty}^0 e^{t^2/2n} e^{-\frac{n}{2}(x - \frac{t}{n})^2} dx$$

$$\text{Let } u = x - \frac{t}{n} \Rightarrow du = dx$$

$$\text{where } \int_{-\infty}^0 e^{-\frac{n}{2}u^2} du = \sqrt{\frac{\pi}{n}}$$

$$\Rightarrow I_1 = \sqrt{\frac{n}{\pi}} e^{t^2/2n} \sqrt{\frac{\pi}{n}} = e^{t^2/2n}$$

Similarly, for the second integral

$$I_2 := \sqrt{\frac{n}{\pi}} \int_{-\infty}^{\infty} e^{tx_n - \frac{n}{2}(x-1)^2} dx,$$

$$tx_n - \frac{n}{2}(x-1)^2 = t(x-1) + t - \frac{n}{2}(x-1)^2.$$

So,

$$\begin{aligned} I_2 &= e^{t^2/2n} \int_{-\infty}^{\infty} e^{t(x-1) - \frac{n}{2}(x-1)^2} dx = e^{t^2/2n} \int_{-\infty}^{\infty} e^{st - \frac{n}{2}s^2} ds, \quad \text{Letting } s = x-1 \\ &= e^{t^2/2n} e^{s^2/2n} \sqrt{\frac{\pi}{n}} \quad \text{by the same above calculation} \\ &= e^{t^2/2n} e^{s^2/2n}. \end{aligned}$$

Thus,

$$M_{X_n}(t) = \frac{1}{2} e^{t^2/2n} + \frac{1}{2} e^{t^2/2n} e^{s^2/2n} = \frac{1}{2} e^{t^2/2n} (1 + e^{s^2}).$$

$$\Rightarrow \lim_{n \rightarrow \infty} M_{X_n}(t) = \lim_{n \rightarrow \infty} \frac{1}{2} e^{t^2/2n} (1 + e^{s^2}) = \frac{1}{2} (1 + e^{t^2}).$$

Now, define

$$M_Y(t) := \frac{1}{2} (1 + e^{t^2}) \quad \text{is the mgf of } Y \sim \text{Bernoulli}(1/2)$$

$$\text{since } E(e^{tX}) = \frac{1}{2} e^{t+1} + \frac{1}{2} e^{t-1}.$$

This implies that

$$X_n \xrightarrow{d} Y \quad \text{for Bernoulli}(1/2) \quad \text{by the mgf convergence theorem.}$$