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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

$$EX = \int_0^{\infty} [1 - F_X(x)] dx,$$

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

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

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

$$\begin{aligned} \Rightarrow \int_0^{\infty} (1 - F_X(x)) dx &= \int_0^{\infty} P(X > x) dx = \int_0^{\infty} \int_x^{\infty} f_X(y) dy dx \\ &= \int_0^{\infty} \left( \int_0^x 1(x \leq y) f_X(y) dy \right) dx \\ &= \int_0^{\infty} \left( \int_0^x 1(x \leq y) f_X(y) dy \right) dy \\ &= \int_0^{\infty} f_X(y) \left( \int_0^y 1(x \leq y) dx \right) dy \\ &= \int_0^{\infty} y f_X(y) dy = EX. \end{aligned}$$

$$\begin{aligned} \Rightarrow EX &= \sum_{k=0}^{\infty} k P(X=k) \\ &= \sum_{k=0}^{\infty} \sum_{j=1}^k P(X=k) \quad , \quad k = \sum_{j=1}^k 1 \\ &= \sum_{k=0}^{\infty} \sum_{j=1}^k 1(x \leq y) P(X=k) \\ &= \sum_{j=1}^{\infty} \sum_{k=j}^{\infty} P(X=k) \\ &= \sum_{j=1}^{\infty} P(X \geq j) \\ &= \sum_{j=1}^{\infty} (1 - P(X \leq j-1)) \\ &= \sum_{j=1}^{\infty} (1 - F_X(j-1)) \\ &= \sum_{k=0}^{\infty} (1 - F_X(k)) \quad , \quad k=j-1. \end{aligned}$$

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

b)  $f(x)$  is the Cauchy pdf and therefore symmetric about  $x=0$ , i.e.

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

To show formally:

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

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  $EX$  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  $\Rightarrow f(x) \geq 0$

ii) Now, I must show  $\int_{-\infty}^{\infty} f(x) dx = 1$

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

let  $u = x, dv = x e^{-x^2/\beta^2} dx$   
 $du = dx, v = \int x e^{-x^2/\beta^2} dx$

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

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

$$= \int_0^{\infty} e^w \left( -\frac{\beta^2}{2} dw \right) = -\frac{\beta^2}{2} \int_0^{\infty} e^w dw = -\frac{\beta^2}{2} e^w \Big|_0^{\infty} = \frac{\beta^2}{2}$$

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

$$\text{where } uv \Big|_0^{\infty} = \left( -\frac{\beta^2}{2} e^{-x^2/\beta^2} \right) \Big|_0^{\infty} = 0 \text{ and } uv \Big|_0^{\infty} = 0 \Rightarrow uv \Big|_0^{\infty} = 0$$

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

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

let  $y = x/\beta \Rightarrow x = \beta y \Rightarrow dx = \beta dy$

$$= \frac{\beta^2}{2} \int_0^{\infty} e^{-\beta^2 y^2 / \beta^2} \beta dy = \beta \int_0^{\infty} e^{-y^2} dy$$

$$= \beta \cdot \frac{\sqrt{\pi}}{2} \quad \text{since } \int_0^{\infty} e^{-y^2} dy = \frac{\sqrt{\pi}}{2}$$

Hence,

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

Thus,

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

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

compute  $B := \int_0^{\infty} x^3 e^{-x^2/\beta^2} dx$  via IBP

$$u = x^2, dv = x e^{-x^2/\beta^2} dx$$

$$du = 2x dx, v = \int x e^{-x^2/\beta^2} dx$$

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

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

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

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

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

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

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

$$= \beta^2 \cdot \frac{\beta^2}{2} = \frac{\beta^4}{2}$$

so,

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

Now compute  $C := \int_0^{\infty} x^4 e^{-x^2/\beta^2} dx$  via IBP

$$\text{let } u = x^3, dv = x e^{-x^2/\beta^2} dx$$

$$du = 3x^2 dx, v = -\frac{\beta^2}{2} e^{-x^2/\beta^2} \text{ as seen previously}$$

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

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

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

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

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

Hence,

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

Therefore,

$$\text{Var } X = E(X^2) - [E(X)]^2 = \frac{3}{2} \beta^2 - \left( \frac{2\beta}{\sqrt{\pi}} \right)^2 = \beta^2 \left( \frac{3}{2} - \frac{4}{\pi} \right)$$

## 2. (Moments) Do problems 2.30, 2.38, 2.39.

**2.30** Find the moment generating function corresponding to

(a)  $f(x) = \frac{1}{c}, \quad 0 < x < c.$

(b)  $f(x) = \frac{2x}{c^2}, \quad 0 < x < c.$

(c)  $f(x) = \frac{1}{2\beta} e^{-|x-\alpha|/\beta}, \quad -\infty < x < \infty, \quad -\infty < \alpha < \infty, \quad \beta > 0.$

(d)  $P(X=x) = \binom{r+x-1}{x} p^r (1-p)^x, \quad x=0,1,\dots, \quad 0 < p < 1, r > 0 \text{ an integer.}$

4)  $E(e^{tx}) = \int_0^c e^{tx} \cdot \frac{1}{c} dx = \frac{1}{c} e^{tx} \Big|_0^c = \frac{1}{c} \frac{e^{tc} - 1}{t} = \frac{e^{tc} - 1}{ct}$

6)  $E(e^{tx}) = \int_0^c \frac{2x}{c^2} e^{tx} dx = \frac{1}{c^2} \int_0^c 2xe^{tx} dx$   
 $\stackrel{\text{IBP}}{=} \frac{1}{c^2} \left[ \frac{2x}{t} e^{tx} - \int_0^c \frac{2}{t} e^{tx} dx \right]$   
 $= \frac{1}{c^2} \left[ \frac{2}{t} e^{tc} - \frac{2}{t^2} (e^{tc} - 1) \right]$   
 $= \frac{2}{c^2} \left[ \frac{tc}{2} e^{tc} - \left( \frac{1}{t^2} e^{tc} - \frac{1}{t^2} \right) \right]$   
 $= \frac{2e^{tc}}{c^2} - \frac{2e^{tc}}{c^2 t^2} + \frac{2}{c^2 t^2} = \frac{1}{c^2 t^2} (2te^{tc} - e^{tc} + 1)$

8)  $E(e^{tx}) = \int_{-\infty}^{\infty} \frac{1}{2\beta} e^{-(|x-\alpha|/\beta)} dx = \int_{-\infty}^{\alpha} \frac{1}{2\beta} e^{-(\alpha-x)/\beta} dx + \int_{\alpha}^{\infty} \frac{1}{2\beta} e^{-(x-\alpha)/\beta} dx$

Since  $\int_{-\infty}^{\infty} f(|x-\alpha|) dx = \int_{-\infty}^{\alpha} f(\alpha-x) dx + \int_{\alpha}^{\infty} f(x-\alpha) dx$  PS

$= \int_{-\infty}^{\alpha} \frac{1}{2\beta} e^{-(\alpha-x)/\beta} dx + \int_{\alpha}^{\infty} \frac{1}{2\beta} e^{-(x-\alpha)/\beta} dx$

$= \frac{e^{-\alpha/\beta}}{2\beta} \left[ e^{x/\beta} \right]_{-\infty}^{\alpha} - \frac{e^{-\alpha/\beta}}{2\beta} \left[ e^{-x/\beta} \right]_{\alpha}^{\infty}$  (1)

$= \left( \frac{e^{-\alpha/\beta + \alpha/\beta + \alpha t}}{2\beta(\frac{1}{\beta} + t)} - 0 \right) - \left( 0 - \frac{e^{-\alpha/\beta - \alpha/\beta + \alpha t}}{2\beta(\frac{1}{\beta} - t)} \right)$

$= \frac{e^{\alpha t}}{2\beta(\frac{1}{\beta} + t)} + \frac{e^{\alpha t}}{2\beta(\frac{1}{\beta} - t)}$

$= \frac{e^{\alpha t}}{2 + 2\beta t} + \frac{e^{\alpha t}}{2 - 2\beta t}$

$= \frac{(2 - 2\beta t) e^{\alpha t} + (2 + 2\beta t) e^{\alpha t}}{(2 + 2\beta t)(2 - 2\beta t)}$

$= \frac{2e^{\alpha t} - 2\beta t e^{\alpha t} + 2e^{\alpha t} + 2\beta t e^{\alpha t}}{4 - 4\beta^2 t^2} = \frac{4e^{\alpha t}}{4 - 4\beta^2 t^2}$

$\frac{4e^{\alpha t}}{4 - 4\beta^2 t^2}, \quad -1/\beta < t < 1/\beta$  s.t. (1) converges, i.e. the integral can only be evaluated approximately when  $e^{-(\frac{1}{\beta} + t)} \Big|_{x=-\infty}^{\alpha} > 0$  iff  $\frac{1}{\beta} + t < 0 \Rightarrow t < -\frac{1}{\beta}$  and  $e^{-x(\frac{1}{\beta} - t)} \Big|_{x=\infty}^{\alpha} > 0$  iff  $\frac{1}{\beta} - t < 0 \Rightarrow t > \frac{1}{\beta}$ .

4)  $E(e^{tx}) = \sum_{x=0}^{\infty} \binom{r+x-1}{x} p^r (1-p)^x e^{tx}$   
 $= \sum_{x=0}^{\infty} \binom{r+x-1}{x} p^r (1-p)^x e^{tx}$   
 $= \sum_{x=0}^{\infty} \binom{r+x-1}{x} p^r (1-p)^x e^{tx}$   
 $= \frac{p^r}{(1-(1-p)e^t)^r} \sum_{x=0}^{\infty} \binom{r+x-1}{x} (1-(1-p)e^t)^x$   
 $\stackrel{\text{Since above Negative Binomial (r, 1-(1-p)e^t) sums to 1}}{=} \frac{p^r}{(1-(1-p)e^t)^r}$  since  $0 < (1-(1-p)e^t) < 1$  w/c  $1-(1-p)e^t$  is a prob.  
 $\Rightarrow 0 < 1-(1-p)e^t < 1$   
 $\Rightarrow 1-(1-p)e^t < 1$  since  $1-(1-p)e^t > 0$   
 $\Rightarrow \log(1-p) + t < \log(1) = 0$   
 $\Rightarrow t < -\log(1-p)$

PS:  $\ln(AB) = \ln(A) + \ln(B)$   
 $\Rightarrow \log(1-p) + t < \log(1) = 0$   
 $\Rightarrow t < -\log(1-p)$

**2.38** Let  $X$  have the negative binomial distribution with pmf

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

where  $0 < p < 1$  and  $r > 0$  is an integer.

(a) Calculate the mgf of  $X$ .

(b) Define a new random variable by  $Y = 2pX$ . Show that, as  $p \downarrow 0$ , the mgf of  $Y$  converges to that of a chi squared random variable with  $2r$  degrees of freedom by showing that

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

4)  $M_X(t) = E(e^{tx}) = \frac{p^r}{(1-(1-p)e^t)^r}$  (as completed above in 2.30d)

6) Let  $Y = 2pX$ . Then,

$$M_Y(t) = M_{2pX}(t) = M_X(2pt) = \frac{p^r}{(1-(1-p)e^{2pt})^r}$$

Since  $\forall a, b \in \mathbb{R}, M_{a+bX}(t) = e^{at} M_X(bt)$

Next,

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

$$= \lim_{p \rightarrow 0} \left( \frac{p}{\frac{d}{dp} [1-(1-p)e^{2pt}] (0)} \right)^r$$

by L'Hopital's rule:  $\lim_{p \rightarrow 0} \frac{f(p)}{g(p)} = \lim_{p \rightarrow 0} \frac{f'(p)}{g'(p)}$  indeterminate

where  $(0) = -(1-p)(2t e^{2pt}) + e^{2pt}(1-p)$   
 $= -e^{2pt}(2t(1-p) - 1) = e^{2pt}(1-2t(1-p))$

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

$$= \left( \frac{1}{1-2t} \right)^r, \quad |t| < \frac{1}{2}.$$

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

ⓐ Verify  $\frac{d}{dx} \left[ \int_0^x e^{-\lambda t} dt \right] = \frac{d}{dx} \left[ -\frac{1}{\lambda} e^{-\lambda t} \Big|_0^x \right] = \frac{d}{dx} \left( -\frac{1}{\lambda} e^{-\lambda x} + \frac{1}{\lambda} \right) = -\frac{1}{\lambda} (-\lambda) e^{-\lambda x} = e^{-\lambda x}$ .

ⓑ Verify  $\frac{d}{d\lambda} \int_0^\infty e^{-\lambda t} dt = \frac{d}{d\lambda} \left[ -\frac{1}{\lambda} e^{-\lambda t} \Big|_0^\infty \right] = \frac{d}{d\lambda} \left( 0 + \frac{1}{\lambda} \right) = \frac{d}{d\lambda} (\lambda^{-1}) = -\frac{1}{\lambda^2}$ .

ⓒ Verify  $\frac{d}{dt} \left[ \int_t^1 \frac{1}{x^2} dx \right] = \frac{d}{dt} \left[ -\frac{1}{x} \Big|_t^1 \right] = \frac{d}{dt} \left( -1 + \frac{1}{t} \right) = -\frac{1}{t^2}$ .

ⓓ Verify  $\frac{d}{dt} \int_1^\infty \frac{1}{(x-t)^2} dx = \frac{d}{dt} \left[ -\frac{1}{(x-t)} \Big|_1^\infty \right] = \frac{d}{dt} \left( -\frac{1}{1-t} \right) = \frac{1}{(1-t)^2}$ .

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  $P(Y = 0) = 2/3$ .

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

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

ⓐ For  $X \sim N(0,1)$ ,

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

Now complete the square in the exponent:

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

So,

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

(Let  $u = x - t$ ,  $du = dx$ )

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

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

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

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

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

$$E(X^4) := M_X^{(4)}(0) = \frac{d}{dt} ((3t+t^3) e^{\frac{t^2}{2}}) = (3+3t^2) e^{\frac{t^2}{2}} + (3t+t^3) (t e^{\frac{t^2}{2}}) = (3+6t^2+t^4) e^{\frac{t^2}{2}} \Big|_{t=0} = 3$$

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

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

$$M_Y(t) = E(e^{tY}) = \frac{1}{6} e^{\sqrt{3}t} + \frac{1}{6} e^{-\sqrt{3}t} + \frac{2}{3} e^0 = \frac{1}{6} (e^{\sqrt{3}t} + e^{-\sqrt{3}t}) + \frac{2}{3},$$

where  $\cosh(u) = \frac{e^u + e^{-u}}{2}$

$$= \frac{1}{3} \cosh(\sqrt{3}t) + \frac{2}{3}.$$

Of interest,

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

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

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

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

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

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

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

ⓑ Define  $Z := P(Z=1) = P(Z=-1) = \frac{1}{6}$ ,  $P(Z=2) = P(Z=-2) = \frac{1}{6}$ , and  $P(Z=0) = \frac{2}{3}$ .

Then,

$$M_Z(t) = E(e^{tZ}) = \frac{1}{6} (e^t + e^{-t}) + \frac{1}{6} (e^{2t} + e^{-2t}) + \frac{2}{3} = \frac{1}{3} \cosh(t) + \frac{1}{3} \cosh(2t) + \frac{2}{3}$$

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

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

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

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

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

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.

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

- 1) 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  $X \sim \text{Bin}(n, p)$ ,  $M_X(t) = [ (1-p) + pe^t ]^n$ .

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

$$M_{X_n}(t) = \frac{1}{2} \left( 1 - \frac{1}{3n} + \frac{e^t}{3n} \right)^n + \frac{1}{2} \left( 1 - \frac{1}{6n} + \frac{e^t}{6n} \right)^n = \frac{1}{2} \left( 1 + \frac{e^t - 1}{3n} \right)^n + \frac{1}{2} \left( 1 + \frac{e^t - 1}{6n} \right)^n.$$

- 2) As  $n \rightarrow \infty$ ,  $(1 + \frac{t}{n})^n \rightarrow e^t$ . Hence,

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

the avg 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  $x \in \mathbb{R}$ ,

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

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

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

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

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

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

- 1) verify that  $f_{X_n}$  is a valid pdf, i.e.

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

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

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

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

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

- ii) By definition, the MGF of  $X_n$  is

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

To compute the first integral

$$I_1 := \sqrt{\frac{n}{2\pi}} \int_{-\infty}^{\infty} e^{tx} e^{-\frac{n}{2}x^2} dx.$$

complete the square in the exponent, i.e.

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

Therefore,

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

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

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

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

Similarly, for the second integral

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

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

So,

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

Thus,

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

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

Now, define

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

$$\text{since } E(e^{tY}) = \frac{1}{2} e^{0 \cdot t} + \frac{1}{2} e^{1 \cdot t}.$$

This implies that

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