# STA2001 Assignment 3

Due Date: June 27, 2023

(3.1-3). Customers arrive randomly at a bank teller's window. Given that one customer arrived during a particular 10-minute period, let X equal the time within the 10 minutes that the customer arrived. If X is U(0, 10), find

- (a) The pdf of X.
- (b)  $P(X \ge 8)$ .
- (c)  $P(2 \le X < 8)$ .
- (d) E(X).
- (e) Var(X).

Solution:

- (a) The pdf is  $f(x) = \frac{1}{10}$ , 0 < x < 10
- (b)  $P(X \ge 8) = \int_8^{10} \frac{1}{10} dx = 0.2$
- (c)  $P(2 \le X < 8) = \int_2^8 \frac{1}{10} dx = 0.6$
- (d)  $E(X) = \frac{0+10}{2} = 5$
- (e)  $Var(X) = \frac{(10-0)^2}{12} = \frac{25}{3}$

(3.1-5). Let Y have a uniform distribution U(0,1), and let

$$W = a + (b - a)Y, \quad a < b.$$

(a) Find the cdf of W.

Hint: Find  $P[a + (b - a)Y \le w]$ .

(b) How is W distributed?

Solution:

(a) Note that pdf of Y is just 1. Let G be the cdf of W, then for a < w < b

$$G(w) = P(W \le w) = P(a + (b - a)Y \le w) = P\left(Y \le \frac{w - a}{b - a}\right) = \int_0^{\frac{w - a}{b - a}} 1 \, dy = \frac{w - a}{b - a}$$

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and G(w) = 0 for  $w \le a$  and G(w) = 1 for  $w \ge b$ .

(b) From the cdf of W in (a), we know that  $W \sim U(a,b)$ .

(3.1-6). A grocery store has n watermelons to sell and makes \$1.00 on each sale. Say the number of consumers of these watermelons is a random variable with a distribution that can be approximated by

$$f(x) = \frac{1}{200}, \quad 0 < x < 200,$$

a pdf of the continuous type. If the grocer does not have enough watermelons to sell to all consumers, she figures that she loses \$5.00 in goodwill from each unhappy customer. But if she has surplus watermelons, she loses 50 cents on each extra watermelon. What should n be to maximize profit?

Hint: If  $X \le n$ , then her profit is (1.00)X + (-0.50)(n - X); but if X > n, her profit is (1.00)n + (-5.00)(X - n). Find the expected value of profit as a function of n, and then select n to maximize that function.

### Solution:

From the hint, we let g(X) be the profit function and write

$$g(X) = \begin{cases} (1.00)X + (-0.50)(n - X), & 0 \le X \le n \\ (1.00)n + (-5.00)(X - n), & 200 \ge X > n \end{cases}$$

where we have used the fact that the continuous random variable X only has positive density on (0, 200).

So the mean of g(X) can be further written as

$$E(g(X)) = \int_0^{200} g(x)f(x)dx$$

$$= \left(\int_0^n (1.5x - 0.5n) dx + \int_n^{200} (-5x + 6n) dx\right) \cdot \frac{1}{200}$$

$$= \frac{1}{200} \left(-\frac{13}{4}n^2 + 1200n - 100000\right)$$

Using the first order condition (i.e. set the first derivative as zero),

$$\frac{d}{dn}E(g(X)) = \frac{d}{dn}\frac{1}{200}\left(-\frac{13}{4}n^2 + 1200n - 100000\right) = \frac{1}{200}\left(-\frac{13}{2}n + 1200\right) = 0$$

and solve the equation we yields

$$n = \frac{2400}{13} = 184.6153 \approx 185$$

Note that we could check that it's indeed a maximum point as the function E(g(X)) is concave in n, i.e. its second derivative is negative.

(3.2-1). What are the pdf, the mean, and the variance of X if the moment-generating function of X is given by the following?

(a) 
$$M(t) = \frac{1}{1-3t}, t < 1/3$$

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(b)  $M(t) = \frac{3}{3-t}, t < 3.$ 

#### Solution:

(a) By matching the moment generating function, we know that X has an exponential distribution with parameter  $\theta = 3$ . The corresponding pdf is given by

$$f(x) = \frac{1}{3}e^{-x/3}, \quad 0 \le x < \infty$$

and zero elsewhere.

Since it's an exponential distribution, its mean and variance are

$$E(X) = \theta = 3, \quad Var(X) = \theta^2 = 9$$

(b) By matching the moment generating function, we know that X has an exponential distribution with parameter  $\theta = 1/3$ . The corresponding pdf is given by

$$f(x) = 3e^{-3x}, \quad 0 \le x < \infty$$

and zero elsewhere.

Since it's an exponential distribution, its mean and variance are

$$E(X) = \theta = \frac{1}{3}, \quad Var(X) = \theta^2 = \frac{1}{9}$$

Note: we could also compute the moments (and thus the variance) by using the property  $E(X^k)$  $M^{(k)}(t)|_{t=0}$ , for positive integers k and a given mgf. This is a general method and it's important in our course.

(3.2-3). Let X have an exponential distribution with mean  $\theta > 0$ . Show that

$$P(X > x + y | X > x) = P(X > y)$$

for any x > 0.

#### Solution:

For x>0 and y<0, it's obvious that P(X>x+y|X>x)=1=P(X>y), as the cdf of an exponential distribution says that P(X > y) = 1 for any y < 0.

Now we consider x > 0 and y > 0. By Bayes' theorem,

$$P(X > x + y | X > x) = \frac{P(X > x + y)}{P(X > x)} = \frac{e^{-(x+y)/\theta}}{e^{-x\theta}} = e^{-y\theta} = P(X > y)$$

which completes the proof.

(3.2-7). Find the moment-generating function for the gamma distribution with parameters  $\alpha$  and  $\theta$ . Hint: In the integral representing  $E(e^{tX})$ , change variables by letting  $y = (1 - \theta t)x/\theta$ , where  $1 - \theta t > 0.$ 

**Solution:** Let f(x) be the pdf of the given Gamma distribution, then by definition of the MGF,

$$\begin{split} M(t) &= E\left[e^{tX}\right] \\ &= \int_0^{+\infty} e^{tx} f(x) dx \\ &= \int_0^{+\infty} e^{tx} \cdot \frac{x^{\alpha - 1} e^{-x/\theta}}{\Gamma(\alpha) \theta^{\alpha}} dx \\ &= \frac{1}{\Gamma(\alpha) \theta^{\alpha}} \int_0^{+\infty} x^{\alpha - 1} e^{-x/\theta} e^{tx} dx \\ &= \frac{1}{\Gamma(\alpha) \theta^{\alpha}} \int_0^{+\infty} x^{\alpha - 1} e^{-x(1 - \theta t)/\theta} dx \\ &= \frac{1}{\Gamma(\alpha) \theta^{\alpha}} \int_0^{+\infty} \left(\frac{\theta u}{1 - \theta t}\right)^{\alpha - 1} e^{-u} \frac{\theta}{(1 - \theta t)} du \\ &= \frac{\theta^{\alpha}}{\Gamma(\alpha) \theta^{\alpha} (1 - \theta t)^{\alpha}} \int_0^{+\infty} u^{\alpha - 1} e^{-u} du \\ &= \frac{\theta^{\alpha}}{\Gamma(\alpha) \theta^{\alpha} (1 - \theta t)^{\alpha}} \Gamma(\alpha) \\ &= \frac{1}{(1 - \theta t)^{\alpha}} \end{split}$$

where we let  $u = (1 - \theta t)x/\theta$  in the 6th equality and thus  $du = \frac{(1 - \theta t)}{\theta}dx$ . The final results is obtained by noticing that  $\Gamma(\alpha) := \int_0^{+\infty} x^{\alpha-1} e^{-x} dx$ .

(3.2-11). If X is  $\chi^2(17)$ , find

- (a) P(X < 7.564)
- (b) P(X > 27.59)
- (c) P(6.408 < X < 27.59)
- (d)  $\chi_{0.95}^2(17)$ (e)  $\chi_{0.025}^2(17)$

#### Solution:

Since X is  $\chi^2(17)$  and degree of freedom is r=17, we can check the chi-square distribution table and find that

- (a) P(X < 7.564) = 0.025
- (b)  $P(X > 27.59) = 1 P(X \le 27.59) = 1 0.95 = 0.05$
- (c) P(6.408 < X < 27.59) = P(X < 27.59) P(X < 6.408) = 0.95 0.01 = 0.94
- (d)  $\chi_{0.95}^2(17) = 8.672$
- (e)  $\chi^2_{0.025}(17) = 30.19$

(3.2-22). Let X have a logistic distribution with pdf

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

Show that

$$Y = \frac{1}{1 + e^{-X}}$$

has a U(0,1) distribution.

Hint: Find  $G(y) = P(Y \le y) = P\left(\frac{1}{1 + e^{-X}} \le y\right)$ , where 0 < y < 1.

## Solution:

We aim to use the cdf of X to compute the cdf of Y according to the hint. So we compute the cdf of X first, which is obtained by

$$F_X(x) = \int_{-\infty}^x \frac{e^{-w}}{(1 + e^{-w})^2} dw = \frac{1}{1 + e^{-x}}, -\infty < x < \infty$$

Note that by definition of Y and  $-\infty < x < \infty$ , we must have 0 < Y < 1.

$$G(y) = P\left(\frac{1}{1 + e^{-X}} \le y\right)$$

$$= P\left(1 + e^{-X} \ge \frac{1}{y}\right)$$

$$= P\left(X \le -\ln\left(\frac{1}{y} - 1\right)\right)$$

$$= F_X\left(-\ln\left(\frac{1}{y} - 1\right)\right)$$

$$= \frac{1}{1 + e^{\ln(1/y - 1)}}$$

$$= y$$

where 0 < y < 1. This shows that  $Y \sim U(0, 1)$ .

(3.3-10). If X is  $N(\mu, \sigma^2)$ , show that the distribution of Y = aX + b is  $N(a\mu + b, a^2\sigma^2)$   $a \neq 0$ . Hint: Find the cdf  $P(Y \leq y)$  of Y, and in the resulting integral, let w = ax + b or, equivalently, x = (w - b)/a.

## Solution:

Let's first consider the case a > 0.

$$G(y) = P(Y \le y)$$

$$= P(aX + b \le y)$$

$$= P\left(X \le \frac{y - b}{a}\right)$$

$$= \int_{-\infty}^{(y-b)/a} \frac{1}{\sigma\sqrt{2\pi}} e^{-(x-\mu)^2/2\sigma^2} dx$$

$$= \int_{-\infty}^{y} \frac{1}{a\sigma\sqrt{2\pi}} e^{-(w-b-a\mu)^2/2\sigma^2a^2} dw$$

where we let w = ax + b in the last equality and thus dw = adx.

For a < 0, the derivation is quite similar, except that the standard deviation should be  $-a\sigma$ .

The final result shows that it is indeed a normal cumulative distribution function of  $N\left(b+a\mu,\sigma^2a^2\right)$ .

(3.3-14). The strength X of a certain material is such that its distribution is found by  $X = e^Y$ , where Y is N(10, 1). Find the cdf and pdf of X, and compute P(10, 000 < X < 20, 000).

Note:  $F(x) = P(X \le x) = P(e^Y \le x) = P(Y \le \ln x)$  so that the random variable X is said to have a lognormal distribution.

#### Solution:

WLOG, let  $Y \sim N(\mu, \sigma^2)$ , then the pdf and cdf of Y are given by

$$f_Y(y) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(y-\mu)^2/2\sigma^2}$$
$$F_Y(y) = \int_{-\infty}^{y} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(z-\mu)^2/2\sigma^2} dz$$

Then, the cdf of X can be computed by

$$F_X(x) = P(X \le x) = P(e^Y \le x) = P(Y \le \ln x) = F_Y(\ln x) = \int_{-\infty}^{\ln x} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(z-\mu)^2/2\sigma^2} dz$$

The pdf of X is the derivative of the cdf of X,

To compute P(10000 < X < 20000), we write

$$\begin{split} P(10000 < X < 20000) &= P(X \le 20000) - P(X \le 10000) \\ &= P\left(e^Y \le 20000\right) - P\left(e^Y \le 10000\right) \\ &= P(Y \le \ln 20000) - P(Y \le \ln 10000) \\ &= P(Z < -0.10) - P(Z < -0.79) \\ &= 1 - P(Z < 0.10) - (1 - P(Z < 0.79)) \\ &= 1 - 0.5398 - (1 - 0.7852) \\ &= 0.2454 \end{split}$$

where we have used the fact  $Z = \frac{Y - \mu}{\sigma} \sim N(0, 1)$ .

$$f_X(x) = F_X'(x)$$

$$= \frac{d}{dx} \left( \int_{-\infty}^{\ln x} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(z-\mu)^2/2\sigma^2} dz \right)$$

$$= \frac{d}{dx} \left( \int_0^x \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(\ln t - \mu)^2/2\sigma^2} \frac{1}{t} dt \right)$$

$$= \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(\ln x - \mu)^2/2\sigma^2} \left( \frac{1}{x} \right)$$

$$= \frac{1}{x\sqrt{2\pi\sigma^2}} e^{-(\ln x - \mu)^2/2\sigma^2}$$

where we have used  $t = e^z$  in the 3rd equality and thus  $dz = 1/u \cdot du$ , and the derivative is obtained by the Fundamental theorem of calculus (or you can finish these two steps in one trial by using the Leibniz integral rule).

Finally we set  $\mu = 10$  and  $\sigma = 1$ ,

$$F_X(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\ln x} e^{-(z-10)^2/2} dz$$
$$f_X(x) = \frac{1}{x\sqrt{2\pi}} e^{-(\ln x - 10)^2/2}$$