

# Solutions to Bain and Engelhardt's Introduction to Probability and Mathematical Statistics

## 06.01

**Given: the pdf of x**  $f_x(x) = \begin{cases} 4x^3 & , \quad 0 < x < 1 \\ 0 & , \quad o/w \end{cases}$

**Find: PDF of a)  $Y = X^4$**

**Setup:** Use the CDF technique to get the CDF of Y in terms of a CDF of X  
 $F_Y(y) = P[Y \leq y] = P[X^4 \leq Y] = P[-y^{\frac{1}{4}} \leq X \leq y^{\frac{1}{4}}] = F_X(y^{\frac{1}{4}}) - F_X(-y^{\frac{1}{4}})$

**Steps: i)** Differentiate with respect to y to find an equation given in terms of the pdf of x:  
 $f_y(y) = \frac{d}{dy}F_X(y^{\frac{1}{4}}) - \frac{d}{dy}F_X(-y^{\frac{1}{4}}) = f_x(y^{\frac{1}{4}})\frac{d}{dy}y^{\frac{1}{4}} - f_x(-y^{\frac{1}{4}})\frac{d}{dy}(-y^{\frac{1}{4}}) = f_x(y^{\frac{1}{4}})\frac{y^{-\frac{3}{4}}}{4} - f_x(-y^{\frac{1}{4}})\frac{-y^{-\frac{3}{4}}}{4}$

**ii)** Plug in the original limits and function for the pdf of x, and compute the cdf for y

**Result:**  $f_y(y) = \begin{cases} 4y^{\frac{3}{4}}\frac{1}{4y^{\frac{3}{4}}} & , \quad 0 < x < 1 \\ 0 & , \quad o/w \end{cases} = \begin{cases} 1 & , \quad 0 < x < 1 \\ 0 & , \quad o/w \end{cases}$

**Find: PDF of b)  $W = e^X$**

**Setup:** Use the CDF technique to get the CDF of W in terms of a CDF of X  
 $F_W(w) = P[W \leq w] = P[e^X \leq W] = P[X \leq \ln W] = F_X(\ln W)$

**Steps: i)** Differentiate with respect to w to find an equation given in terms of the pdf of x:  
 $f_w(w) = \frac{d}{dw}F_X(\ln W)\frac{d}{dw}(\ln w) = f_x(\ln w)\frac{1}{w}$

**ii)** Plug in the original limits and function for the pdf of x, and compute the cdf for y

**Result:**  $f_w(w) = \begin{cases} \frac{4(\ln w)^3}{w} & , \quad 1 < w < e \\ 0 & , \quad o/w \end{cases}$

**Find: PDF of c)  $Z = \ln x$**

**Setup:** Use the CDF technique to get the CDF of Z in terms of a CDF of X  
 $F_Z(z) = P[Z \leq z] = P[\ln x \leq z] = P[X \leq e^z] = F_X(e^z)$

**Steps: i)** Differentiate with respect to z to find an equation given in terms of the pdf of x:  
 $f_z(z) = \frac{d}{dz}F_X(e^z) = f_x(e^z)\frac{de^z}{dz}$

ii) Plug in the original limits and function for the pdf of x, and compute the cdf for y

**Result:** 
$$f_Z(z) = \begin{cases} 4e^{4z} & , \quad -\infty \leq z < 0 \\ 0 & , \quad o/w \end{cases}$$

**Find: PDF of d)  $U = (X - 0.5)^2$**

**Setup:** Use the CDF technique to get the CDF of U in terms of a CDF of X

$$F_U(u) = P[U \leq u] = P[(X - 0.5)^2 \leq u] = P[|X - 0.5| \leq u^{0.5}] = F_X(u^{1/2} + 1/2) - F_X(-u^{1/2} + 1/2)$$

**Steps: i)** Differentiate with respect to u to find an equation given in terms of the pdf of x:

$$f_U(u) = \frac{d}{du} F_X(u^{1/2} + 1/2) = f_x(u^{1/2} + 1/2) \frac{d}{du} (u^{1/2} + 1/2) - f_x(-u^{1/2} + 1/2) \frac{d}{du} (-u^{1/2} + 1/2)$$

$$f_x(u^{1/2} + 1/2) 1/2 u^{-1/2} - f_x(-u^{1/2} + 1/2) 1/2 u^{-1/2}$$

ii) INCOMPLETE

**Result:** 
$$f_Z(z) = \begin{cases} 4e^{4z} & , \quad -\infty \leq z < 0 \\ 0 & , \quad o/w \end{cases}$$

06.02

**Given:**  $X \sim Unif(0, 1)$

**Find: a) PDF of  $Y = X^{1/4}$**

**Setup:**  $F_Y(y) = P[Y \leq y] = P[X^{1/4} \leq y] = P[X \leq y^4] = F_X(y^4)$

**Steps: i)** find the pdf of x. Because X is a Uniform distribution with parameters 1 and 0, the pdf, which for Unif(a,b) is  $1/(b-a)$  where  $a < x < b$ . Here, Unif(0,1) gives  $1/1-0 = 1$

ii) Differentiate with respect to y to find an equation given in terms of the pdf of x:

$$f_Y(y) = \frac{d}{dy} F_X(y^4) = 4y^3$$

**Result:** 
$$f_Y(y) = \begin{cases} 4y^3 & , \quad 0 < y < 1 \\ 0 & , \quad o/w \end{cases}$$

**Find: b) PDF of  $W = e^{-X}$**

**Setup:**  $F_W(z) = P[W \leq w] = P[e^{-X} \leq w] = P[-X \leq \ln w] = P[X \geq -\ln w] = 1 - F_x(-\ln w)$

**Steps: i)** find the pdf of x. See part a) for an explanation of why it is 1 when  $a < x < b$

ii) Differentiate with respect to  $w$  to find an equation given in terms of the pdf of  $x$ :  
 $f_W(w) = -\frac{d}{dw} F_X \frac{d}{dw}(-\ln w) = -f_X(-\ln w) \frac{-1}{w} \quad \text{for } e^{-1} < w < 1 = -\frac{1}{w}$

**Result:**  $f_W(w) = \begin{cases} \frac{1}{w} & , \quad e^{-1} < w < 1 \\ 0 & , \quad o/w \end{cases}$

**Find: c) PDF of  $Z = 1 - e^{-X}$**

**Setup:**  $F_Z(z) = P[Z \leq z] = P[1 - e^{-X} \leq z] = P[-e^{-X} \leq z - 1] = P[e^{-X} \geq 1 - z] = P[-X \geq \ln(1 - z)] = P[X \leq -\ln(1 - z)] = F_X(-\ln(1 - z))$

**Steps: i)** find the pdf of  $x$ . See part a) for an explanation of why it is 1 when  $a < x < b$

ii) Differentiate with respect to  $w$  to find an equation given in terms of the pdf of  $x$ :  
 $f_Z(z) = -\ln(1 - z) = -\frac{-1}{1-z} = \frac{1}{1-z} \quad \text{for } 0 < z < e^{-1}$

**Result:**  $f_W(w) = \begin{cases} \frac{1}{1-z} & , \quad 0 < z < e^{-1} \\ 0 & , \quad o/w \end{cases}$

**Find: d) PDF of  $U = X(1 - X)$**

**Setup:**  $F_U(u) = P[U \leq u] = P[X(1 - x) \leq u] = P[-X^2 + X \leq u] = P[-(X - 1/2)^2 \leq u - 1/4] = P[(X - 1/2)^2 \geq 1/4 - u] = P[|(X - 1/2)| \geq (1/4 - u)^{1/2}] =$

**Steps: i)** find the pdf of  $x$ . See part a) for an explanation of why it is 1 when  $a < x < b$

ii) INCOMPLETE:

$f_Z(z) = -\ln(1 - z) = -\frac{-1}{1-z} = \frac{1}{1-z} \quad \text{for } 0 < z < e^{-1}$

**Result:**  $f_W(w) = \begin{cases} \frac{1}{1-z} & , \quad 0 < z < e^{-1} \\ 0 & , \quad o/w \end{cases}$

06.03

**Given: PDF**  $f_R(r) = \begin{cases} 6r(1 - r) & , \quad 0 < r < 1 \\ 0 & , \quad o/w \end{cases}$

**Find: Distribution of the circumference**

**Setup:** The circumference is  $c = 2\pi r$  - we have the pdf in terms of  $x$ , so this is the transformation

$F_C(c) = P[C \leq c] = P[2\pi r \leq c] = P[r \leq c/2\pi] = F_x(c/2\pi)$

**Steps: i)** Differentiate with respect to  $c$  to find an equation given in terms of the pdf of  $x$ .  
 $f_C(c) = \frac{d}{dc} F_R(c/2\pi) = f_R(c/2\pi) \frac{d}{dc} (c/2\pi) = f_R(c/2\pi)(1/2\pi)$

**ii)** Plug the original pdf back into this new form:

$$f_C(c) = \frac{6c}{2\pi} (1 - (c/2\pi))(1/2\pi) = \frac{6c(2\pi-c)}{2\pi^3} \quad \text{if } 0 < c < 2\pi$$

**Result:** 
$$f_C(c) = \begin{cases} \frac{6c(2\pi-c)}{2\pi^3} & , \quad 0 < c < 2\pi \\ 0 & , \quad o/w \end{cases}$$

## Find: Distribution of the area

**Setup:** The area is  $a = \pi r^2$  so the cdf  $F_A(a) = P[A \leq a] = P[\pi r^2 \leq a] = P[r^2 \leq a/\pi] = P[|r| \leq (a/\pi)^{1/2}] = P[-(a/\pi)^{1/2} \leq c \leq (a/\pi)^{1/2}] = F_r(a/\pi)^{1/2} - F_r(-(a/\pi)^{1/2})$

**Steps: i)** Differentiate with respect to  $a$  to find an equation in terms of the pdf of  $x$ .  
 $f_A(a) = \frac{d}{da} F_R(a/\pi)^{1/2} - \frac{d}{da} F_R(-(a/\pi)^{1/2}) = f_R[(a/\pi)^{1/2}] \frac{d}{da} (a/\pi)^{1/2} - f_R[-(a/\pi)^{1/2}] \frac{d}{da} (-(a/\pi)^{1/2})$

**Result:** 
$$f_A(a) = \begin{cases} \frac{3(\sqrt{\pi}-\sqrt{a})}{\pi^{3/2}} & , \quad 0 < a < \pi \\ 0 & , \quad o/w \end{cases}$$

**06.10** Suppose  $X$  has pdf  $f_X(x) = \frac{1}{2}e^{-|x|}$  for all real  $x$ .

(a) Find the pdf of  $Y = |X|$ .

CDF Method

$$F_Y(y) = P[Y \leq y] = P[|X| \leq y] = P[-y \leq X \leq y] = F_X(y) - F_X(-y)$$

$$f_Y(y) = \frac{dF_X(y)}{dy} - \frac{dF_X(-y)}{dy}$$

$$f_Y(y) = f_X(y) \frac{dy}{dx} - f_X(-y) \left( \frac{-dy}{dy} \right)$$

$$f_y = \frac{1}{2}e^{-y} + \frac{1}{2}e^{-y} = e^{-y} \quad y > 0$$

(b) Let  $W = 0$  if  $X \leq 0$  and  $W = 1$  if  $X > 0$ . Find the CDF of  $W$

$$F_W(w) = P[W = 0] = \frac{1}{2}$$

$$F_W(w) = P[W = 1] = \frac{1}{2}$$

$$F_W(w) =$$

$$\begin{cases} 0 & w \leq 0 \\ \frac{1}{2} & 0 \leq w \leq 1 \\ 1 & w > 1 \end{cases}$$

**06.13**  $X$  has pdf

$$f(x) = \begin{cases} \frac{x^2}{24} & -2 < x < 4 \\ 0 & \text{otherwise} \end{cases}$$

We want pdf of the CDF  $Y = X^2$  with regions:  $(-2, 0) \cup [0, 4)$

$$[F_x(\sqrt{y}) - F_x(-\sqrt{y})] = \left[ f_x(\sqrt{y}) \left( \frac{1}{2} \sqrt{y} \right) - f_x(-\sqrt{y}) \left( -\frac{1}{2} \sqrt{y} \right) \right]$$

$$f_y(y) = \begin{cases} \frac{y}{48\sqrt{y}} + \frac{y}{48\sqrt{y}} & 0 < y < 4 \\ \frac{y}{48\sqrt{y}} & 4 \leq y \leq 16 \\ 0 & \text{otherwise} \end{cases}$$

$$f_y(y) = \begin{cases} \frac{\sqrt{y}}{24} & 0 < y < 4 \\ \frac{\sqrt{y}}{48} & 4 \leq y \leq 16 \\ 0 & \text{otherwise} \end{cases}$$

**06.14**

**Given: Joint PDF**  $f(x, y) = \begin{cases} 4e^{-2(x+y)} & , \quad 0 < x < \infty, 0 < y < \infty \\ 0 & , \quad o/w \end{cases}$

**Find: a) CDF of W=X+Y**

**Setup:**  $F_w(w) = P[W \leq w] = P[X + Y \leq w]$

**Steps:**

i) Express as a sum of probabilities, replace probabilities with binomials

ii) Simplify and Use Combinatorial Identity

**Result:**  $\binom{n+m}{k}$

**06.15** This is a simplified version of example 6.4.5.

$X_1, X_2 \sim POI(\lambda)$  so the MGF of both is  $e^{\lambda(e^t-1)}$ . Thus by theorem 6.4.4

$$M_Y(t) = e^{\lambda(e^t-1)} e^{\lambda(e^t-1)} = e^{2\lambda(e^t-1)} \sim POI(2\lambda)$$

The pdf then of Y is

$$f_Y(y) = \begin{cases} \frac{e^{-2\lambda} (2\lambda)^y}{y!} & y = 0, 1, 2, \dots \\ 0 & \text{otherwise.} \end{cases}$$

**06.16** Note: the pdf of  $f_{x_1, x_2} = \frac{1}{x_1^2} \frac{1}{x_2^2}$

a) We need to find  $f_{u,v} = f_{x_1, x_2}(x_1(u, v), x_2(u, v))|J|$  where J is our jacobian. First we let  $u = x_1 x_2$  and  $v = x_1$  thus  $x_1 = v$  and  $x_2 = \frac{u}{v}$ , now we can find J.

$$J = \begin{vmatrix} 0 & 1 \\ \frac{1}{v} & 0 \end{vmatrix} = \frac{1}{v}$$

Finally, our pdf is:

$$\begin{aligned}
 f_{U,V}(u,v) &= f_{x_1,x_2}(v, \frac{u}{v}) \left| \frac{1}{v} \right| \\
 &= \frac{1}{v^2} \frac{1}{(\frac{u}{v})^2} \left| \frac{1}{v} \right| \\
 &= \frac{1}{u^2 v}, 1 < v < u < \infty
 \end{aligned}$$

**6.18** It is given that  $X$  and  $Y$  have a joint pdf given by

$$f(x, y) = e^{-y} \quad \text{if } 0 < x < y < \infty. \quad (1)$$

**(a): Find the joint pdf of  $S = X + Y$  and  $T = X$ .**

This can be done using the joint transformation method. By rearranging the above formulas we get  $X = T$  and  $Y = S - T$ . Then it is easy to get the jacobian

$$J = \begin{pmatrix} 1 & 0 \\ -1 & 1 \end{pmatrix} \quad (2)$$

whose determinant is clearly one. Note that the order in which you take partial derivatives is unimportant provided you are consistent - you will get the same determinant either way. Then we substitute in  $X = T$  and  $Y = S - T$  into the pdf and multiply by the determinant of the jacobian:

$$f_{S,T}(s, t) = f_{X,Y}(x(s, t), y(s, t)) \times 1 = \begin{cases} e^{t-s} & \text{if } 0 < t < s/2 \\ 0 & \text{otherwise} \end{cases}. \quad (3)$$

The bounds of the function can be found in a few different ways. One way is to consider the bounds of the original function,  $0 < x < y < \infty$ . We can substitute in the new formulas for  $X$  and  $Y$  to get

$$0 < t < s - t < \infty. \quad (4)$$

Then it is apparent that

$$0 < 2t < s < \infty, \quad (5)$$

which then yields

$$0 < t < s/2, \quad (6)$$

the bounds of our new function.

**(b): Find the marginal pdf of  $T$ .**

The easiest way to do this is to "integrate out"  $S$  from the joint pdf we derived:

$$\begin{aligned}
 f_T(t) &= \int_{-\infty}^{\infty} f_{S,T}(s, t) ds = \int_{2t}^{\infty} e^{t-s} ds \\
 &= e^t \int_{2t}^{\infty} e^{-s} ds = e^t (-e^{-s} |_{2t}^{\infty}) \\
 &= e^{-t} \quad \text{if } t > 0.
 \end{aligned} \quad (7)$$

**(c): Find the marginal pdf of S.**

This is just like part (b), except this time "integrate out" T:

$$\begin{aligned} f_S(s) &= \int_{-\infty}^{\infty} f_{S,T}(s,t)dt = \int_0^{s/2} e^{t-s} ds \\ &= e^{-s} \int_0^{s/2} e^t dt = e^{-s} (e^t|_0^{s/2}) \\ &= e^{-s} (e^{s/2} - 1) \quad \text{if } s > 0. \end{aligned} \tag{8}$$

**6.21** Let  $X$  and  $Y$  be continuous random variables with a joint density function given by

$$f_{X,Y}(x,y) = 2(x+y) \quad \text{if } 0 < x < y < 1 \quad \text{and } 0 \quad \text{otherwise.} \tag{9}$$

**(a) Find the joint density function of  $S = X$  and  $T = XY$ .**

We can solve for  $X$  and  $Y$  in terms of the new variables, to get  $X = S$  and  $Y = T/S$ . Then the jacobian is given by

$$J = \begin{pmatrix} 1 & 0 \\ -T/S^2 & 1/S \end{pmatrix}. \tag{10}$$

Then the new pdf is given by

$$f_{T,S}(s,t) = f_{X,Y}(x(s,t), y(s,t)) \times |1/s| = \begin{cases} 2(s+t/s) |1/s|, & 0 < s^2 < t < s < 1 \\ 0 & \text{otherwise.} \end{cases} \tag{11}$$

The bounds of this equation can be interpreted in the following way: the old triangular region in the  $xy$  plane got transformed to the region in the  $st$  plane between the lines  $T = S$  and  $T = S^2$ .

**(b) Find the marginal pdf of T.**

To find the marginal of  $T$ ,  $s$  needs to be "integrated out."

$$\begin{aligned} f_T(t) &= \int_{-\infty}^{\infty} f_{S,T}(s,t)ds \\ &= \int_t^{\sqrt{t}} 2(s+t/s) |1/s| ds = \int_t^{\sqrt{t}} 2(1+t/s^2)ds \\ &= 2 \int_t^{\sqrt{t}} ds + 2t \int_t^{\sqrt{t}} 1/s^2 ds = 2(\sqrt{t} - t) + 2t(-1/s)|_t^{\sqrt{t}} \\ &= 2\sqrt{t} - 2t + 2 - 2\sqrt{t} = \begin{cases} 2 - 2t & t \in (0,1) \\ 0 & \text{otherwise.} \end{cases} \end{aligned} \tag{12}$$

**6.23** We will use the property that independent identically distributed random variables has the form of 6.4.4,  $M_Y(t) = [M_X(t)]^n$  where  $Y = X_1 + X_2 + \dots + X_n$ . then since  $X_i \sim \text{GEO}(p)$

$$\begin{aligned} Mgf(Y) &= M_{X_1}(t)M_{X_2}(t)\dots M_{X_k}(t) \\ &= (M_X(t))^k \\ &= \left(\frac{pe^t}{1-qe^t}\right)^k \sim \text{NegativeBinomial}(k,p) \end{aligned}$$

**06.25** First note,  $X_1, X_2, X_3, X_4$  are all independent, but they are not IID as only  $X_2, X_3, X_4 \sim POI(5)$  with  $X_1$  not being listed. So formula 6.4.5 does not hold. 6.4.4 does though.

A)

$$\begin{aligned} Mgf(Y) &= M_{X_1}(t)M_{X_2+X_3+X_4}(t) \\ &= M_{X_1}(t)(M_{X_i}(t))^3 \end{aligned}$$

Since  $X_2, X_3, X_4$  are iid 6.4.5 holds for moving to this mgf

$$\begin{aligned} &= M_{X_1}(t)(e^{\mu(e^t-1)})^3 \\ &= M_{X_1}(t)e^{3\mu(e^t-1)} \\ &= M_{X_1}(t)e^{15(e^t-1)} \\ e^{25(e^t-1)} &= M_{X_1}(t)e^{15(e^t-1)} \\ \frac{e^{25(e^t-1)}}{e^{15(e^t-1)}} &= M_{X_1}(t) \\ e^{10(e^t-1)} &= M_{X_1}(t) \sim POI(10) \end{aligned}$$

B) For  $W = X_1 + X_2$  we have  $X_1 \sim POI(10)$  and  $X_2 \sim POI(5)$ . So  $POI(10 + 5) = POI(15)$   
**06.29**

**Given: PDF**  $f(x) = \begin{cases} \frac{1}{x^2} & , \quad 1 \leq x < \infty, 0 < y < \infty \\ 0 & , \quad o/w \end{cases}$

**Find: a) Joint PDF of the order statistics**

**Setup:**  $F_w(w) = P[W \leq w] = P[X + Y \leq w]$

**Steps: i)** Differentiate with respect to  $a$  to find an equation in terms of the pdf of  $x$ .

$$f_A(a) = \frac{d}{da}F_R(a/\pi)^{1/2} - \frac{d}{da}F_R - (a/\pi)^{1/2} = f_R[(a/\pi)^{1/2}]\frac{d}{da}(a/\pi)^{1/2}f_R[-(a/\pi)^{1/2}]\frac{d}{da} - (a/\pi)^{1/2}$$

**ii)** Simplify and Use Combinatorial Identity

**Result:**  $\binom{n+m}{k}$

**Find: b) PDF of the smallest order statistic  $Y_1$**

**Setup:**

**Steps: i)**



**Result:**

**Find: c) PDF of the largest order statistic  $Y_n$**

**Setup:**

**Steps: i)**

**Result:**

**Find: d) PDF of the sample range  $R = Y_n - Y_1$ , for  $n = 2$**

**Setup:** The area is  $a = \pi r^2$  so the cdf  $F_A(a) = P[A \leq a] = P[\pi r^2 \leq a] = P[r^2 \leq a/\pi] = P[|r| \leq (a/\pi)^{1/2}] = P[-(a/\pi)^{1/2} \leq c \leq (a/\pi)^{1/2}] = F_r(a/\pi)^{1/2} - F_r(-(a/\pi)^{1/2})$

**Steps: i)**

**Result:**

**Find: e) PDF of the sample median  $R = Y_{(n+1)/2}$ , for  $n$  odd so that  $r = (n+1)/2$**

**Setup:**

**Steps: i)**

**Result: 06.35** Suppose  $X_1, X_2$  are independent exponentially distributed random variables  $X_i \sim \text{EXP}(\theta)$ , and let  $Y = X_1 - X_2$ .

(a) Find the MGF of  $Y$ .

We can think of  $Y = X_1 - X_2$  as  $Y = X_1 + (-1)X_2$ . Then using Theorem 6.4.1,

$$\begin{aligned}M_Y(t) &= (M_{X_1}(t))(M_{-X_2}(t)) \\M_Y(t) &= (M_{X_1}(t))(M_{X_2}(-t)) \\M_Y(t) &= \left(\frac{1}{1-\theta t}\right) \left(\frac{1}{1-\theta(-t)}\right) \\M_Y(t) &= \left(\frac{1}{1-\theta t}\right) \left(\frac{1}{1+\theta t}\right) \\M_Y(t) &= \frac{1}{1-\theta t + \theta t - \theta^2 t^2} \\M_Y(t) &= \frac{1}{1-\theta^2 t^2}\end{aligned}$$

(b) What is the distribution of  $Y$ ?

Since  $\frac{1}{1-\theta^2 t^2}$  is the MGF of a double exponential,  $Y \sim \text{DE}(\theta, 0)$ .

**07.01** Consider a random sample of size  $n$  from a distribution with  $CDF F(x) = 1 - \frac{1}{x}$  if  $1 \leq x \leq \infty$

(a) Derive the CDF of the smallest order statistic,  $X_{1:n}$

Solution:  $G_1(y_1) = 1 - [1 - F_X(y_1)]^n = 1 - [1 - [1 - \frac{1}{y_1}]]^n = 1 - [\frac{1}{y_1}]^n$

$$G_1(y_1) = \begin{cases} 1 - \frac{1}{[y_1]^n} & \text{if } 1 \leq y_1 \\ 0 & \text{if } 0 > y_1. \end{cases}$$

(b) Find the limiting distribution of  $X_{1:n}$  Solution:

$$\lim_{n \rightarrow \infty} 1 - \frac{1}{y_1^n} = \begin{cases} 1 & \text{if } y_1 > 1 \\ 0 & \text{if } y_1 \leq 1 \end{cases}$$

The limiting distribution of  $X_{1:n}$  is degenerate at  $y = 1$

(c) Find the limiting distribution of  $X_{1:n}^n$

Solution:

$$F_{X_{1:n}^n}(y) = P(X_{1:n}^n \leq y) = P(X_{1:n} \leq y^{\frac{1}{n}}) = F_{X_{1:n}}(y^{\frac{1}{n}}) = 1 - \frac{1}{y^{\frac{1}{n}}} = 1 - \frac{1}{y^n}$$

$$\text{then, the limiting distribution of } X_{1:n}^n = \begin{cases} 1 - \frac{1}{y^n} & \text{if } y > 1 \\ 0 & \text{if otherwise} \end{cases}$$

**07.02**

$$F(x) = \begin{cases} -\frac{1}{x^2}, & \text{all real } x \end{cases}$$

2a.  $F_{X_{n:n}}(y) = (\frac{1}{1+e^{-y}})^n$ ;  $\lim_{n \rightarrow \infty} (\frac{1}{1+e^{-y}})^n$  has no limiting distribution.

2b.  $F_{X_{n:n} - \ln(n)}(y) = P[X_{n:n} - \ln(n) \leq y] = P[X_{n:n} \leq y + \ln(n)]$

$$= F_{X_{n:n}}(y + \ln(n))^n = (\frac{1}{1+e^{-(y+\ln(n))}})^n = (\frac{1}{1+\frac{e^{-y}}{n}})^n;$$

$$\lim_{n \rightarrow \infty} (\frac{1}{1+\frac{e^{-y}}{n}})^n = e^{-e^{-y}}$$

**07.03 3a.**  $F(x) = \begin{cases} 1 - \frac{1}{x^2}, & x > 1 \\ 0, & x \leq 0 \end{cases}$

$$F_{X_{1:n}}(y) = P[X_{1:n} \leq y] = 1 - P[X_{1:n} \geq y] = 1 - \frac{1}{y^{2n}}, y > 1$$

$$\lim_{n \rightarrow \infty} (1 - \frac{1}{y^{2n}}) = 1 - 0 = \begin{cases} 1, & y > 1 \\ 0, & y \leq 0 \end{cases}$$

**3b.**

$F_{X_{n:n}}(y) = P[X_{n:n} \leq y] = 1 - P[X_{n:n} \geq y] = 1 - (1 - \frac{1}{y^2})^n = \frac{1}{y^{2n}}; \lim_{n \rightarrow \infty} \frac{1}{y^{2n}} = 0,$   
Therefore  $F_{X_{n:n}}(y)$  has no limiting distribution.

**3c.**

$$F_{n^{-\frac{1}{2}}X_{n:n}}(y) = P[\frac{1}{\sqrt{n}}X_{n:n} \leq y] = P[X_{n:n} \leq \sqrt{n}y] = F_{X_{n:n}}(\sqrt{n}y) = (1 - (\sqrt{n}y)^{-2})^n, \text{ for } y > \frac{1}{\sqrt{n}}$$

$$\lim_{n \rightarrow \infty} (1 - (\sqrt{n}y)^{-2})^n = (1 - \frac{1}{ny^2})^n = \begin{cases} e^{-y^{-2}}, & y > 0 \\ 0, & y \leq 0 \end{cases}$$

**07.07** The WEI (1, 2) distribution has pdf  $f(x) = 2xe^{-x^2}$  for  $x > 0$ , mean  $\mu = \Gamma(\frac{3}{2}) = \frac{1}{2}\Gamma(\frac{1}{2}) = \frac{1}{2}\sqrt{\pi}$  and variance  $\sigma^2 = \Gamma(2) - \Gamma(\frac{3}{2})^2 = 1 - \frac{\pi}{4}$

(a) According to the central limit theorem, this holds with  $a = \mu - \frac{1.96\sigma}{\sqrt{n}}$  and  $b = \mu + \frac{1.96\sigma}{\sqrt{n}}$ , so if  $n = 35$  we have  $a = 0.7328$  and  $b = 1.0397$

(b) For odd  $n$ ,  $X_{\frac{n+1}{2}:n}$  is approximately  $N(x_{\frac{1}{2}}, \frac{c^2}{n})$ , where  $c^2 = \frac{1}{4f(x_{\frac{1}{2}})^2}$ . Now  $F(x_{\frac{1}{2}}) = \frac{1}{2}$ , because  $F(x) = 1 - e^{-x^2}$ , it implies that  $x_{\frac{1}{2}} = \sqrt{\ln 2}$ . Also because  $c^2 = \frac{1}{4\ln 2}$ , we have  $a = x_{\frac{1}{2}} - \frac{1.96c}{\sqrt{n}}$  and  $b = x_{\frac{1}{2}} + \frac{1.96c}{\sqrt{n}}$ , so when  $n = 35$  we have  $a = 0.6336$  and  $b = 1.0315$

**07.11** a) First we need to know the  $\mu$  and the  $\sigma$ . For a Uniform variable with  $a = 0, b = 1$  we have  $\mu = 1/2$  and  $\sigma = 1/\sqrt{12}$  (Note: it is not  $\sigma^2$ ). We also need to know that  $n = 20$  from there we can use the CLT:

$$\begin{aligned} \Pr\left(\sum_{i=1}^{20} X_i < 12\right) &= \Pr\left(\frac{\sum X_i - 10}{\sqrt{20}\frac{1}{\sqrt{12}}} < \frac{12 - 10}{\sqrt{20}\frac{1}{\sqrt{12}}}\right) \\ &= \Phi\left(\frac{12 - 10}{\sqrt{20}\frac{1}{\sqrt{12}}}\right) \\ &\approx .9394 \end{aligned}$$

b) We let  $Y = \sum_{i=1}^{20} X_i$ , let  $Y'$  be our 90th percentile that we want to find. So we setup our probability as  $\Pr(Y \leq Y') = .9$ , .9 as we are interested in the 90th percentile. Using  $\mu, \sigma$ , and  $n$  from part (a) we solve with CLT:

$$\begin{aligned} \Pr(Y \leq Y') &= \Pr\left(\frac{Y - \mu n}{\sigma\sqrt{n}} \leq \frac{Y' - \mu n}{\sigma\sqrt{n}}\right) \\ &= \Pr\left(Z \leq \frac{Y' - 10}{\sqrt{20}\frac{1}{\sqrt{12}}}\right) \text{ Note: } Z \text{ is standard normal due to CLT} \\ .9 &= \Phi\left(\frac{Y' - 10}{\sqrt{20}\frac{1}{\sqrt{12}}}\right) \end{aligned}$$

We now solve for  $Y'$ . We know (from a chart or list) that .9 from  $\Phi$  is  $z \approx 1.285$ . So we set our final equation for finding out  $Y'$  with that in mind.

$$\begin{aligned} \frac{Y' - 10}{1.291} &= 1.285 \\ Y' &\approx 11.658 \end{aligned}$$

**07.12** a) First, an understanding that the wording here implies that  $X$  is actually "failures" of weapons. So the given  $p$  would normally be  $q$  in other contexts. So using the binomial theorem we would have  $p = .05$  and  $q = .95$ . Knowing that we can use the Binomial theorem

easily:

$$\begin{aligned}\Pr(X \geq 1) &= 1 - \Pr(X < 1) \\ &= 1 - \binom{n}{0} (.05)^0 (.95)^n\end{aligned}$$

We now solve for  $n$  from the above equation knowing that the desired probability is .99

$$\begin{aligned}.99 &= 1 - (.95)^n \\ \ln .95^n &= \ln .01 \\ n &= \frac{\ln .01}{\ln .95}\end{aligned}$$

So  $n$ , since it must be an integer, is rounded to 90.

b)

**07.13** From the hint we know that  $Y_n = \sum^n X_i$  where  $X_i \sim Geo(p)$ . So for  $\sum^n X_i$  the  $\mu = \frac{n}{p}$  and  $\sigma^2 = \frac{nq}{p^2}$ . Then by the CLT:

$$\begin{aligned}\Pr(Y_n \leq y) &= \Pr\left(\sum^n X_i \leq y\right) \\ &= \Pr\left(\frac{\sum^n X_i - \frac{n}{p}}{\sqrt{n}\sqrt{\frac{nq}{p^2}}} \leq \frac{y - \frac{n}{p}}{\sqrt{n}\sqrt{\frac{nq}{p^2}}}\right) \\ &= \Phi\left(\frac{y - \frac{n}{p}}{\sqrt{n}\sqrt{\frac{nq}{p^2}}}\right)\end{aligned}$$

**07.16** a) We need two things for this proof. First, we need to know  $\mu$  and  $\sigma^2$  of  $\bar{X}$ . We know this is  $\mu = \mu$  and  $\sigma^2 = \frac{\mu^2}{n}$  from facts of the sample mean distribution of  $POI(\mu)$ . Next the theorems from section 7.6, namely 7.6.2 and from 7.7, 7.7.2. These will let us prove the following:

$$\begin{aligned}\Pr[|\bar{X}_n - \mu| < \epsilon] &\geq 1 - \frac{\mu^2}{\epsilon^2 n} \\ \lim_{n \rightarrow \infty} \Pr[|\bar{X}_n - \mu| < \epsilon] &= 1\end{aligned}$$

From this we now know that  $\bar{X} \xrightarrow{P} \mu$  from 7.6.3. For our goal,  $e^{\bar{X}_n}$  we simply need to know 7.7.2. Since  $\bar{X} \xrightarrow{P} \mu$  then  $e^{\bar{X}} \xrightarrow{P} e^\mu$

b) It has been shown elsewhere in the text that any  $\overline{X_n}$  will converge to  $N(0, 1)$  if standardized. The theorem we need to use then, is 7.7.6 which states that a function of an already convergent series also converges to an asymptotic normal distribution. (For an almost direct example see Example 7.7.3)

Our  $g(y)$  here is  $e^{-\overline{X_n}}$  where  $g(y) = e^y$ . So then  $g'(y) = -e^{-y}$  and using 7.7.6 we can find our distribution if  $\frac{d}{d\mu}e^{-\mu} = -e^{-\mu}$  then  $N(e^{\mu}, \frac{-e^{-2\mu}\mu^2}{n})$

c) From parts (a) we know that  $\overline{X_n} \xrightarrow{P} \mu$  and  $e^{-\overline{X_n}} \xrightarrow{P} e^{-\mu}$ . So we can use theorem 7.7.3 via section (2), which states that  $X_n Y_n \xrightarrow{P} cd$ . In our case we have the prior two found distributions. So then by the theorem  $\overline{X_n} e^{\overline{X_n}} \xrightarrow{P} \mu e^{-\mu}$