

1 Solutions for HW8

1:(a) Since X and Y are independent, $f_{X,Y}(x,y) = f_X(x)f_Y(y) = \frac{1}{24}xy^4e^{-(x+y)}$, $x > 0, y > 0$.

(b) $W > 0, 0 < Z < 1$

(c) $X = WZ, Y = W(1 - Z), |J| = W$,

$$f_{W,Z}(w,z) = w \frac{1}{24} w z w^4 (1-z)^4 e^{-(wz+w(1-z))} = \frac{1}{24} w^6 z (1-z)^4 e^{-w}, w > 0, 0 < z < 1.$$

$$(d) f_W(w) = \int_0^1 f_{W,Z}(w,z) dz = \frac{1}{6!} w^6 e^{-w}, w > 0$$

$$f_Z(z) = \int_0^\infty f_{W,Z}(w,z) dw = \frac{6!}{2!5!} z (1-z)^4, 0 < z < 1.$$

(e) Since $f_{W,Z}(w,z) = f_W(w)f_Z(z)$, W is independent of Z.

2:(a) Since X is independent of Y, $f_{X,Y}(x,y) = f_X(x)f_Y(y) = \frac{1}{2\pi} e^{-\frac{x^2+y^2}{2}}$, $x \in R, y \in R$

(b) Let $X = R \cos \theta, y = R \sin \theta$, then $r > 0, \theta \in [0, 2\pi)$ and $|J| = r$,

$$\text{so } f_{R,\theta}(r,\theta) = \frac{r}{2\pi} e^{-r^2/2}, r > 0,$$

$$\text{so } f_R(r) = \int_0^{2\pi} f_{R,\theta}(r,\theta) d\theta = r e^{-r^2/2}, r > 0.$$

3:(a) $Y|X \sim \text{Bin}(X, 1/3), X \sim \text{Bin}(5, 0.2)$, So

$$P(X = x, Y = y) = P(X = x)P(Y = y|X = x) = \binom{5}{x} 0.2^x 0.8^{4-x} \binom{x}{y} (1/3)^y (2/3)^{x-y}$$

$$0 \leq y \leq x \leq 5$$

(b): the marginal of Y is

$$\sum_y^5 P(X = x, Y = y), \text{ by calculation, } Y \sim \text{Binomial}(5, 1/15).$$

4: Note that for an exponentially distributed random variable with parameter λ , the c.d.f is $F_X(x) = 1 - e^{-\lambda x}$,

$$(a) F_{x(1)}(x) = 1 - \prod_{i=1}^n P(X_i > x) = 1 - e^{-\sum_{i=1}^n \lambda_i x}, x > 0.$$

$$(b) f_{x(1)}(x) = F'_{x(1)}(x) = (\sum_{i=1}^n \lambda_i) e^{-\sum_{i=1}^n \lambda_i x}, x > 0.$$

$$(c) F_{x(n)}(x) = \prod_{i=1}^n P(X_i < x) = \prod_{i=1}^n (1 - e^{-\lambda_i x}), x > 0.$$

$$(d) \text{ Since } \lambda_i = \lambda, \text{ then } F_{x(n)}(x) = (1 - e^{-\lambda x})^n, f_{x(n)}(x) = F'_{x(n)}(x) = n\lambda e^{-\lambda x} (1 - e^{-\lambda x})^{n-1}, x > 0.$$

5:(a) The joint distribution of X and Y is

| | | | | |
|----------|---------|---------|---------|---------|
| | X=0 | 1 | 2 | 3 |
| Y=0 | $1/2^4$ | $1/2^4$ | 0 | 0 |
| 1 | $1/2^4$ | $3/2^4$ | $2/2^4$ | 0 |
| 2 | 0 | $2/2^4$ | $3/2^4$ | $1/2^4$ |
| 3 | 0 | 0 | $1/2^4$ | $1/2^4$ |
| $f_X(x)$ | $2/2^4$ | $6/2^4$ | $6/2^4$ | $2/2^4$ |

(b) The marginal density $P_{Y|X}(y|x) = \frac{P(X=x, Y=y)}{P(X=x)}$

So $P(Y = 0|X = 1) = 1/6, P(Y = 1|X = 1) = 1/2, P(Y = 2|X = 1) = 1/3$.

6:

$$f_{X,Y}(x, y) = \frac{1}{2\pi\sqrt{1-\rho^2}} e^{-\frac{x^2-2\rho xy+y^2}{2(1-\rho^2)}}$$

Note: $x^2 - 2\rho xy + y^2 = (y - \rho x)^2 + (1 - \rho^2)x^2$, so the marginal density of X is

$$\begin{aligned} f_X(x) &= \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy \\ &= \int_{-\infty}^{\infty} \frac{1}{2\pi\sqrt{1-\rho^2}} e^{-\frac{(y-\rho x)^2+(1-\rho^2)x^2}{2(1-\rho^2)}} dy \\ &= \frac{1}{\sqrt{2\pi}} e^{-x^2/2} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi(1-\rho^2)}} e^{-\frac{(y-\rho x)^2}{2(1-\rho^2)}} dy \\ &= \frac{1}{\sqrt{2\pi}} e^{-x^2/2} \text{(Note the integral part is the density of Norm}(\rho x, 1-\rho^2), \text{So the integration is 1)} \end{aligned}$$

$X \sim N(0, 1)$, Similarly, we have $Y \sim N(0, 1)$

"=>" We know X and Y are independent. So $f_{X,Y}(x, y) = f_X(x)f_Y(y)$

$$\frac{1}{2\pi\sqrt{1-\rho^2}} e^{-\frac{x^2-2\rho xy+y^2}{2(1-\rho^2)}} = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} * \frac{1}{\sqrt{2\pi}} e^{-y^2/2}$$

So $\rho = 0$

"<=" , we have $\rho = 0$. So

$$f_{X,Y}(x, y) = \frac{1}{2\pi} e^{-\frac{x^2+y^2}{2}} = f_X(x)f_Y(y)$$

So X and Y are independent

7:(a) $F_{U_{(1)}}(u) = 1 - (1 - u)^n, u \in (0, 1)$,

so $f_{U_{(1)}}(u) = F'_{U_{(1)}}(u) = n(1 - u)^{(n-1)}, u \in (0, 1)$.

(b) $F_{U_{(n)}}(u) = u^n, u \in (0, 1)$,

so $f_{U_{(n)}}(u) = F'_{U_{(n)}}(u) = nu^{(n-1)}, u \in (0, 1)$.

(c) $F_{U_{(1)}, U_{(n)}}(x, y) = P(U_{(1)} \leq x, U_{(n)} \leq y) =$

when $x < 0$ or $y < 0$, 0

when $0 < x < 1$ and $y > 1$, $P(U_{(1)} < x) = 1 - (1 - x)^n$

when $0 < x < y < 1$, $P(U_{(n)} < y) - P(x < U_{(1)} < U_{(n)} < y) = y^n - (y - x)^n$

when $0 < y < 1$ and $x > y$, $P(U_{(n)} < y) = y^n$

when $x > 1$ and $y > 1$, 1.

(d) $f_{U_{(1)}, U_{(n)}}(x, y) = \frac{\partial^2 F(x, y)}{\partial x \partial y} = n(n - 1)(y - x)^{n-2}$, if $0 < x < y < 1$ and 0 otherwise.

8: (a) I_i is i.i.d Bernoulli distribution with $\theta = P(X_i \leq x) = F(x)$. So W is Bin(n, F(x)).

(b)

$$P(W \geq k) = \sum_{i=k}^n P(W = i) = \sum_{i=k}^n (F(x))^i (1 - F(x))^{n-i}$$

(c) $W \geq k$ if and only if at least k X_i 's are less or equal x . It is equivalent to say $X_{(k)} \leq x$

$$G_k(x) = P(X_{(k)} \leq x) = P(W \geq k) = \sum_{i=k}^n (F(x))^i (1 - F(x))^{n-i}$$

(d)

$$\begin{aligned} g_k(x) &= \frac{dG_k(x)}{dx} \\ &= \sum_{i=k}^n \left[\binom{n}{i} i f(x) (F(x))^{i-1} (1 - F(x))^{n-i} - \binom{n}{i} (n-i) f(x) (F(x))^i (1 - F(x))^{n-i-1} \right] \end{aligned}$$

(e) So $a_i = i \binom{n}{i}$, $b_i = (n-i) \binom{n}{i}$. So $b_n = 0$

$$a_{j+1} = (j+1) \frac{n!}{(j+1)!(n-j-1)!} = \frac{n!}{j!(n-j-1)!}$$

$$b_j = (n-j) \frac{n!}{j!(n-j)!} = \frac{n!}{j!(n-j-1)!}$$

So $a_{j+1} = b_j$.

$$g_k(x) = \frac{n!}{(n-k)!(k-1)!} f(x) (F(x))^{k-1} (1 - F(x))^{n-k}$$

(f) If $X_i \sim Uniform[0, 1]$, then $f(x) = 1$, $F(x) = x$, $x \in (0, 1)$. So

$$g_k(x) = \frac{n!}{(n-k)!(k-1)!} (x)^{k-1} (1-x)^{n-k}, x \in (0, 1)$$

For $Beta(k, n-k+1)$, the density is

$$f(x) = \frac{x^{k-1} (1-x)^{n-k}}{B(k, n-k+1)}$$

$B(k, n-k+1) = \frac{\Gamma(k)\Gamma(n-k+1)}{\Gamma(n+1)} = \frac{(k-1)!(n-k)!}{n!}$. So $g_k(x) = f(x)$. $X_{(k)} \sim Beta(k, n-k+1)$

9:(a) $F(x) = \int_{-\infty}^x f(x) = \frac{1}{\pi} \arctan(x) + \frac{1}{2}$, $x \in R$.

(b) $F^{-1}(y) = \tan(\pi y - \frac{\pi}{2})$, $0 \leq y \leq 1$

(c) $X = F^{-1}(U) = \tan(\pi U - \frac{\pi}{2})$, where $U \sim Uniform(0, 1)$.

Since F is the c.d.f of Cauchy distribution and U is uniform between 0 and 1, X has the cauchy distribution.