

ECSE 305A: Probability and Random Signals I
Problem Set 9
solutions

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1. (a)

$$\begin{aligned} F(x, y) &= \int_0^1 \int_0^1 f(x, y) dx dy \\ &= \int_0^1 \int_0^1 cx^k y^l dx dy \\ &= \int_0^1 c \left. \frac{x^{k+1}}{k+1} \right|_0^1 y^l dy \\ &= \left. \frac{c}{k+1} \frac{y^{l+1}}{l+1} \right|_0^1 \\ &= \frac{c}{(k+1)(l+1)} \\ &= 1. \\ &\Rightarrow c = (k+1)(l+1). \end{aligned}$$

(b)

$$\begin{aligned} P(Y > X) &= c \int_0^1 \int_x^1 x^k y^l dx dy \\ &= c \left. 0^1 \frac{y^{l+1}}{l+1} \right|_x^1 x^k dx \\ &= c \int_0^1 \left(\frac{1}{l+1} - \frac{x^{l+1}}{l+1} \right) x^k dx \\ &= (k+1)(l+1) \left[\frac{1}{(k+1)(l+1)} - \frac{1}{(l+1)(k+l+2)} \right]_0^1 \\ &= 1 - \frac{k+1}{k+l+2} \end{aligned}$$

If $l > k$ then $P(Y > X) > 1/2$.

(c)

$$\begin{aligned} f_Y(y) &= \int_0^1 cx^k y^l dx = c \left. \frac{x^{k+1}}{k+1} \right|_0^1 = (l+1)y^l; \\ f_X(x) &= \int_0^1 cx^k y^l dy = c \left. \frac{y^{l+1}}{l+1} \right|_0^1 = (k+1)x^k. \end{aligned}$$

2. (a)

$$\begin{aligned} \sum_{i \in R_I} \sum_{j \in R_J} p(i, j) &= 1 \\ (0 + 1 + 1 + 1 + 2 + 2 + 1 + 2 + 2)c &= 1 \\ &\Rightarrow c = \frac{1}{12} \end{aligned}$$

(b)

$$P_{I|J}(i, j) = \frac{P(I = i, J = j)}{P(J = j)} = \frac{\frac{i^2+j^2}{12}}{P(J = j)}$$

$$P_J(j) = \begin{cases} \frac{1+2+2}{12} = \frac{5}{12}, & j = -1 \\ \frac{0+1+1}{12} = \frac{2}{12}, & j = 0 \\ \frac{1+2+2}{12} = \frac{5}{12}, & j = 1 \end{cases}$$

$$P_{I|J}(i, j) = \begin{cases} \frac{i^2+1}{5}, & j = -1 \\ \frac{i^2}{2}, & j = 0 \\ \frac{i^2+1}{5}, & j = 1 \end{cases}$$

(c)

$$P(|I| = 1|J = 0) = P(I = 1|J = 0) + P(I = -1|J = 0) = 1/2 + 1/2 = 1$$

3. See Fig. 1.

4.

$$\begin{aligned} F_Y(y) &= \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma_X\sigma_Y\sqrt{1-\rho^2}} e^{-\frac{1}{2(1-\rho^2)}[(\frac{x-\mu_X}{\sigma_X})^2 - 2\rho(\frac{x-\mu_X}{\sigma_X})(\frac{y-\mu_Y}{\sigma_Y}) + (\frac{y-\mu_Y}{\sigma_Y})^2]} dx \\ &= \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma_X\sigma_Y\sqrt{1-\rho^2}} e^{-\frac{1}{2(1-\rho^2)}[(\frac{x-\mu_X}{\sigma_X})^2 - 2\rho(\frac{x-\mu_X}{\sigma_X})(\frac{y-\mu_Y}{\sigma_Y}) + \rho^2(\frac{y-\mu_Y}{\sigma_Y})^2 - \rho^2(\frac{y-\mu_Y}{\sigma_Y})^2 + (\frac{y-\mu_Y}{\sigma_Y})^2]} dx \\ &= \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma_X\sigma_Y\sqrt{1-\rho^2}} e^{-\frac{1}{2(1-\rho^2)}[(\frac{x-\mu_X}{\sigma_X}) - \rho(\frac{y-\mu_Y}{\sigma_Y})]^2 + (1-\rho^2)(\frac{y-\mu_Y}{\sigma_Y})^2]} dx \\ &= \frac{e^{-\frac{1}{2}(\frac{y-\mu_Y}{\sigma_Y})^2}}{2\pi\sigma_Y} \int_{-\infty}^{\infty} \frac{1}{\sigma_X\sqrt{1-\rho^2}} e^{-\frac{1}{2(1-\rho^2)}((\frac{x-\mu_X}{\sigma_X}) - \rho(\frac{y-\mu_Y}{\sigma_Y}))^2} dx \\ &= \frac{e^{-\frac{1}{2}(\frac{y-\mu_Y}{\sigma_Y})^2}}{\sqrt{2\pi}\sigma_Y} \end{aligned}$$

5.

$$P(X \leq 0|y = 1) = \Phi\left(\frac{0-1}{1}\right) = \Phi(-1) = 1 - \Phi(1) = 0.1587$$

Experimental value : 0.1570

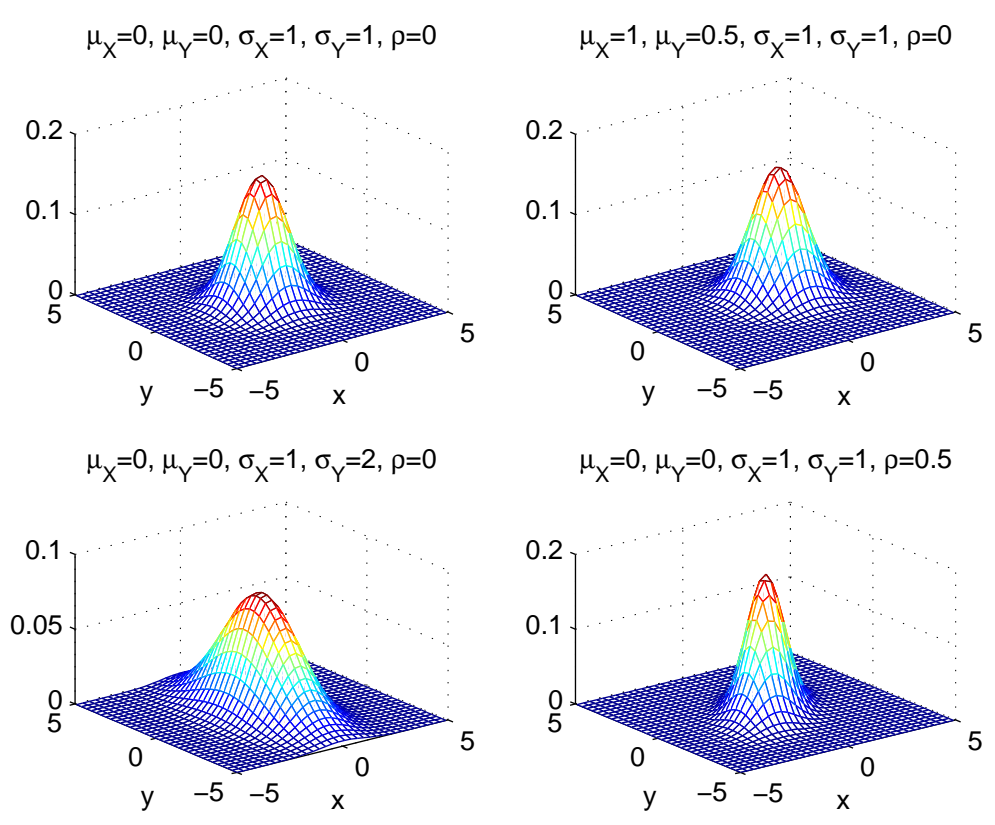


Figure 1: Question 3

6. RV X and Y are dependent. This can be seen by following example:
 Consider $P(X = 8, Y = 8)$ which is equal to zero but $P(X = 8)P(Y = 8) = \frac{\binom{13}{8}}{\binom{52}{8}} \frac{\binom{13}{8}}{\binom{52}{8}} \neq P(X = 8, Y = 8)$.
7. (a)

$$f(x, y) = \frac{1}{(b-a)(d-c)}$$

See Fig. 2.

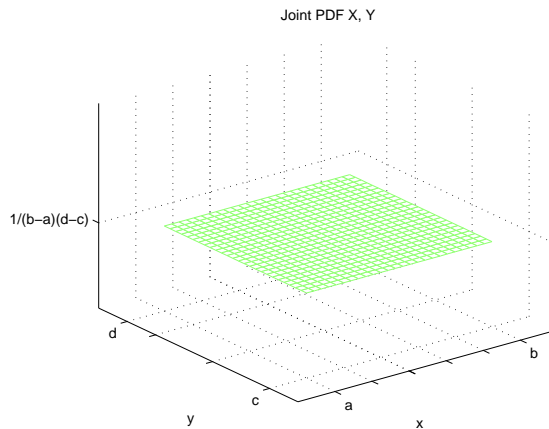


Figure 2: Question 7 (a)

- (b)

$$F(x, y) = \int_a^x dt \int_c^y du f(t, u) = \frac{(y-c)(x-a)}{(b-a)(d-c)}$$

See Fig 3.

- (c)

$$f_Y(y) = \int_a^b \frac{1}{(b-a)(c-d)} dx = \frac{(b-a)}{(b-a)(c-d)} = \frac{1}{d-c}$$

$$f_X(x) = \int_c^d \frac{1}{(b-a)(c-d)} dy = \frac{(d-c)}{(b-a)(c-d)} = \frac{1}{b-a}$$

Therefore, $f(x, y) = f_X(x)f_Y(y)$ and two RV are independent.

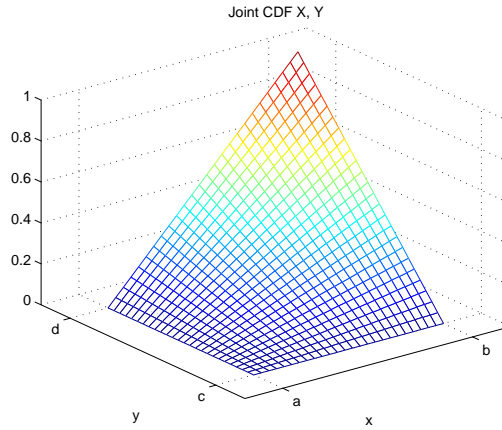


Figure 3: Question 7 (b)

8. Let $f(x, y)$ be the joint probability density function of X and Y . Clearly,

$$f(x, y) = \frac{1}{2\pi\sigma^2} e^{-\frac{x^2}{2\sigma^2}} e^{-\frac{y^2}{2\sigma^2}}$$

Since, $x = r \cos \theta$ and $y = r \sin \theta$

$$J = \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix} = r \neq 0$$

$$g(r, \theta) = f(r \cos \theta, r \sin \theta) |r| = \frac{1}{2\pi\sigma^2} r e^{-\frac{r^2}{2\sigma^2}}$$

Now

$$g_R(r) = \int_0^{2\pi} \frac{1}{2\pi\sigma^2} r e^{-\frac{r^2}{2\sigma^2}} d\theta = \frac{r}{\sigma^2} e^{-\frac{r^2}{2\sigma^2}}, \quad r > 0$$

and

$$g_\Theta(\theta) = \int_0^\infty \frac{1}{2\pi\sigma^2} r e^{-\frac{r^2}{2\sigma^2}} dr = \frac{1}{2\pi}, \quad 0 < \theta < 2\pi$$

Therefore, $g(r, \theta) = g_R(r)g_\Theta(\theta)$, showing that R and Θ are independent random variables

- 9.

$$Z = \max(X, Y) = \begin{cases} X & \text{if } X > Y \\ Y & \text{if } X \leq Y \end{cases}$$

$$\begin{aligned} F_Z(\alpha) &= P(z \leq \alpha) \\ &= P(\max(X, Y) \leq \alpha) \\ &= P(X \leq \alpha, Y \leq \alpha) \\ &= F_{X,Y}(\alpha, \alpha) \\ &= F_X(\alpha)F_Y(\alpha) \\ &= (1 - e^{-\lambda\alpha})^2 \end{aligned}$$