

Tutorial T8

Example T8.1: A particle is known to be in a cylinder. The pdf of the random location (R, Θ, Z) in cylindrical coordinates is

$$f(r, \theta, z) = \begin{cases} c(1-r)rz & 0 \leq r \leq 1, 0 \leq \theta \leq 2\pi, 0 \leq z \leq 1 \\ 0 & \text{else} \end{cases}$$

Are R, Θ and Z independent?

Solution:

Yes. This is because $f(r, \theta, z) = a(r)b(\theta)d(z)$ where

$$\begin{aligned} a(r) &= \begin{cases} r - r^2 & 0 \leq r \leq 1 \\ 0 & \text{else} \end{cases} \\ b(\theta) &= \begin{cases} c & 0 \leq \theta \leq 2\pi \\ 0 & \text{else} \end{cases} \\ c(z) &= \begin{cases} z & 0 \leq z \leq 1 \\ 0 & \text{else} \end{cases} \end{aligned}$$

Example T8.2: Let X and Y be the outcomes of rolling a 4-sided and a 6-sided die. Let $Z = X + Y$. What is the pmf of Z ?

Solution:

$$\begin{aligned} p_X(k) &= \begin{cases} 1/4 & k = 1, \dots, 4 \\ 0 & \text{else} \end{cases} \\ p_Y(k) &= \begin{cases} 1/6 & k = 1, \dots, 6 \\ 0 & \text{else} \end{cases} \end{aligned}$$

$$\begin{aligned} P[Z = n] &= P[X + Y = n] \\ &= P[\cup_{k=-\infty}^{\infty} \{X = k, Y = n - k\}] \\ &= \sum_{k=-\infty}^{\infty} P[X = k, Y = n - k] \\ &= \sum_{k=-\infty}^{\infty} P[X = k]P[Y = n - k] \\ &= \sum_{k=1}^4 P[X = k]P[Y = n - k] \quad [\text{since } X \in \{1, 2, 3, 4\}] \\ &= P[X = 1]P[Y = n - 1] \\ &\quad + P[X = 2]P[Y = n - 2] \\ &\quad + P[X = 3]P[Y = n - 3] \\ &\quad + P[X = 4]P[Y = n - 4] \\ &= \frac{1}{4}P[Y = n - 1] + \frac{1}{4}P[Y = n - 2] \\ &\quad + \frac{1}{4}P[Y = n - 3] + \frac{1}{4}P[Y = n - 4] \end{aligned}$$

If $n \leq 1$, then sum is 0.

If $n = 2$, then sum is $\frac{1}{4}\frac{1}{6} = \frac{1}{24}$

If $n = 3$, then sum is $\frac{2}{24}$

If $n = 4$, then sum is $\frac{3}{24}$

If $n = 5$, then sum is $\frac{4}{24}$

If $n = 6$, then sum is $\frac{4}{24}$

If $n = 7$, then sum is $\frac{4}{24}$

If $n = 8$, then sum is $\frac{3}{24}$

If $n = 9$, then sum is $\frac{2}{24}$

If $n = 10$, then sum is $\frac{1}{24}$

If $n \geq 11$, then sum is 0

Example T8.3: Let X_1, \dots, X_n be independent with $X_k \sim \mathcal{N}(\mu_k, \sigma_k^2)$.

Then, we know (Proposition 26.1) that:

$$X = X_1 + \dots + X_n \sim \mathcal{N}(\mu_1 + \dots + \mu_n, \sigma_1^2 + \dots + \sigma_n^2).$$

a) If $Z_1 = a_1X_1 + b_1, \dots, Z_n = a_nX_n + b_n$, show that Z_1, \dots, Z_n are independent. [Assume $a_1 > 0, \dots, a_n > 0$].

b) What is the pdf of $Z = (a_1X_1 + b_1) + \dots + (a_nX_n + b_n)$?

Solution: a) Since X_1, \dots, X_n are independent:

$$F_{X_1, \dots, X_n}(x_1, \dots, x_n) = F_{X_1}(x_1) \times \dots \times F_{X_n}(x_n)$$

Let $Y \sim \mathcal{N}(0, 1)$. Since $X_k \sim \mathcal{N}(\mu_k, \sigma_k^2)$:

$$\begin{aligned} F_{X_k}(x_k) &= P[X_k \leq x_k] \\ &= P\left[\frac{X_k - \mu_k}{\sigma_k} \leq \frac{x_k - \mu_k}{\sigma_k}\right] \\ &= P\left[Y \leq \frac{x_k - \mu_k}{\sigma_k}\right] \\ &= \Phi\left(\frac{x_k - \mu_k}{\sigma_k}\right) \end{aligned}$$

So

$$F_{X_1, \dots, X_n}(x_1, \dots, x_n) = \Phi\left(\frac{x_1 - \mu_1}{\sigma_1}\right) \times \dots \times \Phi\left(\frac{x_n - \mu_n}{\sigma_n}\right)$$

$$\begin{aligned} F_{Z_1, \dots, Z_n}(z_1, \dots, z_n) &= P[Z_1 \leq z_1, \dots, Z_n \leq z_n] \\ &= P[a_1X_1 + b_1 \leq z_1, \dots, a_nX_n + b_n \leq z_n] \\ &= P\left[X_1 \leq \frac{z_1 - b_1}{a_1}, \dots, X_n \leq \frac{z_n - b_n}{a_n}\right] \\ &= F_{X_1, \dots, X_n}\left(\frac{z_1 - b_1}{a_1}, \dots, \frac{z_n - b_n}{a_n}\right) \\ &= \Phi\left(\frac{\frac{z_1 - b_1}{a_1} - \mu_1}{\sigma_1}\right) \times \dots \times \Phi\left(\frac{\frac{z_n - b_n}{a_n} - \mu_n}{\sigma_n}\right) \\ &= \Phi\left(\frac{z_1 - b_1 - a_1\mu_1}{a_1\sigma_1}\right) \times \dots \times \Phi\left(\frac{z_n - b_n - a_n\mu_n}{a_n\sigma_n}\right) \end{aligned}$$

So Z_1, \dots, Z_n are independent with $Z_k \sim \mathcal{N}(b_k + a_k\mu_k, a_k^2\sigma_k^2)$.

b) Since $Z = Z_1 + \dots + Z_n$ and Z_k 's are Normal and independent, we apply Proposition 26.1: Z is Normal with

$$\mu_Z = E[Z_1] + \dots + E[Z_n] = \sum_{k=1}^n b_k + a_k\mu_k$$

$$\sigma_Z^2 = Var[Z_1] + \dots + Var[Z_n] = \sum_{k=1}^n a_k^2\sigma_k^2$$

Note: if $\rho = 0$, then regardless of what the known value of Y is, $X \sim \mathcal{N}(\mu_0, \sigma_0^2)$. This makes sense since if $\rho = 0$, then X and Y are independent.

Note: if $\rho = 1$, then $\sigma_{X|Y=y}^2 = 0$ and

$$\mu_{X|Y=y} = \mu_X + \sigma_X \frac{y - \mu_Y}{\sigma_Y}$$

$$\sigma_{X|Y=y}^2 = \sigma_X^2(1 - \rho^2).$$

So given $Y = y$ where $y = 79$, X is Normal with mean $80 + 0.5 \frac{10}{8}(79 - 75) = 82.5$ and variance $100(1 - 0.5^2) = 75$.

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