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## Unit 5 HW: Joint Distributions

1.  $W$ : wingspan :  $V$ : velocity  
↓

Normal distribution w/  $\mu_w = 10$ ,  $\sigma_w = 4$

$U$  = R.V. w/ normal distribution  $\mu_u = 0$ ,  $\sigma_u = 1$

$W$  and  $U$  are independent  $\Rightarrow f_{w,u}(w,u) = f_w(w)f_u(u)$

$$f_{w|u}(w,u) = f_w(w)$$

$$f_{u|w}(w,u) = f_u(u)$$

$$V = 0.5W + U$$

$$\text{Variance/covariance matrix} = \begin{bmatrix} \text{Var}(W) & \text{Cov}(W, V) \\ \text{Cov}(V, W) & \text{Var}(V) \end{bmatrix}$$

$$\text{Var}(W) = \sigma_w^2 = 16 \quad W \text{ and } U \text{ are independent}$$

$$\text{Var}(V) = \text{Var}(0.5W + U) = \text{Var}(0.5W) + \text{Var}(U) + 2\text{Cov}(0.5W, U)$$

$$\text{Var}(V) = 0.25\text{Var}(W) + \text{Var}(U) = \frac{16}{4} + 1 = 5$$

$$\text{Cov}(W, V) = \text{Cov}(V, W) = E(WV) - E(W)E(V)$$

$$= E(0.5W^2 + WU) - E(W)E(0.5W + U)$$

$$= 0.5E(W^2) + E(WU) - E(W)[0.5E(W) + E(U)]$$

$$= 0.5E(W^2) + \cancel{E(W)E(U)} - 0.5E(W)^2 - \cancel{E(W)E(U)}$$

because  $W$  and  $U$  are indep.

$$= 0.5(E(W^2) - E(W)^2) = 0.5\text{Var}(W) = \frac{16}{2} = 8$$

Covariance  
Matrix of  $\begin{pmatrix} W \\ V \end{pmatrix}$

$$\begin{bmatrix} 16 & 8 \\ 8 & 5 \end{bmatrix}$$

$$2. \quad f_x(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases} \quad f_{y|x}(x,y) = \begin{cases} \frac{1}{x} & 0 \leq y \leq x \\ 0 & \text{otherwise} \end{cases}$$

a)

$$E(x) = \int_0^1 x \, dx = \left[ \frac{1}{2} x^2 \right]_0^1 = \frac{1}{2}$$

$$E(Y|x) = \int_{-\infty}^{\infty} y f_{y|x}(y|x) \, dy = \int_0^x \frac{y}{x} \, dy = \left[ \frac{1}{2} \frac{y^2}{x} \right]_0^x$$

$$= \frac{1}{2} \frac{x^2}{x} = \frac{x}{2}, \quad \text{if } x = E(x) \text{ then } E(Y|X=E(x)) = \frac{1}{4}$$

$$\therefore E(Y|X=x) = \frac{x}{2} \text{ and } E(Y|X=E(x)) = \frac{1}{4}$$

$$b) \quad E(Y) = E(E(Y|x)) = E\left(\frac{x}{2}\right) = \frac{1}{2} E(x) = \frac{1}{4}$$

$$E(Y) = \frac{1}{4}$$

$$c) \quad E(XY) = E(E(XY|x)) = E(x E(Y|x)) = E\left(\frac{x^2}{2}\right)$$

$$E\left(\frac{x^2}{2}\right) = \int_0^1 \frac{x^2}{2} \, dx = \left[ \frac{x^3}{6} \right]_0^1 = \frac{1}{6}$$

$$d) \quad \text{Cov}(X,Y) = E(XY) - E(X)E(Y) = \frac{1}{6} - \frac{1}{2} \cdot \frac{1}{4} = \frac{1}{24}$$

$$\text{Cov}(X,Y) = \frac{1}{24}$$

3a)  $M$  = wait time in morning  
 $N$  = wait time in evening

$M$  and  $N$  are independent to each other

$$f_M(m) = \begin{cases} \frac{1}{5} & 0 \leq m \leq 5 \\ 0 & \text{otherwise} \end{cases}$$

$$f_N(n) = \begin{cases} \frac{1}{10} & 0 \leq n \leq 10 \\ 0 & \text{otherwise} \end{cases}$$

$$\text{Total Expected wait time} = E(5M + 5N) = 5[E(M) + E(N)]$$

$$E(M) = \int_0^5 \frac{1}{5} m \, dm = \left[ \frac{m^2}{10} \right]_0^5 = 2.5$$

$$E(N) = \int_0^{10} \frac{1}{10} n \, dn = \left[ \frac{n^2}{20} \right]_0^{10} = 5$$

$$\text{Total expected wait time} = 37.5 \text{ minutes}$$

$$b) \text{Var}(5M + 5N) = 25\text{Var}(M) + 25\text{Var}(N)$$

$$\begin{aligned} \text{Var}(M) &= E(m^2) - (E(m))^2 = \int_0^5 f(m) g(m) \, dm - 2.5^2 \\ &= \int_0^5 \frac{1}{5} m^2 \, dm - 6.25 = \left[ \frac{m^3}{15} \right]_0^5 - 6.25 = \frac{25}{12} \end{aligned}$$

$$\text{Var}(N) = \int_0^{10} \frac{1}{10} n^2 \, dn - 25 = \left[ \frac{n^3}{30} \right]_0^{10} - 25 = \frac{25}{3}$$

$$\text{Var}(5M + 5N) = 25 \left[ \frac{25}{12} + \frac{25}{3} \right] = \frac{3125}{12} \approx 260.417$$

$$\begin{aligned} \text{c) } E(5N - 5M) &= 5E(N) - 5E(M) \\ &= 5(5) - 5(2.5) = \underline{12.5} \end{aligned}$$

$$\begin{aligned} \text{d) } \text{Var}(5N - 5M) &= 5^2 \text{Var}(N) + (-5)^2 \text{Var}(M) \\ &= 25 \left( \frac{25}{3} \right) + 25 \left( \frac{25}{12} \right) = \underline{\underline{\frac{3125}{12} \approx 260.417}} \end{aligned}$$

4.  $Y = aX + b$  where  $X, Y$  are random Variables and  $a \neq 0$ ,  
 $\text{corr}(X, Y) = -1$  or  $+1$ .

$$\text{corr}(X, Y) = \rho_{X,Y} = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}$$

$$\begin{aligned} \text{Cov}(X, Y) &= E(XY) - E(X)E(Y) = E[X(ax+b)] - E(X)E(ax+b) \\ &= E(ax^2 + bx) - E(X)[aE(X) + b] \\ &= aE(X^2) + bE(X) - aE(X)^2 - bE(X) \\ &= a[E(X^2) - E(X)^2] = a \text{Var}(X) \end{aligned}$$

$$\therefore \rho_{X,Y} = \frac{a \text{Var}(X)}{\sigma_X \sigma_Y} = \frac{a \sigma_X}{\sigma_Y}$$

$$\begin{aligned} \text{Var}(Y) &= E(Y^2) - E(Y)^2 = E[(ax+b)^2] - [E(ax+b)]^2 \\ &= E[a^2x^2 + 2abx + b^2] - [aE(X) + b]^2 \\ &= a^2E(X^2) + 2abE(X) + b^2 - (a^2E(X)^2 + 2abE(X) + b^2) \\ &= a^2[E(X^2) - E(X)^2] = a^2 \text{Var}(X) \end{aligned}$$

$$\sigma_Y = \sqrt{a^2 \text{Var}(X)} = \pm a \sigma_X$$

$$\therefore \rho_{X,Y} = \frac{a \sigma_X}{\pm a \sigma_X} = \underline{\underline{\pm 1}}$$