

① Buktikan Normal PDF:

$$f(x,y) = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp \left\{ -\frac{\left(\frac{x-\mu_x}{\sigma_x}\right)^2 + \left(\frac{y-\mu_y}{\sigma_y}\right)^2 - 2\rho\left(\frac{x-\mu_x}{\sigma_x}\right)\left(\frac{y-\mu_y}{\sigma_y}\right)}{2(1-\rho^2)} \right\}$$

Marginal

$$\Downarrow \quad f_{X(Y)} = \int_{-\infty}^{\infty} f(x,y) dy = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \int_{-\infty}^{\infty} \exp \left\{ -\frac{\left(\frac{x-\mu_x}{\sigma_x}\right)^2 + \left(\frac{y-\mu_y}{\sigma_y}\right)^2 - 2\rho\left(\frac{x-\mu_x}{\sigma_x}\right)\left(\frac{y-\mu_y}{\sigma_y}\right)}{2(1-\rho^2)} \right\} dy$$

$$\Rightarrow \frac{\exp \left\{ -\frac{\left(\frac{x-\mu_x}{\sigma_x}\right)^2}{2(1-\rho^2)} \right\}}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \int_{-\infty}^{\infty} \exp \left\{ \left(\frac{y-\mu_y}{\sigma_y}\right)^2 - 2\rho\left(\frac{x-\mu_x}{\sigma_x}\right)\left(\frac{y-\mu_y}{\sigma_y}\right) \right\} dy$$

$$\Rightarrow \frac{\exp \left\{ -\frac{\left(\frac{x-\mu_x}{\sigma_x}\right)^2}{2(1-\rho^2)} \right\}}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \int_{-\infty}^{\infty} \exp \left\{ \left(\frac{y-\mu_y}{\sigma_y}\right)^2 - 2\rho\left(\frac{x-\mu_x}{\sigma_x}\right)\left(\frac{y-\mu_y}{\sigma_y}\right) + \rho^2\left(\frac{x-\mu_x}{\sigma_x}\right)^2 - \rho^2\left(\frac{x-\mu_x}{\sigma_x}\right)^2 \right\} dy$$

$$\Rightarrow \frac{\exp\left\{-\left(\frac{x-\mu_x}{\sigma_x}\right)^2\right\}}{2(1-\rho^2)} \exp\left\{\frac{\rho^2}{2(1-\rho^2)}\left(\frac{x-\mu_x}{\sigma_x}\right)^2\right\} \times$$

$$\int_{-\infty}^{\infty} \exp\left\{-\frac{y-\mu_y}{\sigma_y} - \rho\left(\frac{x-\mu_x}{\sigma_x}\right)\right\}^2$$

Let $e = \exp\{ \}$, runny out of room, then

$$\Rightarrow \frac{e^{-\frac{1}{2(1-\rho^2)}\left(\frac{x-\mu_x}{\sigma_x}\right)^2}}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \int_{-\infty}^{\infty} e^{-\frac{1}{2(1-\rho^2)}\left(\frac{y-\mu_y}{\sigma_y} - \rho\left(\frac{x-\mu_x}{\sigma_x}\right)\right)^2} dy$$

Next let $q = \left(\frac{y-\mu_y}{\sigma_y} - \rho\left(\frac{x-\mu_x}{\sigma_x}\right)\right) \cdot \frac{1}{\sqrt{1-\rho^2}} \Rightarrow$ then,

$$\Rightarrow dq = \frac{1}{\sqrt{1-\rho^2}\sigma_y} dy.$$

Simplifying, we get,

$$\Rightarrow \frac{e^{-\frac{1}{2}\left(\frac{x-\mu_x}{\sigma_x}\right)^2}}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}q^2} dq = \frac{e^{-\frac{1}{2}\left(\frac{x-\mu_x}{\sigma_x}\right)^2}}{2\pi\sigma_x} \cdot \sqrt{2\pi}$$

$$\Rightarrow \frac{1}{\sqrt{2\pi}\sigma_x} e^{-\frac{1}{2}\left(\frac{x-\mu_x}{\sigma_x}\right)^2} \equiv \text{Normal Distribution.}$$

\therefore We've shown the marginal distribution for a bivariate normal is a normal distribution.

7b) Find conditional Distribution of $f_{X|Y}(x)$.

Let X and Y be random variables. The conditional distributions are defined

$$\Rightarrow f_{X|Y}(x) = \frac{f(x, y)}{f(y)}, \quad f_{Y|X}(y) = \frac{f(x, y)}{f(x)},$$

where $f(x, y)$ is the joint density function and $f(x), f(y)$ are marginal density functions.

Let $f(x, y)$ be the bivariate normal distribution PDF, then without loss of generality,

$$\begin{aligned} \Rightarrow f_{X|Y} &= \frac{f(x, y)}{f(y)} = \frac{\frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp\left\{-\frac{\left(\frac{x-\mu_x}{\sigma_x}\right)^2 + \left(\frac{y-\mu_y}{\sigma_y}\right)^2 - 2\rho\left(\frac{x-\mu_x}{\sigma_x}\right)\left(\frac{y-\mu_y}{\sigma_y}\right)}{2(1-\rho^2)}\right\}}{\frac{1}{\sqrt{2\pi}\sigma_y} \cdot e^{-\frac{1}{2}\left(\frac{y-\mu_y}{\sigma_y}\right)^2}} \\ &= \frac{1}{\sqrt{2\pi}\sigma_x\sqrt{1-\rho^2}} \exp\left\{-\frac{1}{2(1-\rho^2)}\left(\left(\frac{x-\mu_x}{\sigma_x}\right)^2 - 2\rho\left(\frac{x-\mu_x}{\sigma_x}\right)\left(\frac{y-\mu_y}{\sigma_y}\right) + \left(\frac{y-\mu_y}{\sigma_y}\right)^2(1-\rho^2)\right)\right\} \\ &= \frac{1}{\sqrt{2\pi}\sigma_x\sqrt{1-\rho^2}} \exp\left\{\frac{1}{2(1-\rho^2)}\left(\left(\frac{x-\mu_1}{\sigma_1}\right) - \rho\left(\frac{y-\mu_2}{\sigma_2}\right)\right)^2\right\} \\ &= \frac{1}{\sqrt{2\pi}\sigma_x\sqrt{1-\rho^2}} \exp\left\{\frac{1}{2\sigma_x^2(1-\rho^2)}\left(x - \left(\mu_1 + \sigma_x\rho\left(\frac{y-\mu_2}{\sigma_2}\right)\right)\right)^2\right\} \\ &= N\left(\mu_x + \rho\sigma_x\left(\frac{y-\mu_y}{\sigma_y}\right), \sigma_x^2(1-\rho^2)\right). \end{aligned}$$

\therefore The conditional distribution is normally distributed.

2) Setup

① $Y = \text{"final exam"}$

② $\mu_Y = 73, \sigma_Y = 12$

③ $X = \text{"average of 4"}$

④ $\mu_X = 70, \sigma_X = 15$

⑤ $\rho = .71$

AIM: Find given values,

(a) $E(Y|X) = \mu_{Y|X}$

$$\begin{aligned}\Rightarrow \mu_{Y|X} &= \mu_X + \rho\sigma_X\left(\frac{X - \mu_X}{\sigma_X}\right) \\ &= 70 + (.71)(15)\left(\frac{X - 73}{12}\right) \\ &= \boxed{70 + 10.65\left(\frac{X - 73}{12}\right)}\end{aligned}$$

(b) $V(Y|X) = \sigma_{Y|X}^2$

$$\begin{aligned}\Rightarrow \sigma_{Y|X}^2 &= \sigma_X^2(1 - \rho^2) \\ &= \sigma_X^2(0.4959) \\ &= 225(0.4959) \\ &= \boxed{111.5775}\end{aligned}$$

(c) $P(Y > 90 | X = 80)$

$$\Rightarrow \mu_{Y|X=80} = 70 + 10.65\left(\frac{80 - 73}{12}\right) \approx 76.2125.$$

Thus,

$$\Rightarrow P(Z > \frac{90 - 76.2125}{10.56})$$

$$\Rightarrow P(Z > 1.3056) \approx \boxed{0.0961}.$$

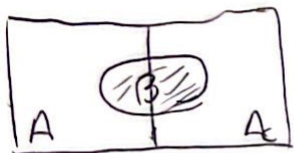
(3) Setup

Event:

- A = "disease"
- A^c = "no disease"
- B = "positive"
- B^c = "negative"

Prob:

- $P(A) = 0.01$
- $P(A^c) = 0.99$
- $P(B|A) = 0.97$
- $P(B|A^c) = 0.06$



(a) find $P(B)$

$$\Rightarrow P(B) = P(B|A)P(A) + P(B|A^c)P(A^c) \\ = (0.97)(0.01) + (0.06)(0.99) \approx \boxed{0.0691}$$

(b) $P(A|B)$

$$\Rightarrow P(A|B) = \frac{P(A)P(B|A)}{P(A)P(B|A) + P(A^c)P(B|A^c)} = \frac{(0.01)(0.97)}{(0.01)(0.97) + (0.99)(0.06)} \\ = \frac{0.0097}{0.0691} \approx \boxed{0.14041}$$

$$(c) P(A^c|B^c) = \frac{P(A^c)[1 - P(B|A^c)]}{P(A)[1 - P(B|A)] + P(A^c)[1 - P(B|A^c)]} \\ = \frac{(0.99)(1 - 0.06)}{(0.01)(1 - 0.97) + (0.99)(1 - 0.06)} \approx \boxed{0.999}$$

4) Setup

$$* P(B) = P(A_1)P(B|A_1) + P(A_2)P(B|A_2) + P(A_3)P(B|A_3) \\ \approx \boxed{0.187}$$

Events:

• $A_1 = \text{Species 1}$

• $A_2 = \text{Species 2}$

• $A_3 = \text{Species 3}$

• $B = \text{logged}$

Prob

• $P(B|A_1) = 1$

• $P(B|A_2) = 0.15$

• $P(B|A_3) = 0.5$

$$(a) P(A_1|B) = \frac{P(A_1)P(B|A_1)}{P(B)} = \frac{(0.45)(1)}{(0.45)(1) + (0.38)(0.15) + (0.17)(0.5)} \approx \boxed{0.2404}$$

$$(b) P(A_2|B) = \frac{P(A_2)P(B|A_2)}{P(B)} = \frac{(0.38)(0.15)}{0.187} \approx \boxed{0.3048}$$

$$(c) P(A_3|B) = \frac{P(A_3)P(B|A_3)}{P(B)} = \frac{(0.17)(0.5)}{0.187} \approx \boxed{0.4545}$$