

H W #1      solutions

Ex. 2.3,

①  $X \perp Y | Z$  if and only if  $p(x, y | z) = p(x | z)p(y | z)$ .

✓ Indep.  $\rightarrow$  factorization

$$p(x, y | z) = \underbrace{p(x | z)}_{g(x, z)} \underbrace{p(y | z)}_{h(y, z)} = g(x, z)h(y, z), \quad \text{as desired}$$

✓ Factorization  $\rightarrow$  indep.

If  $p(x, y | z) = g(x, z)h(y, z)$ ,

$$\begin{aligned} \sum_{x,y} p(x, y | z) &= 1 = \sum_{x,y} g(x, z)h(y, z) = \\ &= \left( \sum_x g(x, z) \right) \left( \sum_y h(y, z) \right). \end{aligned}$$

Moreover,

$$\begin{cases} p(x | z) = \sum_y p(x, y | z) = g(x, z) \sum_y h(y, z), \\ p(y | z) = \sum_x p(x, y | z) = h(y, z) \sum_x g(x, z). \end{cases}$$

Then  $p(x | z)p(y | z) = g(x, z)h(y, z) \times$   
 $\times \underbrace{\left( \sum_x g(x, z) \sum_y h(y, z) \right)}_{\approx 1} = p(x, y | z), \quad \text{as desired}$

(2.) Ex. 2.4,

$$x_1 \sim N(\mu_1, \sigma_1^2), \quad x_2 \sim N(\mu_2, \sigma_2^2)$$

$$p(y) = ?, \text{ where } y = x_1 + x_2$$

$$\text{Note that } N(x|\mu, \sigma^2) = \Phi(x-\mu|\sigma) =$$

$$= \frac{1}{\sigma} \Phi\left(\frac{x-\mu}{\sigma}\right), \text{ where}$$

$\Phi(z) = \text{pdf of } N(0, 1).$   
standard normal

$$\text{Then } p(y) = \int dx_1 \frac{1}{\sigma_1} \Phi\left(\frac{x_1-\mu_1}{\sigma_1}\right) \frac{1}{\sigma_2} \Phi\left(\frac{y-x_1-\mu_2}{\sigma_2}\right) =$$

$$= \frac{1}{2\pi\sigma_1\sigma_2} \int dx_1 e^{-\frac{1}{2}\left(\frac{x_1-\mu_1}{\sigma_1}\right)^2} e^{-\frac{1}{2}\left(\frac{y-x_1-\mu_2}{\sigma_2}\right)^2}.$$

Complete the square:

$$\left(\frac{x_1-\mu_1}{\sigma_1}\right)^2 + \left(\frac{y-x_1-\mu_2}{\sigma_2}\right)^2 =$$

$$= (\omega_1 + \omega_2)(x_1 - \bar{x}) + \frac{\omega_1\omega_2}{\omega_1 + \omega_2} (y - (\mu_1 + \mu_2))^2,$$

$$\text{where } \begin{cases} \omega_1 = \sigma_1^{-2}, \\ \omega_2 = \sigma_2^{-2}. \end{cases} \quad \frac{\omega_1\omega_2}{\omega_1 + \omega_2} = \frac{1}{\sigma_1^2 + \sigma_2^2}$$

$$\bar{x} = \frac{\omega_1\mu_1 + \omega_2(y - \mu_2)}{\omega_1 + \omega_2}$$

Finally,

$$p(y) = \frac{1}{2\pi\sigma_1\sigma_2} e^{-\frac{1}{2(\sigma_1^2+\sigma_2^2)}(y-(\mu_1+\mu_2))^2} \times$$
$$\times \underbrace{\int dx_1 e^{-\frac{1}{2}(\sigma_1^2+\sigma_2^2)(x_1-\bar{x})^2}}_{\sqrt{2\pi} \frac{\sigma_1\sigma_2}{\sqrt{\sigma_1^2+\sigma_2^2}}} \quad \textcircled{=}$$

$$\textcircled{=} \frac{1}{\sqrt{2\pi(\sigma_1^2+\sigma_2^2)}} e^{-\frac{1}{2(\sigma_1^2+\sigma_2^2)}[y-(\mu_1+\mu_2)]^2} =$$
$$= \mathcal{N}(y | \mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2).$$

=====

(3.)

Ex. 2.6,Variance of a sum

$$V[X+Y] = E[(X+Y)^2] - (E[X]+E[Y])^2 =$$

$$= E[X^2 + Y^2 + 2XY] - [E[X]^2 + E[Y]^2 + 2E[X]E[Y]] \quad \text{④}$$

$$\therefore (E[X^2] - E[X]^2) + (E[Y^2] - E[Y]^2) +$$

$$+ 2(E[XY] - E[X]E[Y]) =$$

$$= \text{Var}[X] + \text{Var}[Y] + 2\text{Cov}[X, Y].$$

If  $X$  &  $Y$  are indep.,  $\text{Cov}[X, Y] = 0$

$$\text{and } V[X+Y] = V[X] + V[Y].$$

Q. Ex. 2.11

X	Y	P
G	G	1/4
G	B	1/4
B	G	1/4
B	B	1/4

(b)  $\rightarrow$  (a)

(b)  $\rightarrow$  "  $P(N_g=1, N_B=1)$

(a)  $P(N_g=1 | N_B \geq 1) = \frac{P(N_g=1, N_B \geq 1)}{P(N_B \geq 1)} = \frac{1/2}{3/4} = \frac{2}{3}$ .

or read it off directly from the prob. table

$$(b) P(X=G | Y=B) = \frac{P(X=G, Y=B)}{P(Y=B)} = \frac{1/4}{1/2} = \frac{1}{2}.$$

or directly from prob. table