

1) a)

Gaussian Distribution:

$$N(x|\mu, \sigma^2) = \frac{1}{(2\pi \sigma^2)^{1/2}} \exp \left\{ -\frac{1}{2\sigma^2} (x-\mu)^2 \right\}$$

To the above formula is normalized:

$$\int_{-\infty}^{\infty} N(x|\mu, \sigma^2) dx = 1$$

$$\text{Let } I = \int_{-\infty}^{\infty} \exp \left( -\frac{1}{2\sigma^2} (x-\mu)^2 \right) dx$$

$$I^2 = \int_{-\infty}^{\infty} \exp \left( -\frac{1}{2\sigma^2} (x-\mu)^2 \right) dx$$

$$\Leftrightarrow I^2 = \int_{-\infty}^{\infty} \exp \left( -\frac{1}{2\sigma^2} x^2 \right) dx \cdot \int_{-\infty}^{\infty} \exp \left( -\frac{1}{2\sigma^2} x^2 \right) dx$$

$$\Leftrightarrow I^2 = \int_{-\infty}^{\infty} \exp \left( -\frac{1}{2\sigma^2} x^2 \right) dx \cdot \int_{-\infty}^{\infty} \exp \left( -\frac{1}{2\sigma^2} y^2 \right) dy$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \exp \left( -\frac{1}{2\sigma^2} (x^2 + y^2) \right) dx dy \quad (1)$$

$$\text{Let } x = r \cos \theta \\ y = r \sin \theta$$

$$\Rightarrow x^2 + y^2 = r^2 (\cos^2 \theta + \sin^2 \theta) = r^2.$$

Using the Jacobian determinant:

$$\frac{\partial(x, y)}{\partial(r, \theta)} = \begin{vmatrix} \frac{\partial(x)}{\partial(r)} & \frac{\partial(x)}{\partial(\theta)} \\ \frac{\partial(y)}{\partial(r)} & \frac{\partial(y)}{\partial(\theta)} \end{vmatrix}$$

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

$$= r \cos^2 \theta + r \sin^2 \theta = r.$$

$$(1) \Leftrightarrow \int_0^\infty \int_0^\infty \exp\left(-\frac{1}{2\sigma^2} \cdot r^2\right) r \, dr \, d\theta$$

$$= \int_0^\infty \exp\left(-\frac{x^2}{2\sigma^2}\right) x \, dx = \text{value} = 2\pi \cdot A. \quad (2)$$

$$\text{Let: } u = x^2 \Rightarrow du = 2x \, dx$$

$$A = \int_0^\infty \exp\left(-\frac{u}{2\sigma^2}\right) \frac{du}{2} = \frac{1}{2} \int_0^\infty \exp\left(-\frac{u}{2\sigma^2}\right) du$$

$$(e^u)' = u' \cdot e^u$$

$$\int e^{ax+b} dx = \frac{1}{a} \cdot e^{ax+b}$$

Thứ

Ngày

No.

$$(2) \Leftrightarrow 2\pi \cdot \frac{1}{2} \int_0^\infty \exp\left(\frac{-u}{2\sigma^2}\right) du$$

$$= \pi \left[ \exp\left(\frac{-u}{2\sigma^2}\right) \cdot (-2\sigma^2) \right] \Big|_0^\infty$$

$$= +2\pi\sigma^2$$

$$\Rightarrow I^2 = +2\pi\sigma^2 \Leftrightarrow I = (2\pi\sigma^2)^{\frac{1}{2}}$$

So,

$$\int_{-\infty}^\infty N(x|\mu, \sigma^2) dx = \frac{1}{(2\pi\sigma^2)^{\frac{1}{2}}} \int_{-\infty}^\infty \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) dx$$

$$\Leftrightarrow \int_{-\infty}^\infty N(x|\mu, \sigma^2) dx = \frac{1}{(2\pi\sigma^2)^{\frac{1}{2}}} \cdot I = \frac{1}{(2\pi\sigma^2)^{\frac{1}{2}}} \cdot (2\pi\sigma^2)^{\frac{1}{2}}$$

$$\Leftrightarrow \int_{-\infty}^\infty N(x|\mu, \sigma^2) dx = 1 \quad (\text{proved})$$

b)

We Have:

$$E(x) = \int_{-\infty}^{\infty} N(x|\mu, \delta^2) \cdot x \, dx$$

$$\Leftrightarrow E(x) = \frac{1}{(2\pi\delta^2)^{\frac{1}{2}}} \cdot \int_{-\infty}^{\infty} x \cdot \exp\left(-\frac{1}{2\delta^2}(x-\mu)^2\right) dx$$

$\underbrace{\quad \quad \quad}_{(x-\mu)^2}$

A.

$$\text{Let } x = \frac{x-\mu}{\sqrt{2}\delta} \Rightarrow dx = \frac{\sqrt{2}}{2\delta} dx$$

$$A = \int_{-\infty}^{\infty} (-\sqrt{2}\delta x + \mu) \cdot \exp(-x^2) \cdot \sqrt{2}\delta^2 dx$$

$$A = \sqrt{2}\delta \int_{-\infty}^{\infty} (\sqrt{2}\delta x + \mu) \cdot \exp(-x^2) dx$$

$$= \sqrt{2}\delta \cdot \left[ \int_{-\infty}^{\infty} \sqrt{2}\delta x \cdot \exp(-x^2) dx + \mu \int_{-\infty}^{\infty} \exp(-x^2) dx \right]$$

$$= \sqrt{2}\delta \cdot \left[ \sqrt{2}\delta \int_{-\infty}^{\infty} x \cdot \exp(-x^2) dx + \mu \int_{-\infty}^{\infty} \exp(-x^2) dx \right]$$

KOKUYO

$$\int_0^\infty \exp(-t^2) dt = \frac{\sqrt{\pi}}{2}$$

Thứ

Ngày

No.

We have  $\exp(-x^2)$  is a even function -

$$\Rightarrow \int_{-\infty}^{\infty} \exp(-x^2) dx = 2 \int_0^{\infty} \exp(-x^2) dx.$$

$$= 2 \cdot \frac{\sqrt{\pi}}{2} = \sqrt{\pi}.$$

$$\Rightarrow A = \sqrt{2} \cdot \delta \left( \cdot, \mu, \sqrt{\pi} \right)$$

So,

$$E(x) = \frac{1}{(2\pi\delta^2)^{\frac{1}{2}}} \cdot \sqrt{2}\delta \cdot \mu \cdot \sqrt{\pi}$$

$$\Leftrightarrow E(x) = \mu \quad (\text{Proved})$$

c)

$$\text{Var}(x) = E[x^2] - E[x]^2$$

$$+ E[x^2] = \frac{1}{(2\pi\delta^2)^{\frac{1}{2}}} \cdot \int_{-\infty}^{\infty} x^2 \cdot \exp\left(-\frac{1}{2\delta^2} (x - \mu)^2\right) dx$$

B

$$\text{Let } t = \frac{x - \mu}{\sqrt{2}\delta} \Rightarrow dt = \frac{\sqrt{2}}{2\delta} dx$$

$$\int_a^b f(t) G(t) dt = [F(t)G(t)] \Big|_a^b - \int_a^b F(t) g(t) dt$$

Thứ

Ngày

No.

$$b = \int_{-\infty}^{\infty} (\sqrt{2\delta t} + \mu)^2 \cdot \exp(-t^2) \cdot \sqrt{2\delta} dt$$

$$= \sqrt{2} \cdot \delta \int_{-\infty}^{\infty} (2\delta^2 t^2 + 2\sqrt{2}\delta t \cdot \mu + \mu^2) \exp(-t^2) dt$$

$$= \sqrt{2} \cdot \delta \left[ \left( \int_{-\infty}^{\infty} 2\delta^2 t^2 \cdot \exp(-t^2) dt \right) + \left( \int_{-\infty}^{\infty} 2\sqrt{2}\delta \mu \cdot t \cdot \exp(-t^2) dt \right) \right. \\ \left. + \left( \int_{-\infty}^{\infty} \mu^2 \cdot \exp(-t^2) dt \right) \right]$$

$$= \sqrt{2} \cdot \delta \cdot \left[ \left( 2\delta^2 \int_{-\infty}^{\infty} t^2 \exp(-t^2) dt \right) + \underbrace{\left( 2\sqrt{2}\delta \mu \int_{-\infty}^{\infty} t \cdot \exp(-t^2) dt \right)}_0 \right. \\ \left. + \left( \mu^2 \int_{-\infty}^{\infty} \exp(-t^2) dt \right) \right]$$

$$= \sqrt{2} \delta \cdot \left[ \left( 2\delta^2 \left( \left[ -\frac{t}{2} \exp(-t^2) \right] \Big|_{-\infty}^{\infty} + \int_{-\infty}^{\infty} \frac{1}{2} \exp(-t^2) dt \right) \right) + \right. \\ \left. \mu^2 \cdot \sqrt{\pi} \right]$$

KOKUYO

$$= \sqrt{2} \delta \cdot \left[ 2\delta^2 \left( 0 + \frac{1}{2} \cdot \sqrt{\mu} \right) + \mu^2 \cdot \sqrt{\mu} \right]$$

$$= \sqrt{2} \cdot \delta \left[ \delta^2 \cdot \sqrt{\mu} + \mu^2 \cdot \sqrt{\mu} \right]$$

$$= \sqrt{2} \cdot \delta \cdot \sqrt{\mu} (\delta^2 + \mu^2)$$

$$\Rightarrow E[X^2] = \frac{1}{(\delta^2 + \mu^2)^2} \cdot \sqrt{2} \cdot \delta \cdot \sqrt{\mu} (\delta^2 + \mu^2)$$

$$= \delta^2 + \mu^2$$

$$\begin{aligned} \Rightarrow \text{Var}(x) &= E[X^2] - E[X]^2 \\ &= \delta^2 + \mu^2 - \mu^2 \quad (\text{from b}) \\ &= \delta^2 \quad (\text{proved}) \end{aligned}$$

d)

a) 2)  
Suppose  $x$ : a  $D$ -dimensional vector.

$$x = \begin{pmatrix} x_a \\ x_b \end{pmatrix} \quad \Leftrightarrow \quad \mu = \begin{pmatrix} \mu_a \\ \mu_b \end{pmatrix}$$

Covariance matrix  $\Sigma$ .

$$\Sigma = \begin{pmatrix} \Sigma_{aa} & \Sigma_{ab} \\ \Sigma_{ba} & \Sigma_{bb} \end{pmatrix} \Rightarrow A = \Sigma^{-1} = \begin{pmatrix} A_{aa} & A_{ab} \\ A_{ba} & A_{bb} \end{pmatrix}$$

$\Sigma$  is symmetric  $\Rightarrow \Sigma_{aa}; \Sigma_{bb}$  are symmetric.

$$\Sigma_{ab} = \Sigma_{ba}^T$$

We have:

$$-\frac{1}{2} (x - \mu)^T \cdot \Sigma^{-1} (x - \mu) = -\frac{1}{2} (x - \mu)^T A (x - \mu)$$

$$= -\frac{1}{2} (x_a - \mu_a)^T A_{aa} (x_a - \mu_a) - \frac{1}{2} (x_a - \mu_a)^T A_{ab} (x_b - \mu_b)$$

$$- \frac{1}{2} (x_b - \mu_b)^T A_{ba} (x_a - \mu_a) - \frac{1}{2} (x_b - \mu_b)^T A_{bb} (x_b - \mu_b)$$

$$= -\frac{1}{2} x_a^T A_{aa}^{-1} x_a + x_a^T (A_{aa} \mu_a - A_{ab} (\mu_b - \mu_a)) + c. \quad (1)$$

Compare with: Gauss distribution

$$\Delta^2 = \frac{-1}{2} x^T \tilde{\Sigma}^{-1} n + x^T \tilde{\Sigma}^{-1} \mu + c$$

$$(1) \Rightarrow \tilde{\Sigma}_{ab} = A_{aa}^{-1}$$

$$\begin{aligned} \mu_{ab} &= \sum_{a/b} (A_{aa} \mu_a - A_{ab} (\mu_b - \mu_b)) \\ &= \mu_a - A_{aa}^{-1} A_{ab} (\mu_b - \mu_b) \end{aligned}$$

Using Schur complement.

$$\begin{pmatrix} A & B \\ C & D \end{pmatrix}^{-1} = \begin{pmatrix} M & -MBD^{-1} \\ -D^{-1}CMD^{-1} & D^{-1}CMBD^{-1} \end{pmatrix}, M = (A - BD^{-1}C)^{-1}$$

$$\Rightarrow A_{aa} = (\sum_{aa} - \sum_{ab} \tilde{\Sigma}_{bb}^{-1} \sum_{ba})^{-1}$$

$$A_{ab} = -(\sum_{aa} - \sum_{ab} \tilde{\Sigma}_{bb}^{-1} \sum_{ba})^{-1} \cdot \sum_{ab} \tilde{\Sigma}_{bb}^{-1}$$

So,

$$\mu_{ab} = \mu_a + \sum_{ab} \tilde{\Sigma}_{bb}^{-1} (\mu_b - \mu_b)$$

$$\tilde{\Sigma}_{ab} = \sum_{aa} - \sum_{ab} \tilde{\Sigma}_{bb}^{-1} \sum_{ba}$$

$$\Rightarrow p(x_a | x_b) = N(x_{a|b} | \mu_{a|b}; \Sigma_{a|b})$$

b)

The Marginal Distribution:

$$p(x_a) = \int p(x_a, x_b) dx_b$$

We have quadratic form related to  $x_b$ .

$$\frac{1}{2} x_b^T A_{bb} x_b + x_b^T m = \frac{1}{2} (x_b - A_{bb}^{-1} m)^T A_{bb}$$

$$(x_b - A_{bb}^{-1} m) + \frac{1}{2} m^T A_{bb}^{-1} m$$

$$\text{with } m = A_{bb} \mu_b - A_{ba} (x_a - \mu_a)$$

~~If we integrate over unnormalized Games.~~

$$\int \exp \left( -\frac{1}{2} (x_b - A_{bb}^{-1} m)^T A_{bb} (x_b - A_{bb}^{-1} m) \right) dx_b$$

$\Rightarrow$  The remaining:

$$\begin{aligned} & -\frac{1}{2} x_a^T (A_{aa} - A_{ab} A_{bb}^{-1} A_{ba}) x_a + x_a^T (A_{aa} - A_{ab} A_{bb}^{-1} A_{ba})^{-1} \mu_a \\ & + C. \end{aligned}$$

Similarly,

$$E[x_a] = \mu_a$$

$$\text{Cov}(x_a) = \Sigma_{aa}$$

$$\Rightarrow p(x_a) = N(x_a | \mu_a, \Sigma_{aa})$$