

$$f(x, \tau) = f(z) = N(z | 0, \sigma^2 I) = \frac{1}{2\pi} \frac{1}{\sigma^2} \exp\left\{-\frac{1}{2} z^T \frac{1}{\sigma^2} I z\right\}, \quad z = (x, \tau)$$

2変数 z を成分表示すると

$$f(x, \tau) = \frac{1}{2\pi} \frac{1}{\sigma^2} \exp\left\{-\frac{1}{2} (x, \tau) \frac{1}{\sigma^2} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ \tau \end{pmatrix}\right\} = \frac{1}{2\pi} \frac{1}{\sigma^2} \exp\left\{-\frac{1}{2\sigma^2} (x^2 + \tau^2)\right\}$$

2変数 z の積分 (6.47) F)

$$g(x) = \int_{-\infty}^{\infty} f(x, \tau) d\tau = \int_{-\infty}^{\infty} \frac{1}{2\pi} \frac{1}{\sigma^2} \exp\left\{-\frac{1}{2\sigma^2} (x^2 + \tau^2)\right\} d\tau = \frac{1}{2\pi\sigma^2} \exp\left(-\frac{x^2}{2\sigma^2}\right) \int_{-\infty}^{\infty} \exp\left(-\frac{\tau^2}{2\sigma^2}\right) d\tau$$

$$= \frac{1}{2\pi\sigma^2} \exp\left(-\frac{x^2}{2\sigma^2}\right) \sqrt{2\pi}\sigma = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{x^2}{2\sigma^2}\right)$$

2変数 (6.46) F')

$$k(x, x_n) = \frac{\exp\left\{-\frac{(x-x_n)^2}{2\sigma^2}\right\}}{\sum_m \exp\left\{-\frac{(x-x_m)^2}{2\sigma^2}\right\}} \dots \textcircled{1}$$

(6.48) に基づく積分 $\int_{-\infty}^{\infty} \dots d\tau$

$$\int f(x-x_n, \tau-x_n) d\tau = \int \frac{1}{2\pi} \frac{1}{\sigma^2} \exp\left[-\frac{1}{2\sigma^2} \{(x-x_n)^2 + (\tau-x_n)^2\}\right] d\tau = \frac{1}{2\pi\sigma^2} \exp\left\{-\frac{(x-x_n)^2}{2\sigma^2}\right\} \int \exp\left\{-\frac{(\tau-x_n)^2}{2\sigma^2}\right\} d\tau$$

$$= \frac{1}{2\pi\sigma^2} \exp\left\{-\frac{(x-x_n)^2}{2\sigma^2}\right\} \sqrt{2\pi}\sigma = \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{(x-x_n)^2}{2\sigma^2}\right\}$$

よって (6.48) F')

$$p(\tau|x) = \frac{1}{\sqrt{2\pi}\sigma} \frac{\sum_n \exp\left\{-\frac{(x-x_n)^2 + (\tau-x_n)^2}{2\sigma^2}\right\}}{\sum_m \exp\left\{-\frac{(x-x_m)^2}{2\sigma^2}\right\}} = \sum_n k(x, x_n) \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{(\tau-x_n)^2}{2\sigma^2}\right\} = \sum_n k(x, x_n) \cdot N(\tau|x_n, \sigma^2)$$

を得る。すなわち、

$$E[\tau|x] = \int \tau p(\tau|x) d\tau = \int \tau \sum_n k(x, x_n) N(\tau|x_n, \sigma^2) d\tau = \sum_n k(x, x_n) \int \tau N(\tau|x_n, \sigma^2) d\tau = \sum_n k(x, x_n) x_n$$

また、 $\text{var}[\tau|x] =$

$$\int \tau^2 p(\tau|x) d\tau = \int \tau^2 \sum_n k(x, x_n) N(\tau|x_n, \sigma^2) d\tau = \sum_n k(x, x_n) \int \tau^2 N(\tau|x_n, \sigma^2) d\tau = \sum_n k(x, x_n) \sigma^2$$

を得る。

x_n 期待値 = 分散平均

τ^2 期待値 = 分散分散