

Exercise Round 2

$$1) \quad E[x] = \int x p(x) dx \quad \int := \int_{-\infty}^{+\infty}$$

$$\text{Var}[x] = \int (x - E[x])^2 p(x) dx \quad \leftarrow$$

$$\int_{-\infty}^{+\infty} p(x) dx = 1$$

$$a) \quad \text{Var}[x] = E[x^2] - E^2[x] \quad \leftarrow$$

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

Expand:

$$= \int (x^2 - 2x E[x] + E^2[x]) p(x) dx$$

$$= \underbrace{\int x^2 p(x) dx} - \underbrace{\int 2x E[x] p(x) dx} + \underbrace{\int E^2[x] p(x) dx}$$

$$E[h(x)] = \int h(x) p(x) dx$$

$E[x]$ and $E^2[x]$ are constant

$$= E[x^2] - 2 E[x] \underbrace{\int x p(x) dx}_{E[x]} + \underbrace{E[x]^2}_{E[x]^2} \int p(x) dx$$

$$= E[x^2] - 2 E[x]^2 + E[x]^2$$

$$= E[x^2] - (E[x])^2$$

$$2) \quad \underline{p(x)} = N(x; \overset{\checkmark}{\mu}, \overset{\checkmark}{\sigma^2}) = \frac{1}{\sqrt{2\pi} \sigma} \exp\left(-\frac{1}{2\sigma^2} (x-\mu)^2\right)$$

$$E[x] = \int x \underbrace{\frac{1}{\sqrt{2\pi} \sigma}} \exp\left(-\frac{1}{2\sigma^2} (x-\mu)^2\right) dx$$

$$= \frac{1}{\sqrt{2\pi} \sigma} \int \underbrace{x \exp\left(-\frac{1}{2\sigma^2} (x-\mu)^2\right)} dx \quad \textcircled{1}$$

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

$$1 \cdot \left(\dots \mu \dots \right) \exp\left(-\frac{1}{2\sigma^2} (x-\mu)^2\right) \cdot (-1) \dots \textcircled{2}$$

$$= \frac{1}{\sqrt{2\pi}\sigma} \int \left(\underbrace{x - \mu}_{-1} + 1 \right) \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) dx \leftarrow \text{②}$$

Expand

$$\rightarrow = \frac{1}{\sqrt{2\pi}\sigma} \int \underbrace{\frac{1}{\sigma^2} (x - \mu) \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right)}_{-1} dx \quad \text{③}$$

$$+ \frac{1}{\sqrt{2\pi}\sigma} \int \mu \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) dx \leftarrow$$

$$\frac{\int P(x) dx = 1}{\text{④}}$$

$$= \frac{-\sigma^2}{\sqrt{2\pi}\sigma} \int \frac{d}{dx} \left(\exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) \right) dx \quad \text{⑤}$$

$$+ \mu$$

$$= \frac{-\sigma^2}{\sqrt{2\pi}\sigma} \left(\exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) \right) \Big|_{-\infty}^{+\infty} + \mu$$

$$= 0 + \mu = \mu$$

2) we have: $Z = Lx$, $E[x] = m$, $\text{Cov}[x] = P$ }

show: $E[Z] = Lm$, $Cov[Z] = LP L^T$

$$E[Z] = E[Lx] = \int Lx p(x) dx = L \underbrace{\int x p(x) dx}_{= E[x] = m} = Lm$$

$$\begin{aligned} Cov[Z] &= Cov[Lx] = E[(Lx - E[Lx])(Lx - E[Lx])^T] \\ &= E[(Lx - Lm)(Lx - Lm)^T] \\ &= E[L(x - m)(x - m)^T L^T] \\ &= L \underbrace{E[(x - m)(x - m)^T]}_{Cov[x] = P} L^T = LP L^T \end{aligned}$$

3) $y_n = g(x_n) + r_n$, $Var\{r_n\} = \sigma_n^2$ & ~

a)

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we define $y_n = \frac{y_n}{\sigma_n}$, $g(x_n) = \frac{g(x_n)}{\sigma_n}$, $\hat{r}_n = \frac{\hat{r}_n}{\sigma_n}$

$$\rightarrow \hat{y}_n = \hat{g}(x_n) + \hat{r}_n \quad \text{Var}\{\hat{r}_n\} = ?$$

$$\text{Var}[\hat{r}_n] = \text{Var}\left[\frac{r_n}{\sigma_n}\right] = \frac{1}{\sigma_n} \text{Var}[r_n] \frac{1}{\sigma_n} = \frac{\sigma_n^2}{\sigma_n^2} = 1$$

we saw that $\begin{cases} Z = Lx, & \text{Cov}[x] = P \\ \text{Cov}[Z] = L P L^T \end{cases}$

b) $y_n = g(x_n) + r_n$, $\text{Var}[r_n] = \sigma_n^2$ WLS Problem

$$J_1 = \sum_n \frac{1}{\sigma_n^2} (y_n - g(x_n))^2$$

$$\hat{y}_n = \hat{g}(x_n) + \hat{r}_n, \quad \text{Var}[\hat{r}_n] = 1 \quad \text{LS problem}$$

$$\sum (\hat{y}_n - \hat{g}(x_n))^2 = \sum (y_n - g(x_n))^2$$

$$J_2 = \sum_n (y_n - g(x_n))^2 = \sum_n \left(\frac{y_n}{\sigma_n} - \frac{g(x_n)}{\sigma_n} \right)^2$$

$$= \sum_n \left(\frac{1}{\sigma_n} \right)^2 (y_n - g(x_n))^2$$

$$\boxed{J_1 = J_2}$$