

1. It's given that $w \sim N(0, b^{-1}I)$.

Assume that $y \sim N(Aw, \sigma^2)$ where A is the design matrix and σ is the unknown variable (variance).

Let $\sigma^2 = a^{-1}$.

→ Input data

$\therefore P(w|D) \propto P(D|w) \cdot P(w)$ (Bayes' rule)

$$\propto e^{-\frac{1}{2\sigma^2}(Aw-y)^T(Aw-y)} \cdot e^{-\frac{1}{2}w^T(b^{-1}I)^{-1}w}$$

$$= e^{-\frac{a}{2} \cdot [(Aw-y)^T(Aw-y) + \frac{b}{a} w^T w]}$$

$$= e^{-\frac{a}{2} \cdot (w^T A^T A w - 2w^T A^T y + y^T y + \frac{b}{a} w^T w)}$$

↳ It's a scalar, so $(w^T A^T y)^T = y^T A w = w^T A^T y$

$$= e^{-\frac{1}{2} \cdot [w^T (aA^T A + bI) w - 2aw^T A^T y + ay^T y]}$$

Let $\Lambda = aA^T A + bI$

$$w^T \Lambda \mu = w^T a A^T y$$

$$\Rightarrow \mu = a \Lambda^{-1} A^T y$$

$$\propto e^{-\frac{1}{2}(w^T \Lambda w - 2w^T \Lambda \mu + \mu^T \mu) + \text{constant}}$$

↳ normalization term

$$\therefore P(w|D) \sim N(\mu, \Lambda^{-1}) = N(a \Lambda^{-1} A^T y, (aA^T A + bI)^{-1})$$

$$= N(a(aA^T A + bI)^{-1} A^T y, (aA^T A + bI)^{-1})$$

$$= N((A^T A + \frac{b}{a} I)^{-1} A^T y, (aA^T A + bI)^{-1})$$

which is also a multivariate gaussian distribution.

對應的 quadratic form : $(w - \mu)^T \Lambda (w - \mu)$
 $= w^T \Lambda w - 2w^T \Lambda \mu + \mu^T \mu$

Appendix : Sequential Estimator

$$\mu_n = \frac{x_1 + x_2 + \dots + x_n}{n}, \quad \mu_{n+1} = \frac{x_1 + x_2 + \dots + x_{n+1}}{n+1}$$

$$\Rightarrow \mu_{n+1} = \frac{n\mu_n + x_{n+1}}{n+1} = \frac{(n+1)\mu_n + x_{n+1} - \mu_n}{n+1} = \mu_n + \frac{x_{n+1} - \mu_n}{n+1}$$

→ M2 : 平方和

$$M2_{n+1} = \sum_{i=1}^{n+1} (x_i - \mu_{n+1})^2 = \sum_{i=1}^n (x_i - \mu_n + \mu_n - \mu_{n+1})^2 + (x_{n+1} - \mu_{n+1})^2$$

$$= \sum_{i=1}^n (x_i - \mu_n)^2 + \sum_{i=1}^n 2 \cdot (x_i - \mu_n) (\mu_n - \mu_{n+1}) + \sum_{i=1}^n (\mu_n - \mu_{n+1})^2 + (x_{n+1} - \mu_{n+1})^2$$

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$$= M2_n + \sum_{i=1}^n (\mu_n - \mu_{n+1})^2 + (x_{n+1} - \mu_{n+1})^2$$

$$= M2_n + n \cdot \frac{(x_{n+1} - \mu_n)^2}{(n+1)^2} + \frac{n^2 (x_{n+1} - \mu_n)^2}{(n+1)^2}$$

$$= M2_n + \frac{n \cdot (x_{n+1} - \mu_n)^2}{n+1}$$

$$= M2_n + (x_{n+1} - \mu_n) \cdot \frac{n}{n+1} \cdot (x_{n+1} - \mu_n)$$

$$= M2_n + (x_{n+1} - \mu_n) \cdot \frac{n x_{n+1} - n \mu_n}{n+1}$$

→ $(n+1)\mu_{n+1} = (n+1)\mu_n + x_{n+1} - \mu_n$

$$= M2_n + (x_{n+1} - \mu_n) \cdot \frac{n x_{n+1} - [(n+1)\mu_{n+1} - x_{n+1}]}{n+1}$$

$$= M2_n + (x_{n+1} - \mu_n) \cdot (x_{n+1} - \mu_{n+1})$$

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