

作业六

题目：已知有数据集 $D = \{x_i\}_{i=1}^n$, $x_i \in \mathbb{R}$, $X \sim \mathcal{N}(\mu, \sigma^2)$,

均值 $\mu \sim \mathcal{N}(\mu_0, \sigma_0^2)$, 有 $p(\mu) = \frac{1}{\sigma_0 \sqrt{2\pi}} \exp\left\{-\frac{1}{2\sigma_0^2}(\mu - \mu_0)^2\right\}$,

则似然 $p(D|\mu) = \left(\frac{1}{\sigma \sqrt{2\pi}}\right)^n \exp\left\{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2\right\}$, 根

据贝叶斯公式, 后验概率 $p(\mu|D) \propto p(\mu)p(D|\mu) \propto$

$\exp\left[-\frac{1}{2\sigma_n^2}(\mu - \mu_n)^2\right]$, 其中 $\mu_n = \frac{\sigma^2}{n\sigma_0^2 + \sigma^2} \mu_0 + \frac{\sum_i x_i \sigma_0^2}{n\sigma_0^2 + \sigma^2}$, $\frac{1}{\sigma_n^2} =$

$\frac{1}{\sigma_0^2} + \frac{n}{\sigma^2}$, 即 $p(\mu|D) = \mathcal{N}(\mu_n, \sigma_n^2)$ 。

证明： 验证 $P(x|D)$ 是关于 x 的高斯分布

$$P(x|D) = \int P(x, \mu|D) d\mu = \mathcal{N}(x|\mu_x, \sigma_x^2)$$

提示： 最终 $P(x|D) = A \frac{1}{\sqrt{\sigma^2 + \sigma_n^2} \cdot \sqrt{2\pi}} \cdot \exp\left\{-\frac{(x - \mu_n)^2}{2(\sigma^2 + \sigma_n^2)}\right\}$

$$= A \mathcal{N}(x|\mu_x, \sigma_x^2)$$

其中 A 为常数, $\mu_x = \mu_n$, $\sigma_x^2 = \sigma^2 + \sigma_n^2$

(答案仅供参考)

证明:

$$\begin{aligned}P(x|D) &= \int P(x, \mu|D) d\mu \\&= \int P(x|\mu)P(\mu|D) d\mu \\&= \int N(\mu|\sigma^2)N(\mu_n|\sigma_n^2) d\mu \\&= \int \frac{1}{\sigma\sqrt{2\pi}} \exp\left\{-\frac{1}{2\sigma^2}(x-\mu)^2\right\} \frac{1}{\sigma_n\sqrt{2\pi}} \exp\left\{-\frac{1}{2\sigma_n^2}(\mu-\mu_n)^2\right\} d\mu \\&= \frac{1}{2\pi\sigma\sigma_n} \exp\left\{-\frac{1}{2}\left(\frac{\mu_n^2}{\sigma_n^2} + \frac{x^2}{\sigma^2}\right)\right\} \int \exp\left\{-\frac{1}{2}\left(\frac{1}{\sigma^2} + \frac{1}{\sigma_n^2}\right)\mu^2 + \left(\frac{x}{\sigma^2} + \frac{\mu_n}{\sigma_n^2}\right)\mu\right\} d\mu \\&= \frac{1}{2\pi\sigma\sigma_n} \exp\left\{-\frac{1}{2}\left(\frac{\mu_n^2}{\sigma_n^2} + \frac{x^2}{\sigma^2} + \frac{(\sigma_n^2x + \sigma^2\mu_n)^2}{\sigma_n^2\sigma(\sigma^2 + \sigma_n^2)}\right)\right\} \\&\quad \cdot \int \exp\left\{-\frac{\sigma^2 + \sigma_n^2}{2\sigma_n^2\sigma^2}\left(\mu - \frac{\sigma^2\mu_n + \sigma_n^2x}{\sigma^2 + \sigma_n^2}\right)^2\right\} d\mu \\&= \frac{1}{2\pi\sigma\sigma_n} \exp\left\{-\frac{1}{2(\sigma^2 + \sigma_n^2)}(x - \mu_n)^2\right\} \\&\quad \cdot \int \exp\left\{-\frac{\sigma^2 + \sigma_n^2}{2\sigma_n^2\sigma^2}\left(\mu - \frac{\sigma^2\mu_n + \sigma_n^2x}{\sigma^2 + \sigma_n^2}\right)^2\right\} d\mu\end{aligned}$$

对于 $\int \exp\left\{-\frac{\sigma^2 + \sigma_n^2}{2\sigma_n^2\sigma^2}\left(\mu - \frac{\sigma^2\mu_n + \sigma_n^2x}{\sigma^2 + \sigma_n^2}\right)^2\right\} d\mu$, 由积分公式 $\int \exp\{-x^2\} dx = \sqrt{\pi}$, 用换元法, 可得:

$$\int \exp\left\{-\frac{\sigma^2 + \sigma_n^2}{2\sigma_n^2\sigma^2}\left(\mu - \frac{\sigma^2\mu_n + \sigma_n^2x}{\sigma^2 + \sigma_n^2}\right)^2\right\} d\mu = \frac{\sqrt{2\pi}\sigma_n\sigma}{\sqrt{\sigma_n^2 + \sigma^2}}$$

故原式

$$\begin{aligned}&= \frac{1}{2\pi\sigma\sigma_n} \exp\left\{-\frac{1}{2(\sigma^2 + \sigma_n^2)}(x - \mu_n)^2\right\} \cdot \frac{\sqrt{2\pi}\sigma_n\sigma}{\sqrt{\sigma_n^2 + \sigma^2}} \\&= \frac{1}{\sqrt{2\pi}\sqrt{\sigma^2 + \sigma_n^2}} \exp\left\{-\frac{1}{2(\sigma^2 + \sigma_n^2)}(x - \mu_n)^2\right\}\end{aligned}$$

所以 $P(x|D)$ 是关于 x 的高斯分布

其中 $\mu_x = \mu_n$, $\sigma_x^2 = \sigma^2 + \sigma_n^2$