

SDS 383D, Bayesian Inference in Simple Conjugate Families

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C: Normal prior and sampling distribution with unknown mean and known variance

$$p(\theta|x_1, \dots, x_n) \propto p(\theta|m, v)p(x_1, \dots, x_n|\theta) \quad (1)$$

$$p(\theta|m, v) = \left(\frac{1}{2\pi v}\right)^{1/2} \exp\left(-\frac{1}{2v}(\theta - m)^2\right) \quad (2)$$

$$p(x_i|\theta) = \left(\frac{1}{2\pi\sigma^2}\right)^{1/2} \exp\left(-\frac{1}{2\sigma^2}(x_i - \theta)^2\right) \quad (3)$$

$$p(x_1, \dots, x_n|\theta) = \prod_i^n p(x_i|\theta) = \left(\frac{1}{2\pi\sigma^2}\right)^{n/2} \exp\left(-\frac{1}{2\sigma^2} \sum_i^n (x_i - \theta)^2\right) \quad (4)$$

$$p(\theta|x_1, \dots, x_n) \propto \exp\left(-\frac{1}{2\sigma^2}(x_i - \theta)^2\right) \exp\left(-\frac{1}{2\sigma^2} \sum_i^n (x_i - \theta)^2\right) \quad (5)$$

$$\sum_i^n (x_i - \theta)^2 = \sum_i^n x_i^2 - 2\theta \sum_i^n x_i + \theta^2; \quad \bar{x} = \frac{1}{n} \sum_i^n x_i \quad (6)$$

$$p(\theta|x_1, \dots, x_n) \propto \exp\left(-\frac{n}{2\sigma^2}(-2\theta\bar{x} + \theta^2) - \frac{1}{2v}(\theta^2 - 2\theta m)\right) \quad (7)$$

$$-\frac{1}{2} \left[\left(\frac{n}{\sigma^2} + \frac{1}{v}\right) \theta^2 - 2 \left(\frac{n}{\sigma^2} \bar{x} + \frac{1}{v} m\right) \theta \right] \quad (8)$$

$$a = \frac{n}{\sigma^2} + \frac{1}{v}; \quad b = \frac{n}{\sigma^2} \bar{x} + \frac{1}{v} m \quad (9)$$

$$-\frac{a}{2} \left[\theta^2 - 2\frac{b}{a}\theta \right] = -\frac{a}{2} \left[\theta^2 - 2\frac{b}{a}\theta + \frac{b^2}{a^2} - \frac{b^2}{a^2} \right] \quad (10)$$

$$p(\theta|x_1, ..., x_n) \propto \exp \left[-\frac{a}{2} \left(\theta - \frac{b}{a} \right)^2 \right] \quad (11)$$

$$\theta|x_1, ..., x_n \sim \text{N} \left(\frac{\frac{n}{\sigma^2} \bar{x} + \frac{1}{v} m}{\frac{n}{\sigma^2} + \frac{1}{v}}, \left[\frac{n}{\sigma^2} + \frac{1}{v} \right]^{-1} \right) \quad (12)$$