사회과학자를 위한 데이터 과학 방법론: 보충, 오타, 또는 에러 R을 이용한 사회과학 자료분석

박종희

• 262쪽 평균과 분산에 대한 증명 (by courtesy of 박경태) 기대값은

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

$$\mathbb{E}(e^X) = \int_{\mathbb{R}} e^x \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) dx$$

$$= \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x^2 - 2\mu x + \mu^2 - 2\sigma^2 x)}{2\sigma^2}\right) dx$$

$$= \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{x^2 - 2(\mu + \sigma^2)x + (\mu + \sigma^2)^2 - 2\mu\sigma^2 - \sigma^4}{2\sigma^2}\right) dx$$

$$= \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x - (\mu + \sigma^2))^2}{2\sigma^2}\right) dx \times \exp\left(\mu + \frac{1}{2}\sigma^2\right)$$

$$\mathbb{E}(e^X) = \exp\left(\mu + \frac{1}{2}\sigma^2\right)$$

분산은

$$V(e^X) = \mathbb{E}((e^X)^2) - (\mathbb{E}(e^X))^2$$

$$\mathbb{E}((e^X)^2) = \int_{\mathbb{R}} e^{2x} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) dx = \exp(2\mu + 2\sigma^2)$$

$$= \exp(2\mu + 2\sigma^2) - \exp(2\mu + \sigma^2) = \exp(2\mu + \sigma^2)(\exp(\sigma^2) - 1)$$

309쪽 수식 (9.16)에서 곱셈이 덧셈으로 잘못 표시되었습니다.

$$\mathcal{N}(\mu|\mathbf{y}) \propto \mathcal{N}(\mathbf{y}|\mu, \sigma^2) \mathcal{N}(\mu|\mu_0, \sigma_0^2)$$

$$\propto \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \mu)^2\right) \times \exp\left(-\frac{1}{2\sigma_0^2} \sum_{i=1}^n (\mu - \mu_0)^2\right)$$

$$= \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \mu)^2\right) \times \exp\left(-\frac{1}{2\sigma_0^2} \sum_{i=1}^n (\mu - \mu_0)^2\right)$$

$$= \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \mu)^2\right) \times \exp\left(-\frac{1}{2\sigma_0^2} \sum_{i=1}^n (\mu - \mu_0)^2\right)$$

$$\propto \exp\left[-\frac{1}{2} \left(\frac{\mu^2}{\sigma_0^2} + \frac{\mu^2 n}{\sigma^2}\right) + \mu \left(\frac{\mu_0}{\sigma_0^2} + \frac{\sum_{i=1}^n y_i}{\sigma^2}\right)\right]$$