

STA 3180 Statistical Modelling: Bayesian Inference

1. Suppose we have a sample of size n from a population with mean μ and variance σ^2 . Use Bayes' theorem to calculate the posterior distribution of μ given that the sample mean is \bar{x} and the prior distribution of μ is $N(\mu_0, \sigma_0^2)$.

Solution: The posterior distribution of μ is given by:

$$P(\mu|\bar{x}) = P(\bar{x}|\mu)P(\mu)/P(\bar{x})$$

where $P(\bar{x}|\mu)$ is the likelihood of the sample mean \bar{x} given μ , $P(\mu)$ is the prior distribution of μ , and $P(\bar{x})$ is the marginal probability of the sample mean \bar{x} .

$$P(\bar{x}|\mu) = N(\bar{x}; \mu, \sigma^2/n)$$

$$P(\mu) = N(\mu; \mu_0, \sigma_0^2)$$

$$P(\bar{x}) = \int N(\bar{x}; \mu, \sigma^2/n)N(\mu; \mu_0, \sigma_0^2)d\mu$$

Therefore, the posterior distribution of μ is given by:

$$P(\mu|\bar{x}) = N(\bar{x}; \mu, \sigma^2/n)N(\mu; \mu_0, \sigma_0^2)/P(\bar{x})$$

2. Suppose we have a sample of size n from a population with mean μ and variance σ^2 . Use Bayes' theorem to calculate the posterior distribution of σ^2 given that the sample variance is s^2 and the prior distribution of σ^2 is $IG(\alpha, \beta)$.

Solution: The posterior distribution of σ^2 is given by:

$$P(\sigma^2|s^2) = P(s^2|\sigma^2)P(\sigma^2)/P(s^2)$$

where $P(s^2|\sigma^2)$ is the likelihood of the sample variance s^2 given σ^2 , $P(\sigma^2)$ is the prior distribution of σ^2 , and $P(s^2)$ is the marginal probability of the sample variance s^2 .

$$P(s^2|\sigma^2) = IG(s^2; \alpha + n/2, \beta + n(\bar{x} - \mu)^2/2)$$

$$P(\sigma^2) = IG(\sigma^2; \alpha, \beta)$$

$$P(s^2) = \int IG(s^2; \alpha + n/2, \beta + n(\bar{x} - \mu)^2/2)IG(\sigma^2; \alpha, \beta)d\sigma^2$$

Therefore, the posterior distribution of σ^2 is given by:

$$P(\sigma^2|s^2) = IG(s^2; \alpha + n/2, \beta + n(\bar{x} - \mu)^2/2)IG(\sigma^2; \alpha, \beta)/P(s^2)$$

3. Suppose we have a sample of size n from a population with mean μ and variance σ^2 . Use Bayes' theorem to calculate the posterior distribution of μ and σ^2 given that the sample mean is \bar{x} and the sample variance is s^2 and the prior distributions of μ and σ^2 are $N(\mu_0, \sigma_0^2)$ and $IG(\alpha, \beta)$, respectively.

Solution: The posterior distribution of μ and σ^2 is given by:

$$P(\mu, \sigma^2 | \bar{x}, s^2) = P(\bar{x}, s^2 | \mu, \sigma^2) P(\mu, \sigma^2) / P(\bar{x}, s^2)$$

where $P(\bar{x}, s^2 | \mu, \sigma^2)$ is the joint likelihood of the sample mean \bar{x} and sample variance s^2 given μ and σ^2 , $P(\mu, \sigma^2)$ is the joint prior distribution of μ and σ^2 , and $P(\bar{x}, s^2)$ is the marginal probability of the sample mean \bar{x} and sample variance s^2 .

$$P(\bar{x}, s^2 | \mu, \sigma^2) = N(\bar{x}; \mu, \sigma^2/n) IG(s^2; \alpha + n/2, \beta + n(\bar{x} - \mu)^2/2)$$

$$P(\mu, \sigma^2) = N(\mu; \mu_0, \sigma_0^2) IG(\sigma^2; \alpha, \beta)$$

$$P(\bar{x}, s^2) = \iint N(\bar{x}; \mu, \sigma^2/n) IG(s^2; \alpha + n/2, \beta + n(\bar{x} - \mu)^2/2) N(\mu; \mu_0, \sigma_0^2) IG(\sigma^2; \alpha, \beta) d\mu d\sigma^2$$

Therefore, the posterior distribution of μ and σ^2 is given by:

$$P(\mu, \sigma^2 | \bar{x}, s^2) = N(\bar{x}; \mu, \sigma^2/n) IG(s^2; \alpha + n/2, \beta + n(\bar{x} - \mu)^2/2) N(\mu; \mu_0, \sigma_0^2) IG(\sigma^2; \alpha, \beta) / P(\bar{x}, s^2)$$