

Reliability

It is good practice to report a pt. estimate (e.g. posterior mean, posterior mode), together w/ a measure of reliability e.g. CI, HPD region, posterior variance, Laplace approx).

Confidence Intervals

A Frequentist CI is a probability statement about the interval

$$p(\underbrace{l(Y)}_{\text{random}} < \theta < \underbrace{u(Y)}_{\text{random}} | \theta) = 1 - \alpha$$

A Bayesian CI is a probability statement about θ , i.e.

$$p(\underbrace{l(y)}_{\text{obs. data}} < \theta < \underbrace{u(y)}_{\text{obs. data}} | Y=y) = 1 - \alpha$$

Example: beta-binomial

$$\text{let } y = y_1, \dots, y_n$$

$$\text{Likelihood: } p(y | \theta) = f(\theta) \propto \theta^{\sum y_i} (1-\theta)^{n - \sum y_i}$$

$$\text{Prior: } p(\theta) \propto \theta^{a-1} (1-\theta)^{b-1}$$

$$\text{posterior } (\theta) \propto \text{likelihood} \cdot \text{prior}$$

$$\propto \theta^{a + \sum y_i - 1} (1-\theta)^{b + n - \sum y_i - 1}$$

$$\theta | y \sim \text{beta}(\underbrace{a + \sum y_i}_{\alpha_n}, \underbrace{b + n - \sum y_i}_{\beta_n})$$

$$95\% \text{ Bayesian CI: } \text{qbeta}(c(0.025, 0.975), \alpha_n, \beta_n)$$

HPD Region: $\int_{\theta} p(\theta|y) \cdot \mathbb{1}_{\{p(\theta|y) \geq c\}} d\theta = 1-\alpha$

Find c , then set $\{\theta\}$ s.t. $p(\theta|y) \geq c$.

Laplace Approximation

idea: fit a Gaussian to mode of posterior.

Method: Taylor expand $\log p(\theta|y)$ about $\hat{\theta}_{\text{MAP}}$.

$$\hat{\theta}_{\text{MAP}} = \underset{\theta}{\operatorname{argmax}} p(\theta|y)$$

$$\hat{\theta}_{\text{MAP}}: \frac{d}{d\theta} \log p(\theta|y) = 0 \text{ and} \\ \text{locally concave, } \frac{d^2}{d\theta^2} \log p(\theta|y) \Big|_{\hat{\theta} \pm \varepsilon} < 0$$

Define: $\log p(\theta|y) = L(\theta)$ for convenience

$$L(\theta) \approx L(\hat{\theta}) + \underbrace{L'(\hat{\theta})}_{0} (\theta - \hat{\theta}) + \frac{L''(\hat{\theta})(\theta - \hat{\theta})^2}{2}$$

$$p(\theta|y) = \exp\{L(\theta)\} = \underbrace{e^{L(\hat{\theta})} \cdot e^{\frac{1}{2} L''(\hat{\theta})(\theta - \hat{\theta})^2}}_{\text{kernel of a normal}}$$

w/ mean = $\hat{\theta}$

& variance = $-1/L''(\hat{\theta})$

$$\boxed{\theta|y \approx N(\hat{\theta}_{\text{MAP}}, -1/L''(\hat{\theta}_{\text{MAP}}))}$$

Example beta-binomial

(3)

$$p(\theta|y) = c \cdot \theta^{\alpha_n-1} (1-\theta)^{\beta_n-1}$$

$$L(\theta) = \log p(\theta|y) = \log c + (\alpha_n-1)\log\theta + (\beta_n-1)\log(1-\theta)$$

$$L'(\theta) = \frac{\alpha_n-1}{\theta} - \frac{\beta_n-1}{1-\theta}$$

$$L''(\theta) = -\frac{(\alpha_n-1)}{\theta^2} - \frac{(\beta_n-1)}{(1-\theta)^2}$$

set $L'(\theta) = 0$, then $\hat{\theta}_{\text{MAP}} = \frac{\alpha_n-1}{\beta_n+\alpha_n-2}$

$$p(\theta|y) \approx N(\hat{\theta}_{\text{MAP}}, -1/L''(\theta_{\text{MAP}}))$$