

Reliability

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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

Confidence Interval

A Frequentist CI is a probability statement about the interval

$$P(l(Y) < \theta < u(Y) | \theta) = 1 - \alpha$$

\uparrow
random

\uparrow
random

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

$$P(l(y) < \theta < u(y) | Y=y) = 1-\alpha$$

obs. data

Example: beta-binomial

Let $y = y_1 \cdots y_n$

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

$$\text{Likelihood} = P^T \cdot P^F$$

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

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

$$\propto \frac{1}{a + \sum y_i} \cdot (1 - t)$$

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

95% Bayesian CI: $qbeta(c(0.025, 0.975), x_n, \beta_n)$

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

Find c , i.e., 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 $\hat{\theta}_{MAP}$.

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

$$\hat{\theta}_{MAP}: \frac{d}{d\theta} \log p(\theta|y) = 0 \text{ and}$$

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

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

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

$$p(\theta|y) = \exp\{L(\theta)\} = e^{L(\hat{\theta})} \cdot \underbrace{e^{\frac{1}{2}L''(\hat{\theta})(\theta - \hat{\theta})^2}}$$

Kernel of a normal
w/ mean = $\hat{\theta}$

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

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

(3)

Example beta-binomial

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

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

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

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

set $L'(\theta) = 0$, then

$$\hat{\theta}_{MAP} = \frac{\alpha_n - 1}{\beta_n + \alpha_n - 2}$$

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