

1 Estimating the Precision Parameter of a Rayleigh Distribution.

- (a) As the prior on ξ follows $\mathcal{E}(\lambda)$ and the likelihood is $f(r|\xi) = \xi r \exp\left\{-\frac{\xi r^2}{2}\right\}$, we find the posterior as

$$\begin{aligned}\pi(\xi|r) &\propto f(r|\xi)\pi(\xi) = \xi r \exp\left\{-\frac{\xi r^2}{2}\right\} \cdot \lambda \exp\{-\lambda\xi\} \\ &= r\lambda\xi \exp\left\{-\xi\left(\lambda + \frac{r^2}{2}\right)\right\} \propto \xi \exp\left\{-\xi\left(\lambda + \frac{r^2}{2}\right)\right\}.\end{aligned}$$

Thus, the posterior is gamma $\mathcal{Ga}\left(2, \lambda + \frac{r^2}{2}\right)$.

- (b) We first find the posterior distribution when there are n data points. As the prior on ξ follows $\mathcal{E}(\lambda)$ and the likelihood is

$$L(\xi|r_1, \dots, r_n) = \prod_{i=1}^n f(r_i|\xi) = \left(\prod_{i=1}^n r_i\right) \xi^n \exp\left\{-\frac{\xi}{2} \sum_{i=1}^n r_i^2\right\},$$

we find the posterior as

$$\begin{aligned}\pi(\xi|r) &\propto \left(\prod_{i=1}^n r_i\right) \xi^n \exp\left\{-\frac{\xi}{2} \sum_{i=1}^n r_i^2\right\} \cdot \lambda \exp\{-\lambda\xi\} \\ &= \left(\prod_{i=1}^n r_i\right) \lambda \xi^n \exp\left\{-\xi\left(\lambda + \frac{1}{2} \sum_{i=1}^n r_i^2\right)\right\} \propto \xi^n \exp\left\{-\xi\left(\lambda + \frac{1}{2} \sum_{i=1}^n r_i^2\right)\right\}.\end{aligned}$$

Thus, the posterior is gamma $\mathcal{Ga}\left(n+1, \lambda + \frac{1}{2} \sum_{i=1}^n r_i^2\right)$. The Bayesian estimator of ξ is

$$\mathbb{E}_{\xi|r_1, \dots, r_n} \xi = \frac{n+1}{\lambda + \frac{1}{2} \sum_{i=1}^n r_i^2}$$

Based on the given data, we have $n = 4$ and $\lambda + \frac{1}{2} \sum_{i=1}^n r_i^2 = \lambda + (0.5)(3^2 + 4^2 + 2^2 + 5^2) = \lambda + 27$. The Bayesian estimator of ξ is $\frac{5}{\lambda+27}$.

- (c) For $\lambda = 1$, we know that $[\xi|r] \sim \mathcal{Ga}(5, 28)$. By Rstudio, we obtain that $[0.058, 0.366]$.

2 Estimating Chemotherapy Response Rates.

- (a) We let p be the proportion of the cancer patients that respond to the new chemotherapy treatment. We elicit a beta prior, namely $\mathcal{Beta}(\alpha, \beta)$. By setting the mean μ and

standard deviation σ of the beta prior using $\mu = 0.9, \mu - 2\sigma = 0.8$, which is $\mu = 0.9, \sigma = 0.05$, we have

$$\mu = \frac{\alpha}{\alpha + \beta} = 0.9 \text{ and } \sigma^2 = \frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)} = 0.05^2.$$

By solving the above equation, we get $\alpha = 31.5$ and $\beta = 3.5$. For each single patient, the probability for him responding the treatment with be p . Therefore, the likelihood as follows:

$$\mathcal{L}(p; \mathbf{x}) = p^{\sum_{i=1}^n x_i} (1 - p)^{n - \sum_{i=1}^n x_i}.$$

Furthermore, we can obtain that the posterior is

$$f(p|\mathbf{x}) \propto \mathcal{L}(p; \mathbf{x}) p^{\alpha-1} (1 - p)^{\beta-1} = p^{\alpha-1 + \sum_{i=1}^n x_i} (1 - p)^{\beta-1 + n - \sum_{i=1}^n x_i},$$

which shows that the posterior follows beta distribution $\mathcal{Beta}(\alpha + \sum_{i=1}^n x_i, \beta + n - \sum_{i=1}^n x_i)$.

Thus the Bayes estimate is $\frac{\alpha + \sum_{i=1}^n x_i}{\beta + n - \sum_{i=1}^n x_i + \alpha + \sum_{i=1}^n x_i} = \frac{31.5 + 22}{3.5 + 8 + 31.5 + 22} = 0.823$.

- (b) Note from (a), we have $p|\mathbf{x} \sim \mathcal{Beta}(53.5, 11.5)$. Then we have $[0.722, 0.905]$.
- (c) First we compute the p_0 and p_1 . As $p|\mathbf{x} \sim \mathcal{Beta}(53.5, 11.5)$ from (a), p_0 can be computed as 0.7054 using CDF from beta distribution. Similarly, p_1 is computed as 0.2946.

We then compute π_0 and π_1 . As $p \sim \mathcal{Beta}(31.5, 3.5)$, π_0 can be computed as 0.9585 using CDF from beta distribution. Similarly, π_1 is computed as 0.0415.

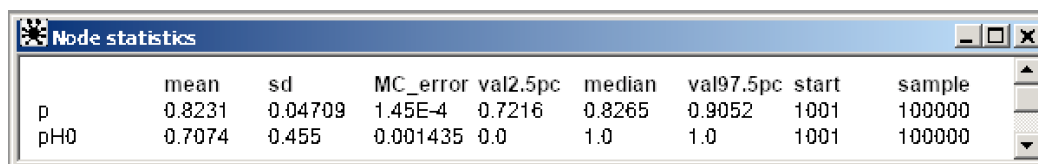
Thus, we find Bayes factor as $B_{10} = \frac{p_1/p_0}{\pi_1/\pi_0} = \frac{0.2946/0.7054}{0.0415/0.9585} = 9.6459$. This follows that $\log_{10} B_{10} = 0.9843$, which shows substantial evidence against H_0 .

- (d) We use the following OpenBUGS code to perform the computation.

```
#Model
model {
  x ~ dbin(p, M);
  p ~ dbeta(31.5, 3.5);
  pH0 <- step(p-0.8);
}

#Data
list(M=30, x=22)
```

We obtain the result, shown in Figure 1, after running the above code. Based on the result, the Bayes estimator is 0.8231 (shown as the mean value of p), the 95% equal-tailed credible set for p is $[0.7216, 0.9052]$, which are both very close the theoretical value computed from (a) and (b).



	mean	sd	MC_error	val2.5pc	median	val97.5pc	start	sample
p	0.8231	0.04709	1.45E-4	0.7216	0.8265	0.9052	1001	100000
pH0	0.7074	0.455	0.001435	0.0	1.0	1.0	1001	100000

Figure 1: Result to OpenBUGS code