## Series 2

## 1. Credible intervals

Unlike the central credible interval, the highest posterior density credible interval is not invariant to transformations as you will show in the following.

Assume that  $X_i$  i.i.d  $\sim \mathcal{N}(0, \sigma^2)$ ,  $i = 1, \ldots, n$  and that  $\sigma$  has the improper prior  $\pi(\sigma) \propto \frac{1}{\sigma}$ . Show that for the transformed parameter  $\sigma^2$ , the 95% highest posterior density (HPD) credible interval is not the same as the interval obtained when taking the square of the endpoints of the HPD credible interval for  $\sigma$ .

## 2. Conjugate priors

In the lecture, we saw the following examples of conjugate priors for exponential family distributions.

| Model                                 | Prior  | Posterior  |
|---------------------------------------|--|--|
| $\operatorname{Binomial}(n,\theta)$   | $\operatorname{Beta}(lpha,eta)$                      | $\mathrm{Beta}(\alpha+x,\beta+n-x)$  |
| Multinomial                           | $\operatorname{Dirichlet}(\alpha_1,\ldots,\alpha_k)$ | $Dirichlet(\alpha_1 + x_1, \dots, \alpha_k + x_k)$   |
| $(n,\theta_1,\ldots,\theta_k)$        |  |  |
| i.i.d. $Poisson(\theta)$              | $\operatorname{Gamma}(\gamma,\lambda)$               | $\operatorname{Gamma}(\gamma + \sum_{i} x_i, \lambda + n)$   |
| i.i.d. Normal $(\mu, \frac{1}{\tau})$ | $Normal(\mu_0, \frac{1}{n_0 \tau}) \times$           | $Normal(\frac{n}{n+n_0}\bar{x} + \frac{n_0}{n+n_0}\mu_0, \frac{1}{(n+n_0)\tau}) \times$                                      |
| $\theta = (\mu, \tau)$                | $\operatorname{Gamma}(\gamma,\lambda)$               | Gamma $(\gamma + \frac{n}{2}, \lambda + \frac{1}{2} \sum_{i} (x_i - \bar{x})^2 + \frac{nn_0}{2(n+n_0)} (\bar{x} - \mu_0)^2)$ |
| $\mathrm{Uniform}(0,\theta)$          | $\mathrm{Pareto}(lpha,\sigma)$                       | $Pareto(\alpha + n, max(\sigma, x_1, \dots, x_n))$   |

Show that one indeed obtains the posteriors in the table when using the priors and likelihoods specified in the table.

## 3. Improper priors

Consider the Poisson model

$$f(x|\theta) = P_{\theta}(X = x) = \frac{\theta^x}{x!}e^{-\theta}, \quad x \in \mathbb{N}_0, \ \theta > 0,$$

and the improper prior

$$\pi(\theta) = \frac{1}{\theta}.$$

Show that the posterior distribution is not well defined for all x.