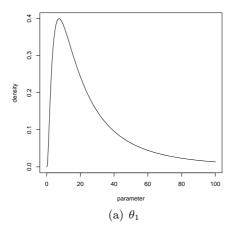
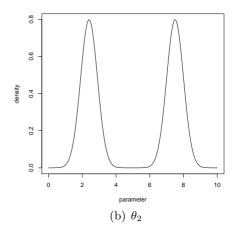
Bayesian Inference: Tutorial 3

1. (From 2009 exam - number in brackets correspond to number of marks - the total exam is out of 50 marks.) Let X_1, \ldots, X_n denote independent random variables that depend on two parameters θ_1 and θ_2 . Suppose that the posterior marginal distributions for the two parameters θ_1 and θ_2 are given in Figures 1(a) and (b) respectively.

Figure 1: Marginal posterior densities of parameters θ_1 and θ_2





Comment on the suitability of summarising these posterior distributions using the following summary statistics:

- (a) posterior mean;
- (b) posterior median;
- (c) 95% symmetric credible interval;
- (d) 95% highest posterior density interval.

[6]

Suggest another summary statistic that could be used in the summarising of the joint posterior distribution $\pi(\theta_1, \theta_2 | \mathbf{x})$, where \mathbf{x} denotes the observed data, to provide information on the posterior relationship between θ_1 and θ_2 . [1]

2. (From December 2014 exam - number in brackets correspond to number of marks - the total exam is out of 50 marks.) A chemist is interested in the maximum possible yield produced by a certain chemical process. Due to the large variability in the data, he assumes that, given a scalar θ , each yield x_i , i = 1, ..., n, is independent of the other yields and follows a uniform distribution $U(0, \theta)$, so that,

$$f(x_i|\theta) = \frac{1}{\theta}, \qquad 0 < x_i < \theta.$$

Before the chemist sees any data, he assumes a Pareto prior distribution for θ , so that,

$$p(\theta) = \begin{cases} \frac{ax_0^a}{\theta^{a+1}} & \theta \ge x_0\\ 0 & \text{otherwise,} \end{cases}$$

where a > 0 and $x_0 > 0$ are known parameters for the prior Pareto distribution. The mean of a Pareto distribution is given by, $\frac{ax_0}{a-1}$, for a > 1, whilst the median of a Pareto distribution is given by, $x_0 \times 2^{1/a}$.

- a) Calculate the posterior distribution for θ . If the posterior is of standard form, state the distributional form (including the associated parameters of the distribution). [4]
- b) Consider a Pareto prior distribution with parameters, $a = 2, x_0 = 0.1$. Consider also observed data $\mathbf{x} = \{x_1, x_2, x_3\} = \{3, 10, 17\}$. Obtain the posterior distribution and indicate how the expert's beliefs have changed after observing the data, using point summaries. Briefly discuss your findings. [2]
- c) Consider an alternative Uniform prior $p(\theta) = U(0, 10)$. Without performing any calculations, discuss the scenario where $p(\theta) = U(0, 10)$, and the observed data are $\mathbf{x} = \{3, 9, 12\}$.
- 3. We observe data $\mathbf{x} = \{x_1, \dots, x_m\}$, from a Multinomial distribution, such that,

$$X \sim MN(N, p),$$

we wish to make inference on the parameters $\mathbf{p} = \{p_1, \dots, p_m\}$. The prior on the unknown parameters \mathbf{p} , is specified to be of the form,

$$\boldsymbol{p} \sim Dir(\alpha_1, \ldots, \alpha_m).$$

What is the corresponding posterior distribution for the parameters p? What is the posterior mean of p_i , i = 1, ..., m?

Note that a list of common probability distributions is provided in appendix A of the lecture notes

4. (From May 2009 exam - number in brackets correspond to number of marks - the total was 50 marks.) Let X_1, \ldots, X_n be independent and identically distributed $N(\mu, \sigma^2)$ random variables, where σ^2 is known. We specify the prior,

$$\mu \sim TN(0, \tau^2),$$

with probability density function,

$$p(\mu) = \frac{\sqrt{2}}{\sqrt{\pi \tau^2}} \exp\left(-\frac{\mu^2}{2\tau^2}\right) \qquad 0 \le \mu < \infty.$$

- (a) Calculate the posterior distribution of μ , given observed data x_1, \ldots, x_n . [5]
- (b) For a given dataset, a practitioner obtains a posterior mean for μ of 3.5, posterior median of 2.96 and 95% highest posterior density interval of (-1.5, 8.5). Without looking at their observed data, describe how the statistician automatically realises that there has been at least one mistake made within the analysis. [2]
- (c) Discuss any disadvantages of specifying a $TN(0, \tau^2)$ prior on μ . [2]

Note that the following distributional information was also provided. A random variable X has a (positive) **Truncated-Normal distribution** $TN(\mu, \sigma^2)$, if its probability density function (p.d.f.) is

$$\frac{1}{(1-\Phi(-\mu))\sqrt{2\pi\sigma^2}}\exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) \qquad 0 \le x < \infty,$$

where Φ denotes the cumulative distribution function of the standard Normal distribution. This result follows from,

$$\frac{N(\mu,\sigma^2)}{\int_0^\infty N(\mu,\sigma^2)dx} = \frac{N(\mu,\sigma^2)}{P(X>0|\mu,\sigma)} = \frac{N(\mu,\sigma^2)}{P(\frac{X-\mu}{\sigma}>\frac{-\mu}{\sigma}|\mu,\sigma)} = \frac{N(\mu,\sigma^2)}{1-P(Z<\frac{-\mu}{\sigma})},$$

where $Z \sim N(0, 1)$.

- 5. (To discuss in groups if possible) You are presented with the following prior beliefs regarding an unknown parameter of interest $\theta \in \mathcal{R}$:
 - (a) Prior information: average of 50 and bounds of [20,80]
 - (b) Prior information: average of 50 and bounds of [25,100]

Use this prior information to propose a sensible prior distribution on θ in each scenario.