# STA623 - Bayesian Data Analysis - Practical 3

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## Practical 3

# Notation

- X, Y, Z random variables
- x, y, z measured / observed values
- $\bar{X}$ ,  $\bar{Y}$ ,  $\bar{Z}$  sample mean estimators for X, Y, Z
- $\bar{x}$ ,  $\bar{y}$ ,  $\bar{z}$  sample mean estimates of X, Y, Z
- $\hat{T}$ ,  $\hat{t}$  given a statistic T, estimator and estimate of T
- P(A) probability of an event A occurring
- $f_X(.)$ ,  $f_Y(.)$ ,  $f_Z(.)$  probability mass / density functions of X, Y, Z; sometimes  $p_X(.)$  etc. rather than  $f_X(.)$
- p(.) used as a shorthand notation for pmfs / pdfs if the use of this is unambiguous (i.e. it is clear which is the random variable)
- $X \sim F$  X distributed according to distribution function F
- E[X], E[Y], E[Z], E[T] the expectation of X, Y, Z, T respectively

## Exercise 1

Show that the Bayes estimator  $\hat{\theta}_B$  for the quadratic loss function  $C(\theta - \hat{\theta}) = (\theta - \hat{\theta})^2$  is given by the posterior mean. In other words, show that:

$$E[\theta|y] = \arg\min_{\hat{\theta}} \int_{\mathcal{Y}} \int_{\Theta} \mathcal{C}(\theta - \hat{\theta}) p(\theta, y) d\theta dy$$

# Exercise 2

Suppose  $\pi \sim \text{Beta}(2,3)$  and  $Y_1, \ldots, Y_n \sim_{\text{iid}} Bernoulli(\pi)$ . Further suppose we observe data  $y_1, \ldots, y_n$  with  $n = 25, k = \sum_i y_i = 16$ .

Find the following:

- posterior distribution  $p(\pi|k)$  and plot it, comparing it to the prior distribution
- posterior predictive distribution  $p(\tilde{y}|y_1,\ldots,y_n)$
- the 95% quantile-based Bayesian confidence interval for  $\pi$
- the 95% HPD interval

#### Further, compute:

- $P(\pi > 0.5|k)$
- For the following 2 hypotheses:  $H_1: \pi \in [0.3, 0.5], H_2: \pi \in [0.5, 0.7]$ , compute the prior and posterior odds and calculate the Bayes factor.

# Exercise 3

Suppose  $\pi \sim \text{Gamma}(5,2)$  and  $Y_1, \ldots, Y_n \sim_{\text{iid}} Poisson(\lambda)$ . Further suppose we observe data  $y_1, \ldots, y_n$  with  $n = 18, k = \sum_i y_i = 40$ .

Find the following:

- posterior distribution  $p(\lambda|y_1,\ldots,y_n)$  and plot it, comparing it to the prior distribution
- posterior predictive distribution  $p(\tilde{y}|y_1,\ldots,y_n)$
- the 95% quantile-based Bayesian confidence interval for  $\lambda$
- the 95% HPD interval

#### Further, compute:

- $P(\lambda \leq 1|y_1,\ldots,n)$
- For the following 2 hypotheses:  $H_1: \lambda \in [0.75, 1.25], H_2: \lambda \in [1.75, 2.25]$ , compute the prior and posterior odds and calculate the Bayes factor.

[end of STA623 BDA Practical 3]