

SEMESTER 2 EXAMINATION 2012/13

MATH6025 Bayesian Methods

Duration: 120 min

Answer ALL questions.

Formula Sheet FS/MATH6025/13 will be available

Only University approved calculators may be used.

A foreign language dictionary (paper version) is permitted provided it contains no notes, additions or annotations.

1. (a) Define the difference between a subjective and an objective Bayesian and give an example of what kind of prior is used in each case. Explain the concepts of a conjugate prior and a non-informative prior.

[7 marks]

Suppose that $\underline{x} = (x_1, \dots, x_n)$ are independent and identically distributed observations from an Exponential distribution with probability density function

$$f(x|\theta) = \theta e^{-\theta x}, \quad x > 0, \theta > 0.$$

- (b) Show that a $\text{Gamma}(\beta, m)$ prior, the form of which can be found in the Poisson section of the summary sheet, is conjugate for θ .

[5 marks]

- (c) Find the Jeffreys prior for θ and obtain the posterior density for $\theta|\underline{x}$ under this prior, up to a normalising constant.

[6 marks]

- (d) Find an appropriate normal approximation to the posterior distribution of $\theta|\underline{x}$.

[7 marks]

2. (a) Show that the mean of the posterior distribution is the Bayes estimator $\tilde{\theta}$ corresponding to the loss function

$$L(\theta, \tilde{\theta}) = (\theta - \tilde{\theta})^2$$

[4 marks]

Suppose that we have a single measurement $X = x_1$ from a Normal distribution $N(\mu, \sigma^2)$, so

$$f(x|\mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right)$$

Also assume that we have a Normal prior for $\mu \sim N(\theta, \tau^2)$ and that σ^2 is known.

- (b) Working to a constant of proportionality and completing the square, show that the posterior distribution is $f(\mu|x) \sim N(\mu^*, \tau^{*2})$ where

$$\mu^* = \frac{\tau^2 x + \sigma^2 \theta}{\sigma^2 + \tau^2}, \quad \tau^{*2} = \frac{\sigma^2 \tau^2}{\sigma^2 + \tau^2}$$

[7 marks]

- (c) Consider a loss function

$$L(\theta, \tilde{\theta}) = \frac{(\theta - \tilde{\theta})^2}{\theta^2}$$

Derive the Bayes estimator that minimises the posterior expectation of this loss function in terms of expectations of the form $E(g(\theta)|\underline{x})$. [4 marks]

- (d) Suppose we now have multiple measurements $X_1 = x_1, X_2 = x_2, \dots, X_n = x_n$ from the Normal distribution.

(i) Using the fact that the exponential of a quadratic form in x_{n+1} and θ is proportional to a bivariate normal distribution and the marginal distributions of a bivariate normal are normal, calculate the density of X_{n+1} with parameters $E(x_{n+1}|\underline{x})$ and $\text{Var}(x_{n+1}|\underline{x})$, given the previous n observations.

(ii) If $\mu = 50, \tau = 2, \sigma = 5$ and we have $X = (49, 48, 45, 40, 50)$ Calculate the mean and variance of X_{n+1} to 1dp.

(iii) Using the values in (ii) find the distribution of the posterior predictive sample mean of m further draws, what happens as $m \rightarrow \infty$? [10 marks]

3. Suppose that we have an industrial process, in which the numbers of failures per batch (X) follows a Poisson distribution with mean $\theta = 1$. Each batch is independent and the company wishes to assess whether this mean has changed.

A change in mean level would mean that the distribution had changed to Poisson(θ), $\theta \neq 1$.

If the mean level has changed, a Gamma(α, β) prior is assumed on θ .

$$X \sim \text{Poisson}(\theta) \quad \text{and} \quad f(\underline{x}) = \prod_{i=1}^n \frac{\theta^{x_i}}{x_i!} e^{-\theta} \quad \theta > 0, x_i \in (0, 1, 2, 3, \dots)$$

The problem of determining if a change in the mean failure rate has occurred can be thought of as choosing between the two models.

$$\begin{aligned} m_1 : X &\sim \text{Po}(\theta) \quad \theta = 1, \\ m_2 : X &\sim \text{Po}(\theta) \quad \theta \sim \text{Gamma}(\alpha, \beta). \end{aligned}$$

- (a) Define the marginal likelihood for a model m , the posterior odds, the Bayes factor and the prior odds in favour of model m_2 .

[4 marks]

- (b) Obtain expressions for the marginal likelihood for models m_1 and m_2 .

[7 marks]

- (c) Write down the posterior odds in favour of m_2 , given that $f(m = 1) = p$.

[4 marks]

- (d) Suppose that $\alpha = 2, \beta = 2, \underline{X} = (1, 1, 2)$ and that the two models are equally likely a priori. Find the posterior odds. What answer would you give the company about whether the mean has changed?

[10 marks]

4. Consider the following hierarchical model,

$$X|\theta \sim \text{Binomial}(n, \theta)$$

$$\theta \sim \text{Beta}(\alpha, \beta)$$

We know that

$$f(x|\theta) = \binom{n}{x} \theta^x (1 - \theta)^{n-x}$$

and the form of the Beta distribution is given as the prior for the Bernoulli distribution on the formula sheet.

(a) Using the above information calculate the joint distribution of $f(X, \theta)$.

[3 marks]

(b) Derive the conditional distributions of $X|\theta$ and $\theta|X$.

[3 marks]

(c) (i) Define a Gibbs sampler giving a general algorithm and discussing why the algorithm would be used.

(ii) Write an example Gibbs sampling routine which could be used to sample from the joint distribution of X and θ .

(iii) Discuss how the sampler would be initialised and how the algorithm would be analysed, include examples of diagnostic plots which should be obtained and show the main features of a Gibbs sampling algorithm, describing the features depicted.

[16 marks]

(d) Describe how the generated samples could be used to estimate $E(\theta|X)$ and $\text{Var}(\theta|X)$.

[3 marks]

END OF PAPER