

# Stochastic Optimisation

## Problem Sheet 3

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1. Let  $U$  be a random variable distributed uniformly on  $[0, 1]$ , and let  $\lambda > 0$  be a given constant. Show that the random variable  $X = (-\log U)/\lambda$  has an exponential distribution with parameter  $\lambda$ . As usual, logarithms are natural, i.e., to base  $e$ .
2. Let  $X$  be a random variable with a  $\text{Beta}(\alpha, \beta)$  distribution, and let  $Y = 1 - X$ . Show that  $Y$  has a  $\text{Beta}(\beta, \alpha)$  distribution.

*Hint.* You have a choice of two different ways to show this, both of which are pretty straightforward.

3. Let  $X$  and  $\Theta$  be independent random variables, where  $X$  has an  $\text{Exp}(\frac{1}{2})$  distribution, and  $\Theta$  is uniformly distributed on  $[0, 2\pi)$ . Let  $V = \sqrt{X} \sin \Theta$  and  $W = \sqrt{X} \cos \Theta$ . Show that  $V$  and  $W$  are independent standard normal random variables, i.e., with zero mean and unit variance.

*Hint.* Compute the joint density of  $V$  and  $W$  and show that it factorises as the product of the densities of two independent standard normal random variables.

4. Consider a Bayesian approach to the problem of inferring the mean of a normal distribution with unknown mean and known variance. More precisely, suppose  $X \sim N(\theta, 1)$ , where  $\theta$  is unknown. Fix  $\mu_0 \in \mathbb{R}$  and  $\sigma_0^2 > 0$ , and let  $\pi_0$  denote the density of a  $N(\mu_0, \sigma_0^2)$  distribution. Suppose  $\pi_0$  denotes the prior distribution of  $\theta$ . Let  $\pi_1(\cdot|x)$  denote the posterior distribution, conditional on observing a sample of  $X$  which takes the value  $x$ . In other words,

$$\pi_1(\theta|x) \propto \pi_0(\theta)f_\theta(x),$$

where  $\propto$  stands for ‘proportional to’ and  $f_\theta$  denotes the density of a  $N(\theta, 1)$  random variable. The constant of proportionality will be determined by the requirement that  $\pi_1(\cdot|x)$  be a probability density, i.e., that it integrate to 1.

Show that  $\pi_1(\cdot|x)$  is the density of a  $N(\mu_1, \sigma_1^2)$  random variable, where

$$\mu_1 = \frac{\mu_0 + x\sigma_0^2}{1 + \sigma_0^2}, \quad \sigma_1^2 = \frac{\sigma_0^2}{1 + \sigma_0^2}.$$

**Remark.** This shows that the family of normal distributions is a family of conjugate priors for the mean of a normal random variable with known variance.

*Hint.* Unfortunately, this is a calculation-heavy exercise. You can lighten the burden somewhat by ignoring constants wherever possible.

5. Use the answer to the last question to formulate a version of the Thompson sampling algorithm for a two-armed bandit, where the rewards from arm  $i$  are iid, normally distributed, with unknown mean  $\mu_i$  and unit variance. Explain your algorithm in sufficient detail to enable a non-expert to implement it. You may assume that the non-expert has access to a software package that will generate independent random variables with specified parameters, from any of a family of commonly used probability distributions.