

SMF SOLUTIONS 5. 25.5.2012

Q1. (i) With X the number of successes, and as the prior in p is uniform,

$$P(X = x|p) = \binom{n}{x} p^x (1-p)^{n-x}, \quad P(X = x) = \int_0^1 \binom{n}{x} p^x (1-p)^{n-x} dp,$$

so

$$\begin{aligned} P(a < p < b|X = x) &= \int_a^b \binom{n}{x} p^x (1-p)^{n-x} dp / \int_0^1 \binom{n}{x} p^x (1-p)^{n-x} dp \\ &= \int_a^b \binom{n}{x} p^x (1-p)^{n-x} dp / B(x+1, n-x+1). \end{aligned}$$

So the posterior is $B(x+1, n-x+1)$.

(ii) If the prior is now $B(\alpha, \beta)$, as in (i)

$$\begin{aligned} P(a < p < b|X = x) &\propto \int_a^b \binom{n}{x} p^x (1-p)^{n-x} \cdot p^{\alpha-1} (1-p)^{\beta-1} dp \\ &= \int_a^b \binom{n}{x} p^{x+\alpha-1} (1-p)^{n-x+\beta-1} dp. \end{aligned}$$

So the posterior is $B(x+\alpha, n-x+\beta)$ (observe that $U(0, 1) = B(1, 1)$, so (i) is the case $\alpha = \beta = 1$).

Q2. (i) For the Bernoulli distribution $B(p)$, $f(x; p) = p^x (1-p)^{1-x}$,

$$\ell = x \log p + (1-x) \log(1-p), \quad \ell' = \frac{x}{p} - \frac{1-x}{1-p},$$

$$\ell'' = -\frac{x}{p^2} - \frac{1-x}{(1-p)^2},$$

$$I(p) = -E[\ell''] = \frac{(1-p)}{(1-p)^2} + \frac{p}{p^2} = \frac{1}{p} + \frac{1}{1-p} = \frac{1}{p(1-p)}.$$

(ii) So the Jeffreys prior is $\pi(p) \propto \sqrt{I(p)} = 1/\sqrt{p(1-p)}$. This is the Beta distribution $B(\frac{1}{2}, \frac{1}{2})$, and $B(\frac{1}{2}, \frac{1}{2}) = \Gamma(\frac{1}{2})\Gamma(\frac{1}{2})/\Gamma(1) = \pi$, as $\Gamma(\frac{1}{2}) = \sqrt{\pi}$. So the Jeffreys prior is

$$\pi(x) = \frac{1}{\pi\sqrt{x(1-x)}} \quad (x \in [0, 1]).$$

This is the *arc-sine law* (so called because the corresponding distribution function involves an arcsine). It is important in, e.g., fluctuations in coin-tossing. See e.g.

W. FELLER, *An introduction to probability theory and its applications*, Vol. 1, 3rd ed., Wiley, 1968, Ch. III.

Q3. Recall $\Gamma(z+1) = z\Gamma(z)$. $B(\alpha, \beta)$ has mean

$$\begin{aligned} E[X] &= \int_0^1 x \cdot x^{\alpha-1}(1-x)^{\beta-1} dx / B(\alpha, \beta) = \int_0^1 x^\alpha(1-x)^{\beta-1} dx / B(\alpha, \beta) \\ &= B(\alpha+1, \beta) / B(\alpha, \beta) = \frac{\Gamma(\alpha+1)\Gamma(\beta)}{\Gamma(\alpha+\beta)} / \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)} = \alpha/(\alpha+\beta), \end{aligned}$$

Q4. So the posterior mean in Q1(ii) is $(x+\alpha)/(n+\alpha+\beta)$. As the amount of data increases, $n \rightarrow \infty$, and by SLLN $x/n \rightarrow p$ a.s., where p is the true parameter value. With no data, $x=n=0$, and the mean is the prior mean $\alpha/(\alpha+\beta)$. The value above is a compromise between these two.

Q5.

$$(f_\alpha * f_\beta)(x) = \frac{1}{\Gamma(\alpha)\Gamma(\beta)} \int_0^x y^{\alpha-1} e^{-y} \cdot (x-y)^{\beta-1} e^{-(x-y)} dy = \frac{1}{\Gamma(\alpha)\Gamma(\beta)} \cdot e^{-x} \int_0^x y^{\alpha-1} (x-y)^{\beta-1} dy.$$

In the integral, I say, substitute $y = xu$. Then $I = x^{\alpha+\beta-1} \int_0^1 u^{\alpha-1} (1-u)^{\beta-1} du = x^{\alpha+\beta-1} B(\alpha, \beta)$. Combining, the RHS has the form of $f_{\alpha+\beta}(x)$ (to within constants!):

$$(f_\alpha * f_\beta)(x) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \cdot B(\alpha, \beta) f_{\alpha+\beta}(x).$$

As both sides are densities, both integrate to 1. So the constant on the RHS is 1, which gives Euler's integral for the Beta function. NHB