## MA322: Advanced Statistical Methods - Work Sheet 2

1. Suppose that  $X_1, X_2, \ldots$ , are Bernoulli trials with success probability  $\theta$ . Let t denote the observed number of successes in n such trials. Assume the beta prior distribution with parameters  $\alpha$  and  $\beta$ . Show that

$$Pr(X_{n+1} = 1|t) = \frac{t+\alpha}{n+\alpha+\beta}.$$

## Solution:

We know that the posterior of  $\theta$  is the beta distribution:

$$\pi(\theta|t) = \frac{1}{B(t+\alpha, n-t+\beta)} \theta^{t+\alpha-1} (1-\theta)^{n-t+\beta}, 0 < \theta < 1.$$

Now

$$Pr(X_{n+1} = x|t) = \int_0^1 \theta^x (1-\theta)^{(1-x)} \frac{1}{B(t+\alpha,n-t+\beta)} \theta^{t+\alpha-1} (1-\theta)^{n-t+\beta}$$
$$= \frac{B(x+t+\alpha,1-x+n-t+\beta)}{B(t+\alpha,n-t+\beta)}.$$

When x = 1, we have

$$Pr(X_{n+1} = 1|t) = \frac{B(1+t+\alpha, n-t+\beta)}{B(t+\alpha, n-t+\beta)}$$
$$= \frac{t+\alpha}{n+\alpha+\beta}$$

by using the relation:

$$B(a,b) = \frac{\Gamma(a) \Gamma(b)}{\Gamma(a+b)}.$$

2. Suppose that the number of defects on a roll of magnetic recording tape has a Poisson distribution for which the mean  $\theta$  is unknown and that the prior distribution of  $\theta$  is a gamma distribution with parameters  $\alpha = 3$  and  $\beta = 1$ . When five rolls of this tape are selected at random and inspected, the number of defects found on the rolls are 2, 2, 6, 0, and 3. Let t denote the sum of the observations. Show that

$$Pr(X_{n+1} = x|t) = \frac{(n+\beta)^{t+\alpha}}{(n+1+\beta)^{t+\alpha+x}} \frac{\Gamma(t+\alpha+x)}{\Gamma(t+\alpha)} \frac{1}{x!}$$

Write a S-Plus function to calculate the probability. Obtain the answers for x = 0, 1 and 2.

**Solution:** The posterior here is the gamma distribution with parameters  $t + \alpha$  and  $n + \beta$ . Now

$$Pr(X_{n+1} = x|t) = \int_0^\infty e^{-\theta} \frac{\theta^x}{x!} \frac{(n+\beta)^{t+\alpha}}{\Gamma(t+\alpha)} \theta^{t+\alpha-1} e^{-\theta(n+\beta)} d\theta$$

$$= \frac{(n+\beta)^{t+\alpha}}{\Gamma(t+\alpha)} \frac{1}{x!} \int_0^\infty \theta^{x+t+\alpha-1} e^{-\theta(1+n+\beta)} d\theta$$

$$= \frac{(n+\beta)^{t+\alpha}}{\Gamma(t+\alpha)} \frac{1}{x!} \frac{\Gamma(x+t+\alpha)}{(1+n+\beta)^{(x+t+\alpha)}}.$$

3. Suppose  $X_1, \ldots, X_n$  is a random sample from the distribution with pdf  $f(x|\theta) = \theta e^{-\theta x}$ . Suppose the prior for  $\theta$  is given by  $\pi(\theta) = \beta e^{-\beta \theta}$  for some known  $\beta > 0$ . Let t denote the sum of the observations. Show that

$$f(x_{n+1}|t) = (n+1)\frac{(t+\beta)^{n+1}}{(x_{n+1}+t+\beta+)^{n+2}}.$$

Suppose that  $n = 5, t = 1.36, \beta = 1$ . Calculate the probability  $Pr(X_{n+1} < a|t)$ .

**Solution:** The posterior here is the gamma distribution with parameters n+1 and  $t+\beta$ . Now

$$f(x_{n+1}|t) = \int_0^\infty \theta e^{-\theta x_{n+1}} \frac{(t+\beta)^{n+1}}{\Gamma(n+1)} \theta^{n+1-1} e^{-\theta(t+\beta)} d\theta$$

$$= \frac{(t+\beta)^{n+1}}{\Gamma(n+1)} \int_0^\infty \theta^{n+2-1} e^{-\theta(x_{n+1}+t+\beta)} d\theta$$

$$= \frac{(t+\beta)^{n+1}}{\Gamma(n+1)} \frac{\Gamma(n+2)}{(x_{n+1}+t+\beta)^{n+2}}$$

$$= (n+1) \frac{(t+\beta)^{n+1}}{(x_{n+1}+t+\beta)^{n+2}},$$

using the relation

$$\Gamma(a+1) = a \, \Gamma(a).$$

Now

$$Pr(X_{n+1} < a|t) = \int_0^a (n+1) \frac{(t+\beta)^{n+1}}{(x_{n+1}+t+\beta)^{n+2}} dx_{n+1}$$

$$= (n+1)(t+\beta)^{n+1} \int_0^a (x_{n+1}+t+\beta)^{-(n+2)} dx_{n+1}$$

$$= (n+1)(t+\beta)^{n+1} \left[ \frac{(x_{n+1}+t+\beta)^{-(n+2)+1}}{-(n+2)+1} \right]_0^a$$

$$= (t+\beta)^{n+1} \left\{ \frac{1}{(t+\beta)^{n+1}} - \frac{1}{(a+t+\beta)^{n+1}} \right\}.$$

4. Suppose that  $X_1, \ldots, X_n$  is a random sample from  $N(\theta, \sigma^2)$  population where  $\sigma^2 = 100$ . Let the observed value of  $\bar{X}$  be 118.7 where n = 10. Assume the prior distribution N(100, 225) for  $\theta$ . Obtain the posterior predictive distribution of  $X_{n+1}$  and hence obtain the best 95% predictive interval for  $X_{n+1}$ .

**Solution:** Here the posterior distribution of  $\theta|x_1,\ldots,x_n$  is the normal distribution with

$$\mathrm{mean} = \frac{n\bar{x}/\sigma^2 + \mu/\tau^2}{n/\sigma^2 + 1/\tau^2} \text{ and Variance} = \frac{1}{n/\sigma^2 + 1/\tau^2}.$$

The predictive distribution is also normal with

mean = 
$$\frac{n\bar{x}/\sigma^2 + \mu/\tau^2}{n/\sigma^2 + 1/\tau^2}$$
 and Variance =  $\sigma^2 + \frac{1}{n/\sigma^2 + 1/\tau^2}$ .

Substituting all the numbers we have that the predictive distribution is normal with

mean = 
$$117.9$$
 and Variance =  $109.6$ .

Therfore, the 95% predictive interval is  $(117.9 \pm 1.96\sqrt{109.6})$  which is (97.38, 138.42).