

Problem 1

Answer to the problem goes here.

1. Problem 1 part 1 answer here.

As the mean rate of defection in plant A is 0.02. $\frac{\alpha_0}{\alpha_0 + \beta_0} = 0.02, \alpha_0 = 0.02 * 100 = 2, \beta_0 = 98$

I postulate beta prior for the defective rate for plant A to be $Beta(\alpha_0 = 2, \beta_0 = 98)$

As the mean rate of defection in plant B is 0.1. $\frac{\alpha_0}{\alpha_0 + \beta_0} = 0.1, \alpha_0 = 0.1 * 100 = 10, \beta_0 = 90$

I postulate beta prior for the defective rate for plant A to be $Beta(\alpha_1 = 10, \beta_1 = 90)$

2. Problem 1 part 2 answer here.

Denote $P(D|A) = 0.02, P(D|B) = 0.1$

During inspection, $P(A) = 0.6, P(B) = 0.4$

$$P(D) = P(D|A)P(A) + P(D|B)P(B) = 0.02 * 0.6 + 0.4 * 0.1 = 0.052$$

Denote number of defections during inspection follow $Bin(n, p)$

$$\pi(p|x) \propto f(x|p)\pi(p) \propto p^{x+\alpha-1}(1-p)^{n-x+\beta-1}$$

$$p|X \sim Beta(x + \alpha, n - x + \beta)$$

In plant A:

$$p|X \sim Beta(x + \alpha_0, n - x + \beta_0), \text{plugging in } n = 200, x = 15 \Rightarrow p|X \sim Beta(17, 283)$$

In plant B:

$$p|X \sim Beta(x + \alpha_1, n - x + \beta_1), \text{plugging in } n = 200, x = 15 \Rightarrow p|X \sim Beta(25, 275)$$

3. Problem 1 part 3 :

In plant A: [0.03346604, 0.08544675]

In plant B: [0.05483958, 0.11707831]

4. Problem 1 part 4:

In plant A: [0.03175030, 0.08313973]

In plant B: [0.05307317, 0.11483440]

Problem 2

$$\begin{aligned}
p(\theta_2|y) &\propto \int f(\theta_1)f(\theta_2)f(y|\theta_1, \theta_2)d\theta_1 \\
&= f(\theta_2) \int_{-\infty}^{+\infty} f(y|\theta_1, \theta_2)f(\theta_1)d\theta_1 \\
&= f(\theta_2) \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(1 - (\theta_1 + \theta_2))^2}{2}\right) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(\theta_1)^2}{2}\right) d\theta_1 \\
&= f(\theta_2) \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(1 - (\theta_1 + \theta_2))^2}{2}\right) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(\theta_1)^2}{2}\right) d\theta_1 \\
&= f(\theta_2) \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(1 - 2\theta_1 - 2\theta_2 + 2\theta_1^2 + 2\theta_1\theta_2 + \theta_2^2)}{2}\right) \frac{1}{\sqrt{2\pi}} d\theta_1 \\
&= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(1 - 2\theta_2 + \theta_2^2)}{2}\right) f(\theta_2) \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(-2\theta_1 + 2\theta_1^2 + 2\theta_1\theta_2)}{2}\right) d\theta_1 = \\
&= \frac{1}{2\pi} \exp\left(-\frac{(1 - 2\theta_2 + \theta_2^2)}{2}\right) f(\theta_2) \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi * \frac{1}{2}}} \exp\left(-\frac{(\theta_1 - \frac{(1 - \theta_2)}{2})^2}{2 * \frac{1}{2}}\right) \exp\left(\frac{(1 - \theta_2)^2}{4}\right) \sqrt{2\pi * \frac{1}{2}} d\theta_1 \\
&= f(\theta_2) \frac{1}{2\pi} \exp\left(-\frac{(1 - 2\theta_2 + \theta_2^2)}{2}\right) \exp\left(\frac{(1 - \theta_2)^2}{4}\right) \sqrt{2\pi * \frac{1}{2}} \\
&= \frac{1}{2\sqrt{\pi}} \exp\left(-\frac{(1 - \theta_2)^2}{4}\right) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(\theta_2)^2}{2}\right) \propto \exp\left(-\frac{(\theta_2 - \frac{1}{3})^2}{4/3}\right)
\end{aligned}$$

Which is the density of $N(\frac{1}{3}, \frac{2}{3})$

Similarly, $p(\theta_1|y) \propto \int f(\theta_1)f(\theta_2)f(y|\theta_1, \theta_2)d\theta_2$

Using the similar integration approach as above, the marginal distribution of θ_1 is $N(\frac{1}{3}, \frac{2}{3})$

Problem 3

$$\begin{aligned}
f(y, \theta, \tau^2) &\propto \left(\prod_{i=1}^n f(y_i | \theta) \right) \pi(\theta | \tau^2) \pi(\tau^2) \\
&= \left(\prod_{i=1}^n \frac{1}{\sqrt{2\pi}} e^{-\frac{(y_i - \theta)^2}{2}} \right) \frac{1}{\sqrt{2\pi\tau^2}} e^{-\frac{\theta^2}{2\tau^2}} \frac{1}{\Gamma(1)} (\tau^2)^{-2} e^{-\frac{1}{\tau^2}} \\
&\propto e^{-\frac{\sum_{i=1}^n (y_i - \theta)^2}{2}} (\tau^2)^{-\frac{1}{2}} e^{-\frac{\theta^2}{2\tau^2}} (\tau^2)^{-2} e^{-\frac{1}{\tau^2}}
\end{aligned}$$

Thus,

$$\pi(\theta|y, \tau^2) \propto e^{-\frac{\sum_{i=1}^n (y_i - \theta)^2}{2}} e^{-\frac{\theta^2}{2\tau^2}} = e^{-\frac{\tau^2 \sum_{i=1}^n y_i^2 - 2\theta \tau^2 \sum_{i=1}^n y_i + n\theta^2 \tau^2}{2\tau^2}} e^{-\frac{\theta^2}{2\tau^2}} \propto e^{-\frac{\theta - \frac{\tau^2 \sum_{i=1}^n y_i}{n\tau^2 + 1}}{(\frac{\tau^2}{n\tau^2 + 1})^2}}$$

Which is the kernel of Normal $N(\frac{\tau^2 \sum_{i=1}^n y_i}{n\tau^2 + 1}, \frac{\tau^2}{n\tau^2 + 1})$

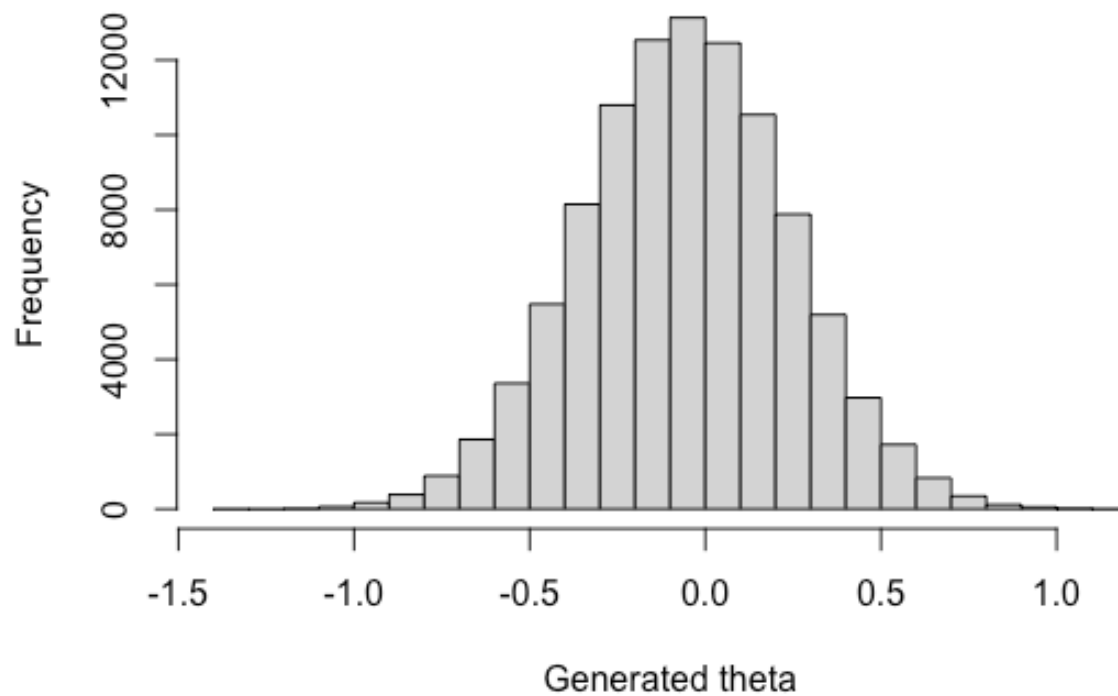
As $\sum_{i=1}^n y_i = -0.63$, $n = 10$, the conditional distribution follows Normal $N(\frac{-0.63\tau^2}{10\tau^2 + 1}, \frac{\tau^2}{10\tau^2 + 1})$

$$\pi(\tau^2|\theta, y) \propto (\tau^2)^{-\frac{1}{2}} e^{-\frac{\theta^2}{2\tau^2}} (\tau^2)^{-2} e^{-\frac{1}{\tau^2}} = (\tau^2)^{-\frac{5}{2}} e^{-\frac{\theta^2 + 2}{2\tau^2}} = (\tau^2)^{-\frac{3}{2} - 1} e^{-\frac{\theta^2 + 2}{2}} / \tau^2$$

Which is the kernel of inverse gamma $(\frac{3}{2}, \frac{\theta^2 + 2}{2})$

By implementing the Gibbs sampling (generate 100,000 samples and use 1,000 samples as burn-in), I have the results below.

Posterior density plot of theta:



Posterior mean: -0.05653624

95% equi-tailed credible interval of θ : [-0.6406627, 0.5286508]