Assignment #4

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Problem 1

Part i

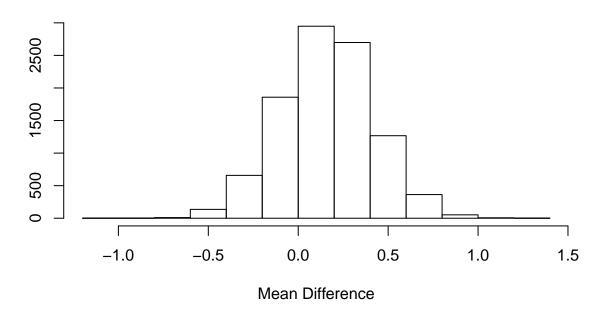
Please see the Code Appendix for the sampling technique.

Part ii

Please see the Code Appendix for the sampling technique.

Part iii

Histogram: Mean Difference



Part iv

HPD Interval: (0.7911925 , 1.476375)

Since out interval does not contain zero, we can be certain that the treatment group mean is larger than the control group mean, thus implying that the treatment does have a material affect as compared to the control.

Problem 2

Part i

Prior Distribution:

$$p(\theta) \propto exp \left[-\frac{1}{2} (\theta - \mu_{\theta})^{T} \Sigma_{\theta}^{-1} (\theta - \mu_{\theta}) \right]$$

$$\propto exp \left[-\frac{1}{2} (\theta^{T} \Sigma_{\theta}^{-1} \theta - 2\theta^{T} \Sigma_{\theta}^{-1} \mu_{\theta} + \mu_{\theta}^{T} \Sigma_{\theta}^{-1} \mu_{\theta}) \right]$$

$$\propto exp \left[-\frac{1}{2} (\theta^{T} \Sigma_{\theta}^{-1} \theta - 2\theta^{T} \Sigma_{\theta}^{-1} \mu_{\theta} + \mu_{\theta}^{T} \Sigma_{\theta}^{-1} \mu_{\theta}) \right]$$

$$\propto exp \left[-\frac{1}{2} (\theta^{T} \Sigma_{\theta}^{-1} \theta - 2\theta^{T} \Sigma_{\theta}^{-1} \mu_{\theta}) \right]$$

$$\propto exp \left[-\frac{1}{2} (\theta^{T} \Sigma_{\theta}^{-1} \theta - 2\theta^{T} \Sigma_{\theta}^{-1} \mu_{\theta}) \right]$$

Where:

$$\begin{aligned} Q_{\theta_1} &= \Sigma_{\theta}^{-1} \\ \ell_{\theta_1} &= \Sigma_{\theta}^{-1} \mu_{\theta} \end{aligned}$$

Likelihood:

$$p(y|\theta) \propto exp \left[-\frac{1}{2} \left(\sum_{i=1}^{n} y_i^T \Sigma^{-1} y_i - 2 \sum_{i=1}^{n} \theta^T \Sigma^{-1} y_i + \sum_{i=1}^{n} \theta^T \Sigma^{-1} \theta \right) \right]$$

$$\propto exp \left[-\frac{1}{2} \left(\theta^T n \Sigma^{-1} \theta - 2 \theta^T \Sigma^{-1} n \bar{y} \right) \right]$$

$$\propto exp \left[-\frac{1}{2} \left(\theta^T Q_{\theta_2} \theta - 2 \theta^T \ell_{\theta_2} \right) \right]$$

Where:

$$\begin{aligned} Q_{\theta_2} &= n \Sigma^{-1} \\ \ell_{\theta_2} &= \Sigma^{-1} n \bar{y} \end{aligned}$$

Posterior:

$$\begin{split} p(\theta|y) &\propto exp \bigg[-\frac{1}{2} \Big(\theta^T Q_{\theta_1} \theta - 2 \theta^T \ell_{\theta_1} \Big) \bigg] \times exp \bigg[-\frac{1}{2} \bigg(\theta^T Q_{\theta_2} \theta - 2 \theta^T \ell_{\theta_2} \Big) \bigg] \\ &\propto exp \bigg[-\frac{1}{2} \theta^T Q_{\theta_1} \theta + \theta^T \ell_{\theta_1} - \frac{1}{2} \theta^T Q_{\theta_2} \theta + \theta^T \ell_{\theta_2} \bigg] \\ &\propto exp \bigg[\theta^T \Big(\ell_{\theta_1} + \ell_{\theta_2} \Big) - \frac{1}{2} \theta^T \Big(Q_{\theta_1} + Q_{\theta_2} \Big) \theta \bigg] \\ &\propto exp \bigg[-\frac{1}{2} \Big(\theta^T Q_{\theta} \theta - 2 \theta^T \ell_{\theta} \Big) \bigg] \end{split}$$

Where:

$$\begin{aligned} Q_{\theta} &= Q_{\theta_1} + Q_{\theta_2} = \Sigma_{\theta}^{-1} + n\Sigma_{\theta}^{-1} \\ \ell_{\theta} &= \ell_{\theta_1} + \ell_{\theta_2} = \Sigma_{\theta}^{-1}\mu_{\theta} + \Sigma^{-1}n\bar{y} \end{aligned}$$

Thus:

$$\theta | y \sim MVN(Q_{\theta}^{-1}\ell_{\theta}, Q_{\theta}^{-1})$$

Part ii

Posterior Mean, $Q_{\theta}^{-1}\ell_{\theta}$:

$$\begin{pmatrix} \frac{1}{\sigma_{\theta_{1}}^{2}} & 0 & 0 & 0 & \cdots \\ 0 & \frac{1}{\sigma_{\theta_{2}}^{2}} & 0 & 0 & \cdots \\ 0 & 0 & \frac{1}{\sigma_{\theta_{3}}^{2}} & 0 & \cdots \\ 0 & 0 & 0 & \frac{1}{\sigma_{\theta_{4}}^{2}} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}^{-1} + \begin{pmatrix} \frac{n}{\sigma_{1}^{2}} & 0 & 0 & 0 & \cdots \\ 0 & \frac{n}{\sigma_{2}^{2}} & 0 & 0 & \cdots \\ 0 & 0 & \frac{n}{\sigma_{3}^{2}} & 0 & \cdots \\ 0 & 0 & 0 & \frac{n}{\sigma_{4}^{2}} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}^{-1} \times \begin{pmatrix} \frac{1}{\sigma_{\theta_{1}}^{2}} & 0 & 0 & 0 & \cdots \\ 0 & \frac{1}{\sigma_{\theta_{2}}^{2}} & 0 & 0 & \cdots \\ 0 & 0 & \frac{1}{\sigma_{\theta_{4}}^{2}} & \cdots \\ 0 & 0 & 0 & \frac{1}{\sigma_{\theta_{4}}^{2}} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix} \\ \mu_{\theta} + \begin{pmatrix} \frac{n\bar{y}}{\sigma_{1}^{2}} & 0 & 0 & 0 & \cdots \\ 0 & \frac{n\bar{y}}{\sigma_{2}^{2}} & 0 & 0 & \cdots \\ 0 & 0 & \frac{n\bar{y}}{\sigma_{4}^{2}} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

Posterior Covariance, Q_{θ}^{-1} :

$$\begin{pmatrix} \frac{1}{\sigma_{\theta_{1}}^{2}} & 0 & 0 & 0 & \dots \\ 0 & \frac{1}{\sigma_{\theta_{2}}^{2}} & 0 & 0 & \dots \\ 0 & 0 & \frac{1}{\sigma_{\theta_{3}}^{2}} & 0 & \dots \\ 0 & 0 & 0 & \frac{1}{\sigma_{\theta_{4}}^{2}} & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}^{-1} + \begin{pmatrix} \frac{n}{\sigma_{1}^{2}} & 0 & 0 & 0 & \dots \\ 0 & \frac{n}{\sigma_{2}^{2}} & 0 & 0 & \dots \\ 0 & 0 & \frac{n}{\sigma_{3}^{2}} & 0 & \dots \\ 0 & 0 & 0 & \frac{n}{\sigma_{4}^{2}} & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}^{-1}$$

Part iii

Posterior Mean, $Q_{\theta}^{-1}\ell_{\theta}$:

$$\begin{pmatrix} \frac{1}{\sigma_{\theta}^{2}} & 0 & 0 & 0 & \cdots \\ 0 & \frac{1}{\sigma_{\theta}^{2}} & 0 & 0 & \cdots \\ 0 & 0 & \frac{1}{\sigma_{\theta}^{2}} & 0 & \cdots \\ 0 & 0 & 0 & \frac{1}{\sigma_{\theta}^{2}} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}^{-1} + \begin{pmatrix} \frac{n}{\sigma^{2}} & 0 & 0 & 0 & \cdots \\ 0 & \frac{n}{\sigma^{2}} & 0 & 0 & \cdots \\ 0 & 0 & 0 & \frac{n}{\sigma^{2}} & 0 & \cdots \\ 0 & 0 & 0 & \frac{n}{\sigma^{2}} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}^{-1} \times \begin{pmatrix} \frac{1}{\sigma_{\theta}^{2}} & 0 & 0 & 0 & \cdots \\ 0 & \frac{1}{\sigma_{\theta}^{2}} & 0 & 0 & \cdots \\ 0 & 0 & \frac{1}{\sigma_{\theta}^{2}} & \cdots \\ 0 & 0 & 0 & \frac{1}{\sigma_{\theta}^{2}} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix} \\ \mu_{\theta} + \begin{pmatrix} \frac{n\bar{y}}{\sigma^{2}} & 0 & 0 & 0 & \cdots \\ 0 & \frac{n\bar{y}}{\sigma^{2}} & 0 & 0 & \cdots \\ 0 & 0 & \frac{n\bar{y}}{\sigma^{2}} & \cdots \\ 0 & 0 & 0 & \frac{n\bar{y}}{\sigma^{2}} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

Posterior Covariance, Q_{θ}^{-1} :

$$\begin{pmatrix} \frac{1}{\sigma_{\theta}^{2}} & 0 & 0 & 0 & \dots \\ 0 & \frac{1}{\sigma_{\theta}^{2}} & 0 & 0 & \dots \\ 0 & 0 & \frac{1}{\sigma_{\theta}^{2}} & 0 & \dots \\ 0 & 0 & 0 & \frac{1}{\sigma_{\theta}^{2}} & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}^{-1} + \begin{pmatrix} \frac{n}{\sigma^{2}} & 0 & 0 & 0 & \dots \\ 0 & \frac{n}{\sigma^{2}} & 0 & 0 & \dots \\ 0 & 0 & \frac{n}{\sigma^{2}} & 0 & \dots \\ 0 & 0 & 0 & \frac{n}{\sigma^{2}} & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}^{-1}$$

Problem 3

Solving for a in equation (b) yields:

$$La = \ell_{\theta}$$
$$a = \frac{\ell_{\theta}}{L}$$

Plugging in our result for a into equation (c) yields:

$$\begin{split} L^T \theta^* &= a + e \\ &= \frac{\ell_{\theta}}{L} + e \\ LL^T \theta^* &= \ell_{\theta} + Le \\ Q_{\theta} \theta^* &= \ell_{\theta} + Le \\ \theta^* &= \frac{\ell_{\theta} + Le}{Q_{\theta}} \\ &= Q_{\theta}^{-1}(\ell_{\theta} + Le) \\ &= Q_{\theta}^{-1}\ell_{\theta} + Q_{\theta}^{-1}Le \\ &= Q_{\theta}^{-1}\ell_{\theta} + N\left[0, (Q_{\theta}^{-1}L)^T I_d(Q_{\theta}^{-1}L)\right] \\ &= Q_{\theta}^{-1}\ell_{\theta} + N\left[0, L^T Q_{\theta}^{-1}^T Q_{\theta}^{-1}L\right] \\ &= Q_{\theta}^{-1}\ell_{\theta} + N\left[0, Q_{\theta}^{-1}\right] \\ &= Q_{\theta}^{-1}\ell_{\theta} + N\left[0, Q_{\theta}^{-1}\right] \\ &\sim N(Q_{\theta}^{-1}\ell_{\theta}, Q_{\theta}^{-1}) \end{split}$$

Thus we see that $\theta^* \sim N(Q_{\theta}^{-1}\ell_{\theta}, Q_{\theta}^{-1})$.

Code Appendix

```
## Load necessary packages
library(coda)
library(asbio)

########## Problem 1 ########

##### Part i #####

num_draws <- 10^4</pre>
```

```
mean_c <- 1.013
mean_t <- 1.173
## Draw from the scaled inverse chi-square
n_1 < -32
df <- n_1 - 1
samp_i_1 <- rinvchisq(num_draws, df, scale=1/df)</pre>
## Draw from the normal distribution
samp_i_2 <- rnorm(num_draws, mean = mean_c, sqrt(samp_i_1))</pre>
##### Part ii #####
## Draw from the scaled inverse chi-square
n_2 < -36
df <- n_2 - 1
samp_ii_1 <- rinvchisq(num_draws, df, scale=1/df)</pre>
## Draw from the normal distribution
samp_ii_2 <- rnorm(num_draws, mean = mean_t, sqrt(samp_ii_1))</pre>
##### Part iii #####
## Generate the histogram of mean differences
hist(samp_ii_2 - samp_i_2,
     xlab = "Mean Difference",
    ylab = "",
     main ="Histogram: Mean Difference")
##### Part iv #####
## Construct the HPD interval of mean differences
interval <- as.vector(HPDinterval(as.mcmc(samp_ii_2 - samp_ii_1), prob = 0.95)[1, 1:2])</pre>
## Output the result
cat("HPD Interval: ", "(", interval[1], ",", interval[2], ")")
```