Errata File for Second Edition

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This errata file contains a list of errors, short clarifications and modifications to the manuscript's second edition.

1. Page 170, bottom half of the page, bracket missing: CHANGE:

$$\operatorname{Var}(\hat{a} - a) = \operatorname{E}_{y}(\operatorname{Var}(a|y))$$

TO.

$$\operatorname{Var}(\hat{a} - a) = \operatorname{E}_{y}(\operatorname{Var}(a|y))$$

2. Page 171, the first partial derivative below Eq. (4.62). CHANGE:

$$\frac{\partial \ln p(a, b|y, b_0, a_0, B, G, R)}{\partial b} = -B^{-1}(b - b_0) + XR^{-1}(y - Xb - Za)$$

TO:

$$\frac{\partial \ln p (a, b | y, b_0, a_0, B, G, R)}{\partial b} = -B^{-1}(b - b_0) + X'R^{-1} (y - Xb - Za)$$

3. Page 172, top. CHANGE: "When B^{-1} implying $B = \infty$..." TO: "When and if $B^{-1} = 0$ implying $B = \infty$..."

4. Page 172,

CHANGE:

Finally, add and subtract $\hat{\theta}'(W'R^{-1}W + \Sigma)\hat{\theta}$ and rearrange to yield

$$y'R^{-1}y + \left(\theta - \widehat{\theta}\right)' \left(W'R^{-1}W + \Sigma\right) \left(\theta - \widehat{\theta}\right)' - \widehat{\theta}' \left(W'R^{-1}W + \Sigma\right) \widehat{\theta} + \theta_0' \Sigma \theta_0$$

TO:

Finally, add and subtract $\hat{\theta}'(W'R^{-1}W + \Sigma)\hat{\theta}$ and rearrange to yield

$$y'R^{-1}y + \left(\theta - \widehat{\theta}\right)' \left(W'R^{-1}W + \Sigma\right) \left(\theta - \widehat{\theta}\right) - \widehat{\theta}' \left(W'R^{-1}W + \Sigma\right) \widehat{\theta} + \theta_0' \Sigma \theta_0$$

5. Page 172, equation at the bottom, CHANGE:

$$\ln p\left(\theta|y,\theta_0,B,G,R\right) \propto -\frac{1}{2} \left(\theta - \widehat{\theta}\right)' \left(W'R^{-1}W + \Sigma\right) \left(\theta - \widehat{\theta}\right)'$$

TO:

$$\ln p\left(\theta|y,\theta_0,B,G,R\right) \propto -\frac{1}{2} \left(\theta - \widehat{\theta}\right)' \left(W'R^{-1}W + \Sigma\right) \left(\theta - \widehat{\theta}\right)$$

6. Page 353, Eq. (7.41), remove "comma" before " $i=1,\ldots,m$ " CHANGE:

$$y|\mu, \alpha, \delta, \sigma^2 \sim N\left(1\mu + \sum_{i=1}^m X_i\left(\alpha_i\delta_i\right), I\sigma^2\right), \quad , i = 1, \dots, m$$
 (7.41)

TO:

$$y|\mu,\alpha,\delta,\sigma^2 \sim N\left(1\mu + \sum_{i=1}^m X_i\left(\alpha_i\delta_i\right),I\sigma^2\right), \quad i=1,\ldots,m$$
 (7.41)

7. Page 373, paragraph above Eq. (8.23), defines the number of false discoveries V(A), given that k hypotheses are rejected, as binomially distributed $Bi(k, \Pr(H=0|z\in A))$. Another definition can be found on page 376, paragraph above Eq. (8.33), concerning the expected number of false discoveries. Both may benefit from a little explanation that is provided below.

Background

Let the binary random variable $V_i = 1$ if hypothesis i is rejected and if hypothesis i is a true null; $V_i = 0$, otherwise. Using the notation on page 363, we can write $V_i = I(H_i = 0, R_i = 1)$, where $R_i = 1$ if $z_i \in A$ and z_i is the observed value of the scalar random variable Z_i . Treating the hypotheses as random, it follows that V_i is a Bernoulli random variable with probability $\Pr(H_i = 0, Z_i \in A)$; that is $V_i \sim Br(\Pr(H_i = 0, Z_i \in A))$.

$$V_i \sim Br(\Pr(H_i = 0, Z_i \in A).$$

• Page 373. Consider the binary random variable $V_i|R_i=1$ that takes the value 1 if $H_i=0$ and 0 otherwise. Then $[V_i|R_i=1] \sim Br(\Pr(H_i=0|Z_i \in A))$, a Bernoulli distributed random variable. If in a series of independent trials the total number of rejections is R=k, among these k rejections, the total number of false discoveries is binomially distributed:

$$[V|R=k] \sim Bi(k, \Pr(H=0|Z \in A).$$

The expected number of false discoveries among the k rejections is

$$E(V|R = k) = k \Pr(H = 0|Z \in A).$$

The subscript i has been omitted because $\Pr(H_i = 0 | Z_i \in A)$ is the same for each i.

• Page 376. Consider an experiment where m independent hypotheses are tested. Given the definition of V_i and the independence assumption, the total number of false discoveries, $V = \sum_{i=1}^{m} V_i$, is then binomially distributed

$$V \sim Bi(m, \Pr(H = 0, Z \in A)),$$

with expected value

$$E(V) = m \Pr(H = 0, Z \in A) = m \Pr(Z \in A | H = 0) \pi_0$$

where $\pi_0 = \Pr(H = 0)$, the prior probability of a true null.

Note that: (i) hypotheses H_i are treated as random (with a common prior π_0), (ii) random vectors (H_i, Z_i) , i = 1, 2, ..., m, are independently distributed, (iii) the rejection region A is data-independent and fixed in advance.