

Bayesian Time-Series Econometrics

Book 2 - algebraic derivations

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First edition

Bayesian Time-Series Econometrics

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Cover illustration: Thomas Bayes (d. 1761) in Terence O'Donnell, *History of Life Insurance in Its Formative Years* (Chicago: American Conservation Co., 1936), p. 335.

To my wife, Mélanie.

To my sons, Tristan and Arnaud.

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PART I

Bayesian statistics

Three applied examples

derivations for equation (1.3.3)

$$f(y|p) = \prod_{i=1}^n p^{y_i} (1-p)^{1-y_i} = p^{\sum_{i=1}^n y_i} (1-p)^{\sum_{i=1}^n 1-y_i} = p^m (1-p)^{n-m} \quad (\text{a.1.3.1})$$

derivations for equation (1.3.5)

The derivative is given by:

$$\frac{d \log(f(y|p))}{dp} = \frac{m}{p} - \frac{n-m}{1-p} \quad (\text{a.1.3.2})$$

Set the value to 0 and solve for p :

$$\begin{aligned} \frac{m}{p} - \frac{n-m}{1-p} &= 0 \\ \Leftrightarrow \frac{m}{p} &= \frac{n-m}{1-p} \\ \Leftrightarrow m(1-p) &= p(n-m) \\ \Leftrightarrow m - mp &= np - mp \\ \Leftrightarrow m &= np \\ \Leftrightarrow p &= \frac{m}{n} \end{aligned} \quad (\text{a.1.3.3})$$

derivations for equation (1.3.11)

$$f(y|p) = \prod_{i=1}^n \frac{\lambda^{y_i} e^{-\lambda}}{y_i!} = \frac{\prod_{i=1}^n \lambda^{y_i} \prod_{i=1}^n e^{-\lambda}}{\prod_{i=1}^n y_i!} = \frac{\lambda^{\sum_{i=1}^n y_i} e^{-n\lambda}}{\prod_{i=1}^n y_i!} \quad (\text{a.1.3.4})$$

derivations for equation (1.3.13)

The derivative is given by:

$$\frac{d \log(f(y|\lambda))}{d\lambda} = \frac{\sum_{i=1}^n y_i}{\lambda} - n \quad (\text{a.1.3.5})$$

Set the value to 0 and solve for λ :

$$\begin{aligned} \frac{\sum_{i=1}^n y_i}{\lambda} - n &= 0 \\ \Leftrightarrow \frac{\sum_{i=1}^n y_i}{\lambda} &= n \\ \Leftrightarrow \lambda &= \frac{1}{n} \sum_{i=1}^n y_i \end{aligned} \quad (\text{a.1.3.6})$$

derivations for equation (1.3.16)

$$\begin{aligned}
& \lambda^{\sum_{i=1}^n y_i} e^{-n\lambda} \times \lambda^{a-1} e^{-\lambda/b} \\
&= \lambda^{a+\sum_{i=1}^n y_i-1} e^{-\lambda(n+1/b)} \\
&= \lambda^{a+\sum_{i=1}^n y_i-1} e^{-\lambda/(n+1/b)^{-1}}
\end{aligned} \tag{a.1.3.7}$$

Now:

$$(n+1/b)^{-1} = \frac{1}{n+1/b} = \frac{b}{bn+1} \tag{a.1.3.8}$$

Hence:

$$\begin{aligned}
& \lambda^{\sum_{i=1}^n y_i} e^{-n\lambda} \times \lambda^{a-1} e^{-\lambda/b} \\
&= \lambda^{a+\sum_{i=1}^n y_i-1} e^{-\lambda/\frac{b}{bn+1}}
\end{aligned} \tag{a.1.3.9}$$

derivations for equation (1.3.19)

$$\begin{aligned}
& f(y|\mu) \\
&= \prod_{i=1}^n (2\pi\sigma)^{-1/2} \exp\left(-\frac{1}{2} \frac{(y_i - \mu)^2}{\sigma}\right) \\
&= \prod_{i=1}^n (2\pi\sigma)^{-1/2} \prod_{i=1}^n \exp\left(-\frac{1}{2} \frac{(y_i - \mu)^2}{\sigma}\right) \\
&= (2\pi\sigma)^{-n/2} \exp\left(-\frac{1}{2} \sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right)
\end{aligned} \tag{a.1.3.10}$$

derivations for equation (1.3.21)

The derivative is given by:

$$\frac{d \log(f(y|\mu))}{d\mu} = \sum_{i=1}^n \frac{(y_i - \mu)}{\sigma} \tag{a.1.3.11}$$

Set the value to 0 and solve for μ :

$$\begin{aligned}
& \sum_{i=1}^n \frac{(y_i - \mu)}{\sigma} = 0 \\
& \Leftrightarrow \sum_{i=1}^n (y_i - \mu) = 0 \\
& \Leftrightarrow \sum_{i=1}^n y_i - n\mu = 0 \\
& \Leftrightarrow \sum_{i=1}^n y_i = n\mu \\
& \Leftrightarrow \mu = \frac{1}{n} \sum_{i=1}^n y_i
\end{aligned} \tag{a.1.3.12}$$

derivations for equation (1.3.24)

First group the exponential terms:

$$\pi(\mu|y) \propto \exp\left(-\frac{1}{2} \sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \times \exp\left(-\frac{1}{2} \frac{(\mu - m)^2}{\nu}\right) = \exp\left(-\frac{1}{2} \left[\sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma} + \frac{(\mu - m)^2}{\nu} \right]\right) \quad (\text{a.1.3.13})$$

Develop the term within the square bracket:

$$\begin{aligned} & \sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma} + \frac{(\mu - m)^2}{\nu} \\ &= \frac{1}{\sigma} \sum_{i=1}^n (y_i^2 + \mu^2 - 2\mu y_i) + \frac{1}{\nu} (\mu^2 + m^2 - 2\mu m) \\ &= \frac{1}{\sigma} \left(\sum_{i=1}^n y_i^2 + n\mu^2 - 2\mu \sum_{i=1}^n y_i \right) + \frac{1}{\nu} (\mu^2 + m^2 - 2\mu m) \end{aligned} \quad (\text{a.1.3.14})$$

Group the terms:

$$= \mu^2 \left(\frac{n}{\sigma} + \frac{1}{\nu} \right) - 2\mu \left(\frac{1}{\sigma} \sum_{i=1}^n y_i + \frac{m}{\nu} \right) + \frac{1}{\sigma} \sum_{i=1}^n y_i^2 + \frac{m^2}{\nu} \quad (\text{a.1.3.15})$$

Set back in (a.1.3.13):

$$\pi(\mu|y) \propto \exp\left(-\frac{1}{2} \left[\mu^2 \left(\frac{n}{\sigma} + \frac{1}{\nu} \right) - 2\mu \left(\frac{1}{\sigma} \sum_{i=1}^n y_i + \frac{m}{\nu} \right) + \frac{1}{\sigma} \sum_{i=1}^n y_i^2 + \frac{m^2}{\nu} \right]\right) \quad (\text{a.1.3.16})$$

Further aspects of Bayesian priors and posteriors

derivations for equation (1.4.5)

First group the terms:

$$\begin{aligned}
 & \pi(\mu, \sigma|y) \\
 \propto & \sigma^{-n/2} \exp\left(-\frac{1}{2} \sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \times \sigma^{-1/2} \exp\left(-\frac{1}{2} \frac{(\mu - m)^2}{v\sigma}\right) \times \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right) \\
 = & \sigma^{-(n+\alpha)/2-1} \times \sigma^{-1/2} \times \exp\left(-\frac{1}{2\sigma} \left[\sum_{i=1}^n (y_i - \mu)^2 + \frac{(\mu - m)^2}{v} + \delta \right]\right)
 \end{aligned} \tag{a.1.4.1}$$

Develop the term in the square bracket:

$$\begin{aligned}
 & \sum_{i=1}^n (y_i - \mu)^2 + \frac{(\mu - m)^2}{v} + \delta \\
 = & \sum_{i=1}^n (y_i^2 + \mu^2 - 2\mu y_i) + \frac{\mu^2}{v} + \frac{m^2}{v} - 2\mu \frac{m}{v} + \delta \\
 = & \sum_{i=1}^n y_i^2 + n\mu^2 - 2\mu \sum_{i=1}^n y_i + \frac{\mu^2}{v} + \frac{m^2}{v} - 2\mu \frac{m}{v} + \delta \\
 = & \mu^2 \left(n + \frac{1}{v}\right) - 2\mu \left(\sum_{i=1}^n y_i + \frac{m}{v}\right) + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} + \delta
 \end{aligned} \tag{a.1.4.2}$$

Complete the squares:

$$= \mu^2 \left(n + \frac{1}{v}\right) - 2\mu \frac{\bar{v}}{\bar{v}} \left(\sum_{i=1}^n y_i + \frac{m}{v}\right) + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} + \delta + \frac{\bar{m}^2}{\bar{v}} - \frac{\bar{m}^2}{\bar{v}} \tag{a.1.4.3}$$

Define:

$$\bar{v} = \left(n + \frac{1}{v}\right)^{-1} \quad \bar{m} = \bar{v} \left(\sum_{i=1}^n y_i + \frac{m}{v}\right) \tag{a.1.4.4}$$

Then (a.1.4.3) rewrites:

$$\begin{aligned}
 = & \frac{\mu^2}{\bar{v}} + \frac{\bar{m}^2}{\bar{v}} - 2\mu \frac{\bar{m}}{\bar{v}} + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} + \delta - \frac{\bar{m}^2}{\bar{v}} \\
 = & \frac{(\mu - \bar{m})^2}{\bar{v}} + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} + \delta - \frac{\bar{m}^2}{\bar{v}}
 \end{aligned} \tag{a.1.4.5}$$

Substituting back (a.1.4.5) in (a.1.4.1) eventually yields:

$$\begin{aligned}
& \pi(\mu, \sigma | y) \\
& \propto \sigma^{-(n+\alpha)/2-1} \times \sigma^{-1/2} \times \exp \left(-\frac{1}{2\sigma} \left[\frac{(\mu - \bar{m})^2}{\bar{v}} + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} + \delta - \frac{\bar{m}^2}{\bar{v}} \right] \right) \\
& = \sigma^{-(n+\alpha)/2-1} \times \sigma^{-1/2} \times \exp \left(-\frac{1}{2} \frac{(\mu - \bar{m})^2}{\sigma \bar{v}} \right) \times \exp \left(-\frac{1}{2\sigma} \left[\sum_{i=1}^n y_i^2 + \frac{m^2}{v} + \delta - \frac{\bar{m}^2}{\bar{v}} \right] \right) \\
& = \sigma^{-\bar{\alpha}/2-1} \times \sigma^{-1/2} \times \exp \left(-\frac{1}{2} \frac{(\mu - \bar{m})^2}{\sigma \bar{v}} \right) \times \exp \left(-\frac{\bar{\delta}}{2\sigma} \right) \tag{a.1.4.6}
\end{aligned}$$

with:

$$\bar{\alpha} = n + \alpha \quad \bar{\delta} = \sum_{i=1}^n y_i^2 + \frac{m^2}{v} + \delta - \frac{\bar{m}^2}{\bar{v}} \tag{a.1.4.7}$$

derivations for equation (1.4.10)

Rearrange the terms:

$$\begin{aligned}
& \pi(\mu | y) \\
& \propto \Gamma \left(\frac{\bar{\alpha} + 1}{2} \right) \left(\frac{\bar{\delta} + (\mu - \bar{m})^2 / \bar{v}}{2} \right)^{-\frac{\bar{\alpha}+1}{2}} \\
& \propto \left(\frac{\bar{\delta} + (\mu - \bar{m})^2 / \bar{v}}{2} \right)^{-\frac{\bar{\alpha}+1}{2}} \\
& \propto \left(\bar{\delta} + \frac{(\mu - \bar{m})^2}{\bar{v}} \right)^{-\frac{\bar{\alpha}+1}{2}} \\
& = \bar{\delta} \left(1 + \frac{(\mu - \bar{m})^2}{\bar{\delta} \bar{v}} \right)^{-\frac{\bar{\alpha}+1}{2}} \\
& \propto \left(1 + \frac{(\mu - \bar{m})^2}{\bar{\delta} \bar{v}} \right)^{-\frac{\bar{\alpha}+1}{2}} \\
& = \left(1 + \frac{1}{\bar{\alpha}} \frac{(\mu - \bar{m})^2}{\bar{\delta} \bar{v} / \bar{\alpha}} \right)^{-\frac{\bar{\alpha}+1}{2}} \tag{a.1.4.8}
\end{aligned}$$

derivations for equation (1.4.13)

Solve for the derivative:

$$\begin{aligned}
& 2 \int (\hat{\theta} - \theta) \pi(\theta | y) d\theta = 0 \\
& \Leftrightarrow \int (\hat{\theta} - \theta) \pi(\theta | y) d\theta = 0 \\
& \Leftrightarrow \int \hat{\theta} \pi(\theta | y) d\theta - \int \theta \pi(\theta | y) d\theta = 0 \\
& \Leftrightarrow \hat{\theta} \int \pi(\theta | y) d\theta = \int \theta \pi(\theta | y) d\theta \\
& \Leftrightarrow \hat{\theta} = \int \theta \pi(\theta | y) d\theta \tag{a.1.4.9}
\end{aligned}$$

derivations for equation (1.4.16)

Rearrange the expression:

$$\begin{aligned}
 f(y) &= \int \int (2\pi)^{-n/2} (2\pi)^{-1/2} v^{-1/2} \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \\
 &\quad \times \sigma^{-n/2} \exp\left(-\frac{1}{2} \sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \times \sigma^{-1/2} \exp\left(-\frac{1}{2} \frac{(\mu - m)^2}{v\sigma}\right) \times \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right) d\mu d\sigma
 \end{aligned} \tag{a.1.4.10}$$

The second row can be recognised as equation (a.1.4.1). Using the same manipulations, one obtains equation (a.1.4.6), and thus the previous expression rewrites as:

$$\begin{aligned}
 f(y) &= \int \int (2\pi)^{-n/2} (2\pi)^{-1/2} v^{-1/2} \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \\
 &\quad \times \sigma^{-1/2} \exp\left(-\frac{1}{2} \frac{(\mu - \bar{m})^2}{\sigma \bar{v}}\right) \times \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right) d\mu d\sigma
 \end{aligned} \tag{a.1.4.11}$$

with $\bar{m}, \bar{v}, \bar{\alpha}$ and $\bar{\delta}$ defined as in (a.1.4.4) and (a.1.4.7). Now add multiplicative terms to obtain normal and inverse Gamma probability density functions, and take constants out of the integral:

$$\begin{aligned}
 f(y) &= (2\pi)^{-n/2} v^{-1/2} \bar{v}^{1/2} \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \frac{\Gamma(\bar{\alpha}/2)}{\bar{\delta}/2^{\bar{\alpha}/2}} \\
 &\quad \times \int \int (2\pi \bar{v} \sigma)^{-1/2} \exp\left(-\frac{1}{2} \frac{(\mu - \bar{m})^2}{\sigma \bar{v}}\right) \times \frac{\bar{\delta}/2^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)} \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right) d\mu d\sigma
 \end{aligned} \tag{a.1.4.12}$$

The expression can simplify further. Consider only the constant on the first line:

$$\begin{aligned}
 &(2\pi)^{-n/2} v^{-1/2} \bar{v}^{1/2} \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \frac{\Gamma(\bar{\alpha}/2)}{\bar{\delta}/2^{\bar{\alpha}/2}} \\
 &= 2^{-n/2} \pi^{-n/2} v^{-1/2} ((n+1/v)^{-1})^{1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\bar{\alpha}/2}} \frac{2^{\bar{\alpha}/2}}{2^{\alpha/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \\
 &= 2^{-n/2} \pi^{-n/2} v^{-1/2} (n+1/v)^{-1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\bar{\alpha}/2}} \frac{2^{(\alpha+n)/2}}{2^{\alpha/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \\
 &= \pi^{-n/2} (1+vn)^{-1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\bar{\alpha}/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)}
 \end{aligned} \tag{a.1.4.13}$$

Substitute back in (a.1.4.12):

$$\begin{aligned}
 f(y) &= \pi^{-n/2} (1+vn)^{-1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\bar{\alpha}/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \\
 &\quad \times \int \int (2\pi \bar{v} \sigma)^{-1/2} \exp\left(-\frac{1}{2} \frac{(\mu - \bar{m})^2}{\sigma \bar{v}}\right) \times \frac{\bar{\delta}/2^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)} \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right) d\mu d\sigma
 \end{aligned} \tag{a.1.4.14}$$

derivations for equation (1.4.19)

Rearrange the expression:

$$\begin{aligned}
\mathbb{P}(M_i|y) &= \frac{f(y|M_i) \mathbb{P}(M_i)}{f(y)} \\
\Leftrightarrow \mathbb{P}(M_i|y) &= \frac{f(y, M_i) / \pi(M_i) \mathbb{P}(M_i)}{f(y)} \\
\Leftrightarrow \mathbb{P}(M_i|y) &= \frac{\int f(y, M_i, \theta_i) / \pi(M_i) d\theta_i \mathbb{P}(M_i)}{f(y)} \\
\Leftrightarrow \mathbb{P}(M_i|y) &= \frac{\int \frac{f(y, M_i, \theta_i)}{\pi(M_i, \theta)} \frac{\pi(M_i, \theta)}{\pi(M_i)} d\theta_i \mathbb{P}(M_i)}{f(y)} \\
\Leftrightarrow \mathbb{P}(M_i|y) &= \frac{\int f(y|M_i, \theta_i) \pi(\theta_i|M_i) d\theta_i \mathbb{P}(M_i)}{f(y)} \tag{a.1.4.15}
\end{aligned}$$

derivations for equation (1.4.24)

Rearrange the expression to obtain:

$$\begin{aligned}
& f(\hat{y}|y) \\
&= \int \int \sigma^{-1/2} \exp\left(-\frac{1}{2} \frac{(\hat{y} - \mu)^2}{\sigma}\right) \times \sigma^{-n/2} \exp\left(-\frac{1}{2} \sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \\
&\quad \times \sigma^{-1/2} \exp\left(-\frac{1}{2} \frac{(\mu - m)^2}{\nu \sigma}\right) \times \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right) d\mu d\sigma \\
&= \int \int \sigma^{-1/2} \exp\left(-\frac{1}{2\sigma} \left[(\hat{y} - \mu)^2 + \sum_{i=1}^n (y_i - \mu)^2 + \frac{(\mu - m)^2}{\nu} + \delta \right]\right) \sigma^{-(\alpha+n+1)/2-1} d\mu d\sigma \\
&= \int \int \sigma^{-1/2} \exp\left(-\frac{1}{2\sigma} \left[(\hat{y} - \mu)^2 + \sum_{i=1}^n (y_i - \mu)^2 + \frac{(\mu - m)^2}{\nu} + \delta \right]\right) \sigma^{-\hat{\alpha}/2-1} d\mu d\sigma \tag{a.1.4.16}
\end{aligned}$$

with:

$$\hat{\alpha} = \alpha + n + 1 \tag{a.1.4.17}$$

Consider the term in square brackets:

$$\begin{aligned}
& (\hat{y} - \mu)^2 + \sum_{i=1}^n (y_i - \mu)^2 + \frac{(\mu - m)^2}{v} + \delta \\
&= \hat{y}^2 + \mu^2 - 2\hat{y}\mu + \sum_{i=1}^n (y_i^2 + \mu^2 - 2y_i\mu) + \frac{\mu^2}{v} + \frac{m^2}{v} - 2\mu\frac{m}{v} + \delta \\
&= \hat{y}^2 + \mu^2 - 2\hat{y}\mu + \sum_{i=1}^n y_i^2 + n\mu^2 - 2\mu\sum_{i=1}^n y_i + \frac{\mu^2}{v} + \frac{m^2}{v} - 2\mu\frac{m}{v} + \delta \\
&= \left(\delta + \hat{y}^2 + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} \right) + \mu^2 \left(1 + n + \frac{1}{v} \right) - 2\mu \left(\hat{y} + \sum_{i=1}^n y_i + \frac{m}{v} \right) \\
&= \left(\delta + \hat{y}^2 + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} \right) + \mu^2 \left(1 + n + \frac{1}{v} \right) - 2\mu\frac{\hat{v}}{\hat{v}} \left(\hat{y} + \sum_{i=1}^n y_i + \frac{m}{v} \right) + \frac{\hat{m}^2}{\hat{v}} - \frac{\hat{m}^2}{\hat{v}} \\
&= \left(\delta + \hat{y}^2 + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \frac{\hat{m}^2}{\hat{v}} \right) + \mu^2 \left(1 + n + \frac{1}{v} \right) - 2\mu\frac{\hat{v}}{\hat{v}} \left(\hat{y} + \sum_{i=1}^n y_i + \frac{m}{v} \right) + \frac{\hat{m}^2}{\hat{v}} \tag{a.1.4.18}
\end{aligned}$$

Define:

$$\hat{\delta} = \left(\delta + \hat{y}^2 + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \frac{\hat{m}^2}{\hat{v}} \right) \quad \hat{v} = \left(1 + n + \frac{1}{v} \right)^{-1} \quad \hat{m} = \hat{v} \left(\hat{y} + \sum_{i=1}^n y_i + \frac{m}{v} \right) \tag{a.1.4.19}$$

Then (a.1.4.17) becomes:

$$\begin{aligned}
&= \hat{\delta} + \frac{\mu^2}{\hat{v}} - 2\mu\frac{\hat{m}}{\hat{v}} + \frac{\hat{m}^2}{\hat{v}} \\
&= \hat{\delta} + \frac{(\mu - \hat{m})^2}{\hat{v}} \tag{a.1.4.20}
\end{aligned}$$

Substitute back in (a.1.4.16):

$$\begin{aligned}
& f(\hat{y}|y) \\
&= \int \int \sigma^{-1/2} \exp \left(-\frac{1}{2\sigma} \left[\hat{\delta} + \frac{(\mu - \hat{m})^2}{\hat{v}} \right] \right) \sigma^{-\hat{\alpha}/2-1} d\mu d\sigma \\
&= \int \int \sigma^{-1/2} \exp \left(-\frac{1}{2} \frac{(\mu - \hat{m})^2}{\hat{v}\sigma} \right) \times \sigma^{-\hat{\alpha}/2-1} \exp \left(-\frac{\hat{\delta}}{2\sigma} \right) d\mu d\sigma \\
&= \int \sigma^{-\hat{\alpha}/2-1} \exp \left(-\frac{\hat{\delta}}{2\sigma} \right) \int \sigma^{-1/2} \exp \left(-\frac{1}{2} \frac{(\mu - \hat{m})^2}{\hat{v}\sigma} \right) d\mu d\sigma \tag{a.1.4.21}
\end{aligned}$$

The second integral contains the kernel of a normal distribution with mean \hat{m} and variance $\hat{v}\sigma$. It thus integrates to a constant (not involving \hat{y}) and can be relegated to the normalization constant, yielding:

$$\propto \int \sigma^{-(\hat{\alpha}+1)/2-1} \exp \left(-\frac{\hat{\delta}}{2\sigma} \right) d\sigma \tag{a.1.4.22}$$

The remaining integral contains the kernel of an inverse Gamma distribution with shape $\hat{\alpha}$ and scale $\hat{\delta}$. It integrates to the reciprocal of the normalization constant of the inverse Gamma distribution (see book1, section 4.3), which does involve \hat{y} . The term must thus be retained, yielding:

$$\begin{aligned}
& f(\hat{y}|y) \\
& \propto \Gamma(\hat{\alpha})(\hat{\delta}/2)^{-\hat{\alpha}/2} \\
& \propto (\hat{\delta}/2)^{-\hat{\alpha}/2} \\
& \propto (\hat{\delta})^{-\hat{\alpha}/2} \\
& = \left(\delta + \hat{y}^2 + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \frac{\hat{m}^2}{\hat{v}} \right)^{-\hat{\alpha}/2} \\
& = \left(\delta + \hat{y}^2 + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \hat{v} \left[\hat{y} + \sum_{i=1}^n y_i + \frac{m}{v} \right]^2 \right)^{-\hat{\alpha}/2}
\end{aligned} \tag{a.1.4.23}$$

Define:

$$\tilde{m} = \sum_{i=1}^n y_i + \frac{m}{v} \tag{a.1.4.24}$$

Then (a.1.4.23) becomes:

$$\begin{aligned}
& = \left(\delta + \hat{y}^2 + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \hat{v}[\hat{y} + \tilde{m}]^2 \right)^{-\hat{\alpha}/2} \\
& = \left(\delta + \hat{y}^2 + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \hat{v}\hat{y}^2 - \hat{v}\tilde{m}^2 - 2\hat{v}\tilde{m}\hat{y} \right)^{-\hat{\alpha}/2} \\
& = \left(\delta + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \hat{v}\tilde{m}^2 + \hat{y}^2(1 - \hat{v}) - 2\hat{v}\tilde{m}\hat{y} \right)^{-\hat{\alpha}/2}
\end{aligned} \tag{a.1.4.25}$$

Complete the squares:

$$\begin{aligned}
& = \left(\delta + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \hat{v}\tilde{m}^2 + \hat{y}^2(1 - \hat{v}) - 2\hat{v}\frac{\hat{v}}{\hat{v}}\tilde{m}\hat{y} + \frac{\hat{m}^2}{\hat{v}} - \frac{\hat{m}^2}{\hat{v}} \right)^{-\hat{\alpha}/2} \\
& = \left(\left[\delta + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \hat{v}\tilde{m}^2 - \frac{\hat{m}^2}{\hat{v}} \right] + \hat{y}^2(1 - \hat{v}) - 2\hat{v}\frac{\hat{v}}{\hat{v}}\tilde{m}\hat{y} + \frac{\hat{m}^2}{\hat{v}} \right)^{-\hat{\alpha}/2}
\end{aligned} \tag{a.1.4.26}$$

Define:

$$\bar{\alpha} = \alpha + n \quad \bar{\delta} = \delta + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \hat{v}\tilde{m}^2 - \frac{\hat{m}^2}{\hat{v}} \quad \bar{v} = (1 - \hat{v})^{-1} \quad \bar{m} = \hat{v}\tilde{m} \tag{a.1.4.27}$$

Then (a.1.4.26) becomes:

$$\begin{aligned}
&= \left(\ddot{\delta} + \frac{\hat{y}^2}{\hat{v}} - 2\hat{y}\frac{\dot{m}}{\hat{v}} + \frac{\dot{m}^2}{\hat{v}} \right)^{-(\bar{\alpha}+1)/2} \\
&= \left(\ddot{\delta} + \frac{(\hat{y} - \dot{m})^2}{\hat{v}} \right)^{-(\bar{\alpha}+1)/2} \\
&= \ddot{\delta}^{-(\bar{\alpha}+1)/2} \left(1 + \frac{(\hat{y} - \dot{m})^2}{\ddot{\delta}\hat{v}} \right)^{-(\bar{\alpha}+1)/2} \\
&\propto \left(1 + \frac{(\hat{y} - \dot{m})^2}{\ddot{\delta}\hat{v}} \right)^{-(\bar{\alpha}+1)/2} \\
&= \left(1 + \frac{1}{\bar{\alpha}} \frac{(\hat{y} - \dot{m})^2}{\ddot{\delta}\hat{v}/\bar{\alpha}} \right)^{-(\bar{\alpha}+1)/2} \tag{a.1.4.28}
\end{aligned}$$

Finally, reformulate all the messy terms:

$$\begin{aligned}
\hat{v} &= (1 - \hat{v})^{-1} = \frac{1}{1 - \hat{v}} = \frac{1}{1 - \frac{1}{1+n+\frac{1}{v}}} = \frac{1}{\frac{1+n+\frac{1}{v}-1}{1+n+\frac{1}{v}}} = \frac{1}{\frac{n+\frac{1}{v}}{1+n+\frac{1}{v}}} = \frac{1+n+\frac{1}{v}}{n+\frac{1}{v}} = \frac{v+vn+1}{vn+1} \\
&= 1 + \frac{v}{vn+1} = 1 + \frac{1}{n+1/v} = 1 + \left(n + \frac{1}{v} \right)^{-1} = 1 + \bar{v} \tag{a.1.4.29}
\end{aligned}$$

with \bar{v} defined as in (a.1.4.4).

Also:

$$\frac{\hat{v}}{1 - \hat{v}} = \frac{\frac{1}{1+n+\frac{1}{v}}}{1 - \frac{1}{1+n+\frac{1}{v}}} = \frac{\frac{1}{1+n+\frac{1}{v}}}{\frac{1+n+\frac{1}{v}-1}{1+n+\frac{1}{v}}} = \frac{1}{\frac{n+\frac{1}{v}}{1+n+\frac{1}{v}}} = \frac{1}{n+\frac{1}{v}} = \left(n + \frac{1}{v} \right)^{-1} = \bar{v} \tag{a.1.4.30}$$

Then:

$$\dot{m} = \hat{v}\dot{m} = \frac{\hat{v}}{1 - \hat{v}}\dot{m} = \bar{v}\dot{m} = \bar{v} \left(\sum_{i=1}^n y_i + \frac{m}{v} \right) = \bar{m} \tag{a.1.4.31}$$

with \bar{m} defined as in (a.1.4.4).

Finally:

$$\begin{aligned}
\hat{v}\dot{m}^2 + \frac{\dot{m}^2}{\hat{v}} &= \hat{v}\dot{m}^2 + (\hat{v}\dot{m})^2/\hat{v} = \hat{v}\dot{m}^2 + \hat{v}^2\dot{m}^2 = \hat{v}\dot{m}^2(1 + \hat{v}) = \hat{v}\dot{m}^2 \left(1 + \frac{\hat{v}}{1 - \hat{v}} \right) \\
&= \hat{v}\dot{m}^2 \left(\frac{1 - \hat{v} + \hat{v}}{1 - \hat{v}} \right) = \hat{v}\dot{m}^2 \left(\frac{1}{1 - \hat{v}} \right) = \dot{m}^2 \left(\frac{\hat{v}}{1 - \hat{v}} \right) = \dot{m}^2\bar{v} = \bar{m}\dot{m} = \frac{\bar{m}}{\bar{v}}\dot{m} = \frac{\dot{m}^2}{\bar{v}} \tag{a.1.4.32}
\end{aligned}$$

Substitute in (a.1.4.27) to obtain:

$$\ddot{\delta} = \delta + \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \frac{\dot{m}^2}{\hat{v}} = \bar{\delta} \tag{a.1.4.33}$$

with $\bar{\delta}$ defined as in (a.1.4.7).

Substitute (a.1.4.29), (a.1.4.31) and (a.1.4.33) in (a.1.4.28) to eventually obtain:

$$f(\hat{y}|y) \propto \left(1 + \frac{1}{\bar{\alpha}} \frac{(\hat{y} - \bar{m})^2}{\bar{\delta}(1 + \bar{v})/\bar{\alpha}} \right)^{-(\bar{\alpha}+1)/2} \tag{a.1.4.34}$$

Properties of Bayesian estimates

derivations for equation (1.5.1)

The mean of a Beta distribution with shapes a and b is given by $\frac{a}{a+b}$. Given the posterior hyperparameters $\bar{\alpha} = \alpha + m$ and $\bar{\beta} = \beta + n - m$, the posterior mean writes as:

$$\begin{aligned}
 & \mathbb{E}(p|y) \\
 &= \frac{\bar{\alpha}}{\bar{\alpha} + \bar{\beta}} \\
 &= \frac{\alpha + m}{\alpha + m + \beta + n - m} \\
 &= \frac{\alpha + m}{\alpha + \beta + n} \\
 &= \frac{\alpha}{\alpha + \beta + n} + \frac{m}{\alpha + \beta + n} \\
 &= \frac{\alpha}{\alpha + \beta} \frac{\alpha + \beta}{\alpha + \beta + n} + \frac{m}{n} \frac{n}{\alpha + \beta + n} \\
 &= \gamma \mathbb{E}(p) + (1 - \gamma) \hat{p}
 \end{aligned} \tag{a.1.5.1}$$

with:

$$\mathbb{E}(p) = \frac{\alpha}{\alpha + \beta} \quad \hat{p} = \frac{m}{n} \quad \gamma = \frac{\alpha + \beta}{\alpha + \beta + n} \tag{a.1.5.2}$$

derivations for equation (1.5.2)

The mean of a Gamma distribution with shape a and scale b is given by ab . Given the posterior hyperparameters $\bar{a} = a + \sum_{i=1}^n y_i$ and $\bar{b} = \frac{b}{bn+1}$, the posterior mean writes as:

$$\begin{aligned}
 & \mathbb{E}(\lambda|y) \\
 &= \frac{(a + \sum_{i=1}^n y_i)b}{bn + 1} \\
 &= \frac{ab}{bn + 1} + \frac{b \sum_{i=1}^n y_i}{bn + 1} \\
 &= ab \left(\frac{1}{bn + 1} \right) + \frac{\sum_{i=1}^n y_i}{n} \left(\frac{bn}{bn + 1} \right) \\
 &= \gamma \mathbb{E}(\lambda) + (1 - \gamma) \hat{\lambda}
 \end{aligned} \tag{a.1.5.3}$$

with:

$$\mathbb{E}(\lambda) = ab \quad \hat{\lambda} = \frac{\sum_{i=1}^n y_i}{n} \quad \gamma = \frac{1}{bn + 1} \tag{a.1.5.4}$$

derivations for equation (1.5.3)

The mean of a normal distribution with mean μ and variance σ is given by μ . Given the posterior hyperparameters $\bar{v} = \left(\frac{n}{\sigma} + \frac{1}{v}\right)^{-1}$ and $\bar{m} = \bar{v}\left(\frac{1}{\sigma}\sum_{i=1}^n y_i + \frac{m}{v}\right)$, the posterior variance writes as:

$$\bar{v} = \left(\frac{n}{\sigma} + \frac{1}{v}\right)^{-1} = \frac{1}{n/\sigma + 1/v} = \frac{\sigma}{n + \sigma/v} \quad (\text{a.1.5.5})$$

Then the posterior mean can be expressed as:

$$\begin{aligned} & \mathbb{E}(\mu|y) \\ &= \frac{\sigma}{n + \sigma/v} \left(\frac{1}{\sigma} \sum_{i=1}^n y_i + \frac{m}{v} \right) \\ &= \frac{1}{n + \sigma/v} \left(\sum_{i=1}^n y_i \right) + \frac{\sigma}{n + \sigma/v} \left(\frac{m}{v} \right) \\ &= \frac{n}{n + \sigma/v} \left(\frac{\sum_{i=1}^n y_i}{n} \right) + \frac{\sigma/v}{n + \sigma/v} m \\ &= \frac{vn}{vn + \sigma} \left(\frac{\sum_{i=1}^n y_i}{n} \right) + \frac{\sigma}{vn + \sigma} m \\ &= \gamma \mathbb{E}(\mu) + (1 - \gamma) \hat{\mu} \end{aligned} \quad (\text{a.1.5.6})$$

with:

$$\mathbb{E}(\mu) = m \quad \hat{\mu} = \frac{\sum_{i=1}^n y_i}{n} \quad \gamma = \frac{\sigma}{vn + \sigma} \quad (\text{a.1.5.7})$$

PART II

Simulation methods

The Gibbs sampling algorithm

derivations for equation (2.6.17)

Combine all the terms to obtain:

$$\begin{aligned}
 & f(y) \\
 & \approx (2\pi\sigma)^{-n/2} \exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \frac{(2\pi v)^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - m)^2}{v}\right)}{\frac{1}{J}\sum_{j=1}^J (2\pi\bar{v})^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right)} \frac{\frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right)}{\frac{\bar{\delta}/2^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)} \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right)} \\
 & = (2\pi)^{-n/2} \sigma^{-n/2} \exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \frac{v^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - m)^2}{v}\right)}{\frac{1}{J}\sum_{j=1}^J \bar{v}^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right)} \frac{\frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right)}{\frac{\bar{\delta}/2^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)} \sigma^{-(\alpha+n)/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right)} \\
 & = (2\pi)^{-n/2} \exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \frac{v^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - m)^2}{v}\right)}{\frac{1}{J}\sum_{j=1}^J \bar{v}^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right)} \frac{\frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \exp\left(-\frac{1}{2\sigma} [\delta + \sum_{i=1}^n (y_i - \mu)^2]\right)}{\frac{\bar{\delta}/2^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right)} \\
 & = (2\pi)^{-n/2} \exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \frac{v^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - m)^2}{v}\right)}{\frac{1}{J}\sum_{j=1}^J \bar{v}^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right)} \frac{\frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)}}{\frac{\bar{\delta}/2^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)}} \\
 & = 2^{-n/2} \pi^{-n/2} \exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \frac{v^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - m)^2}{v}\right)}{\frac{1}{J}\sum_{j=1}^J \bar{v}^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right)} \frac{\frac{\delta^{\alpha/2}}{\Gamma(\alpha/2)} 2^{\alpha/2}}{\frac{\bar{\delta}^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)} 2^{\bar{\alpha}/2}} \\
 & = \pi^{-n/2} \exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\sigma}\right) \frac{v^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - m)^2}{v}\right)}{\frac{1}{J}\sum_{j=1}^J \bar{v}^{-1/2} \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right)} \frac{\frac{\delta^{\alpha/2}}{\Gamma(\alpha/2)}}{\frac{\bar{\delta}^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)}} \\
 & = \pi^{-n/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\bar{\alpha}/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \frac{\exp\left(-\frac{1}{2}\frac{(\mu - m)^2}{v}\right)}{\frac{1}{J}\sum_{j=1}^J (v/\bar{v})^{1/2} \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right)} \tag{a.2.6.1}
 \end{aligned}$$

Now:

$$v/\bar{v} = v(n/\sigma + 1/v) = vn/\sigma + 1 \tag{a.2.6.2}$$

Hence:

$$f(y) \approx \pi^{-n/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\bar{\alpha}/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \frac{\exp\left(-\frac{1}{2}\frac{(\mu - m)^2}{v}\right)}{\frac{1}{J}\sum_{j=1}^J (1 + vn/\sigma)^{1/2} \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right)} \tag{a.2.6.3}$$

The Metropolis-Hastings algorithm

derivations for equation (2.7.7)

Rearrange:

$$\begin{aligned}
 & \pi(\mu|y, \lambda) \\
 \propto & \exp\left(-\frac{1}{2} \sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda)}\right) \times \exp\left(-\frac{1}{2} \frac{(\mu - m)^2}{v}\right) \\
 = & \exp\left(-\frac{1}{2} \left[\sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda)} + \frac{(\mu - m)^2}{v} \right]\right) \tag{a.2.7.1}
 \end{aligned}$$

Develop the term within the square bracket:

$$\begin{aligned}
 & \sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda)} + \frac{(\mu - m)^2}{v} \\
 = & \frac{1}{\exp(\lambda)} \sum_{i=1}^n (y_i^2 + \mu^2 - 2\mu y_i) + \frac{1}{v} (\mu^2 + m^2 - 2\mu m) \\
 = & \frac{1}{\exp(\lambda)} \left(\sum_{i=1}^n y_i^2 + n\mu^2 - 2\mu \sum_{i=1}^n y_i \right) + \frac{1}{v} (\mu^2 + m^2 - 2\mu m) \tag{a.2.7.2}
 \end{aligned}$$

Group the terms and complete the squares:

$$\begin{aligned}
 = & \mu^2 \left(\frac{n}{\exp(\lambda)} + \frac{1}{v} \right) - 2\mu \left(\frac{1}{\exp(\lambda)} \sum_{i=1}^n y_i + \frac{m}{v} \right) + \frac{1}{\exp(\lambda)} \sum_{i=1}^n y_i^2 + \frac{m^2}{v} \\
 = & \mu^2 \left(\frac{n}{\exp(\lambda)} + \frac{1}{v} \right) - 2\mu \frac{\bar{v}}{\bar{v}} \left(\frac{1}{\exp(\lambda)} \sum_{i=1}^n y_i + \frac{m}{v} \right) + \frac{1}{\exp(\lambda)} \sum_{i=1}^n y_i^2 + \frac{m^2}{v} + \frac{\bar{m}^2}{\bar{v}} - \frac{\bar{m}^2}{\bar{v}} \tag{a.2.7.3}
 \end{aligned}$$

Define:

$$\bar{v} = \left(\frac{n}{\exp(\lambda)} + \frac{1}{v} \right)^{-1} \quad \bar{m} = \bar{v} \left(\frac{1}{\exp(\lambda)} \sum_{i=1}^n y_i + \frac{m}{v} \right) \tag{a.2.7.4}$$

Then (a.2.7.3) rewrites:

$$\begin{aligned}
 = & \frac{\mu^2}{\bar{v}} + \frac{\bar{m}^2}{\bar{v}} - 2\mu \frac{\bar{m}}{\bar{v}} + \frac{1}{\exp(\lambda)} \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \frac{\bar{m}^2}{\bar{v}} \\
 = & \frac{(\mu - \bar{m})^2}{\bar{v}} + \frac{1}{\exp(\lambda)} \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \frac{\bar{m}^2}{\bar{v}} \tag{a.2.7.5}
 \end{aligned}$$

Substitute back in (a.2.7.1):

$$\begin{aligned}
& \pi(\mu|y, \lambda) \\
& \propto \exp\left(-\frac{1}{2}\left[\frac{(\mu - \bar{m})^2}{\bar{v}} + \frac{1}{\exp(\lambda)} \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \frac{\bar{m}^2}{\bar{v}}\right]\right) \\
& = \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right) \exp\left(-\frac{1}{2}\left[\frac{1}{\exp(\lambda)} \sum_{i=1}^n y_i^2 + \frac{m^2}{v} - \frac{\bar{m}^2}{\bar{v}}\right]\right) \\
& \propto \exp\left(-\frac{1}{2}\frac{(\mu - \bar{m})^2}{\bar{v}}\right) \tag{a.2.7.6}
\end{aligned}$$

derivations for equation (2.7.14)

$$\begin{aligned}
& \alpha(\lambda^{(j-1)}, \lambda^{(j)}) \\
& = \frac{\exp(\lambda^{(j)})^{-n/2}}{\exp(\lambda^{(j-1)})^{-n/2}} \frac{\exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda^{(j)})}\right)}{\exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda^{(j-1)})}\right)} \frac{\exp\left(-\frac{1}{2}\frac{(\lambda^{(j)} - g)^2}{z}\right)}{\exp\left(-\frac{1}{2}\frac{(\lambda^{(j-1)} - g)^2}{z}\right)} \tag{a.2.7.7}
\end{aligned}$$

Consider the first term:

$$\begin{aligned}
& \frac{\exp(\lambda^{(j)})^{-n/2}}{\exp(\lambda^{(j-1)})^{-n/2}} \\
& = \exp(\lambda^{(j)} - \lambda^{(j-1)})^{-n/2} \\
& = \exp\left(\frac{n}{2}(\lambda^{(j-1)} - \lambda^{(j)})\right) \tag{a.2.7.8}
\end{aligned}$$

Consider the second term:

$$\begin{aligned}
& \frac{\exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda^{(j)})}\right)}{\exp\left(-\frac{1}{2}\sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda^{(j-1)})}\right)} \\
& = \frac{\exp\left(-\frac{1}{2}\exp(-\lambda^{(j)})\sum_{i=1}^n (y_i - \mu)^2\right)}{\exp\left(-\frac{1}{2}\exp(-\lambda^{(j-1)})\sum_{i=1}^n (y_i - \mu)^2\right)} \\
& = \exp\left(-\frac{1}{2}\left[\exp(-\lambda^{(j)}) - \exp(-\lambda^{(j-1)})\right]\sum_{i=1}^n (y_i - \mu)^2\right) \\
& = \exp\left(\frac{1}{2}\left[\exp(-\lambda^{(j-1)}) - \exp(-\lambda^{(j)})\right]\sum_{i=1}^n (y_i - \mu)^2\right) \tag{a.2.7.9}
\end{aligned}$$

Consider the third term:

$$\begin{aligned}
& \frac{\exp\left(-\frac{1}{2}\frac{(\lambda^{(j)} - g)^2}{z}\right)}{\exp\left(-\frac{1}{2}\frac{(\lambda^{(j-1)} - g)^2}{z}\right)} \\
& = \exp\left(-\frac{1}{2}\left[\frac{(\lambda^{(j)} - g)^2 - (\lambda^{(j-1)} - g)^2}{z}\right]\right) \\
& = \exp\left(\frac{1}{2}\left[\frac{(\lambda^{(j-1)} - g)^2 - (\lambda^{(j)} - g)^2}{z}\right]\right) \tag{a.2.7.10}
\end{aligned}$$

Substitute back in (a.2.7.7):

$$\begin{aligned} & \alpha(\lambda^{(j-1)}, \lambda^{(j)}) \\ = & \exp \left(\frac{1}{2} \left[\frac{n(\lambda^{(j-1)} - \lambda^{(j)}) + [\exp(-\lambda^{(j-1)}) - \exp(-\lambda^{(j)})] \sum_{i=1}^n (y_i - \mu)^2}{(\lambda^{(j-1)} - g)^2 - (\lambda^{(j)} - g)^2} \right] \right) \end{aligned} \quad (\text{a.2.7.11})$$

derivations for equation (2.7.21)

Rearrange the expression:

$$\begin{aligned} & \frac{1}{f(y)} \\ \approx & \frac{1}{J} \sum_{j=1}^J \frac{g(\theta^{(j)})}{f(y|\mu^{(j)}, \lambda^{(j)}) \pi(\mu^{(j)}) \pi(\lambda^{(j)})} \\ = & \frac{1}{J} \sum_{j=1}^J \frac{\mathbb{1}(\theta \in \hat{\Theta}) \times \omega^{-1} (2\pi)^{-k/2} |\hat{\Sigma}|^{-1/2} \exp\left(-\frac{1}{2}(\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta})\right) \mathbb{1}(\theta \in \hat{\Theta})}{(2\pi \exp(\lambda))^{-n/2} \exp\left(-\frac{1}{2} \sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda)}\right) (2\pi v)^{-1/2} \exp\left(-\frac{1}{2} \frac{(\mu - m)^2}{v}\right) (2\pi z)^{-1/2} \exp\left(-\frac{1}{2} \frac{(\lambda - g)^2}{z}\right)} \\ = & \mathbb{1}(\theta \in \hat{\Theta}) \times \omega^{-1} (2\pi)^{(n+2-k)/2} |\hat{\Sigma}|^{-1/2} (vz)^{1/2} \\ & \times \frac{1}{J} \sum_{j=1}^J \exp \left(\frac{1}{2} \left[n\lambda + \sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda)} + \frac{(\mu - m)^2}{v} + \frac{(\lambda - g)^2}{z} - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right) \\ = & \mathbb{1}(\theta \in \hat{\Theta}) \times (\omega J)^{-1} (2\pi)^{n/2} |\hat{\Sigma}|^{-1/2} (vz)^{1/2} \\ & \times \sum_{j=1}^J \exp \left(\frac{1}{2} \left[n\lambda + \sum_{i=1}^n \frac{(y_i - \mu)^2}{\exp(\lambda)} + \frac{(\mu - m)^2}{v} + \frac{(\lambda - g)^2}{z} - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right) \end{aligned} \quad (\text{a.2.7.12})$$

Mathematical theory

derivations for equation (2.8.11)

The definition of an invariant distribution implies that:

$$(\pi_1 \ \pi_2 \ \pi_3 \ \pi_4 \ \dots) \begin{pmatrix} p+q & r & 0 & 0 & 0 & \dots \\ p & q & r & 0 & 0 & \dots \\ 0 & p & q & r & 0 & \dots \\ 0 & 0 & p & q & r & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix} = (\pi_1 \ \pi_2 \ \pi_3 \ \pi_4 \ \dots) \quad (\text{a.2.8.1})$$

The product with the first column of P yields:

$$\begin{aligned} \pi_1(p+q) + \pi_2p &= \pi_1 \\ \Leftrightarrow \pi_2p &= \pi_1(1-p-q) \\ \Leftrightarrow \pi_2p &= \pi_1r \\ \Leftrightarrow \pi_2 &= (r/p)\pi_1 \end{aligned} \quad (\text{a.2.8.2})$$

The second column yields:

$$\begin{aligned} \pi_1r + \pi_2q + \pi_3p &= \pi_2 \\ \Leftrightarrow \pi_1r + \pi_3p &= \pi_2(1-q) \\ \Leftrightarrow \pi_1r + \pi_3p &= \pi_1(r/p)(1-q) \\ \Leftrightarrow \pi_1 + \pi_3(p/r) &= \pi_1(1-q)/p \\ \Leftrightarrow \pi_3(p/r) &= \pi_1(1-q-p)/p \\ \Leftrightarrow \pi_3(p/r) &= \pi_1(r/p) \\ \Leftrightarrow \pi_3(p/r) &= \pi_2 \\ \Leftrightarrow \pi_3 &= (r/p)\pi_2 \\ \Leftrightarrow \pi_3 &= (r/p)^2\pi_1 \end{aligned} \quad (\text{a.2.8.3})$$

The third column yields:

$$\begin{aligned}
& \pi_2 r + \pi_3 q + \pi_4 p = \pi_3 \\
\Leftrightarrow & \pi_2 r + \pi_4 p = \pi_3 (1 - q) \\
\Leftrightarrow & \pi_2 r + \pi_4 p = \pi_2 (r/p)(1 - q) \\
\Leftrightarrow & \pi_2 + \pi_4 (p/r) = \pi_2 (1 - q)/p \\
\Leftrightarrow & \pi_4 (p/r) = \pi_2 (1 - q - p)/p \\
\Leftrightarrow & \pi_4 (p/r) = \pi_2 (r/p) \\
\Leftrightarrow & \pi_4 (p/r) = \pi_3 \\
\Leftrightarrow & \pi_4 = (r/p) \pi_3 \\
\Leftrightarrow & \pi_4 = (r/p)^2 \pi_2 \\
\Leftrightarrow & \pi_4 = (r/p)^3 \pi_1
\end{aligned} \tag{a.2.8.4}$$

Continuing this way, one obtains in general that $\pi_j = (r/p)^{j-1} \pi_1$. If an invariant distribution exists, we must have $\pi_1 + \pi_2 + \pi_3 \cdots = 1$. Hence:

$$\begin{aligned}
& \pi_1 + \pi_2 + \pi_3 + \cdots = 1 \\
\Leftrightarrow & \pi_1 + (r/p) \pi_1 + (r/p)^2 \pi_1 + \cdots = 1 \\
\Leftrightarrow & \pi_1 (1 + (r/p) + (r/p)^2 + \cdots) = 1 \\
\Leftrightarrow & \pi_1 \frac{1}{1 - r/p} = 1 \\
\Leftrightarrow & \pi_1 = 1 - r/p
\end{aligned} \tag{a.2.8.5}$$

derivations for equation (2.8.16)

Start from the definition and rearrange:

$$\begin{aligned}
\pi(y_t) &= \int \pi(y_{t-1}) p(y_{t-1}, y_t) dy_{t-1} \\
&\propto \int \exp\left(-\frac{1}{2} \frac{(y_{t-1} - \mu)^2}{\sigma}\right) \exp\left(-\frac{1}{2} \frac{(y_t - c - \gamma y_{t-1})^2}{s}\right) dy_{t-1} \\
&= \int \exp\left(-\frac{1}{2} \left[\frac{(y_{t-1} - \mu)^2}{\sigma} + \frac{(y_t - c - \gamma y_{t-1})^2}{s} \right]\right) dy_{t-1} \\
&= \int \exp\left(-\frac{1}{2} \left[\frac{(y_{t-1} - \mu)^2 (1 - \gamma^2)}{s} + \frac{(y_t - c - \gamma y_{t-1})^2}{s} \right]\right) dy_{t-1} \\
&= \int \exp\left(-\frac{1}{2s} [(y_{t-1} - \mu)^2 (1 - \gamma^2) + (y_t - \mu(1 - \gamma) - \gamma y_{t-1})^2]\right) dy_{t-1}
\end{aligned} \tag{a.2.8.6}$$

Consider the term within the square brackets:

$$\begin{aligned}
& (y_{t-1} - \mu)^2 (1 - \gamma^2) + (y_t - \mu(1 - \gamma) - \gamma y_{t-1})^2 \\
&= (1 - \gamma^2) y_{t-1}^2 + (1 - \gamma^2) \mu^2 - 2(1 - \gamma^2) \mu y_{t-1} + y_t^2 + \mu^2 (1 - \gamma)^2 + \gamma^2 y_{t-1}^2 \\
&\quad - 2\mu(1 - \gamma) y_t - 2\gamma y_t y_{t-1} + 2\mu\gamma(1 - \gamma) y_{t-1} \\
&= y_{t-1}^2 + (1 - \gamma^2) \mu^2 - 2(1 - \gamma^2) \mu y_{t-1} + y_t^2 + \mu^2 (1 - \gamma)^2 \\
&\quad - 2\mu(1 - \gamma) y_t - 2\gamma y_t y_{t-1} + 2\mu\gamma(1 - \gamma) y_{t-1} \\
&= y_{t-1}^2 + (1 - \gamma^2) \mu^2 - 2\mu y_{t-1} + 2\gamma^2 \mu y_{t-1} + y_t^2 + \mu^2 (1 - \gamma)^2 \\
&\quad - 2\mu(1 - \gamma) y_t - 2\gamma y_t y_{t-1} + 2\mu\gamma y_{t-1} - 2\mu\gamma^2 y_{t-1}
\end{aligned} \tag{a.2.8.7}$$

$$\begin{aligned}
&= y_{t-1}^2 + (1 - \gamma^2)\mu^2 - 2\mu y_{t-1} + y_t^2 + \mu^2(1 - \gamma)^2 - 2\mu(1 - \gamma)y_t - 2\gamma y_t y_{t-1} + 2\mu\gamma y_{t-1} \\
&= y_{t-1}^2 + (1 - \gamma^2)\mu^2 - 2\mu y_{t-1}(1 - \gamma) + y_t^2 + \mu^2(1 - \gamma)^2 - 2\mu(1 - \gamma)y_t - 2\gamma y_t y_{t-1} \\
&= y_{t-1}^2 + (1 - \gamma^2)\mu^2 - 2\mu y_{t-1}(1 - \gamma) + y_t^2 + \mu^2(1 - \gamma)^2 - 2\mu(1 - \gamma^2)y_t + 2\mu(1 - \gamma)\gamma y_t - 2\gamma y_t y_{t-1} \\
&= y_{t-1}^2 + (1 - \gamma^2)\mu^2 - 2\mu y_{t-1}(1 - \gamma) + (1 - \gamma^2)y_t^2 + \gamma^2 y_t^2 + \mu^2(1 - \gamma)^2 \\
&\quad - 2\mu(1 - \gamma^2)y_t + 2\mu(1 - \gamma)\gamma y_t - 2\gamma y_t y_{t-1} \\
&= (1 - \gamma^2)y_t^2 + (1 - \gamma^2)\mu^2 - 2\mu(1 - \gamma^2)y_t \\
&\quad + y_{t-1}^2 + \mu^2(1 - \gamma)^2 + \gamma^2 y_t^2 - 2\gamma y_t y_{t-1} - 2\mu y_{t-1}(1 - \gamma) + 2\mu(1 - \gamma)\gamma y_t \\
&= (1 - \gamma^2)y_t^2 + (1 - \gamma^2)\mu^2 - 2\mu(1 - \gamma^2)y_t \\
&\quad + y_{t-1}^2 + c^2 + \gamma^2 y_t^2 - 2\gamma y_t y_{t-1} - 2c y_{t-1} + 2c\gamma y_t \\
&= (1 - \gamma^2)(y_t - \mu)^2 + (y_{t-1} - c - \gamma y_t)^2 \tag{a.2.8.8}
\end{aligned}$$

Substitute back in (a.2.8.6):

$$\pi(y_t) = \int \exp\left(-\frac{1}{2s} [(1 - \gamma^2)(y_t - \mu)^2 + (y_{t-1} - c - \gamma y_t)^2]\right) dy_{t-1} \tag{a.2.8.9}$$

and this eventually reformulates as:

$$\pi(y_t) = \exp\left(-\frac{1}{2} \frac{(y_t - \mu)^2}{\sigma}\right) \int \exp\left(-\frac{1}{2} \frac{(y_{t-1} - c - \gamma y_t)^2}{s}\right) dy_{t-1} \tag{a.2.8.10}$$

PART III

Econometrics

The linear regression model

derivations for equation (3.9.7)

Consider first β . To do so, rewrite the likelihood function as:

$$\log(f(y|\beta, \sigma)) = -\frac{n}{2} \log(2\pi) - \frac{n}{2} \log(\sigma) - \frac{1}{2\sigma} (y'y + \beta'X'X\beta - 2\beta'X'y) \quad (\text{a.3.9.1})$$

Then solve for the partial derivative:

$$\begin{aligned} \frac{\partial \log(f(y|\beta, \sigma))}{\partial \beta} &= 0 \\ \Leftrightarrow -\frac{1}{2\sigma} (2X'X\beta - 2X'y) &= 0 \\ \Leftrightarrow 2X'X\beta - 2X'y &= 0 \\ \Leftrightarrow X'X\beta - X'y &= 0 \\ \Leftrightarrow \beta &= (X'X)^{-1}X'y \end{aligned} \quad (\text{a.3.9.2})$$

Hence the estimate is $\hat{\beta} = (X'X)^{-1}X'y$. Consider now σ . Solve for the partial derivative:

$$\begin{aligned} \frac{\partial \log(f(y|\beta, \sigma))}{\partial \sigma} &= 0 \\ \Leftrightarrow -\frac{n}{2} \frac{1}{\sigma} + \frac{1}{2} \frac{(y - X\beta)'(y - X\beta)}{\sigma^2} &= 0 \\ \Leftrightarrow -n + \frac{(y - X\beta)'(y - X\beta)}{\sigma} &= 0 \\ \Leftrightarrow \frac{(y - X\beta)'(y - X\beta)}{\sigma} &= n \\ \Leftrightarrow \sigma &= \frac{(y - X\hat{\beta})'(y - X\hat{\beta})}{n} \end{aligned} \quad (\text{a.3.9.3})$$

This expression gives the optimum for any value of β . To obtain a global maximum we must choose the value of β that maximizes the likelihood, namely $\hat{\beta}$. Therefore, the estimate for σ is given by $\hat{\sigma} = (y - X\hat{\beta})'(y - X\hat{\beta})/n$.

derivations for equation (3.9.12)

Develop and group:

$$\begin{aligned} &\exp\left(-\frac{1}{2} \frac{(y - X\beta)'(y - X\beta)}{\sigma}\right) \times \exp\left(-\frac{1}{2}(\beta - b)'V^{-1}(\beta - b)\right) \\ &= \exp\left(-\frac{1}{2} [(y - X\beta)' \sigma^{-1} (y - X\beta) + (\beta - b)'V^{-1}(\beta - b)]\right) \end{aligned} \quad (\text{a.3.9.4})$$

Consider the term in square brackets:

$$\begin{aligned}
& (y - X\beta)' \sigma^{-1} (y - X\beta) + (\beta - b)' V^{-1} (\beta - b) \\
&= y' \sigma^{-1} y + \beta' X' \sigma^{-1} X \beta - 2\beta' X' \sigma^{-1} y + \beta' V^{-1} \beta + b' V^{-1} b - 2\beta' V^{-1} b \\
&= \beta' (V^{-1} + \sigma^{-1} X' X) \beta - 2\beta' (V^{-1} b + \sigma^{-1} X' y) + b' V^{-1} b + y' \sigma^{-1} y
\end{aligned} \tag{a.3.9.5}$$

Substitute back in (a.3.9.4):

$$= \exp \left(-\frac{1}{2} [\beta' (V^{-1} + \sigma^{-1} X' X) \beta - 2\beta' (V^{-1} b + \sigma^{-1} X' y) + b' V^{-1} b + y' \sigma^{-1} y] \right) \tag{a.3.9.6}$$

derivations for equation (3.9.23)

Group and rearrange:

$$\begin{aligned}
& \pi(\beta, \sigma | y) \\
&\propto \sigma^{-n/2} \exp \left(-\frac{1}{2} \frac{(y - X\beta)' (y - X\beta)}{\sigma} \right) \times |\sigma V|^{-1/2} \exp \left(-\frac{1}{2} (\beta - b)' (\sigma V)^{-1} (\beta - b) \right) \\
&\times \sigma^{-\alpha/2-1} \exp \left(-\frac{\delta}{2\sigma} \right) \\
&= \sigma^{-k/2} \exp \left(-\frac{1}{2\sigma} [(y - X\beta)' (y - X\beta) + (\beta - b)' V^{-1} (\beta - b)] \right) \sigma^{-(\alpha+n)/2-1} \exp \left(-\frac{\delta}{2\sigma} \right)
\end{aligned} \tag{a.3.9.7}$$

Define:

$$\bar{\alpha} = \alpha + n \tag{a.3.9.8}$$

Then:

$$= \sigma^{-k/2} \exp \left(-\frac{1}{2\sigma} [(y - X\beta)' (y - X\beta) + (\beta - b)' V^{-1} (\beta - b)] \right) \sigma^{-\bar{\alpha}/2-1} \exp \left(-\frac{\delta}{2\sigma} \right) \tag{a.3.9.9}$$

Consider the term between the square brackets:

$$\begin{aligned}
& (y - X\beta)' (y - X\beta) + (\beta - b)' V^{-1} (\beta - b) \\
&= y' y + \beta' X' X \beta - 2\beta' X' y + \beta' V^{-1} \beta + b' V^{-1} b - 2\beta' V^{-1} b \\
&= \beta' (V^{-1} + X' X) \beta - 2\beta' (V^{-1} b + X' y) + y' y + b' V^{-1} b \\
&= \beta' (V^{-1} + X' X) \beta - 2\beta' \bar{V}^{-1} \bar{V} (V^{-1} b + X' y) + y' y + b' V^{-1} b + \bar{b}' \bar{V}^{-1} \bar{b} - \bar{b}' \bar{V}^{-1} \bar{b}
\end{aligned} \tag{a.3.9.10}$$

Define:

$$\bar{V} = (V^{-1} + X' X)^{-1} \quad \bar{b} = \bar{V} (V^{-1} b + X' y) \tag{a.3.9.11}$$

Then (a.3.9.10) becomes:

$$\begin{aligned}
&= \beta' \bar{V}^{-1} \beta - 2\beta' \bar{V}^{-1} \bar{b} + y' y + b' V^{-1} b + \bar{b}' \bar{V}^{-1} \bar{b} - \bar{b}' \bar{V}^{-1} \bar{b} \\
&= (\beta' \bar{V}^{-1} \beta - 2\beta' \bar{V}^{-1} \bar{b} + \bar{b}' \bar{V}^{-1} \bar{b}) + y' y + b' V^{-1} b - \bar{b}' \bar{V}^{-1} \bar{b} \\
&= (\beta - \bar{b})' \bar{V}^{-1} (\beta - \bar{b}) + y' y + b' V^{-1} b - \bar{b}' \bar{V}^{-1} \bar{b}
\end{aligned} \tag{a.3.9.12}$$

Substitute back in (a.3.9.9):

$$= \sigma^{-k/2} \exp \left(-\frac{1}{2} (\beta - \bar{b})' (\sigma \bar{V})^{-1} (\beta - \bar{b}) \right) \sigma^{-\bar{\alpha}/2-1} \exp \left(-\frac{\delta + y' y + b' V^{-1} b - \bar{b}' \bar{V}^{-1} \bar{b}}{2\sigma} \right) \tag{a.3.9.13}$$

Define:

$$\bar{\delta} = \delta + y'y + b'V^{-1}b - \bar{b}'\bar{V}^{-1}\bar{b} \quad (\text{a.3.9.14})$$

Then (a.3.9.13) eventually rewrites:

$$\pi(\beta, \sigma|y) \propto \sigma^{-k/2} \exp\left(-\frac{1}{2}(\beta - \bar{b})'(\sigma\bar{V})^{-1}(\beta - \bar{b})\right) \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right) \quad (\text{a.3.9.15})$$

derivations for equation (3.9.28)

Rearrange:

$$\begin{aligned} & \Gamma\left(\frac{\bar{\alpha}+k}{2}\right) \left(\frac{\bar{\delta} + (\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b})}{2}\right)^{-\frac{\bar{\alpha}+k}{2}} \\ & \propto \left(\frac{\bar{\delta} + (\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b})}{2}\right)^{-\frac{\bar{\alpha}+k}{2}} \\ & \propto (\bar{\delta} + (\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b}))^{-\frac{\bar{\alpha}+k}{2}} \\ & \propto (1 + (\beta - \bar{b})'(\bar{\delta}\bar{V})^{-1}(\beta - \bar{b}))^{-\frac{\bar{\alpha}+k}{2}} \\ & \propto \left(1 + \frac{1}{\bar{\alpha}}(\beta - \bar{b})'(\bar{\delta}\bar{V}/\bar{\alpha})^{-1}(\beta - \bar{b})\right)^{-\frac{\bar{\alpha}+k}{2}} \end{aligned} \quad (\text{a.3.9.16})$$

derivations for equation (3.9.38)

Rearrange the likelihood function:

$$\begin{aligned} & f(y|\beta, \sigma) \\ & = (2\pi)^{-n/2} |\sigma W|^{-1/2} \exp\left(-\frac{1}{2}(y - X\beta)'(\sigma W)^{-1}(y - X\beta)\right) \\ & = (2\pi\sigma)^{-n/2} |W|^{-1/2} \exp\left(-\frac{1}{2} \frac{(y - X\beta)'W^{-1}(y - X\beta)}{\sigma}\right) \end{aligned} \quad (\text{a.3.9.17})$$

This reformulates further as:

$$\begin{aligned} & = (2\pi\sigma)^{-n/2} \left(\prod_{i=1}^n w_i^{-1/2}\right) \exp\left(-\frac{1}{2} \frac{(y - X\beta)' \text{diag}(\exp(-Z\gamma)) (y - X\beta)}{\sigma}\right) \\ & = (2\pi\sigma)^{-n/2} \left(\prod_{i=1}^n \exp(z_i'\gamma)\right)^{-1/2} \exp\left(-\frac{1}{2} \frac{(y - X\beta)' \text{diag}(\exp(-Z\gamma)) (y - X\beta)}{\sigma}\right) \\ & = (2\pi\sigma)^{-n/2} \left(\exp\left(\sum_{i=1}^n z_i'\gamma\right)\right)^{-1/2} \exp\left(-\frac{1}{2} \frac{(y - X\beta)' \text{diag}(\exp(-Z\gamma)) (y - X\beta)}{\sigma}\right) \\ & = (2\pi\sigma)^{-n/2} (\exp(1_n'Z\gamma))^{-1/2} \exp\left(-\frac{1}{2} \frac{(y - X\beta)' \text{diag}(\exp(-Z\gamma)) (y - X\beta)}{\sigma}\right) \\ & = (2\pi\sigma)^{-n/2} \exp\left(-\frac{1}{2}1_n'Z\gamma\right) \exp\left(-\frac{1}{2} \frac{(y - X\beta)' \text{diag}(\exp(-Z\gamma)) (y - X\beta)}{\sigma}\right) \\ & = (2\pi\sigma)^{-n/2} \exp\left(-\frac{1}{2} [1_n'Z\gamma + (y - X\beta)' \text{diag}(\exp(-Z\gamma)) (y - X\beta)/\sigma]\right) \end{aligned} \quad (\text{a.3.9.18})$$

derivations for equation (3.9.44)

Rearrange the terms:

$$\begin{aligned}
& \pi(\beta|y, \sigma, w) \\
&= \exp\left(-\frac{1}{2} \frac{(y - X\beta)'W^{-1}(y - X\beta)}{\sigma}\right) \times \exp\left(-\frac{1}{2}(\beta - b)'V^{-1}(\beta - b)\right) \\
&= \exp\left(-\frac{1}{2} [(y - X\beta)'(\sigma W)^{-1}(y - X\beta) + (\beta - b)'V^{-1}(\beta - b)]\right) \tag{a.3.9.19}
\end{aligned}$$

Consider the term in square brackets and complete the squares:

$$\begin{aligned}
& (y - X\beta)'(\sigma W)^{-1}(y - X\beta) + (\beta - b)'V^{-1}(\beta - b) \\
&= y'(\sigma W)^{-1}y + \beta'X'(\sigma W)^{-1}X\beta - 2\beta'X'(\sigma W)^{-1}y + \beta'V^{-1}\beta + b'V^{-1}b - 2\beta'V^{-1}b \\
&= \beta'(V^{-1} + \sigma^{-1}X'W^{-1}X)\beta - 2\beta'(V^{-1}b + \sigma^{-1}X'W^{-1}y) + y'(\sigma W)^{-1}y + b'V^{-1}b \\
&= \beta'(V^{-1} + \sigma^{-1}X'W^{-1}X)\beta - 2\beta'\bar{V}^{-1}\bar{V}(V^{-1}b + \sigma^{-1}X'W^{-1}y) \\
&\quad + y'(\sigma W)^{-1}y + b'V^{-1}b + \bar{b}'\bar{V}^{-1}\bar{b} - \bar{b}'\bar{V}^{-1}\bar{b} \tag{a.3.9.20}
\end{aligned}$$

Define:

$$\bar{V} = (V^{-1} + \sigma^{-1}X'W^{-1}X)^{-1} \quad \bar{b} = \bar{V}(V^{-1}b + \sigma^{-1}X'W^{-1}y) \tag{a.3.9.21}$$

Then (a.3.9.20) rewrites:

$$\begin{aligned}
&= \beta'\bar{V}^{-1}\beta - 2\beta'\bar{V}^{-1}\bar{b} + \bar{b}'\bar{V}^{-1}\bar{b} + y'(\sigma W)^{-1}y + b'V^{-1}b - \bar{b}'\bar{V}^{-1}\bar{b} \\
&= (\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b}) + y'(\sigma W)^{-1}y + b'V^{-1}b - \bar{b}'\bar{V}^{-1}\bar{b} \tag{a.3.9.22}
\end{aligned}$$

Substitute back in (a.3.9.19):

$$\begin{aligned}
& \pi(\beta|y, \sigma, w) \\
&= \exp\left(-\frac{1}{2} [(\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b}) + y'(\sigma W)^{-1}y + b'V^{-1}b - \bar{b}'\bar{V}^{-1}\bar{b}]\right) \\
&= \exp\left(-\frac{1}{2}(\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b})\right) \exp\left(-\frac{1}{2} [y'(\sigma W)^{-1}y + b'V^{-1}b - \bar{b}'\bar{V}^{-1}\bar{b}]\right) \\
&\propto \exp\left(-\frac{1}{2}(\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b})\right) \tag{a.3.9.23}
\end{aligned}$$

derivations for equation (3.9.69)

Rearrange the terms:

$$\begin{aligned}
& \pi(\phi|y, \beta, \sigma) \\
&\propto \exp\left(-\frac{1}{2} \frac{(\varepsilon - E\phi)'(\varepsilon - E\phi)}{\sigma}\right) \times \exp\left(-\frac{1}{2}(\phi - p)'H^{-1}(\phi - p)\right) \\
&= \exp\left(-\frac{1}{2} [(\varepsilon - E\phi)'\sigma^{-1}(\varepsilon - E\phi) + (\phi - p)'H^{-1}(\phi - p)]\right) \tag{a.3.9.24}
\end{aligned}$$

Consider the term in square brackets and complete the squares:

$$\begin{aligned}
& (\varepsilon - E\phi)' \sigma^{-1} (\varepsilon - E\phi) + (\phi - p)' H^{-1} (\phi - p) \\
&= \varepsilon' \sigma^{-1} \varepsilon + \phi' E' \sigma^{-1} E \phi - 2\phi' E' \sigma^{-1} \varepsilon + \phi' H^{-1} \phi + p' H^{-1} p - 2\phi' H^{-1} p \\
&= \phi' (H^{-1} + \sigma^{-1} E' E) \phi - 2\phi' (H^{-1} p + \sigma^{-1} E' \varepsilon) + \varepsilon' \sigma^{-1} \varepsilon + p' H^{-1} p \\
&= \phi' (H^{-1} + \sigma^{-1} E' E) \phi - 2\phi' \bar{H}^{-1} \bar{H} (H^{-1} p + \sigma^{-1} E' \varepsilon) + \varepsilon' \sigma^{-1} \varepsilon + p' H^{-1} p + \bar{p}' \bar{H}^{-1} \bar{p} - \bar{p}' \bar{H}^{-1} \bar{p}
\end{aligned} \tag{a.3.9.25}$$

Define:

$$\bar{H} = (H^{-1} + \sigma^{-1} E' E)^{-1} \quad \bar{p} = \bar{H} (H^{-1} p + \sigma^{-1} E' \varepsilon) \tag{a.3.9.26}$$

Then (a.3.9.25) becomes:

$$\begin{aligned}
&= \phi' \bar{H}^{-1} \phi - 2\phi' \bar{H}^{-1} \bar{p} + \bar{p}' \bar{H}^{-1} \bar{p} + \varepsilon' \sigma^{-1} \varepsilon + p' H^{-1} p - \bar{p}' \bar{H}^{-1} \bar{p} \\
&= (\phi - \bar{p})' \bar{H}^{-1} (\phi - \bar{p}) + \varepsilon' \sigma^{-1} \varepsilon + p' H^{-1} p - \bar{p}' \bar{H}^{-1} \bar{p}
\end{aligned} \tag{a.3.9.27}$$

Substitute back in (a.3.9.24) to obtain:

$$\begin{aligned}
& \pi(\phi|y, \beta, \sigma) \\
&= \exp \left(-\frac{1}{2} [(\phi - \bar{p})' \bar{H}^{-1} (\phi - \bar{p}) + \varepsilon' \sigma^{-1} \varepsilon + p' H^{-1} p - \bar{p}' \bar{H}^{-1} \bar{p}] \right) \\
&= \exp \left(-\frac{1}{2} (\phi - \bar{p})' \bar{H}^{-1} (\phi - \bar{p}) \right) \exp \left(-\frac{1}{2} [\varepsilon' \sigma^{-1} \varepsilon + p' H^{-1} p - \bar{p}' \bar{H}^{-1} \bar{p}] \right) \\
&\propto \exp \left(-\frac{1}{2} (\phi - \bar{p})' \bar{H}^{-1} (\phi - \bar{p}) \right)
\end{aligned} \tag{a.3.9.28}$$

Applications with the linear regression model

derivations for equation (3.10.4)

Rearrange the expression:

$$\begin{aligned}
 & f(\hat{y}|y) \\
 \propto & \int \exp\left(-\frac{1}{2} \frac{(\hat{y} - \hat{X}\beta)'(\hat{y} - \hat{X}\beta)}{\sigma}\right) \exp\left(-\frac{1}{2}(\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b})\right) d\beta \\
 = & \int \exp\left(-\frac{1}{2} [\sigma^{-1}(\hat{y} - \hat{X}\beta)'(\hat{y} - \hat{X}\beta) + (\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b})]\right) d\beta \tag{a.3.10.1}
 \end{aligned}$$

Consider the term in square brackets:

$$\begin{aligned}
 & \sigma^{-1}(\hat{y} - \hat{X}\beta)'(\hat{y} - \hat{X}\beta) + (\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b}) \\
 = & \sigma^{-1}\hat{y}'\hat{y} + \sigma^{-1}\beta'\hat{X}'\hat{X}\beta - 2\sigma^{-1}\beta'\hat{X}'\hat{y} + \beta'\bar{V}^{-1}\beta + \bar{b}'\bar{V}^{-1}\bar{b} - 2\beta'\bar{V}^{-1}\bar{b} \\
 = & \beta'(\bar{V}^{-1} + \sigma^{-1}\hat{X}'\hat{X})\beta - 2\beta'(\bar{V}^{-1}\bar{b} + \sigma^{-1}\hat{X}'\hat{y}) + \sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} \\
 = & \beta'(\hat{V}^{-1} + \sigma^{-1}\hat{X}'\hat{X})\beta - 2\beta'\hat{V}^{-1}\hat{V}(\bar{V}^{-1}\bar{b} + \sigma^{-1}\hat{X}'\hat{y}) + \sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} + \hat{b}'\hat{V}^{-1}\hat{b} - \hat{b}'\hat{V}^{-1}\hat{b} \tag{a.3.10.2}
 \end{aligned}$$

Define:

$$\hat{V} = (\bar{V}^{-1} + \sigma^{-1}\hat{X}'\hat{X})^{-1} \quad \hat{b} = \hat{V}(\bar{V}^{-1}\bar{b} + \sigma^{-1}\hat{X}'\hat{y}) \tag{a.3.10.3}$$

Then (a.3.10.2) becomes:

$$\begin{aligned}
 & = \beta'\hat{V}^{-1}\beta - 2\beta'\hat{V}^{-1}\hat{b} + \hat{b}'\hat{V}^{-1}\hat{b} + \sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \hat{b}'\hat{V}^{-1}\hat{b} \\
 & = (\beta - \hat{b})'\hat{V}^{-1}(\beta - \hat{b}) + \sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \hat{b}'\hat{V}^{-1}\hat{b} \tag{a.3.10.4}
 \end{aligned}$$

Substituting back in (a.3.10.1):

$$\begin{aligned}
 & f(\hat{y}|y) \\
 \propto & \int \exp\left(-\frac{1}{2} [(\beta - \hat{b})'\hat{V}^{-1}(\beta - \hat{b}) + \sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \hat{b}'\hat{V}^{-1}\hat{b}]\right) d\beta \\
 = & \int \exp\left(-\frac{1}{2}(\beta - \hat{b})'\hat{V}^{-1}(\beta - \hat{b})\right) \exp\left(-\frac{1}{2} [\sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \hat{b}'\hat{V}^{-1}\hat{b}]\right) d\beta \\
 = & \exp\left(-\frac{1}{2} [\sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \hat{b}'\hat{V}^{-1}\hat{b}]\right) \int \exp\left(-\frac{1}{2}(\beta - \hat{b})'\hat{V}^{-1}(\beta - \hat{b})\right) d\beta \\
 \propto & \exp\left(-\frac{1}{2} [\sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \hat{b}'\hat{V}^{-1}\hat{b}]\right) \tag{a.3.10.5}
 \end{aligned}$$

Consider the term in square brackets:

$$\begin{aligned}
& \sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \hat{b}'\hat{V}^{-1}\hat{b} \\
&= \sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - (\bar{V}^{-1}\bar{b} + \sigma^{-1}\hat{X}'\hat{y})'\hat{V}\hat{V}^{-1}\hat{V}(\bar{V}^{-1}\bar{b} + \sigma^{-1}\hat{X}'\hat{y}) \\
&= \sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - (\bar{V}^{-1}\bar{b} + \sigma^{-1}\hat{X}'\hat{y})'\hat{V}(\bar{V}^{-1}\bar{b} + \sigma^{-1}\hat{X}'\hat{y}) \\
&= \sigma^{-1}\hat{y}'\hat{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \bar{b}'\bar{V}^{-1}\hat{V}\bar{V}^{-1}\bar{b} - \sigma^{-2}\hat{y}'\hat{X}\hat{V}\hat{X}'\hat{y} - 2\sigma^{-1}\hat{y}'\hat{X}\hat{V}\bar{V}^{-1}\bar{b} \\
&= \hat{y}'(\sigma^{-1}I_m - \sigma^{-2}\hat{X}\hat{V}\hat{X}')\hat{y} - \bar{b}'(\bar{V}^{-1} - \bar{V}^{-1}\hat{V}\bar{V}^{-1})\bar{b} - 2\sigma^{-1}\hat{y}'\hat{X}\hat{V}\bar{V}^{-1}\bar{b}
\end{aligned} \tag{a.3.10.6}$$

In what follows, we make use of property m.13 (the Sherman-Woodbury-Morrison identity):
 $(A + BDC)^{-1} = A^{-1} - A^{-1}B(D^{-1} + CA^{-1}B)^{-1}CA^{-1}$.

Consider the central part of the second term in (a.3.10.6). Rearrange and use the identity twice to obtain:

$$\begin{aligned}
& \bar{V}^{-1} - \bar{V}^{-1}\hat{V}\bar{V}^{-1} \\
&= \bar{V}^{-1} - \bar{V}^{-1}(\bar{V}^{-1} + \sigma^{-1}\hat{X}'\hat{X})^{-1}\bar{V}^{-1} \\
&= (\bar{V} + \sigma(\hat{X}'\hat{X}))^{-1} \\
&= \sigma^{-1}\hat{X}'\hat{X} - \sigma^{-1}\hat{X}'\hat{X}(\sigma^{-1}\hat{X}'\hat{X} + \bar{V}^{-1})^{-1}\sigma^{-1}\hat{X}'\hat{X} \\
&= \sigma^{-1}\hat{X}'\hat{X} - \sigma^{-2}\hat{X}'\hat{X}\hat{V}\hat{X}'\hat{X} \\
&= \hat{X}'(\sigma^{-1}I_m - \sigma^{-2}\hat{X}\hat{V}\hat{X}')\hat{X}
\end{aligned} \tag{a.3.10.7}$$

Consider finally the central part of the third term in (a.3.10.6). We note that $\hat{V}(\bar{V}^{-1} + \sigma^{-1}\hat{X}'\hat{X}) = I_m$ so $\hat{V}\bar{V}^{-1} = I_m - \hat{V}\sigma^{-1}\hat{X}'\hat{X}$. Following:

$$\hat{X}\hat{V}\bar{V}^{-1} = \hat{X} - \hat{X}\hat{V}\sigma^{-1}\hat{X}'\hat{X} = (I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}')\hat{X} \tag{a.3.10.8}$$

Substitute (a.3.10.7) and (a.3.10.8) back in (a.3.10.6) to obtain:

$$\begin{aligned}
&= \hat{y}'(\sigma^{-1}I_m - \sigma^{-2}\hat{X}\hat{V}\hat{X}')\hat{y} - \bar{b}'\hat{X}'(\sigma^{-1}I_m - \sigma^{-2}\hat{X}\hat{V}\hat{X}')\hat{X}\bar{b} - 2\sigma^{-1}\hat{y}'(I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}')\hat{X}\bar{b} \\
&= \sigma^{-1}\hat{y}'(I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}')\hat{y} - \sigma^{-1}\bar{b}'\hat{X}'(I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}')\hat{X}\bar{b} - 2\sigma^{-1}\hat{y}'(I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}')\hat{X}\bar{b} \\
&= \sigma^{-1}[\hat{y}'(I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}')\hat{y} - \bar{b}'\hat{X}'(I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}')\hat{X}\bar{b} - 2\hat{y}'(I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}')\hat{X}\bar{b}] \\
&= \sigma^{-1}(\hat{y} - \hat{X}\bar{b})'(I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}')(\hat{y} - \hat{X}\bar{b})
\end{aligned} \tag{a.3.10.9}$$

We use one last time the Sherman-Woodbury-Morrison identity on the central term to obtain:

$$I_m - \sigma^{-1}\hat{X}\hat{V}\hat{X}' = I_m - \sigma^{-1}\hat{X}(\bar{V}^{-1} + \sigma^{-1}\hat{X}'\hat{X})^{-1}\hat{X}' = (I_m + \sigma^{-1}\hat{X}\bar{V}\hat{X}')^{-1} \tag{a.3.10.10}$$

Substituting back in (a.3.10.9):

$$\begin{aligned}
&= \sigma^{-1}(\hat{y} - \hat{X}\bar{b})'(I_m + \sigma^{-1}\hat{X}\bar{V}\hat{X}')^{-1}(\hat{y} - \hat{X}\bar{b}) \\
&= (\hat{y} - \hat{X}\bar{b})'(\sigma I_m + \hat{X}\bar{V}\hat{X}')^{-1}(\hat{y} - \hat{X}\bar{b})
\end{aligned} \tag{a.3.10.11}$$

Eventually substituting back in (a.3.10.5), we conclude:

$$f(\hat{y}|y) \propto \exp\left(-\frac{1}{2}(\hat{y} - \hat{X}\bar{b})'(\sigma I_m + \hat{X}\bar{V}\hat{X}')^{-1}(\hat{y} - \hat{X}\bar{b})\right) \tag{a.3.10.12}$$

derivations for equation (3.10.6)

Rearrange the expression:

$$\begin{aligned}
& f(\hat{y}|y) \\
& \propto \int \int \sigma^{-m/2} \exp\left(-\frac{1}{2} \frac{(\hat{y} - \hat{X}\beta)'(\hat{y} - \hat{X}\beta)}{\sigma}\right) \exp\left(-\frac{1}{2} \frac{(y - X\beta)'(y - X\beta)}{\sigma}\right) \\
& \times \sigma^{-k/2} \exp\left(-\frac{1}{2}(\beta - b)'(\sigma V)^{-1}(\beta - b)\right) \times \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right) \\
& = \int \int \sigma^{-k/2} \exp\left(-\frac{1}{2\sigma} [(\hat{y} - \hat{X}\beta)'(\hat{y} - \hat{X}\beta) + (y - X\beta)'(y - X\beta) + (\beta - b)'(\sigma V)^{-1}(\beta - b) + \delta]\right) \\
& \times \sigma^{-(\alpha+n+m)/2-1} d\beta d\sigma \tag{a.3.10.13}
\end{aligned}$$

Consider the term in the square bracket:

$$\begin{aligned}
& (\hat{y} - \hat{X}\beta)'(\hat{y} - \hat{X}\beta) + (y - X\beta)'(y - X\beta) + (\beta - b)'(\sigma V)^{-1}(\beta - b) + \delta \\
& = \hat{y}'\hat{y} + \beta'\hat{X}'\hat{X}\beta - 2\beta'\hat{X}'\hat{y} + y'y + \beta'X'X\beta - 2\beta'X'y + \beta'V^{-1}\beta + b'V^{-1}b - 2\beta'V^{-1}b + \delta \\
& = (\delta + \hat{y}'\hat{y} + y'y + b'V^{-1}b) + \beta'(V^{-1} + X'X + \hat{X}'\hat{X})\beta - 2\beta'(V^{-1}b + X'y + \hat{X}'\hat{y}) \\
& = (\delta + \hat{y}'\hat{y} + y'y + b'V^{-1}b) + \beta'(V^{-1} + X'X + \hat{X}'\hat{X})\beta - 2\beta'\hat{V}^{-1}\hat{V}(V^{-1}b + X'y + \hat{X}'\hat{y}) + \hat{b}'\hat{V}^{-1}\hat{b} - \hat{b}'\hat{V}^{-1}\hat{b} \\
& = (\delta + \hat{y}'\hat{y} + y'y + b'V^{-1}b - \hat{b}'\hat{V}^{-1}\hat{b}) + \beta'(V^{-1} + X'X + \hat{X}'\hat{X})\beta - 2\beta'\hat{V}^{-1}\hat{V}(V^{-1}b + X'y + \hat{X}'\hat{y}) + \hat{b}'\hat{V}^{-1}\hat{b} \\
& \tag{a.3.10.14}
\end{aligned}$$

Define:

$$\hat{\delta} = \delta + \hat{y}'\hat{y} + y'y + b'V^{-1}b - \hat{b}'\hat{V}^{-1}\hat{b} \quad \hat{V} = (V^{-1} + X'X + \hat{X}'\hat{X})^{-1} \quad \hat{b} = \hat{V}(V^{-1}b + X'y + \hat{X}'\hat{y}) \tag{a.3.10.15}$$

Then (a.3.10.14) rewrites:

$$\begin{aligned}
& = \hat{\delta} + \beta'\hat{V}^{-1}\beta - 2\beta'\hat{V}^{-1}\hat{b} + \hat{b}'\hat{V}^{-1}\hat{b} \\
& = \hat{\delta} + (\beta - \hat{b})'\hat{V}^{-1}(\beta - \hat{b}) \tag{a.3.10.16}
\end{aligned}$$

Substitute back in (a.3.10.13):

$$\begin{aligned}
& f(\hat{y}|y) \\
& \propto \int \int \sigma^{-k/2} \exp\left(-\frac{1}{2\sigma} [\hat{\delta} + (\beta - \hat{b})'\hat{V}^{-1}(\beta - \hat{b})]\right) \sigma^{-(\alpha+n+m)/2-1} d\beta d\sigma \\
& = \int \int \sigma^{-k/2} \exp\left(-\frac{1}{2\sigma} (\beta - \hat{b})'\hat{V}^{-1}(\beta - \hat{b})\right) d\beta \sigma^{-(\alpha+n+m)/2-1} \exp\left(-\frac{\hat{\delta}}{2\sigma}\right) d\sigma \\
& = \int \int \sigma^{-k/2} \exp\left(-\frac{1}{2\sigma} (\beta - \hat{b})'\hat{V}^{-1}(\beta - \hat{b})\right) d\beta \sigma^{-(\hat{\alpha}+m)/2-1} \exp\left(-\frac{\hat{\delta}}{2\sigma}\right) d\sigma \tag{a.3.10.17}
\end{aligned}$$

with:

$$\hat{\alpha} = \alpha + n + m \tag{a.3.10.18}$$

The first term is the kernel of a multivariate normal distribution; integration hence yields a constant:

$$= \int \sigma^{-\hat{\alpha}/2-1} \exp\left(-\frac{\hat{\delta}}{2\sigma}\right) d\sigma \tag{a.3.10.19}$$

The remaining term is the kernel of an inverse gamma distribution; integration thus yields the reciprocal of the normalization constant:

$$\begin{aligned}
&= \Gamma(\hat{\alpha}/2)(\hat{\delta}/2)^{-\hat{\alpha}/2} \\
&\propto (\hat{\delta}/2)^{-\hat{\alpha}/2} \\
&\propto \hat{\delta}^{-\hat{\alpha}/2} \\
&= (\delta + \hat{y}'\hat{y} + y'y + b'V^{-1}b - \hat{b}'\hat{V}^{-1}\hat{b})^{-\hat{\alpha}/2} \\
&= (\delta + \hat{y}'\hat{y} + y'y + b'V^{-1}b - (V^{-1}b + X'y + \hat{X}'\hat{y})'\hat{V}'\hat{V}^{-1}\hat{V}(V^{-1}b + X'y + \hat{X}'\hat{y}))^{-\hat{\alpha}/2} \\
&= (\delta + \hat{y}'\hat{y} + y'y + b'V^{-1}b - (V^{-1}b + X'y + \hat{X}'\hat{y})'\hat{V}(V^{-1}b + X'y + \hat{X}'\hat{y}))^{-\hat{\alpha}/2} \tag{a.3.10.20}
\end{aligned}$$

Define:

$$\tilde{b} = V^{-1}b + X'y \tag{a.3.10.21}$$

Then (a.3.10.20) rewrites:

$$\begin{aligned}
&= (\delta + \hat{y}'\hat{y} + y'y + b'V^{-1}b - (\tilde{b} + \hat{X}'\hat{y})'\hat{V}(\tilde{b} + \hat{X}'\hat{y}))^{-\hat{\alpha}/2} \\
&= (\delta + \hat{y}'\hat{y} + y'y + b'V^{-1}b - \tilde{b}'\hat{V}\tilde{b} - \hat{y}'\hat{X}\hat{V}\hat{X}'\hat{y} - 2\hat{y}'\hat{X}\hat{V}\tilde{b})^{-\hat{\alpha}/2} \\
&= (\delta + y'y + b'V^{-1}b + \hat{y}'(I_m - \hat{X}\hat{V}\hat{X}')\hat{y} - \tilde{b}'\hat{V}\tilde{b} - 2\hat{y}'\hat{X}\hat{V}\tilde{b})^{-\hat{\alpha}/2} \\
&= ([\delta + y'y + b'V^{-1}b - \tilde{b}'\hat{V}\tilde{b} - \hat{y}'\hat{V}^{-1}\hat{y}] + \hat{y}'(I_m - \hat{X}\hat{V}\hat{X}')\hat{y} - 2\hat{y}'\hat{V}^{-1}\hat{V}\hat{X}\hat{V}\tilde{b} + \hat{y}'\hat{V}^{-1}\hat{y})^{-\hat{\alpha}/2} \tag{a.3.10.22}
\end{aligned}$$

Define:

$$\check{\delta} = \delta + y'y + b'V^{-1}b - \tilde{b}'\hat{V}\tilde{b} - \hat{y}'\hat{V}^{-1}\hat{y} \quad \check{V} = (I_m - \hat{X}\hat{V}\hat{X}')^{-1} \quad \check{y} = \hat{V}\hat{X}\hat{V}\tilde{b} \tag{a.3.10.23}$$

Then (a.3.10.22) rewrites:

$$\begin{aligned}
&= (\check{\delta} + \hat{y}'\check{V}^{-1}\hat{y} - 2\hat{y}'\check{V}^{-1}\hat{y} + \hat{y}'\check{V}^{-1}\hat{y})^{-\hat{\alpha}/2} \\
&= (\check{\delta} + (\hat{y} - \check{y})'\check{V}^{-1}(\hat{y} - \check{y}))^{-\hat{\alpha}/2} \\
&= \check{\delta}^{-\hat{\alpha}/2}(1 + (\hat{y} - \check{y})'[\check{\delta}\check{V}]^{-1}(\hat{y} - \check{y}))^{-\hat{\alpha}/2} \\
&\propto (1 + (\hat{y} - \check{y})'[\check{\delta}\check{V}]^{-1}(\hat{y} - \check{y}))^{-\hat{\alpha}/2} \\
&= \left(1 + \frac{1}{\check{\alpha}}(\hat{y} - \check{y})'[\check{\delta}\check{V}/\check{\alpha}]^{-1}(\hat{y} - \check{y})\right)^{-(\check{\alpha}+m)/2} \tag{a.3.10.24}
\end{aligned}$$

Thus we finally conclude:

$$f(\hat{y}|y) \propto \left(1 + \frac{1}{\check{\alpha}}(\hat{y} - \check{y})'[\check{\delta}\check{V}/\check{\alpha}]^{-1}(\hat{y} - \check{y})\right)^{-(\check{\alpha}+m)/2} \tag{a.3.10.25}$$

Finally, reformulate the messy terms. First, reformulate \check{V} . For this, we make again use of property m.13 (the Sherman-Woodbury-Morrison identity): $(A + BDC)^{-1} = A^{-1} - A^{-1}B(D^{-1} + CA^{-1}B)^{-1}CA^{-1}$.

Then, starting from (a.3.10.23):

$$\begin{aligned}
&\check{V} \\
&= (I_m - \hat{X}\hat{V}\hat{X}')^{-1} \\
&= (I_m - \hat{X}(V^{-1} + X'X + \hat{X}'\hat{X})^{-1}\hat{X}')^{-1} \\
&= I_m + \hat{X}(V^{-1} + X'X)^{-1}\hat{X}' \\
&= I_m + \hat{X}\check{V}\hat{X}' \tag{a.3.10.26}
\end{aligned}$$

Now consider the term \ddot{y} . Start from:

$$\begin{aligned}
& \dot{V}\hat{X}\hat{V} \\
&= (I_m + \hat{X}\bar{V}\hat{X}')\hat{X}\hat{V} \\
&= \hat{X}\hat{V} + \hat{X}\bar{V}\hat{X}'\hat{X}\hat{V} \\
&= \hat{X}(\hat{V} + \bar{V}\hat{X}'\hat{X}\hat{V}) \\
&= \hat{X}(I_m + \bar{V}\hat{X}'\hat{X})\hat{V}
\end{aligned} \tag{a.3.10.27}$$

We then note that (a.3.10.15) implies:

$$\hat{V} = (V^{-1} + X'X + \hat{X}'\hat{X})^{-1} \Leftrightarrow \hat{V} = (\bar{V}^{-1} + \hat{X}'\hat{X})^{-1} \tag{a.3.10.28}$$

Hence:

$$\begin{aligned}
&= \hat{X}(I_m + \bar{V}\hat{X}'\hat{X})(\bar{V}^{-1} + \hat{X}'\hat{X})^{-1} \\
&= \hat{X}\bar{V}(\bar{V}^{-1} + \hat{X}'\hat{X})(\bar{V}^{-1} + \hat{X}'\hat{X})^{-1} \\
&= \hat{X}\bar{V}
\end{aligned} \tag{a.3.10.29}$$

Using this result in (a.3.10.23), and combining with definition (a.3.10.21), we obtain:

$$\ddot{y} = \dot{V}\hat{X}\hat{V}\tilde{b} = \hat{X}\bar{V}\tilde{b} = \hat{X}\bar{b} \tag{a.3.10.30}$$

Finally, reformulate $\ddot{\delta}$. First, note that:

$$\begin{aligned}
& \tilde{b}'\hat{V}\tilde{b} + \dot{y}'\bar{V}^{-1}\dot{y} \\
&= \bar{b}'\bar{V}^{-1}\hat{V}\bar{V}^{-1}\bar{b} + \bar{b}'\hat{X}'(I_m - \hat{X}\hat{V}\hat{X}')\hat{X}\bar{b} \\
&= \bar{b}'[(\hat{V}^{-1} - \hat{X}'\hat{X})'\hat{V}(\hat{V}^{-1} - \hat{X}'\hat{X}) + \hat{X}'\hat{X} - \hat{X}'\hat{X}\hat{V}\hat{X}'\hat{X}]\bar{b} \\
&= \bar{b}'[(\hat{V}^{-1} - \hat{X}'\hat{X})'(I_k - \hat{V}\hat{X}'\hat{X}) + \hat{X}'\hat{X} - \hat{X}'\hat{X}\hat{V}\hat{X}'\hat{X}]\bar{b} \\
&= \bar{b}'[\hat{V}^{-1} - \hat{X}'\hat{X} - \hat{X}'\hat{X} + \hat{X}'\hat{X}\hat{V}\hat{X}'\hat{X} + \hat{X}'\hat{X} - \hat{X}'\hat{X}\hat{V}\hat{X}'\hat{X}]\bar{b} \\
&= \bar{b}'[\hat{V}^{-1} - \hat{X}'\hat{X}]\bar{b} \\
&= \bar{b}'[\bar{V}^{-1} + \hat{X}'\hat{X} - \hat{X}'\hat{X}]\bar{b} \\
&= \bar{b}'\bar{V}^{-1}\bar{b}
\end{aligned} \tag{a.3.10.31}$$

Substituting this in (a.3.10.23) to obtain:

$$\ddot{\delta} = \delta + \dot{y}'\dot{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \tilde{b}'\hat{V}\tilde{b} - \dot{y}'\bar{V}^{-1}\dot{y} = \delta + \dot{y}'\dot{y} + \bar{b}'\bar{V}^{-1}\bar{b} - \bar{b}'\bar{V}^{-1}\bar{b} = \delta \tag{a.3.10.32}$$

Eventually substituting for (a.3.10.26), (a.3.10.30) and (a.3.10.32) in (a.3.10.25) yields:

$$f(\hat{y}|y) \propto \left(1 + \frac{1}{\bar{\alpha}}(\hat{y} - \hat{X}\bar{b})'[\bar{\delta}(I_m + \hat{X}\bar{V}\hat{X}')/\bar{\alpha}]^{-1}(\hat{y} - \hat{X}\bar{b})\right)^{-(\bar{\alpha}+m)/2} \tag{a.3.10.33}$$

derivations for equation (3.10.10)

The log likelihood function is given by:

$$\log(f(y|\beta, \sigma)) = -\frac{n}{2} \log(2\pi) - \frac{n}{2} \log(\sigma) - \frac{1}{2} \frac{(y - X\beta)'(y - X\beta)}{\sigma} \quad (\text{a.3.10.34})$$

The function is estimated at the maximum likelihood values. Hence $\beta = \hat{\beta}$ and $\sigma = \hat{\sigma} = \frac{\hat{\epsilon}'\hat{\epsilon}}{n}$.

Substituting in (a.3.10.34):

$$\begin{aligned} & \log(f(y|\hat{\beta}, \hat{\sigma})) \\ &= -\frac{n}{2} \log(2\pi) - \frac{n}{2} \log(\hat{\sigma}) - \frac{1}{2} \frac{(y - X\hat{\beta})'(y - X\hat{\beta})}{\hat{\sigma}} \\ &= -\frac{n}{2} \log(2\pi) - \frac{n}{2} \log\left(\frac{\hat{\epsilon}'\hat{\epsilon}}{n}\right) - \frac{1}{2} \frac{n \hat{\epsilon}'\hat{\epsilon}}{\hat{\epsilon}'\hat{\epsilon}} \\ &= -\frac{n}{2} \log(2\pi) - \frac{n}{2} \log(\hat{\epsilon}'\hat{\epsilon}/n) - \frac{n}{2} \\ &= -\frac{n}{2} [\log(2\pi) + \log(\hat{\epsilon}'\hat{\epsilon}/n) + 1] \end{aligned} \quad (\text{a.3.10.35})$$

Noting then that $|\theta| = k + 1$ (corresponding to k coefficients for β and one for σ), AIC obtains as:

$$\begin{aligned} AIC &= 2|\theta|/n - 2\hat{L}/n \\ &= 2(k+1)/n - 2\left(-\frac{n}{2} [\log(2\pi) + \log(\hat{\epsilon}'\hat{\epsilon}/n) + 1]\right)/n \\ &= 2(k+1)/n + \log(2\pi) + \log(\hat{\epsilon}'\hat{\epsilon}/n) + 1 \end{aligned} \quad (\text{a.3.10.36})$$

Using similar calculations, BIC immediately obtains as:

$$BIC = (k+1) \log(n)/n + \log(2\pi) + \log(\hat{\epsilon}'\hat{\epsilon}/n) + 1 \quad (\text{a.3.10.37})$$

derivations for equation (3.10.19)

Rearrange:

$$\begin{aligned}
& f(y) \\
&= \int (2\pi\sigma)^{-n/2} \exp\left(-\frac{1}{2} \frac{(y-X\beta)'(y-X\beta)}{\sigma}\right) \times (2\pi)^{-k/2} |V|^{-1/2} \exp\left(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b)\right) d\beta \\
&= \int (2\pi)^{-(n+k)/2} \sigma^{-n/2} |V|^{-1/2} \times \exp\left(-\frac{1}{2} [(y-X\beta)'\sigma^{-1}(y-X\beta) + (\beta-b)'V^{-1}(\beta-b)]\right) d\beta
\end{aligned} \tag{a.3.10.38}$$

Consider the term square brackets:

$$\begin{aligned}
& (y-X\beta)'\sigma^{-1}(y-X\beta) + (\beta-b)'V^{-1}(\beta-b) \\
&= y'\sigma^{-1}y + \beta'X'\sigma^{-1}X\beta - 2\beta'X'\sigma^{-1}y + \beta'V^{-1}\beta + b'V^{-1}b - 2\beta'V^{-1}b \\
&= \beta'(V^{-1} + \sigma^{-1}X'X)\beta - 2\beta'(V^{-1}b + \sigma^{-1}X'y) + y'\sigma^{-1}y + b'V^{-1}b \\
&= \beta'(V^{-1} + \sigma^{-1}X'X)\beta - 2\beta'\bar{V}^{-1}\bar{V}(V^{-1}b + \sigma^{-1}X'y) + \bar{b}\bar{V}^{-1}\bar{b} + y'\sigma^{-1}y + b'V^{-1}b - \bar{b}\bar{V}^{-1}\bar{b} \tag{a.3.10.39}
\end{aligned}$$

Define:

$$\bar{V} = (V^{-1} + \sigma^{-1}X'X)^{-1} \quad \bar{b} = \bar{V}(V^{-1}b + \sigma^{-1}X'y) \tag{a.3.10.40}$$

Then (a.3.10.39) reformulates:

$$\begin{aligned}
&= \beta'\bar{V}^{-1}\beta - 2\beta'\bar{V}^{-1}\bar{b} + \bar{b}\bar{V}^{-1}\bar{b} + y'\sigma^{-1}y + b'V^{-1}b - \bar{b}\bar{V}^{-1}\bar{b} \\
&= (\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b}) + y'\sigma^{-1}y + b'V^{-1}b - \bar{b}\bar{V}^{-1}\bar{b}
\end{aligned} \tag{a.3.10.41}$$

Substitute back in (a.3.10.38):

$$\begin{aligned}
& f(y) \\
&= \int (2\pi)^{-(n+k)/2} \sigma^{-n/2} |V|^{-1/2} \times \exp\left(-\frac{1}{2} [(\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b}) + y'\sigma^{-1}y + b'V^{-1}b - \bar{b}\bar{V}^{-1}\bar{b}]\right) d\beta \\
&= (2\pi)^{-(n+k)/2} \sigma^{-n/2} |V|^{-1/2} (2\pi)^{k/2} |\bar{V}|^{1/2} \times \exp\left(-\frac{1}{2} [y'\sigma^{-1}y + b'V^{-1}b - \bar{b}\bar{V}^{-1}\bar{b}]\right) \\
&\times \int (2\pi)^{-k/2} |\bar{V}|^{-1/2} \exp\left(-\frac{1}{2}(\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b})\right) d\beta \\
&= (2\pi)^{-n/2} \sigma^{-n/2} |\bar{V}|^{1/2} |V|^{-1/2} \times \exp\left(-\frac{1}{2} [y'\sigma^{-1}y + b'V^{-1}b - \bar{b}\bar{V}^{-1}\bar{b}]\right) \\
&\times \int (2\pi)^{-k/2} |\bar{V}|^{-1/2} \exp\left(-\frac{1}{2}(\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b})\right) d\beta
\end{aligned} \tag{a.3.10.42}$$

derivations for equation (3.10.21)

Consider the term:

$$\begin{aligned}
& (2\pi)^{-n/2} \sigma^{-n/2} |\bar{V}|^{1/2} |V|^{-1/2} \\
&= (2\pi)^{-n/2} \sigma^{-n/2} |(V^{-1} + \sigma^{-1}X'X)^{-1}|^{1/2} |V|^{-1/2} \\
&= (2\pi)^{-n/2} \sigma^{-n/2} |V^{-1} + \sigma^{-1}X'X|^{-1/2} |V|^{-1/2} \\
&= (2\pi)^{-n/2} \sigma^{-n/2} (|V||V^{-1} + \sigma^{-1}X'X|)^{-1/2} \\
&= (2\pi)^{-n/2} \sigma^{-n/2} |I_k + \sigma^{-1}VX'X|^{-1/2}
\end{aligned} \tag{a.3.10.43}$$

Hence:

$$f(y) = (2\pi)^{-n/2} \sigma^{-n/2} |I_k + \sigma^{-1}VX'X|^{-1/2} \exp\left(-\frac{1}{2} [y'\sigma^{-1}y + b'V^{-1}b - \bar{b}\bar{V}^{-1}\bar{b}]\right) \tag{a.3.10.44}$$

derivations for equation (3.10.23)

Rearrange the expression:

$$\begin{aligned}
& f(y) \\
&= \int \int (2\pi\sigma)^{-n/2} \exp\left(-\frac{1}{2} \frac{(y - X\beta)'(y - X\beta)}{\sigma}\right) \\
&\times (2\pi)^{-k/2} |\sigma V|^{-1/2} \exp\left(-\frac{1}{2} (\beta - b)'(\sigma V)^{-1}(\beta - b)\right) \times \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right) d\beta d\sigma \\
&= \int \int (2\pi\sigma)^{-n/2} (2\pi)^{-k/2} |\sigma V|^{-1/2} \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \sigma^{-\alpha/2-1} \\
&\times \exp\left(-\frac{1}{2\sigma} [(y - X\beta)'(y - X\beta) + (\beta - b)'V^{-1}(\beta - b) + \delta]\right) d\beta d\sigma
\end{aligned} \tag{a.3.10.45}$$

Consider the term in square brackets and complete the squares:

$$\begin{aligned}
& (y - X\beta)'(y - X\beta) + (\beta - b)'V^{-1}(\beta - b) + \delta \\
&= y'y + \beta'X'X\beta - 2\beta'X'y + \beta'V^{-1}\beta + b'V^{-1}b - 2\beta'V^{-1}b + \delta \\
&= \beta'(V^{-1} + X'X)\beta - 2\beta'(V^{-1}b + X'y) + \delta + y'y + b'V^{-1}b \\
&= \beta'(V^{-1} + X'X)\beta - 2\beta'\bar{V}^{-1}\bar{V}(V^{-1}b + X'y) + \delta + y'y + b'V^{-1}b + \bar{b}'\bar{V}^{-1}\bar{b} - \bar{b}'\bar{V}^{-1}\bar{b}
\end{aligned} \tag{a.3.10.46}$$

Define:

$$\bar{V} = (V^{-1} + X'X)^{-1} \quad \bar{b} = \bar{V}(V^{-1}b + X'y) \quad \bar{\delta} = \delta + y'y + b'V^{-1}b - \bar{b}'\bar{V}^{-1}\bar{b} \tag{a.3.10.47}$$

Then (a.3.10.46) rewrites:

$$\begin{aligned}
&= \beta'\bar{V}^{-1}\beta - 2\beta'\bar{V}^{-1}\bar{b} + \bar{b}'\bar{V}^{-1}\bar{b} + \bar{\delta} \\
&= (\beta - \bar{b})'\bar{V}^{-1}(\beta - \bar{b}) + \bar{\delta}
\end{aligned} \tag{a.3.10.48}$$

Substituting back in (a.3.10.45):

$$\begin{aligned}
& f(y) \\
&= \int \int (2\pi\sigma)^{-n/2} (2\pi)^{-k/2} |\sigma V|^{-1/2} \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \sigma^{-\alpha/2-1} \\
&\times \exp\left(-\frac{1}{2\sigma} [(\beta - \bar{b})' \bar{V}^{-1} (\beta - \bar{b}) + \bar{\delta}]\right) d\beta d\sigma \\
&= \int \int (2\pi)^{-n/2} (2\pi)^{-k/2} |\sigma V|^{-1/2} \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \\
&\times \sigma^{-(\alpha+n)/2-1} \exp\left(-\frac{1}{2\sigma} [(\beta - \bar{b})' \bar{V}^{-1} (\beta - \bar{b}) + \bar{\delta}]\right) d\beta d\sigma
\end{aligned} \tag{a.3.10.49}$$

define:

$$\bar{\alpha} = \alpha + n \tag{a.3.10.50}$$

Then (a.3.10.49) rewrites:

$$\begin{aligned}
&= \int \int (2\pi)^{-n/2} (2\pi)^{-k/2} |\sigma V|^{-1/2} \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \\
&\times \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{1}{2\sigma} [(\beta - \bar{b})' \bar{V}^{-1} (\beta - \bar{b}) + \bar{\delta}]\right) d\beta d\sigma \\
&= (2\pi)^{-n/2} |\sigma V|^{-1/2} |\sigma \bar{V}|^{1/2} \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \frac{\Gamma(\bar{\alpha}/2)}{\bar{\delta}/2^{\bar{\alpha}/2}} \\
&\times \int \int (2\pi)^{-k/2} |\sigma \bar{V}|^{-1/2} \exp\left(-\frac{1}{2} (\beta - \bar{b})' (\sigma \bar{V})^{-1} (\beta - \bar{b})\right) \times \frac{\bar{\delta}/2^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)} \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right) d\beta d\sigma \\
&= 2^{-n/2} \pi^{-n/2} |V|^{-1/2} |\bar{V}|^{1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\bar{\alpha}/2}} \frac{2^{(\alpha+n)/2}}{2^{\alpha/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \\
&\times \int \int (2\pi)^{-k/2} |\sigma \bar{V}|^{-1/2} \exp\left(-\frac{1}{2} (\beta - \bar{b})' (\sigma \bar{V})^{-1} (\beta - \bar{b})\right) \times \frac{\bar{\delta}/2^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)} \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right) d\beta d\sigma \\
&= \pi^{-n/2} |V|^{-1/2} |\bar{V}|^{1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\bar{\alpha}/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \\
&\times \int \int (2\pi)^{-k/2} |\sigma \bar{V}|^{-1/2} \exp\left(-\frac{1}{2} (\beta - \bar{b})' (\sigma \bar{V})^{-1} (\beta - \bar{b})\right) \times \frac{\bar{\delta}/2^{\bar{\alpha}/2}}{\Gamma(\bar{\alpha}/2)} \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right) d\beta d\sigma
\end{aligned} \tag{a.3.10.51}$$

derivations for equation (3.10.25)

Reformulate the expression:

$$\begin{aligned}
& \pi^{-n/2} |V|^{-1/2} |\bar{V}|^{1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\alpha/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \\
= & \pi^{-n/2} |V|^{-1/2} |(V^{-1} + X'X)^{-1}|^{1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\alpha/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \\
= & \pi^{-n/2} |V|^{-1/2} |(V^{-1} + X'X)|^{-1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\alpha/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \\
= & \pi^{-n/2} |V(V^{-1} + X'X)|^{-1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\alpha/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \\
= & \pi^{-n/2} |I_k + VX'X|^{-1/2} \frac{\delta^{\alpha/2}}{\bar{\delta}^{\alpha/2}} \frac{\Gamma(\bar{\alpha}/2)}{\Gamma(\alpha/2)} \tag{a.3.10.52}
\end{aligned}$$

derivations for equation (3.10.27)

Rearrange the expression:

$$\begin{aligned}
& f(y) \\
& \approx \frac{f(y|\beta^*, \sigma^*)\pi(\beta^*, \sigma^*)}{\pi(\sigma^*|y, \beta^*) \times \frac{1}{j} \sum_{j=1}^J \pi(\beta^*|\sigma^{(j)}, y)} \\
& = (2\pi\sigma)^{-n/2} \exp\left(-\frac{1}{2} \frac{(y-X\beta)'(y-X\beta)}{\sigma}\right) \times \frac{(2\pi)^{-k/2}|V|^{-1/2} \exp(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b))}{\frac{1}{j} \sum_{j=1}^J (2\pi)^{-k/2}|\bar{V}|^{-1/2} \exp(-\frac{1}{2}(\beta-\bar{b})'\bar{V}^{-1}(\beta-\bar{b}))} \\
& \times \frac{\frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right)}{\frac{\bar{\delta}/2^{\alpha/2}}{\Gamma(\bar{\alpha}/2)} \sigma^{-\bar{\alpha}/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right)} \\
& = 2^{-n/2} \pi^{-n/2} \sigma^{-n/2} \exp\left(-\frac{1}{2} \frac{(y-X\beta)'(y-X\beta)}{\sigma}\right) \times \frac{(2\pi)^{-k/2}|V|^{-1/2} \exp(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b))}{\frac{1}{j} \sum_{j=1}^J (2\pi)^{-k/2}|\bar{V}|^{-1/2} \exp(-\frac{1}{2}(\beta-\bar{b})'\bar{V}^{-1}(\beta-\bar{b}))} \\
& \times \frac{\Gamma(\bar{\alpha}/2) 2^{(\alpha+n)/2}}{\Gamma(\alpha/2) 2^{\alpha/2}} \frac{\delta^{\alpha/2} \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right)}{\bar{\delta}^{\bar{\alpha}/2} \sigma^{-(\alpha+n)/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right)} \\
& = 2^{-n/2} \pi^{-n/2} \sigma^{-n/2} \times \frac{(2\pi)^{-k/2}|V|^{-1/2} \exp(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b))}{\frac{1}{j} \sum_{j=1}^J (2\pi)^{-k/2}|\bar{V}|^{-1/2} \exp(-\frac{1}{2}(\beta-\bar{b})'\bar{V}^{-1}(\beta-\bar{b}))} \\
& \times \frac{\Gamma(\bar{\alpha}/2) 2^{(\alpha+n)/2}}{\Gamma(\alpha/2) 2^{\alpha/2}} \frac{\delta^{\alpha/2} \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta+(y-X\beta)'(y-X\beta)}{2\sigma}\right)}{\bar{\delta}^{\bar{\alpha}/2} \sigma^{-(\alpha+n)/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right)} \\
& = 2^{-n/2} \pi^{-n/2} \sigma^{-n/2} \times \frac{(2\pi)^{-k/2}|V|^{-1/2} \exp(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b))}{\frac{1}{j} \sum_{j=1}^J (2\pi)^{-k/2}|\bar{V}|^{-1/2} \exp(-\frac{1}{2}(\beta-\bar{b})'\bar{V}^{-1}(\beta-\bar{b}))} \\
& \times \frac{\Gamma(\bar{\alpha}/2) 2^{(\alpha+n)/2}}{\Gamma(\alpha/2) 2^{\alpha/2}} \frac{\delta^{\alpha/2} \sigma^{-\alpha/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right)}{\bar{\delta}^{\bar{\alpha}/2} \sigma^{-(\alpha+n)/2-1} \exp\left(-\frac{\bar{\delta}}{2\sigma}\right)} \\
& = \pi^{-n/2} \frac{|V|^{-1/2} \exp(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b))}{\frac{1}{j} \sum_{j=1}^J |\bar{V}|^{-1/2} \exp(-\frac{1}{2}(\beta-\bar{b})'\bar{V}^{-1}(\beta-\bar{b}))} \frac{\Gamma(\bar{\alpha}/2) \delta^{\alpha/2}}{\Gamma(\alpha/2) \bar{\delta}^{\bar{\alpha}/2}} \\
& = \pi^{-n/2} \frac{\exp(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b))}{\frac{1}{j} \sum_{j=1}^J |V|^{1/2} |\bar{V}|^{-1/2} \exp(-\frac{1}{2}(\beta-\bar{b})'\bar{V}^{-1}(\beta-\bar{b}))} \frac{\Gamma(\bar{\alpha}/2) \delta^{\alpha/2}}{\Gamma(\alpha/2) \bar{\delta}^{\bar{\alpha}/2}} \\
& = \pi^{-n/2} \frac{\exp(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b))}{\frac{1}{j} \sum_{j=1}^J |V|^{1/2} |V^{-1} + \sigma^{-1}X'X|^{-1/2} \exp(-\frac{1}{2}(\beta-\bar{b})'\bar{V}^{-1}(\beta-\bar{b}))} \frac{\Gamma(\bar{\alpha}/2) \delta^{\alpha/2}}{\Gamma(\alpha/2) \bar{\delta}^{\bar{\alpha}/2}} \\
& = \pi^{-n/2} \frac{\exp(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b))}{\frac{1}{j} \sum_{j=1}^J |V|^{1/2} |V^{-1} + \sigma^{-1}X'X|^{1/2} \exp(-\frac{1}{2}(\beta-\bar{b})'\bar{V}^{-1}(\beta-\bar{b}))} \frac{\Gamma(\bar{\alpha}/2) \delta^{\alpha/2}}{\Gamma(\alpha/2) \bar{\delta}^{\bar{\alpha}/2}} \\
& = \pi^{-n/2} \frac{\exp(-\frac{1}{2}(\beta-b)'V^{-1}(\beta-b))}{\frac{1}{j} \sum_{j=1}^J |I_k + \sigma^{-1}VX'X|^{1/2} \exp(-\frac{1}{2}(\beta-\bar{b})'\bar{V}^{-1}(\beta-\bar{b}))} \frac{\Gamma(\bar{\alpha}/2) \delta^{\alpha/2}}{\Gamma(\alpha/2) \bar{\delta}^{\bar{\alpha}/2}} \tag{a.3.10.53}
\end{aligned}$$

derivations for equation (3.10.29)

Substitute for the functions and rearrange:

$$\begin{aligned}
& \frac{1}{f(y)} \\
& \approx \frac{1}{J} \sum_{j=1}^J \frac{g(\theta^{(j)})}{f(y|\beta^{(j)}, \sigma^{(j)}, \gamma^{(j)}) \pi(\beta^{(j)}) \pi(\sigma^{(j)}) \pi(\gamma^{(j)})} \\
& = \frac{1}{J} \sum_{j=1}^J \frac{\omega^{-1} (2\pi)^{-(k+h+1)/2} |\hat{\Sigma}|^{-1/2} \exp\left(-\frac{1}{2}(\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta})\right) \mathbb{1}(\theta \in \hat{\Theta})}{\left[(2\pi\sigma)^{-n/2} |W|^{-1/2} \exp\left(-\frac{1}{2} \frac{(y-X\beta)' W^{-1} (y-X\beta)}{\sigma}\right) \times (2\pi)^{-k/2} |V|^{-1/2} \exp\left(-\frac{1}{2}(\beta - b)' V^{-1} (\beta - b)\right) \right.} \\
& \quad \left. \times \frac{\delta/2\alpha/2}{\Gamma(\alpha/2)} \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right) \times (2\pi)^{-h/2} |Q|^{-1/2} \exp\left(-\frac{1}{2}(\gamma - g)' Q^{-1} (\gamma - g)\right) \right]} \\
& = \frac{1}{J} \sum_{j=1}^J \mathbb{1}(\theta \in \hat{\Theta}) \omega^{-1} (2\pi)^{(n+k+h-(k+h+1))/2} |\hat{\Sigma}|^{-1/2} |W|^{1/2} |V|^{1/2} |Q|^{1/2} \frac{\Gamma(\alpha/2)}{\delta/2\alpha/2} \sigma^{(\alpha+n)/2+1} \\
& \quad \times \exp\left(\frac{1}{2} \left[(y-X\beta)' (\sigma W)^{-1} (y-X\beta) + (\beta - b)' V^{-1} (\beta - b) + \delta\sigma^{-1} \right. \right. \\
& \quad \left. \left. + (\gamma - g)' Q^{-1} (\gamma - g) - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right) \\
& = (\omega J)^{-1} (2\pi)^{(n-1)/2} |\hat{\Sigma}|^{-1/2} |V|^{1/2} |Q|^{1/2} \frac{\Gamma(\alpha/2)}{\delta/2\alpha/2} \\
& \quad \times \sum_{j=1}^J \mathbb{1}(\theta \in \hat{\Theta}) |W|^{1/2} \sigma^{(\alpha+n)/2+1} \exp\left(\frac{1}{2} \left[(y-X\beta)' (\sigma W)^{-1} (y-X\beta) + (\beta - b)' V^{-1} (\beta - b) \right. \right. \\
& \quad \left. \left. + \delta\sigma^{-1} + (\gamma - g)' Q^{-1} (\gamma - g) - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right) \tag{a.3.10.54}
\end{aligned}$$

Using logs on both sides yields:

$$\begin{aligned}
& -\log(f(y)) \approx \log\left((\omega J)^{-1} (2\pi)^{(n-1)/2} |\hat{\Sigma}|^{-1/2} |V|^{1/2} |Q|^{1/2} \frac{\Gamma(\alpha/2)}{\delta/2\alpha/2}\right) \\
& + \log\left(\sum_{j=1}^J \mathbb{1}(\theta \in \hat{\Theta}) |W|^{1/2} \sigma^{(\alpha+n)/2+1} \exp\left(\frac{1}{2} \left[(y-X\beta)' (\sigma W)^{-1} (y-X\beta) + (\beta - b)' V^{-1} (\beta - b) \right. \right. \right. \\
& \quad \left. \left. \left. + \delta\sigma^{-1} + (\gamma - g)' Q^{-1} (\gamma - g) - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right)\right) \tag{a.3.10.55}
\end{aligned}$$

or:

$$\begin{aligned}
& \log(f(y)) \approx -\log\left((\omega J)^{-1} (2\pi)^{(n-1)/2} |\hat{\Sigma}|^{-1/2} |V|^{1/2} |Q|^{1/2} \frac{\Gamma(\alpha/2)}{\delta/2\alpha/2}\right) \\
& -\log\left(\sum_{j=1}^J \mathbb{1}(\theta \in \hat{\Theta}) |W|^{1/2} \sigma^{(\alpha+n)/2+1} \exp\left(\frac{1}{2} \left[(y-X\beta)' (\sigma W)^{-1} (y-X\beta) + (\beta - b)' V^{-1} (\beta - b) \right. \right. \right. \\
& \quad \left. \left. \left. + \delta\sigma^{-1} + (\gamma - g)' Q^{-1} (\gamma - g) - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right)\right) \tag{a.3.10.56}
\end{aligned}$$

derivations for equation (3.10.32)

Substitute for the functions and rearrange:

$$\begin{aligned}
& \frac{1}{f(y)} \\
& \approx \frac{1}{J} \sum_{j=1}^J \frac{g(\theta^{(j)})}{f(y|\beta^{(j)}, \sigma^{(j)}, \phi^{(j)}) \pi(\beta^{(j)}) \pi(\sigma^{(j)}) \pi(\phi^{(j)})} \\
& = \frac{1}{J} \sum_{j=1}^J \frac{\omega^{-1} (2\pi)^{-(k+q+1)/2} |\hat{\Sigma}|^{-1/2} \exp(-\frac{1}{2}(\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta})) \mathbb{1}(\theta \in \hat{\Theta})}{\left[(2\pi\sigma)^{-T/2} \exp(-\frac{1}{2}(\varepsilon - E\phi)' \sigma^{-1} (\varepsilon - E\phi)) \times (2\pi)^{-k/2} |V|^{-1/2} \exp(-\frac{1}{2}(\beta - b)' V^{-1} (\beta - b)) \right.} \\
& \quad \left. \times \frac{\delta/2^{\alpha/2}}{\Gamma(\alpha/2)} \sigma^{-\alpha/2-1} \exp\left(-\frac{\delta}{2\sigma}\right) \times (2\pi)^{-q/2} |Z|^{-1/2} \exp(-\frac{1}{2}(\phi - p)' Z^{-1} (\phi - p)) \right]} \\
& = \frac{1}{J} \sum_{j=1}^J \mathbb{1}(\theta \in \hat{\Theta}) \omega^{-1} (2\pi)^{(T+k+q-(k+q+1))/2} |\hat{\Sigma}|^{-1/2} |V|^{1/2} |Z|^{1/2} \frac{\Gamma(\alpha/2)}{\delta/2^{\alpha/2}} \sigma^{(\alpha+T)/2+1} \\
& \quad \times \exp\left(\frac{1}{2} \left[(\varepsilon - E\phi)' \sigma^{-1} (\varepsilon - E\phi) + (\beta - b)' V^{-1} (\beta - b) \right. \right. \\
& \quad \left. \left. + \delta \sigma^{-1} + (\phi - p)' Z^{-1} (\phi - p) - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right) \\
& = (\omega J)^{-1} (2\pi)^{(T-1)/2} |\hat{\Sigma}|^{-1/2} |V|^{1/2} |Z|^{1/2} \frac{\Gamma(\alpha/2)}{\delta/2^{\alpha/2}} \\
& \quad \times \sum_{j=1}^J \mathbb{1}(\theta \in \hat{\Theta}) \sigma^{(\alpha+T)/2+1} \exp\left(\frac{1}{2} \left[(\varepsilon - E\phi)' \sigma^{-1} (\varepsilon - E\phi) + (\beta - b)' V^{-1} (\beta - b) \right. \right. \\
& \quad \left. \left. + \delta \sigma^{-1} + (\phi - p)' Z^{-1} (\phi - p) - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right)
\end{aligned} \tag{a.3.10.57}$$

Using logs on both sides yields:

$$\begin{aligned}
& -\log(f(y)) \approx \log\left((\omega J)^{-1} (2\pi)^{(T-1)/2} |\hat{\Sigma}|^{-1/2} |V|^{1/2} |Z|^{1/2} \frac{\Gamma(\alpha/2)}{\delta/2^{\alpha/2}}\right) \\
& + \log\left(\sum_{j=1}^J \mathbb{1}(\theta \in \hat{\Theta}) \sigma^{(\alpha+T)/2+1} \exp\left(\frac{1}{2} \left[(\varepsilon - E\phi)' \sigma^{-1} (\varepsilon - E\phi) + (\beta - b)' V^{-1} (\beta - b) \right. \right. \right. \\
& \quad \left. \left. + \delta \sigma^{-1} + (\phi - p)' Z^{-1} (\phi - p) - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right)\right)
\end{aligned} \tag{a.3.10.58}$$

or:

$$\begin{aligned}
& \log(f(y)) \approx -\log\left((\omega J)^{-1} (2\pi)^{(T-1)/2} |\hat{\Sigma}|^{-1/2} |V|^{1/2} |Z|^{1/2} \frac{\Gamma(\alpha/2)}{\delta/2^{\alpha/2}}\right) \\
& -\log\left(\sum_{j=1}^J \mathbb{1}(\theta \in \hat{\Theta}) \sigma^{(\alpha+T)/2+1} \exp\left(\frac{1}{2} \left[(\varepsilon - E\phi)' \sigma^{-1} (\varepsilon - E\phi) + (\beta - b)' V^{-1} (\beta - b) \right. \right. \right. \\
& \quad \left. \left. + \delta \sigma^{-1} + (\phi - p)' Z^{-1} (\phi - p) - (\theta - \hat{\theta})' \hat{\Sigma}^{-1} (\theta - \hat{\theta}) \right] \right)\right)
\end{aligned} \tag{a.3.10.59}$$

Bibliography

