

# Supplementary material

October 7, 2025

## 1 Gauss-Hermite quadrature

The univariate Gauss-Hermite quadrature approximates any integral of the form

$$\int_{-\infty}^{+\infty} \exp(-z^2)g(z) \, dz,$$

by a weighted sum,

$$\int_{-\infty}^{+\infty} \exp(-z^2)g(z) \, dz \approx \sum_{k=1}^Q w_k g(z_k), \quad (1)$$

where  $g(z) : \mathbb{R} \rightarrow \mathbb{R}$ ,  $Q$  denotes the order of the approximation, the  $z_k$  are the zeros of the  $Q$ th order Hermite polynomial,

$$H_Q(z) = (-1)^Q e^{z^2/2} \frac{d^Q e^{-z^2/2}}{dz^Q},$$

and the  $w_k$  are the corresponding weights.

The multivariate Gauss-Hermite quadrature considers the  $m$ -dimensional vector  $\mathbf{z} = (z_1, \dots, z_m)$  and the function  $g(\mathbf{z}) : \mathbb{R}^m \rightarrow \mathbb{R}$ . The expression (1) may be applied to approximate the  $m$ -dimensional integral

$$\int_{\mathbb{R}^m} w(\mathbf{z})g(\mathbf{z}) \, d\mathbf{z} = \int_{\mathbb{R}} \cdots \int_{\mathbb{R}} w(z_1, \dots, z_m)g(z_1, \dots, z_m) \, dz_1 \cdots dz_m, \quad (2)$$

where  $w(z_1, \dots, z_m) = \exp(-z_1^2) \cdots \exp(-z_m^2)$ .

Then, the univariate Gauss-Hermite quadrature applies to each of the components of  $\mathbf{z}$  in turn. Using  $Q_r$  nodes in the  $r$ th dimension,  $r = 1, \dots, m$ , yields the approximation

$$\int_{\mathbb{R}^m} w(\mathbf{z})g(\mathbf{z}) \, d\mathbf{z} \approx \sum_{k_1=1}^{Q_1} w_{k_1}^{(1)} \cdots \sum_{k_m=1}^{Q_m} w_{k_m}^{(m)} g(z_{k_1}^{(1)}, \dots, z_{k_m}^{(m)}), \quad (3)$$

where  $z_{k_r}^{(r)}$  is the  $k_r$ th zero of the Hermite polynomial with degree  $Q_r$  and  $w_{k_r}^{(r)}$  is the corresponding weight. The  $z_{k_r}^{(r)}$  are also denoted as quadrature nodes or quadrature points.

In equation (6), we need to evaluate the following integral:

$$\begin{aligned} I_i &= \int \int_{\mathbb{R}^2} \prod_{j=1}^{n_i} f(y_{ij} \mid \gamma_{i1}, \gamma_{i2}; \beta_1, \beta_2) \cdot f(\gamma_{i1}; \lambda_1) f(\gamma_{i2}; \lambda_2) d\gamma_{i1} d\gamma_{i2} \quad (4) \\ &= \int \int_{\mathbb{R}^2} \prod_{j=1}^{n_i} f(y_{ij} \mid \gamma_{i1}, \gamma_{i2}; \beta_1, \beta_2) \frac{e^{-\frac{\gamma_{i1}^2}{2\lambda_1^2}}}{\sqrt{2\pi\lambda_1}} \frac{e^{-\frac{\gamma_{i2}^2}{2\lambda_2^2}}}{\sqrt{2\pi\lambda_2}} d\gamma_{i1} d\gamma_{i2}. \end{aligned}$$

Performing the following change of variables:

$$b_{i1} = \frac{\gamma_{i1}}{\sqrt{2}\lambda_1}, \quad b_{i2} = \frac{\gamma_{i2}}{\sqrt{2}\lambda_2},$$

we obtain the integral  $I_i$  as follows:

$$I_i = \int \int_{\mathbb{R}^2} \prod_{j=1}^{n_i} f(y_{ij} \mid \sqrt{2}\lambda_1 b_{i1}, \sqrt{2}\lambda_2 b_{i2}; \beta_1, \beta_2) \frac{\exp(-b_{i1}^2) \exp(-b_{i2}^2)}{\pi} db_{i1} db_{i2}, \quad (5)$$

which has the form (2). Using the multivariate Gauss-Hermite quadrature, the integral  $I_i$  is approximated as

$$I_i \cong \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} \prod_{j=1}^{n_i} f(y_{ij} \mid \sqrt{2}\lambda_1 z_{k_1}, \sqrt{2}\lambda_2 z_{k_2}; \beta_1, \beta_2) \frac{w_{k_1} w_{k_2}}{\pi},$$

where  $z_{k_1}$  and  $z_{k_2}$  are the quadrature nodes and  $w_{k_1}$  and  $w_{k_2}$  are the corresponding weights.

## 2 Score function

The score function is given by  $\mathbf{U}(\boldsymbol{\theta}) = (\mathbf{U}_{\beta_1}^T(\boldsymbol{\theta}), \mathbf{U}_{\beta_2}^T(\boldsymbol{\theta}), U_{\lambda_1}(\boldsymbol{\theta}), U_{\lambda_2}(\boldsymbol{\theta}))^T$ , with

$$\begin{aligned}
\mathbf{U}_{\beta_1}(\boldsymbol{\theta}) &= \sum_{i=1}^N \frac{\mathbf{X}_{i1}^T \mathbf{D}_{i1}}{B_i}, \\
\mathbf{U}_{\beta_2}(\boldsymbol{\theta}) &= \sum_{i=1}^N \frac{\mathbf{X}_{i2}^T \mathbf{D}_{i2}}{B_i}, \\
U_{\lambda_1}(\boldsymbol{\theta}) &= \sum_{i=1}^N \frac{D_{i3}}{B_i}, \\
U_{\lambda_2}(\boldsymbol{\theta}) &= \sum_{i=1}^N \frac{D_{i4}}{B_i},
\end{aligned}$$

where  $\mathbf{X}_{i1}$  and  $\mathbf{X}_{i2}$  are the  $(n_i \times p_1)$  and  $(n_i \times p_2)$  matrices containing co-variates related to the models for  $\mu$  and  $\sigma$ , respectively. The  $n_i$ -dimensional vectors  $\mathbf{D}_{i1}$  and  $\mathbf{D}_{i2}$  and the quantities  $D_{i3}$ ,  $D_{i4}$ , and  $B_i$  are defined as follows.

$$\text{Let } B_i = \left( \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i w_{k_1} w_{k_2} / \pi \right) \text{ and}$$

$$\begin{aligned}
A_i &= \exp \left( \sum_{j=1}^{n_i} \log \Gamma \left( \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \right) - \log \Gamma \left( \mu_{ij} \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \right) - \log \Gamma \left( (1 - \mu_{ij}) \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \right) \right. \\
&\quad \left. + \left( \mu_{ij} \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} - 1 \right) \log y_{ij} + \left( (1 - \mu_{ij}) \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} - 1 \right) \log (1 - y_{ij}) \right),
\end{aligned}$$

with  $g_1(\mu_{ij}) = \eta_{ij1} = \mathbf{x}_{ij1}^T \boldsymbol{\beta}_1 + \sqrt{2} \lambda_1 z_{k_1}$  and  $g_2(\sigma_{ij}) = \eta_{ij2} = \mathbf{x}_{ij2}^T \boldsymbol{\beta}_2 + \sqrt{2} \lambda_2 z_{k_2}$ .

The  $n_i$ -dimensional vector  $\mathbf{D}_{i1}$  defined in  $\mathbf{U}_{\beta_1}(\boldsymbol{\theta})$  is expressed as

$$\mathbf{D}_{i1} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i \mathbf{Q}_{i1} \mathbf{P}_i \frac{w_{k_1} w_{k_2}}{\pi},$$

with  $\mathbf{Q}_{i1} = \text{diag}\{d\mu_{i1}/d\eta_{i11}, \dots, d\mu_{in_i}/d\eta_{in_i1}\}$  and  $\mathbf{P}_i = (p_{i1}, \dots, p_{in_i})^T$ , where  $p_{ij} = ((1 - \sigma_{ij}^2)/\sigma_{ij}^2)(y_{ij}^* - \mu_{ij}^*)$ ,  $y_{ij}^* = \log\{y_{ij}/(1 - y_{ij})\}$ , and  $\mu_{ij}^* = \psi(\mu_{ij}((1 - \sigma_{ij}^2)/\sigma_{ij}^2)) - \psi((1 - \mu_{ij})((1 - \sigma_{ij}^2)/\sigma_{ij}^2))$  and  $\psi(\cdot)$  is the digamma function, i.e.,  $\psi(z) = d\log\Gamma(z)/dz$ , for  $z > 0$ .

The  $n_i$ -dimensional vector  $\mathbf{D}_{i2}$  defined in  $\mathbf{U}_{\beta_2}(\boldsymbol{\theta})$  and is given by

$$\mathbf{D}_{i2} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i \mathbf{Q}_{i2} \mathbf{R}_i \frac{w_{k_1} w_{k_2}}{\pi},$$

with  $\mathbf{Q}_{i2} = \text{diag}\{\text{d}\sigma_{i1}/\text{d}\eta_{i12}, \dots, \text{d}\sigma_{in_i}/\text{d}\eta_{in_i2}\}$  and  $\mathbf{R}_i = (r_{i1}, \dots, r_{in_i})^T$ , where  $r_{ij} = -2a_{ij}/\sigma_{ij}^3$  and

$$a_{ij} = \mu_{ij}(y_{ij}^* - \mu_{ij}^*) + \log(1 - y_{ij}) + \psi\left(\frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2}\right) - \psi\left((1 - \mu_{ij})\frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2}\right).$$

The quantity  $D_{i3}$  defined in  $U_{\lambda_1}(\boldsymbol{\theta})$  is expressed as:

$$D_{i3} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i \mathbf{Q}_{i1}^* \mathbf{P}_i \frac{\sqrt{2} z_{k_1} w_{k_1} w_{k_2}}{\pi},$$

with  $\mathbf{Q}_{i1}^* = (\text{d}\mu_{i1}/\text{d}\eta_{i11}, \dots, \text{d}\mu_{in_i}/\text{d}\eta_{in_i1})$ .

The quantity  $D_{i3}$  defined in  $U_{\lambda_2}(\boldsymbol{\theta})$  is given by

$$D_{i4} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i \mathbf{Q}_{i2}^* \mathbf{R}_i \frac{\sqrt{2} z_{k_2} w_{k_1} w_{k_2}}{\pi},$$

with  $\mathbf{Q}_{i2}^* = (\text{d}\sigma_{i1}/\text{d}\eta_{i12}, \dots, \text{d}\sigma_{in_i}/\text{d}\eta_{in_i2})$ .

### 3 Observed information matrix

The elements of the  $(p \times p)$  observed information matrix are defined as follows:

$$\begin{aligned}
\mathbf{J}_{\beta_1\beta_1} &= -\sum_{i=1}^N \mathbf{X}_{i1}^T \left( \frac{\mathbf{D}_{i1}\mathbf{D}_{i1}^T}{B_i^2} - \frac{\mathbf{D}_{i5}}{B_i} \right) \mathbf{X}_{i1}, \\
\mathbf{J}_{\beta_1\beta_2} &= -\sum_{i=1}^N \mathbf{X}_{i1}^T \left( \frac{\mathbf{D}_{i1}\mathbf{D}_{i2}^T}{B_i^2} - \frac{\mathbf{D}_{i6}}{B_i} \right) \mathbf{X}_{i2}, \\
\mathbf{J}_{\beta_1\lambda_1} &= -\sum_{i=1}^N \mathbf{X}_{i1}^T \left( \frac{\mathbf{D}_{i1}D_{i3}}{B_i^2} - \frac{\mathbf{D}_{i7}}{B_i} \right), \\
\mathbf{J}_{\beta_1\lambda_2} &= -\sum_{i=1}^N \mathbf{X}_{i1}^T \left( \frac{\mathbf{D}_{i1}D_{i4}}{B_i^2} - \frac{\mathbf{D}_{i8}}{B_i} \right), \\
\mathbf{J}_{\beta_2\beta_2} &= -\sum_{i=1}^N \mathbf{X}_{i2}^T \left( \frac{\mathbf{D}_{i2}\mathbf{D}_{i2}^T}{B_i^2} - \frac{\mathbf{D}_{i9}}{B_i} \right) \mathbf{X}_{i2}, \\
\mathbf{J}_{\beta_2\lambda_1} &= -\sum_{i=1}^N \mathbf{X}_{i2}^T \left( \frac{\mathbf{D}_{i2}D_{i3}}{B_i^2} - \frac{\mathbf{D}_{i10}}{B_i} \right), \\
\mathbf{J}_{\beta_2\lambda_2} &= -\sum_{i=1}^N \mathbf{X}_{i2}^T \left( \frac{\mathbf{D}_{i2}D_{i4}}{B_i^2} - \frac{\mathbf{D}_{i11}}{B_i} \right), \\
\mathbf{J}_{\lambda_1\lambda_1} &= -\sum_{i=1}^N \left( \frac{D_{i3}^2}{B_i^2} - \frac{D_{i12}}{B_i} \right), \\
\mathbf{J}_{\lambda_1\lambda_2} &= -\sum_{i=1}^N \left( \frac{D_{i3}D_{i4}}{B_i^2} - \frac{D_{i13}}{B_i} \right), \\
\mathbf{J}_{\lambda_2\lambda_2} &= -\sum_{i=1}^N \left( \frac{D_{i4}^2}{B_i^2} - \frac{D_{i14}}{B_i} \right).
\end{aligned}$$

The matrix  $\mathbf{D}_{i5}$  of order  $(n_i \times n_i)$  defined in  $\mathbf{J}_{\beta_1\beta_1}$  is given by

$$\mathbf{D}_{i5} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i (\mathbf{Q}_{i1}\mathbf{P}_i(\mathbf{Q}_{i1}\mathbf{P}_i)^T + \mathbf{T}_i) \frac{w_{k_1}w_{k_2}}{\pi},$$

with  $\mathbf{T}_i = \text{diag}\{t_{i1}, \dots, t_{in_i}\}$  and

$$t_{ij} = \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \left( (y_{ij}^* - \mu_{ij}^*) \frac{d^2 \mu_{ij}}{d\eta_{ij1}^2} - b_{ij} \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \left( \frac{d\mu_{ij}}{d\eta_{ij1}} \right)^2 \right),$$

where  $b_{ij} = \psi'(\mu_{ij}((1 - \sigma_{ij}^2)/\sigma_{ij}^2)) + \psi'((1 - \mu_{ij})((1 - \sigma_{ij}^2)/\sigma_{ij}^2))$  and  $\psi'(\cdot)$  is the trigamma function.

The matrix  $\mathbf{D}_{i6}$  of order  $(n_i \times n_i)$  defined in  $\mathbf{J}_{\beta_1\beta_2}$  is expressed as

$$\mathbf{D}_{i6} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i (\mathbf{Q}_{i1} \mathbf{P}_i (\mathbf{Q}_{i2} \mathbf{R}_i)^T + \mathbf{V}_i \mathbf{Q}_{i1} \mathbf{Q}_{i2}) \frac{w_{k_1} w_{k_2}}{\pi},$$

with  $\mathbf{V}_i = \text{diag}\{v_{i1}, \dots, v_{in_i}\}$  and

$$v_{ij} = -\frac{2}{\sigma_{ij}^3} \left( (y_{ij}^* - \mu_{ij}^*) - \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \left( \mu_{ij} b_{ij} - \psi' \left( (1 - \mu_{ij}) \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \right) \right) \right).$$

The  $n_i$ -dimensional vector  $\mathbf{D}_{i7}$  defined in  $\mathbf{J}_{\beta_1 \lambda_1}$  is given by

$$\mathbf{D}_{i7} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i (\mathbf{Q}_{i1} \mathbf{P}_i \mathbf{Q}_{i1}^* \mathbf{P}_i + \mathbf{T}_i^*) \frac{\sqrt{2} z_{k_1} w_{k_1} w_{k_2}}{\pi},$$

with  $\mathbf{T}_i^* = (t_{i1}, \dots, t_{in_i})^T$ .

The  $n_i$ -dimensional vector  $\mathbf{D}_{i8}$  defined in  $\mathbf{J}_{\beta_1 \lambda_2}$  is expressed as

$$\mathbf{D}_{i8} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i (\mathbf{Q}_{i1} \mathbf{P}_i \mathbf{Q}_{i2}^* \mathbf{R}_i + \mathbf{Q}_{i1} \mathbf{Q}_{i2} \mathbf{V}_i^*) \frac{\sqrt{2} z_{k_2} w_{k_1} w_{k_2}}{\pi},$$

with  $\mathbf{V}_i^* = (v_{i1}, \dots, v_{in_i})^T$ .

The matrix  $\mathbf{D}_{i9}$  of order  $(n_i \times n_i)$  defined in  $\mathbf{J}_{\beta_2 \beta_2}$  is given by

$$\mathbf{D}_{i9} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i (\mathbf{Q}_{i2} \mathbf{R}_i (\mathbf{Q}_{i2} \mathbf{R}_i)^T + \mathbf{Z}_i) \frac{w_{k_1} w_{k_2}}{\pi},$$

with  $\mathbf{Z}_i = \text{diag}\{z_{i1}, \dots, z_{in_i}\}$  and

$$z_{ij} = \frac{-2}{\sigma_{ij}^3} \left( a_{ij} \frac{d^2 \sigma_{ij}}{d\eta_{ij2}^2} - \frac{1}{\sigma_{ij}} \left( 3a_{ij} - \frac{2}{\sigma_{ij}^2} c_{ij} \right) \left( \frac{d\sigma_{ij}}{d\eta_{ij2}} \right)^2 \right),$$

where

$$c_{ij} = \mu_{ij}^2 b_{ij} - 2\mu_{ij} \psi' \left( (1 - \mu_{ij}) \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \right) - \psi' \left( \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \right) + \psi' \left( (1 - \mu_{ij}) \frac{1 - \sigma_{ij}^2}{\sigma_{ij}^2} \right).$$

The  $n_i$ -dimensional vector  $\mathbf{D}_{i10}$  defined in  $\mathbf{J}_{\beta_2 \lambda_1}$  is expressed as

$$\mathbf{D}_{i10} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i (\mathbf{Q}_{i2} \mathbf{R}_i \mathbf{Q}_{i1}^* \mathbf{P}_i + \mathbf{Q}_{i1} \mathbf{Q}_{i2} \mathbf{V}_i^*) \frac{\sqrt{2} z_{k_1} w_{k_1} w_{k_2}}{\pi}.$$

The  $n_i$ -dimensional vector  $\mathbf{D}_{i11}$  defined in  $\mathbf{J}_{\beta_2 \lambda_2}$  is given by

$$\mathbf{D}_{i11} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i \left( \mathbf{Q}_{i2} \mathbf{R}_{\mathbf{Q}_{i2}}^* \mathbf{R}_i + \mathbf{Z}_i^* \right) \frac{\sqrt{2} z_{k_1} w_{k_1} w_{k_2}}{\pi},$$

where  $\mathbf{Z}_i^* = (z_{i1}, \dots, z_{in_i})^T$ .

The quantity  $D_{i12}$  defined in  $J_{\lambda_1 \lambda_1}$  is given by

$$D_{i12} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i \left( \mathbf{Q}_{i1}^* \mathbf{P}_i \mathbf{Q}_{i1}^* \mathbf{P}_i + \text{trace}(\mathbf{T}_i^*) \right) \frac{2z_{k_1}^2 w_{k_1} w_{k_2}}{\pi}.$$

The quantity  $D_{i13}$  defined in  $J_{\lambda_1 \lambda_2}$  is expressed as

$$D_{i13} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i \left( \mathbf{Q}_{i1}^* \mathbf{P}_i \mathbf{Q}_{i2}^* \mathbf{R}_i + \text{trace}(\mathbf{V}_i^*) \right) \frac{2z_{k_1} z_{k_2} w_{k_1} w_{k_2}}{\pi}.$$

The quantity  $D_{i14}$  defined in  $J_{\lambda_2 \lambda_2}$  is given by

$$D_{i14} = \sum_{k_1=1}^{Q_1} \sum_{k_2=1}^{Q_2} A_i \left( \mathbf{Q}_{i2}^* \mathbf{R}_i \mathbf{Q}_{i2}^* \mathbf{R}_i + \text{trace}(\mathbf{Z}_i^*) \right) \frac{2z_{k_2}^2 w_{k_1} w_{k_2}}{\pi}.$$