

1. The risk modeling

Given

$$y_n = \beta_n^T \mathcal{Z} + \sqrt{1 - \beta_n^T \beta_n} \epsilon_n$$

where $\mathcal{Z} \sim \mathcal{N}(0, I_S)$ and $\epsilon \sim \mathcal{N}(0, I_N)$, the resulting $y_n \sim \mathcal{N}(0, 1)$ as shown below

$$\mathbb{E} \{y_n\} = \sum_i \beta_{n,i} \mathbb{E} \{\mathcal{Z}_i\} + \sqrt{1 - \beta_n^T \beta_n} \mathbb{E} \{\epsilon_n\} = 0$$

$$\begin{aligned} \text{var} \{y_n\} &= \mathbb{E} \{(y_n - \mathbb{E} \{y_n\})^2\} \\ &= \mathbb{E} \left\{ (\beta_n^T \mathcal{Z} + \sqrt{1 - \beta_n^T \beta_n} \epsilon_n)^2 \right\} \\ &= \beta_n^T \beta_n \mathbb{E} \{(\mathcal{Z} - 0)^2\} + (1 - \beta_n^T \beta_n) \mathbb{E} \{(\epsilon_n - 0)^2\} \\ &= \beta_n^T \beta_n \text{var} \{\mathcal{Z}\} + (1 - \beta_n^T \beta_n) \text{var} \{\epsilon_n\} \\ &= 1 \end{aligned}$$

2. Motivation for threshold between different states $H_{c(n)}^c$

The motivation is to model discrete probability in the credit state matrix with a continuous distribution such as the gaussian. In this case, we want to set $H_{c(n)}^c$ such that

$$p(H_{c(n)}^{c-1} \leq y_n \leq H_{c(n)}^c) = p_{c(n)}^c$$

therefore we can write

$$p(y_n \leq H_{c(n)}^c) = \sum_{\gamma=1}^c p_{c(n)}^\gamma \quad y_n \xrightarrow{\mathcal{N}(0,1)} \quad \Phi(H_{c(n)}^c) = \sum_{\gamma=1}^c p_{c(n)}^\gamma$$

3. Confidence interval for monte carlo estimation

$$p(L_N(\mathcal{Z}, \epsilon) \geq l) \in p(L_N(\mathcal{Z}, \epsilon) \geq l) \pm CI$$

Idea is that for the two naive algorithms that are purported to be equivalent, the CI should be approximately the same

1. Likelihood Function Note

$$Z \sim \mathcal{N}(0, I) \quad \leftarrow \quad Z \sim \mathcal{N}(\mu, I)$$

The quotient is then

$$\frac{\sqrt{2\pi}^S e^{-Z^T Z/2}}{\sqrt{2\pi}^S e^{-(Z-\mu)^T (Z-\mu)/2}}$$