1. The risk modeling

Given

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

where $\mathcal{Z} \sim \mathcal{N}\left(0, I_{S}\right)$ and $\epsilon \sim \mathcal{N}\left(0, I_{N}\right)$, the resulting $y_{n} \sim \mathcal{N}\left(0, 1\right)$ as shown below

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

$$var \{y_n\} = \mathbb{E} \left\{ (y_n - \mathbb{E} \{y_n\})^2 \right\}$$

$$= \mathbb{E} \left\{ (\beta_n^T \mathcal{Z} + \sqrt{1 - \beta_n^T \beta_n} \epsilon_n)^2 \right\}$$

$$= \beta_n^T \beta_n \mathbb{E} \left\{ (\mathcal{Z} - 0)^2 \right\} + (1 - \beta_n^T \beta_n) \mathbb{E} \left\{ (\epsilon_n - 0)^2 \right\}$$

$$= \beta_n^T \beta_n var \{\mathcal{Z}\} + (1 - \beta_n^T \beta_n) var \{\epsilon_n\}$$

$$= 1$$

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 cawe, we want to set $H_{c(n)}^c$ such that

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

therefore we can write

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

3. Confidence interval for monte carlo estimation

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

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