

Appendix

Appendix

The following paragraphs contain additional information about the **simulation studies** that were covered in chapter ??.

Section ??: Correlated Predictor Variables

The simulation design was chosen in the following way:

- The design matrix $\mathbf{X} = (\mathbf{x}_1 \ \mathbf{x}_2 \ \mathbf{x}_3)$ is simulated from a three dimensional normal distribution $\mathcal{N}_3(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ with mean vector $\boldsymbol{\mu} = (-5 \ 2 \ 0)^T$ and covariance matrix $\begin{pmatrix} 1 & \rho & \rho \\ \rho & 3 & \rho \\ \rho & \rho & 5 \end{pmatrix}$. Hence, the dependence among the regressors is fully determined by the parameter ρ .
- The design matrix $\mathbf{Z} = (\mathbf{z}_1 \ \mathbf{z}_2)$ consists of linear combinations of the regressors \mathbf{x}_1 up to \mathbf{x}_3 , more specifically $\mathbf{z}_1 = 0.8 \cdot \mathbf{x}_1 + 0.2 \cdot \mathbf{x}_2$ and $\mathbf{z}_2 = \mathbf{x}_2 - 0.5 \cdot \mathbf{x}_3$.
- In both design matrices intercept columns are added for estimation purposes. Moreover, all columns in \mathbf{X} and \mathbf{Z} are standardized, i.e. mean-centered around 0 and scaled to unit variance.
- The true coefficient vectors are given by $\boldsymbol{\beta} = (\beta_0 \ \beta_1 \ \beta_2 \ \beta_3)^T = (0 \ 3 \ -1 \ 1)^T$ and $\boldsymbol{\gamma} = (\gamma_0 \ \gamma_1 \ \gamma_2)^T = (0 \ 2 \ 0)^T$.
- The outcome variable \mathbf{y} is generated according to the correctly specified location-scale model $y_i \stackrel{iid}{\sim} \mathcal{N}(\mathbf{x}_i^T \boldsymbol{\beta}, \exp(\mathbf{z}_i^T \boldsymbol{\gamma})^2)$ for $i = 1, \dots, n$ with sample size $n = 50$.
- Three different values were chosen for $\rho \in \{0, -0.5, 0.9\}$ to compare the ‘nice’ case of uncorrelated predictors with the performance for negative and positive dependence. For each covariance structure the three models `mcmc_ridge()`, `mcmc()` and `lmls()` were fitted, where each Posterior Mean estimate from both of the Markov Chain Monte Carlo samplers is based on 10000 samples.

Section ??: Challenging the Model Assumptions

The data for this second simulation study is generated by the following conventions:

- The design matrix $\mathbf{X} = (\mathbf{1}_n \ \mathbf{x}_1 \ \mathbf{x}_2 \ \mathbf{x}_3 \ \mathbf{x}_4)$ contains four independently sampled regressor variables plus one intercept column:
 - $\mathbf{x}_1 \stackrel{iid}{\sim} \mathcal{N}(5, 16)$,
 - $\mathbf{x}_2 \stackrel{iid}{\sim} \text{Exp}(5)$,
 - $\mathbf{x}_3 \stackrel{iid}{\sim} \mathcal{U}([-2, 12])$,
 - $\mathbf{x}_4 \stackrel{iid}{\sim} \text{Ber}(0.3)$.
- The design matrix $\mathbf{Z} = (\mathbf{1}_n \ \mathbf{x}_1 \ \mathbf{x}_2 \ \mathbf{z}_3)$ contains the additional regressor variable $\mathbf{z}_3 \stackrel{iid}{\sim} t_{10}$, which is independently sampled from all other columns.
- All covariate vectors in both design matrices (except for the intercept columns) are standardized before generating the values for \mathbf{y} .

- The true coefficient vectors are given by $\beta = (\beta_0 \ \beta_1 \ \beta_2 \ \beta_3 \ \beta_4)^T = (0 \ -3 \ -1 \ -1 \ 2)^T$ and $\gamma = (\gamma_0 \ \gamma_1 \ \gamma_2 \ \gamma_3)^T = (0 \ 1 \ 2 \ 3)^T$.
- Three different specifications for the outcome distribution were chosen:
 - $y_i \sim \mathcal{N}(\mu, \sigma^2)$,
 - $y_i \sim \mu + \left(\sigma \cdot \sqrt{\frac{3}{5}}\right) T$, where $T \sim t_5$,
 - $y_i \sim \mu + \sigma \cdot U$, where $U \sim \mathcal{U}([0, 1])$.
- In order to isolate the impact of the different shapes of the three probability distributions, the mean $\mu = \mathbf{x}_i^T \beta$ and the variance $\sigma^2 = \exp(\mathbf{z}_i^T \gamma)^2$ are held constant across the models.
- The sample size is set to $n = 50$ and the `mcmc()` as well as the `mcmc_ridge()` results are based on 1000 simulations.

Section ??: Redundant Covariates

We again state the conditions that the simulation study is based on:

- The design matrix $\mathbf{X} = (\mathbf{1}_n \ \mathbf{x}_1 \ \dots \ \mathbf{x}_{20})$ consists of one intercept column plus 10 *pairs* of successive regressors, starting with the pair $(\mathbf{x}_1, \mathbf{x}_2)$. Each pair $(\mathbf{x}_i, \mathbf{x}_{i+1})$ for $i \in \{1, 3, \dots, 19\}$ is (independently from all remaining pairs) drawn from a bivariate normal distribution with mean vector $\mu = (0 \ 0)^T$ and correlation matrix $\begin{pmatrix} 1 & 0.9 \\ 0.9 & 1 \end{pmatrix}$.

Afterwards, each column of \mathbf{X} except the intercept column is standardized to zero mean and unit variance.

- The design matrix $\mathbf{Z} = (\mathbf{1}_n \ \mathbf{x}_1 \ \mathbf{x}_3)$ is of minor interest in this case and consists of an intercept column plus two uncorrelated columns chosen from \mathbf{X} .
- The true coefficients of β are determined by the pattern $\beta_i = 0$, if i is even and $\beta_i = 1$, if i is odd. Thus, all covariates with even subscript are redundant, whereas those with odd subscript contribute to \mathbf{y} . The true γ , again of minor interest here, is given by $\gamma = (0 \ 1 \ 1)^T$.
- The outcome variable \mathbf{y} is generated according to the correctly specified location-scale model $y_i \stackrel{iid}{\sim} \mathcal{N}(\mathbf{x}_i^T \beta, \exp(\mathbf{z}_i^T \gamma)^2)$ for $i = 1, \dots, n$.
- The sample size $n = 50$ is deliberately chosen small compared to the number of regressors. Both of the Bayesian models generate 10000 values for each coefficient.

Section ??: Sample Size

Section ??: Hyperparameters