

# Simulation Scheme

The simulation study contains two parts

- $Y$  is continuous
- $Y$  is binary

## Continuous

For the first type (i.e.,  $Y$  is continuous), we consider the following two cases

- Linear regression
- Nonlinear regression (tentatively)

The linear regression setting use Gaussian error terms, and we consider both the correct model and the misspecified model. The label shift assumption is that  $p(\mathbf{x}|y)$  remains the same, however, deriving the explicit form of  $p(\mathbf{x}|y)$  can be difficult. For example, for the following model

$$Y \sim X_1 + X_2 + X_3 + X_1X_2 + X_1X_3 + \epsilon,$$

where  $\epsilon \sim \text{Norm}(0, \sigma^2)$ . Even if  $\mathbf{X}$  has a multivariate normal distribution, deriving the closed form expression for  $p(\mathbf{x}|y)$  is non-trivial. Thus, we used the rejection sampling algorithm to generate samples. We need to find a constant  $C_{\max}$  such that  $\rho(y; \boldsymbol{\beta})/C_{\max}$  is bounded by 1. Here, we outline the procedure to generate samples from the target distribution as follows,

1. Generate a sample of  $\mathbf{X} = \mathbf{x}$  from the marginal distribution of  $p_s(\mathbf{x})$ .
2. Given the sample  $\mathbf{X} = \mathbf{x}$ , generate  $Y = y$  using the specified model  $p_s(y|\mathbf{x})$ .
3. Generate a number  $U = u$  from the uniform distribution  $U(0, 1)$ .
4. If  $U = u < \rho(y; \boldsymbol{\beta})/C_{\max}$ , accept this pair  $(\mathbf{x}, y)$ ; otherwise reject this pair.
5. Repeat the previous steps until accumulate the desired sample size.

## Setting 1

The marginal distribution of  $\mathbf{X} = (X_1, X_2, X_3)^T$  is

$$\mathbf{X} \sim \text{MVN}(\mu, \Sigma),$$

where

$$\mu = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \quad \Sigma = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

The model  $p_s(y|\mathbf{x})$  is

$$Y = 1 + 2X_1 + 2X_2 + 2X_3 + X_1X_2 + X_2X_3 + \epsilon,$$

where  $\epsilon \sim \text{Norm}(0, 1)$ .

## References