## Bayesian mean-only linear model with Stan

In this program, we assume that we have a vector of N observations that can be reasonably modeled by a  $N(\mu, \sigma)$  distribution.

Our objective is to fit the mean-only linear model:

$$Y = \beta_0 + e$$

where:

- Y is a vector of observed measurements (Sepal.Width from the iris dataset).
- $\beta_0$  is a parameter, assumed to be a random variable and given a prior distribution.
- e is a vector of independent normal random variables with mean zero and standard deviation  $\sigma_e$ .

In the Bayesian formulation,  $\beta_0$  and  $\sigma_e$  are treated as parameters, which are random variables in the Bayesian framework.

The Bayes equation is:

$$f(\beta_0, \sigma_e|Y) \propto f(Y|\beta_0, \sigma_e) \cdot f(\beta_0) \cdot f(\sigma_e)$$

where:

- $f(\beta_0, \sigma_e|Y)$  is the (joint) posterior distribution of  $\beta_0$  and  $\sigma_e$  given Y
- $f(Y|\beta_0, \sigma_e)$  is the conditional likelihood of Y given  $\beta_0$  and  $\sigma_e$
- $f(\beta_0)$  is the prior distribution for  $\beta_0$
- $f(\sigma_e)$  is the prior distribution for  $\sigma_e$

The likelihood and priors are specified in mean\_only.stan:

With this model file, we have the following specifications:

- The likelihood  $f(Y|\beta_0,\sigma_e)$  is  $N(\beta_0,\sigma_e)$ , coded as y  $\sim$  normal(beta\_0,sigma)
- The prior  $f(\beta_0)$  is N(0,100), coded as beta\_0  $\sim$  normal(0,100)
- The prior  $f(\sigma_e)$  is half-cauchy (because of < lower = 0 >) coded as sigma\_e  $\sim$  cauchy(0,10)