

## **Simulation study**

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## Simulation study: why we need it?

**Target:** Evaluate the performance of the methodology using the **user-generated** data sets from the **true model**.

1. We are not sure of the real performance of a newly proposed method.
2. With the generated data sets, we can test whether the method can reproduce the true parameters.

## Simulation study: procedures

1. specify your model
2. give true parameter values
3. generate data sets (usually 100 replications)
4. implement your method
5. summarize the estimation results (Bias, RMS, etc.)

## 1. Specify my model

Let  $y_i$  denotes the count outcome that follows  $Pois(\mu_i)$ , and the loglinear model can be expressed as follows:

$$y_i \sim Pois(\mu_i)$$
$$\log(\mu_i) = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i}$$

where  $x_1$  and  $x_2$  denote observed covariates, and  $\beta_0, \beta_1, \beta_2$  are unknown parameters.

## 2. Give true parameter values

$$y_i \sim \text{Pois}(\mu_i)$$

$$\log(\mu_i) = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i}$$

- sample size:  $n = 300$
- number of covariates:  $p = 2$
- parameter true values:  $\beta_0 = 0.5$ ,  $\beta_1 = 1$ ,  $\beta_2 = -1$
- number of replications:  $n_{\text{rep}} = 10$

### 3. Generate data sets

$$y_i \sim \text{Pois}(\mu_i)$$

$$\log(\mu_i) = 0.5 + x_{1i} - x_{2i}$$

1. generate  $x_{1i}, x_{2i}$  from  $N(0, 1)$  (try other distributions!)
2. calculate  $\mu_i$  based on the true model
3. generate  $y_i$  from poisson distribution with parameter  $\mu_i$
4. replicate for  $n_{\text{rep}} = 10$  times and **save**

see [examples/simu/glm\\_count.R](#) generate data part

## 4. Implement your method

see [examples/simu/glm\\_count.R](#) estimation part

## 5. Summarize the estimation results

see [examples/simu/glm\\_count.R](#) summary part