## design and generating data

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## Design of simulation settings

We conducted simulation studies to assess the performance of three survival models. In total, we created 7 simulation settings by mixing the event time generated from the three specified baseline hazard function: exponential, Weibull, and Gompertz hazard function. Then the generated data were fitted to exponential, Weibull, and Cox proportional hazard models. The parameters we applied in the models are constants:  $\alpha = 0.5, \gamma = 1.5, \beta = -0.5$ . Since the shape of Weibull distribution has great difference between  $\gamma > 1$  and  $\gamma < 1$ , thus we simulated in scenarios where  $\gamma = 0.5$  and  $\gamma = 1.5$ ,

We simulated 500 data sets in each simulation setting. After running the models, a set of  $\beta$  was extracted and used to calculate the mean bias, variance, and squared error. To evaluate the efficiency performances of models, we simulated data of different sample sizes: 20, 40, 60, 80, 100, 200, 400. Similarly, bias, variance, MSE are calculated.

All the simulation processes were performed in R.

## Methods for generating data

The survival dataset contains treatment assignment, status indicator, and observed time. Treatment assignment variable  $X_i$  is generated from a Bernoulli distribution with p = 0.5. By utilizing the inverse transformation method, we can obtain event time T:

$$T = H_0^{-1} \left( \frac{-\log(u)}{e^{x^T \beta}} \right)$$
, where  $U \sim U(0, 1)$ 

The followings are specific baseline hazard functions we applied:

- 1. Survival time under Exponential distribution:  $T = -\frac{\log(u)}{\lambda e^{x^T \beta}}$
- 2. Survival time under Weibull distribution:  $T = \left(-\frac{\log(u)}{\lambda e^{x^T \beta}}\right)^{\frac{1}{\gamma}}$
- 3. Survival time under Gompertz distribution:  $T = \frac{1}{\alpha} \log \left(1 \frac{\alpha \log(u)}{\lambda e^{x^T \beta}}\right)$

We simulated survival data by the event time generated from the mixtures of three baseline distributions.

Mixture of Exponential and Weibull distribution:

$$T = p * (-\frac{\log(u)}{\lambda e^{x^T \beta}}) + (1 - p) * (-\frac{\log(u)}{\lambda e^{x^T \beta}})^{1/\gamma}$$

We take values of p as 0, 0.5 and 1. When p = 1, event time is generated from exponential baseline; when p = 0, event time is generated from Weibull distribution.

Mixture of Exponential and Gompertz distribution:

$$T = p * \left(-\frac{\log(u)}{\lambda e^{x^T \beta}}\right) + (1 - p) * \left(\frac{1}{\alpha} \log\left(1 - \frac{\alpha \log(u)}{\lambda e^{x^T \beta}}\right)\right)$$

Similary, we take values of p as 0, 0.5 and 1. When p=1, event time is generated from exponential baseline; when p=0, event time is generated from Gompertz distribution. Finally, make event indicator variable by applying administrative censoring at t=5.

By repeating each of the above simulation process 500 times, we get survival datasets with sample size ranging from 20 to 400.