

# Untitled

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More details for the data generation.

## example 1

CDF:

$$S(x, y) = \begin{cases} e^{-\theta x} e^{-(e^{\theta y} - 1)(x - y + 1)} & x \geq y \\ e^{-\theta x} e^{-(e^{\theta y} - 1)} & x < y \end{cases}$$

PDF:

$$f(x, y) = \begin{cases} [(\theta(x - y + 1) - 1)e^{2\theta y} + (\theta^2(x - y + 1) - \theta(x - y + 1) - 2(\theta + 1))e^{\theta y} + \theta - 1] \\ \times e^{-\theta x} e^{-(e^{\theta y} - 1)(x - y + 1)} & x \geq y \\ \theta^2 e^{(-e^{\theta y} + \theta y - \theta x + 1)} & x < y \end{cases}$$

Conditional CDF:

$$F_{C|T}(x, y) = \begin{cases} 1 - \frac{1}{\theta}(e^{\theta y} + \theta - 1)e^{-(e^{\theta y} - 1)(x - y + 1)} & x \geq y \\ e^{1 - e^{\theta x}} - e^{1 - e^{\theta y}} & x < y \end{cases}$$

$$F_{C|T}(x, y) = \int_0^y f(y|x)dy = \int_0^y \frac{f(x, y)}{f(x)}dy = \int_0^y \frac{\partial F(x, y)}{\partial x \partial y f(x)}dy = \frac{1}{f(x)} \frac{\partial F(x, y)}{\partial x} \Big|_0^u$$

$$F_{C|T}(x, y) = \begin{cases} 1 - \frac{1}{\theta_1}(2(e^{\theta_2 y} - 1)(x - y) + \theta_1)e^{-(e^{\theta_2 y} - 1)(x - y + 1)} & x \geq y \\ e^{1 - e^{\theta_2 x}} - e^{1 - e^{\theta_1 x}} & x < y \end{cases}$$

## example 2

CDF:

$$S(x, y) = \begin{cases} e^{-\theta x} e^{-(e^{2\theta y} - 1)(x - y + 1)} & x \geq y \\ e^{-\theta x} e^{-(e^{2\theta y} - 1)} & x < y \end{cases}$$

PDF:

$$f(x, y) = \begin{cases} [(2\theta(x - y + 1) - 1)e^{4\theta y} + 2\theta(\theta - 1)(x - y)e^{2\theta y} + (2\theta - 1)(\theta - 2)e^{2\theta y} + \theta - 1] \\ \times e^{-\theta x} e^{-(e^{2\theta y} - 1)(x - y + 1)} & x \geq y \\ 2\theta^2 e^{(-e^{2\theta y} + 2\theta y - \theta x + 1)} & x < y \end{cases}$$

Conditional CDF:

$$F_{C|T}(x, y) = \begin{cases} 1 - \frac{1}{\theta}(e^{2\theta y} + \theta - 1)e^{-(e^{2\theta y} - 1)(x - y + 1)} & x \geq y \\ e^{1 - e^{2\theta x}} - e^{1 - e^{2\theta y}} & x < y \end{cases}$$

### example 3

$$F_{C|T}(x, y) = \begin{cases} 1 - \frac{1}{\theta_1}(e^{\theta_2 y} + \theta_1 - 1)e^{-(e^{\theta_2 y} - 1)(x - y + 1)} & x \geq y \\ e^{1 - e^{\theta_2 x}} - e^{1 - e^{\theta_2 y}} & x < y \end{cases}$$

all different

$e^{(by - 1)}$

example

$$e^{(-ax)} e^{(e^y - 1)(x - y + 1)} ((x - y)e^{(2y)} + (a - 1)(x - y)e^y + a - 1)$$

Consider joint distribution of  $T$  and  $C$ ,

$$S_{T,C}(t, s; X) = \begin{cases} S_T(t; X)K(t, s; X) & t \geq s \\ S_T(t; X)S_C(s; X) & t < s \end{cases}$$

where  $F_T(t; X) = 1 - S_T(t; X)$  and  $F_C(s; X) = 1 - S_C(s; X)$  are CDF functions. The  $X$  is the baseline vectors, which serves as parameters for the joint distribution. Besides,  $K(t, s; X)$  is a joint function of  $T$  and  $C$ , where

- 1. cannot be factored as a production of a function that only contain  $T$  and a function that only contain  $C$
- 2.  $K(t, s; X) \geq 0$  when  $t, s \geq 0$
- 3.  $K(t, 0; X) = 1$
- 4.  $K(s, s; X) = S_C(s; X)$
- 5.  $K(t, s; X) = 0$  as  $t, s \rightarrow \infty$

Then, the marginal distribution for event time is

$$\begin{aligned} P(T > t; X) &= \int_t^\infty f_T(t; X)dt \\ &= \int_t^\infty \left\{ \int_0^\infty f_{T,C}(t, s; X)ds \right\}dt \\ &= P(T > t, C > 0; X) = S_T(t; X)K(t, 0; X) \\ &= S_T(t; X) \end{aligned}$$

The marginal distribution for the censoring time is

$$\begin{aligned}
P(C > s; X) &= \int_s^\infty f_C(s; X) ds \\
&= \int_s^\infty \left\{ \int_0^\infty f_{T,C}(t, s; X) dt \right\} ds \\
&= P(T > 0, C > s; X) = S_T(0; X) S_C(s; X) \\
&= S_C(s; X)
\end{aligned}$$

The distribution for the observed time  $Z = T \wedge C$  is

$$S_Z(t; X) = P(T > t, C > t; X) = S_T(t; X) S_C(t; X)$$

with pdf:

$$f_Z(t; X) = -\frac{\partial[S_T(t; X) S_C(t; X)]}{\partial t} = f_T(t; X) S_C(t; X) + S_T(t; X) f_C(t; X)$$

Suppose the hazard for event is

$$\lambda_T(t; X) = \frac{f_T(t; X)}{S_T(t; X)}$$

The hazard for the censoring is

$$\lambda_C(t; X) = \frac{f_C(t; X)}{S_C(t; X)}$$

Then the  $m(t; X)$  is

$$m(t; X) = \frac{\frac{f_T(t; X)}{S_T(t; X)}}{\frac{f_Z(t; X)}{S_Z(t; X)}} = \frac{S_T(t; X) S_C(t; X) \frac{f_T(t; X)}{S_T(t; X)}}{f_T(t; X) S_C(t; X) + S_T(t; X) f_C(t; X)} = \frac{\lambda_T(t; X)}{\lambda_T(t; X) + \lambda_C(t; X)}$$

Suppose the event time is from a cox PH model, then  $S_T(t; X) = S_0(t)^{\exp(\beta X)}$  and  $\lambda_T(t; X) = \lambda_0(t) \exp(\beta X)$ . Then

$$m(t; X) = \frac{\lambda_T(t; X)}{\lambda_T(t; X) + \lambda_C(t; X)} = \frac{1}{1 + \frac{\lambda_C(t; X)}{\lambda_0(t)} \exp(-\beta X)} = \frac{1}{1 + \exp(\ln(\lambda_C(t; X)) - \ln(\lambda_0(t)) - \beta X)}$$

which can be treated as a logistic regression, with independent variables, transformed  $X$ , the baseline covariates and transformed  $t$ , the observed time.

To make the model more clear, let's look at two examples with the above model structure.

### Example 1

Let construct a joint distribution as following

$$S_{T,C}(t, s; \theta_1, \theta_2) = \begin{cases} e^{-\theta_1 t} e^{-(e^{\theta_2 s} - 1)((t-s)^2 + 1)} & \text{when } t \geq s \\ e^{-\theta_1 t} e^{-(e^{\theta_2 s} - 1)} & \text{when } t < s \end{cases}$$

where

$$S_T(t; X) = e^{-\theta_1 t}, S_C(s; X) = e^{-(e^{\theta_2 s} - 1)}; K(t, s; X) = e^{-(e^{\theta_2 s} - 1)((t-s)^2 + 1)}$$

and  $\theta_1, \theta_2$  are parameters associated with  $X$ . Suppose  $x_1$  has four levels 1,2,3,4 and  $x_2$  has two levels 0,1, which is an indicator for gender,

$$x_2 = \begin{cases} 1 & \text{female} \\ 0 & \text{male} \end{cases}$$

Let  $\theta_1 = \beta x_1, \theta_2 = \beta x_1 x_2$ . Therefore,

$$S_{T,C}(t, s; x_1, x_2) = \begin{cases} e^{-\beta x_1 t} e^{-(e^{(\beta x_1 x_2)s} - 1)((t-s)^2 + 1)} & \text{when } t \geq s \\ e^{-\beta x_1 t} e^{-(e^{(\beta x_1 x_2)s} - 1)} & \text{when } t < s \end{cases}$$

That is, for females:

$$S_{T,C}(t, s; x_1, x_2) = \begin{cases} e^{-\beta x_1 t} e^{-(e^{(\beta x_1)s} - 1)((t-s)^2 + 1)} & \text{when } t \geq s \\ e^{-\beta x_1 t} e^{-(e^{(\beta x_1)s} - 1)} & \text{when } t < s \end{cases}$$

For males:

$$S_{T,C}(t, s; x_1, x_2) = \begin{cases} e^{-\beta x_1 t} & \text{when } t \geq s \\ e^{-\beta x_1 t} & \text{when } t < s \end{cases} = e^{-\beta x_1 t}$$

There is no censorship in males.

For the event time distribution,

$$S_T(t; x_1, x_2) = e^{-\beta x_1 t} = [e^{-t}]^{\beta x_1} = [e^{-t}]^{\exp(\ln(\beta x_1))} = [e^{-t}]^{\exp(\ln(\beta) + \ln(x_1))}$$

Therefore,  $S_T(t; x_1, x_2)$  is a cox PH model. And

$$\begin{aligned} f_T(t; x_1, x_2) &= \beta x_1 e^{-\beta x_1 t}, \\ \lambda_T(t; x_1, x_2) &= \frac{f_T(t)}{S_T(t)} = \frac{\beta x_1 e^{-\beta x_1 t}}{e^{-\beta x_1 t}} = \beta x_1 = \lambda_{T0}(t) \exp(\ln(\beta) + \ln(x_1)), \quad \lambda_{T0}(t) = 1 \end{aligned}$$

The censoring time is

$$\begin{aligned} S_C(t; x_1, x_2) &= e^{-(e^{(\beta x_1 x_2)t} - 1)} \\ f_C(s; x_1, x_2) &= (\beta x_1 x_2) e^{-e^{(\beta x_1 x_2)s} + (\beta x_1 x_2)s + 1} \\ \lambda_C(s; x_1, x_2) &= \frac{f_C(s; x_1, x_2)}{S_C(s; x_1, x_2)} = (\beta x_1 x_2) e^{(\beta x_1 x_2)s} \end{aligned}$$

The associated  $m()$  function is

$$\begin{aligned} m(t; x_1, x_2) &= \frac{\lambda_T(t; x_1, x_2)}{\lambda_T(t; x_1, x_2) + \lambda_C(t; x_1, x_2)} \\ &= \frac{\beta x_1}{\beta x_1 + (\beta x_1 x_2) e^{(\beta x_1 x_2)t}} \\ &= \frac{1}{1 + x_2 e^{(\beta x_1 x_2)t}} \\ &= \begin{cases} \frac{1}{1 + e^{(\beta x_1)t}} & x_2 = 1, \text{female} \\ \frac{1}{1 + 0} = 1 & x_2 = 0, \text{male} \end{cases} \end{aligned}$$

## Example 2

To avoid  $\theta_1, \theta_2$  generating 0, we reset the range for  $x_2$ , where

$$x_2 = \begin{cases} 2 & \text{female} \\ 1 & \text{male} \end{cases}$$

The other functions are the same. That is

$$S_{T,C}(t, s; x_1, x_2) = \begin{cases} e^{-\beta x_1 t} e^{-(e^{(\beta x_1 x_2)s} - 1)((t-s)^2 + 1)} & \text{when } t \geq s \\ e^{-\beta x_1 t} e^{-(e^{(\beta x_1 x_2)s} - 1)} & \text{when } t < s \end{cases}$$

For females:

$$S_{T,C}(t, s; x_1, x_2) = \begin{cases} e^{-\beta x_1 t} e^{-(e^{(2\beta x_1)s} - 1)((t-s)^2 + 1)} & \text{when } t \geq s \\ e^{-\beta x_1 t} e^{-(e^{(2\beta x_1)s} - 1)} & \text{when } t < s \end{cases}$$

For males:

$$S_{T,C}(t, s; x_1, x_2) = \begin{cases} e^{-\beta x_1 t} e^{-(e^{(\beta x_1)s} - 1)((t-s)^2 + 1)} & \text{when } t \geq s \\ e^{-\beta x_1 t} e^{-(e^{(\beta x_1)s} - 1)} & \text{when } t < s \end{cases}$$

And the associated  $m()$  function is

$$\begin{aligned} m(t; x_1, x_2) &= \frac{\lambda_T(t; x_1, x_2)}{\lambda_T(t; x_1, x_2) + \lambda_C(t; x_1, x_2)} \\ &= \frac{\beta x_1}{\beta x_1 + (\beta x_1 x_2) e^{(\beta x_1 x_2)t}} \\ &= \frac{1}{1 + x_2 e^{(\beta x_1 x_2)t}} \\ &= \frac{1}{1 + e^{\ln(x_2) + (\beta x_1 x_2)t}} \\ &= \begin{cases} \frac{1}{1 + e^{\ln(2) + (2\beta x_1)t}} & x_2 = 2, \text{ female} \\ \frac{1}{1 + e^{(\beta x_1)t}} & x_2 = 1, \text{ male} \end{cases} \end{aligned}$$

## Example 3

To mimic the cox PH model situation, we set  $\theta_1 = \exp(\beta x_1)$ ,  $\theta_2 = \exp(\beta x_1 x_2)$ , where  $x_1$  has four levels 1,2,3,4 and  $x_2$  has two levels 0,1, which is an indicator for gender,

$$x_2 = \begin{cases} 1 & \text{female} \\ 0 & \text{male} \end{cases}, \quad \theta_2 = \begin{cases} \exp(\beta x_1) & \text{female} \\ 1 & \text{male} \end{cases}$$

$$S_{T,C}(t, s; x_1, x_2) = \begin{cases} e^{-(e^{\beta x_1})t} e^{-(e^{(e^{\beta x_1 x_2})^s} - 1)((t-s)^2 + 1)} & \text{when } t \geq s \\ e^{-(e^{\beta x_1})t} e^{-(e^{(e^{\beta x_1 x_2})^s} - 1)} & \text{when } t < s \end{cases}$$

That is, for females:

$$S_{T,C}(t, s; x_1, x_2) = \begin{cases} e^{-\left(e^{\beta x_1}\right)t} e^{-\left(e^{\left(e^{\beta x_1}\right)^s - 1}\right)((t-s)^2 + 1)} & \text{when } t \geq s \\ e^{-\left(e^{\beta x_1}\right)t} e^{-\left(e^{\left(e^{\beta x_1}\right)^s - 1}\right)} & \text{when } t < s \end{cases}$$

For males:

$$S_{T,C}(t, s; x_1, x_2) = \begin{cases} e^{-\left(e^{\beta x_1}\right)t} e^{-\left(e^s - 1\right)((t-s)^2 + 1)} & \text{when } t \geq s \\ e^{-\left(e^{\beta x_1}\right)t} e^{-\left(e^s - 1\right)} & \text{when } t < s \end{cases}$$

where

$$S_T(t; x_1, x_2) = e^{-\left(e^{\beta x_1}\right)t} = [e^{-t}]^{\left(e^{\beta x_1}\right)} = [e^{-t}]^{\exp(\beta x_1)}$$

where  $S_0(t) = e^{-t}$ . Therefore,  $S_T(t; x_1, x_2)$  is a cox PH model. And

$$f_T(t; x_1, x_2) = \exp(\beta x_1) e^{-\exp(\beta x_1)t},$$

$$\lambda_T(t; x_1, x_2) = \frac{f_T(t)}{S_T(t)} = \exp(\beta x_1) = \lambda_{T0}(t) \exp(\beta x_1), \quad \lambda_{T0}(t) = 1$$

The censoring time is

$$S_C(t; x_1, x_2) = e^{-(e^{\exp(\beta x_1 x_2)s} - 1)}$$

$$f_C(s; x_1, x_2) = \exp(\beta x_1 x_2) e^{-e^{\exp(\beta x_1 x_2)s} + \exp(\beta x_1 x_2)s + 1}$$

$$\lambda_C(s; x_1, x_2) = \frac{f_C(s; x_1, x_2)}{S_C(s; x_1, x_2)} = \exp(\beta x_1 x_2) e^{\exp(\beta x_1 x_2)s}$$

The associated  $m()$  function is

$$\begin{aligned} m(t; x_1, x_2) &= \frac{\lambda_T(t; x_1, x_2)}{\lambda_T(t; x_1, x_2) + \lambda_C(t; x_1, x_2)} \\ &= \frac{\exp(\beta x_1)}{\exp(\beta x_1) + \exp(\beta x_1 x_2) e^{\exp(\beta x_1 x_2)t}} \\ &= \frac{1}{1 + \exp(-\beta x_1 + \beta x_1 x_2) \exp((\beta x_1 x_2)t)} \\ &= \frac{1}{1 + \exp(-\beta x_1 + \beta x_1 x_2 + (\beta x_1 x_2)t)} \\ &= \begin{cases} \frac{1}{1 + e^{(\beta x_1)t}} & x_2 = 1, \text{ female} \\ \frac{1}{1 + e^{-\beta x_1}} & x_2 = 0, \text{ male} \end{cases} \end{aligned}$$

## Joint CDF

We denote  $T_i, i = 1, \dots, N$  are the independent, identically, distributed (iid) lifetimes, whose corresponding cumulative distribution function (CDF) is  $F$ , probability distribution function (PDF) is  $f$ ; the censoring time is defined as  $C_i, i = 1, \dots, N$ .  $C_i$ s are also iid, with CDF denoted as  $G$  and PDF denoted as  $g$ . We set the censors happen on the right and the observed time is  $Z_i = T_i \wedge C_i$ , whose CDF is  $H$  and PDF is  $h$ . The  $\delta_i = I_{[T_i \leq C_i]}$  is the status indicator, which shows whether the event of the  $i$ th subject is censored ( $\delta_i = 0$ ) or observed ( $\delta_i = 1$ ).

$$S_{T,C}(t, s; \theta_1, \theta_2) = \begin{cases} e^{-\theta_1 t} e^{-(e^{\theta_2 s} - 1)(t-s+1)} & \text{when } t \geq s \\ e^{-\theta_1 t} e^{-(e^{\theta_2 s} - 1)} & \text{when } t < s \end{cases}$$

where

$$\begin{aligned} \theta_1 &= \beta_1^T X_1 = \beta_{11}x_{11} + \beta_{12}x_{12} + \dots + \beta_{1n_1}x_{1n_1}, \\ \theta_2 &= \beta_2^T X_2 = \beta_{21}x_{21} + \beta_{22}x_{22} + \dots + \beta_{2n_2}x_{2n_2} \end{aligned}$$

And its associated pdf is

$$f_{T,C}(t, s; \theta_1, \theta_2) = \begin{cases} \left\{ (\theta_2(t-s+1) - 1)e^{2\theta_2 y} + \theta_2(\theta_1 - 1)(t-s)e^{\theta_2 s} + (\theta_2 - 1)(\theta_1 - 2)e^{\theta_2 s} + \theta_1 - 1 \right\} \\ \times e^{-\theta_1 t} e^{-(e^{\theta_2 t} - 1)(t-s+1)} \\ \theta_1 \theta_2 e^{-e^{\theta_2 s} + \theta_2 s - \theta_1 t + 1} \end{cases}$$

when  
when

The marginal CDF and PDF of survival time and censoring time are:

$$1\text{- CDF: } S_T(t; \theta_1, \theta_2) = P(T > t) = P(T > t, C > 0) = e^{-\theta_1 t}, \quad \text{PDF: } f_T(x) = \theta_1 e^{-\theta_1 t}$$

$$1\text{- CDF: } S_C(s; \theta_1, \theta_2) = P(C > s) = P(T > 0, C > s) = e^{-(e^{\theta_2 s} - 1)}, \quad \text{PDF: } f_C(x) = \theta_2 e^{-e^{\theta_2 s} + \theta_2 s + 1}$$

Then the conditional pdf of censoring time given the death time is:

- When  $t \geq s$

$$f_{C|T=t}(s; \theta_1, \theta_2) = \frac{f_{T,C}(s, t; \theta_1, \theta_2)}{f_T(t; \theta_1, \theta_2)} =$$

$$\frac{1}{\theta_1} \left\{ (\theta_2(t-s+1) - 1)e^{2\theta_2 y} + \theta_2(\theta_1 - 1)(t-s)e^{\theta_2 s} + (\theta_2 - 1)(\theta_1 - 2)e^{\theta_2 s} + \theta_1 - 1 \right\} e^{-(e^{\theta_2 s} - 1)(t-s+1)}$$

$$F_{C|T=t}(s; \theta_1, \theta_2) = \int_0^s f_{C|T=t}(s; \theta_1, \theta_2) ds =$$

- When  $t < s$

$$f_{C|T=t}(s; \theta_1, \theta_2) = \frac{f_{T,C}(s, t; \theta_1, \theta_2)}{f_T(t; \theta_1, \theta_2)} = \frac{\theta_1 \theta_2 e^{-e^{\theta_2 s} + \theta_2 s - \theta_1 t + 1}}{\theta_1 e^{-\theta_1 t}} = \theta_2 e^{-e^{\theta_2 s} + \theta_2 s + 1}$$

$$F_{C|T=t}(s; \theta_1, \theta_2) = \int_t^s f_{C|T=t}(s; \theta_1, \theta_2) ds = e^{1-e^{\theta_2 t}} - e^{1-e^{\theta_2 s}}$$

### parameter setting

For example, if we set

$$\theta_1 = \exp(\beta_0 + \beta_1 x_1) = \exp(1 + x_1), \theta_2 = 1$$

where  $\beta_0 = \beta_1 = 1$

$$S_{T,C}(t, s; \theta_1, \theta_2) = \begin{cases} e^{-\theta_1 t} e^{-(e^s - 1)(t - s + 1)} & \text{when } t \geq s \\ e^{-\theta_1 t} e^{-(e^s - 1)} & \text{when } t < s \end{cases}$$

And its associated pdf is

$$f_{T,C}(t, s; \theta_1, \theta_2) = \begin{cases} \left\{ ((t - s + 1) - 1)e^{2y} + (\theta_1 - 1)(t - s)e^s + \theta_1 - 1 \right\} \\ \times e^{-\theta_1 t} e^{-(e^s - 1)(t - s + 1)} & \text{when } t \geq s \\ \theta_1 \theta_2 e^{-e^{\theta_2 s} + \theta_2 s - \theta_1 t + 1} & \text{when } t < s \end{cases}$$

$$(((x-y + 1) - 1)\exp(2y) + (a-1)(x-y)\exp(y) + a-1) \exp(-(\exp(t) - 1)(x-y + 1))$$

The marginal CDF and PDF of survival time and censoring time are:

$$1\text{- CDF: } S_T(t; \theta_1, \theta_2) = P(T > t) = P(T > t, C > 0) = e^{-\theta_1 t}, \quad \text{PDF: } f_T(x) = \theta_1 e^{-\theta_1 x}$$

$$1\text{- CDF: } S_C(s; \theta_1, \theta_2) = P(C > s) = P(T > 0, C > s) = e^{-(e^s - 1)}, \quad \text{PDF: } f_C(x) = e^{-e^s + s + 1}$$

Then the conditional pdf of censoring time given the death time is:

- When  $t \geq s$

$$f_{C|T=t}(s; \theta_1, \theta_2) = \frac{f_{T,C}(s, t; \theta_1, \theta_2)}{f_T(t; \theta_1, \theta_2)} = \frac{1}{\theta_1} \left\{ (t - s)e^{2y} + (\theta_1 - 1)(t - s)e^s + \theta_1 - 1 \right\} e^{-(e^s - 1)(t - s + 1)}$$

$$F_{C|T=t}(s; \theta_1, \theta_2) = \int_0^s f_{C|T=t}(s; \theta_1, \theta_2) ds = 1 - \frac{1}{\theta_1} (e^s + \theta_1 - 1) e^{-(e^s - 1)(t - s + 1)}$$

- When  $t < s$

$$f_{C|T=t}(s; \theta_1, \theta_2) = \frac{f_{T,C}(s, t; \theta_1, \theta_2)}{f_T(t; \theta_1, \theta_2)} = \frac{\theta_1 e^{-e^s + s - \theta_1 t + 1}}{\theta_1 e^{-\theta_1 t}} = e^{-e^s + s + 1}$$

$$F_{C|T=t}(s; \theta_1, \theta_2) = \int_t^s f_{C|T=t}(s; \theta_1, \theta_2) ds = e^{1 - e^t} - e^{1 - e^s}$$

$$m(t; \theta_1, \theta_2) = \frac{\theta_1}{\theta_1 + \theta_2 e^{\theta_2 t}} = \frac{1}{1 + \frac{\theta_2}{\theta_1} e^{\theta_2 t}} = \frac{1}{1 + e^{-\beta_0 - \beta_1 x_1} e^t} = \frac{1}{1 + e^{-\beta_0 - \beta_1 x_1 + t}}$$