example 6 two weibull

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Two weibull distribution, mimic Dikta's example

The joint distribution

$$S_{T,C}(x,y) = \begin{cases} \exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1}) \exp\left(-(\frac{\theta_2}{\lambda_2}y)^{k_2}(x-y+1)\right) & x \ge y\\ \exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1}) \exp(-(\frac{\theta_2}{\lambda_2}y)^{k_2}) & x < y \end{cases}$$

Then the joint pdf when x < y is

$$f_{T,C}(x,y) = k_1 k_2 (\frac{\theta_1}{\lambda_1})^{k_1} (\frac{\theta_2}{\lambda_2})^{k_2} x^{k_1 - 1} y^{k_2 - 1} \exp(-(\frac{\theta_1}{\lambda_1} x)^{k_1} - (\frac{\theta_2}{\lambda_2} y)^{k_2})$$

And then the survival function is

$$S_T(x) = P(T > x, C > 0) = \exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})$$

which is a weibull distribution with shape parameter k_1 and scale parameter λ_1 . The θ_1 is the parameter associated with the covariates X.

The pdf of the survival function is

$$f_T(x) = k_1 (\frac{\theta_1}{\lambda_1})^{k_1} x^{k_1 - 1} \exp(-(\frac{\theta_1}{\lambda_1} x)^{k_1})$$

The CDF of the observed time is

$$S_H(x) = P(T > x, C > x) = \exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})\exp(-(\frac{\theta_2}{\lambda_2}x)^{k_2})$$

The pdf of the observed time is

$$f_H(x) = \left(k_1 \left(\frac{\theta_1}{\lambda_1}\right)^{k_1} x^{k_1 - 1} + k_2 \left(\frac{\theta_2}{\lambda_2}\right)^{k_2} x^{k_2 - 1}\right) \exp\left(-\left(\frac{\theta_1}{\lambda_1}x\right)^{k_1} - \left(\frac{\theta_2}{\lambda_2}x\right)^{k_2}\right)$$

The
$$\psi(x) = \int_x^\infty f_{T,C}(x,y) dy = k_1(\frac{\theta_1}{\lambda_1})^{k_1} x^{k_1 - 1} \exp(-(\frac{\theta_1}{\lambda_1} x)^{k_1}) \exp(-(\frac{\theta_2}{\lambda_2} x)^{k_2})$$

Therefore, the $\rho(x)$ function equals to 1.

$$\rho(x) = \frac{f_T(x)/\psi(x) - 1}{S_T(x)/S_H(x) - 1} = \frac{\frac{k_1(\frac{\theta_1}{\lambda_1})^{k_1}x^{k_1 - 1}\exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})}{\frac{k_1(\frac{\theta_1}{\lambda_1})^{k_1}x^{k_1 - 1}\exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})\exp(-(\frac{\theta_2}{\lambda_2}x)^{k_2})}} - 1}{\frac{\exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})\exp(-(\frac{\theta_2}{\lambda_2}x)^{k_2})}{\exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})\exp(-(\frac{\theta_2}{\lambda_2}x)^{k_2})}} - 1} = 1$$

The m(x) function is

$$m(x) = \frac{f_T(x)/S_T(x)}{f_H(x)/S_H(x)} = \frac{\frac{k_1(\frac{\theta_1}{\lambda_1})^{k_1}x^{k_1-1} \exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})}{\exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})}}{\frac{\exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})}{\exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1})}} = \frac{k_1(\frac{\theta_1}{\lambda_1})^{k_1}x^{k_1-1} + k_2(\frac{\theta_2}{\lambda_2})^{k_2}x^{k_2-1}) \exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1}) \exp(-(\frac{\theta_2}{\lambda_2}x)^{k_2})}}{\exp(-(\frac{\theta_1}{\lambda_1}x)^{k_1}) \exp(-(\frac{\theta_2}{\lambda_2}x)^{k_2})}} = \frac{k_1(\frac{\theta_1}{\lambda_1})^{k_1}x^{k_1-1}}{k_1(\frac{\theta_1}{\lambda_1})^{k_1}x^{k_1-1} + k_2(\frac{\theta_2}{\lambda_2})^{k_2}x^{k_2-1}}} = \frac{1}{1 + \frac{k_2}{k_1}(\frac{\theta_2}{\lambda_2})^{k_2}(\frac{\theta_1}{\lambda_1})^{k_1}x^{k_2-k_1}}}$$

The censoring percentage is

$$P(T < C) = \int_0^\infty \int_0^y k_1 k_2 (\frac{\theta_1}{\lambda_1})^{k_1} (\frac{\theta_2}{\lambda_2})^{k_2} x^{k_1 - 1} y^{k_2 - 1} \exp(-(\frac{\theta_1}{\lambda_1} x)^{k_1} - (\frac{\theta_2}{\lambda_2} y)^{k_2}) dx dy$$

To make it sample, if $k_1 = k_2 = 1$, the missing precentage is

$$\frac{\frac{\theta_1}{\lambda_1}}{\frac{\theta_1}{\lambda_1} + \frac{\theta_2}{\lambda_2}}$$

Let's then plug in numbers and make a simple example.

$$S_{T,C}(t,s|x) = \begin{cases} \exp(-(\beta^T x)t) \exp\left(-((\beta^T x)s)(t-s+1)\right) & x \ge y\\ \exp(-(\beta^T x)t) \exp(-(\beta^T x)s) & x < y \end{cases}$$

The censoring percentage is 0.5.

The m function is

$$m(t|x) = \frac{1}{1 + (\beta^T x)^2}$$