Parameterization of copula functions for bivariate survival data in the **surrosurv** package.

Modelling and simulation

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Let define the joint survival function of S and T via a copula function:

$$S(s,t) = P(S > s, T > t) = C(u,v)|_{u=S_S(s),v=S_T(t)},$$
(1)

where $S_S(\cdot) = P(S > s)$ and $S_T(\cdot) = P(T > t)$ are the marginal survival functions of S and T.

Modelling

In the case of possibly right-censored data, the individual contribution to the likelihood is

- $S(s,t) = C(u,v)|_{S_S(s),S_T(t)}$ if S is censored at time s and T is censored at time t,
- $-\frac{\partial}{\partial t}S(s,t) = \frac{\partial}{\partial v}C(u,v)\big|_{S_S(s),S_T(t)} f_S(s)$ if S is censored at time s and T=t,
- $-\frac{\partial}{\partial s}S(s,t) = \frac{\partial}{\partial v} \left. C(u,v) \right|_{S_S(s),S_T(t)} f_T(t)$ if S=s and T is censored at time t,
- $\frac{\partial^2}{\partial s \partial t} S(s,t) = \frac{\partial^2}{\partial u \partial v} C(u,v) \Big|_{S_S(s),S_T(t)} f_S(s) f_t(t) \text{ if } S = s \text{ and } T = t.$

Clayton copula

The bivariate Clayton [1978] copula is defined as

$$C(u,v) = (u^{-\theta} + v^{-\theta} - 1)^{-1/\theta}, \qquad \theta > 0.$$
 (2)

The first derivative with respect to u is

$$\frac{\partial}{\partial u}C(u,v) = \left(u^{-\theta} + v^{-\theta} - 1\right)^{-\frac{1+\theta}{\theta}}u^{-(1+\theta)}$$

$$= \left[\frac{C(u,v)}{u}\right]^{1+\theta}.$$
(3)

The second derivative with respect to u and v is

$$\frac{\partial^2}{\partial u \partial v} C(u, v) = (1 + \theta) \frac{C(u, v)^{1+2\theta}}{(uv)^{1+\theta}}.$$
(4)

The Kendall [1938]'s tau for the Clayton copula is

$$\tau = \frac{\theta}{\theta + 2}. ag{5}$$

Plackett copula

The bivariate Plackett [1965] copula is defined as

$$C(u,v) = \frac{[Q - R^{1/2}]}{2(\theta - 1)}, \qquad \theta > 0,$$
(6)

with

$$Q = 1 + (\theta - 1)(u + v),$$

$$R = Q^2 - 4\theta(\theta - 1)uv.$$
(7)

Given that

$$\frac{\partial}{\partial u}Q = \theta - 1,\tag{8}$$

$$\frac{\partial}{\partial u}R = 2(\theta - 1)\left(1 - (\theta + 1)v + (\theta - 1)u\right),\tag{9}$$

the first derivative of C(u, v) with respect to u is

$$\frac{\partial}{\partial u}C(u,v) = \frac{1}{2} \left[1 - \frac{1 - (\theta + 1)v + (\theta - 1)u}{R^{1/2}} \right]
= \frac{1}{2} \left[1 - \frac{Q - 2\theta v}{R^{1/2}} \right].$$
(10)

By defining

$$f = 1 - (\theta + 1)v + (\theta - 1)u, (11)$$

$$g = R^{1/2} \tag{12}$$

and given that

$$f' = \frac{\partial}{\partial u} f = -(\theta + 1),\tag{13}$$

$$g' = \frac{\partial}{\partial u}g = \frac{\theta - 1}{R^{1/2}} \Big(1 - (\theta + 1)u + (\theta - 1)v \Big),\tag{14}$$

then, the second derivative with respect to u and v is

$$\frac{\partial^2}{\partial u \partial v} C(u, v) = -\frac{f'g - fg'}{2g^2}$$

$$= -\frac{\theta}{R^{3/2}} \left[1 + (\theta - 1)(u + v - 2uv) \right]$$

$$= -\frac{\theta}{R^{3/2}} \left[Q - 2(\theta - 1)uv \right].$$
(15)

The Kendall's tau for the Plackett copula cannot be computed analytically and is obtained numerically.

Gumbel-Hougaard copula

The bivariate ?-Hougaard [1986] copula is defined as

$$C(u,v) = \exp\left(-Q^{\theta}\right), \qquad \theta \in (0,1), \tag{16}$$

with

$$Q = (-\ln u)^{1/\theta} + (-\ln v)^{1/\theta}.$$
(17)

Given that

$$\frac{\partial}{\partial u}Q = -\frac{(-\ln u)^{1/\theta - 1}}{\theta u},\tag{18}$$

then, the first derivative with respect to u is

$$\frac{\partial}{\partial u}C(u,v) = \frac{(-\ln u)^{1/\theta - 1}}{u}C(u,v)Q^{\theta - 1} \tag{19}$$

and the second derivative with respect to u and v is

$$\frac{\partial^2}{\partial u \partial v} C(u, v) = \frac{\left[(-\ln u)(-\ln v) \right]^{1/\theta - 1}}{uv} C(u, v) Q^{\theta - 2} \left[\frac{1}{\theta} - 1 + Q^{\theta} \right]. \tag{20}$$

The Kendall's tau for the Gumbel-Hougaard copula is

$$\tau = 1 - \theta. \tag{21}$$

Simulation

The function simData.cc() generates data from a Clayton copula model. First, the time value for the surrogate endpoint S is generated from its (exponential) marginal survival function:

$$S = -\log(U_S/\lambda_S), \quad \text{with } U_S := S_S(S) \sim U(0, 1). \tag{22}$$

Then, the time value for the true endpoint T is generated conditionally on the value s of S. The conditional survival function of $T \mid S$ is

$$S_{T|S}(t \mid s) = \frac{-\frac{\partial}{\partial s}S(s,t)}{-\frac{\partial}{\partial s}S(s,0)} = \frac{\frac{\partial}{\partial u}C(u,v)}{\frac{\partial}{\partial u}C(u,1)}$$
(23)

As the Clayton copula is used, we get (see Equtaion 3)

$$S_{T|S}(t \mid s) = \left[\frac{C(S_S(s), S_T(t))}{C(S_S(s), 1)}\right]^{1+\theta} = \left[\frac{U_S^{-\theta} + S_T(t)^{-\theta} - 1}{U_S^{-\theta}}\right]^{\frac{1+\theta}{\theta}}$$
$$= \left[1 + U_S^{\theta}(S_T(t)^{-\theta} - 1)\right]^{\frac{1+\theta}{\theta}}$$
(24)

By generating uniform random values for $U_T := S_{T|S}(T \mid s) \sim U(0,1)$, the values for $T \mid S$ are obtained as follows:

$$U_{T} = \left[1 + U_{S}^{\theta}(S_{T}(T)^{-\theta} - 1)\right]^{\frac{1+\theta}{\theta}}$$

$$S_{T}(T) = \left[\left(U_{T}^{-\frac{\theta}{1+\theta}} - 1\right)U_{S}^{-\theta} + 1\right]^{-1/\theta}$$

$$T = -\log(U_{T}'/\lambda_{T}), \quad \text{with } U_{T}' = \left[\left(U_{T}^{-\frac{\theta}{1+\theta}} - 1\right)U_{S}^{-\theta} + 1\right]^{-1/\theta}.$$
(25)

References

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