

of the unit square, and is such that $\alpha(C)$ is well defined for any bivariate copula C . Examples are measures of monotone association such as Spearman's ρ_S and the correlation of normal scores ρ_N . The theory below can be adapted to other types of functionals.

Assume that a is twice continuously differentiable in the region where it is non-zero, and that $a_{12} = \frac{\partial^2 a}{\partial u \partial v}$ is integrable on $(0, 1)^2$. With (u_0, v_0) and (u, v) in the interior of the unit square (or on the boundary if $a(\cdot)$ is finite everywhere),

$$a(u, v) - a(u_0, v) - a(u, v_0) + a(u_0, v_0) = \int_{u_0}^u \int_{v_0}^v a_{12}(s, t) \, ds dt.$$

After substituting for $a(u, v)$ and changing the order of integration, (5.15) is the same as:

$$\begin{aligned} \alpha^*(C) &= \int_0^1 \int_0^1 a_{12}(s, t) C(s, t) \, ds dt + \int_{u_0}^1 \int_{v_0}^1 a_{12}(s, t) \, ds dt \\ &\quad - \int_{u_0}^1 \int_0^1 t a_{12}(s, t) \, ds dt - \int_0^1 \int_{v_0}^1 s a_{12}(s, t) \, ds dt \\ &\quad + \int_0^1 a(u_0, v) \, dv + \int_0^1 a(u, v_0) \, du - a(u_0, v_0). \end{aligned} \tag{5.16}$$

The sample version of (5.15) is

$$\begin{aligned} \hat{\alpha} &= \alpha(\hat{C}_n) = \int_0^1 \int_0^1 a(u, v) \, d\hat{C}_n(u, v) = n^{-1} \sum_{i=1}^n a\left(\frac{r_{i1} - \frac{1}{2}}{n}, \frac{r_{i2} - \frac{1}{2}}{n}\right) \\ &= \alpha^*(\hat{C}_n) + O_p(n^{-1}) \\ &= \int_0^1 \int_0^1 a(u, v) \, d\tilde{C}_n(u, v) + O_p(n^{-1}). \end{aligned}$$

Note that with $S_{in} = [(r_{i1} - 1)/n, r_{i1}/n] \times [(r_{i2} - 1)/n, r_{i2}/n]$,

$$\int_{(r_{i1}-1)/n}^{r_{i1}/n} \int_{(r_{i2}-1)/n}^{r_{i2}/n} a(u, v) \, d\hat{C}_n(u, v) = n^{-1} a\left(\frac{r_{i1} - \frac{1}{2}}{n}, \frac{r_{i2} - \frac{1}{2}}{n}\right),$$

and from a Taylor expansion to first order,

$$\left| \int_{S_{in}} a(u, v) \tilde{C}_n(u, v) \, dudv - n^{-1} a\left(\frac{r_{i1} - \frac{1}{2}}{n}, \frac{r_{i2} - \frac{1}{2}}{n}\right) \right| \leq n^{-2} \max_{(u,v) \in S_{in}} \left\{ \frac{\partial a}{\partial u}, \frac{\partial a}{\partial v} \right\}.$$

For asymptotics, with the continuous mapping theorem,

$$\begin{aligned} n^{1/2}[\alpha(\hat{C}_n) - \alpha(C)] &= \int_{[0,1]^2} a_{12}(u, v) n^{1/2} [\hat{C}_n(u, v) - C(u, v)] \, dudv + O_p(n^{-1/2}) \\ &\rightarrow_d \int_{[0,1]^2} a_{12}(u, v) \mathcal{G}_C(u, v) \, dudv. \end{aligned}$$

Hence $\alpha(\hat{C}_n)$ is asymptotically normal. If (5.16) doesn't hold, asymptotic normality holds under some conditions and the asymptotic representation has a different form.

An example of a tail-weighted measure of dependence is $\text{Cor}[b(U), b(V)|U > \frac{1}{2}, V > \frac{1}{2}]$ if b is increasing on $[\frac{1}{2}, 1]$ with $b(\frac{1}{2}) = 0$. For the upper semi-correlation of normal scores, $b(u) = \Phi^{-1}(u)$ for $\frac{1}{2} \leq u < 1$ (but the second version α^* after integration by parts might not be valid). This conditional correlation involves $\mathbb{E}[b^{m_1}(U)b^{m_2}(V)|U > \frac{1}{2}, V > \frac{1}{2}]$. with $(m_1, m_2) \in \{(1, 1), (2, 0), (0, 2), (1, 0), (0, 1)\}$, that is, (5.15) for five different $a(\cdot)$ functions