

# NOTES: INTRODUCTION TO COPULA THEORY

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## CHAPTER 1

### Preliminaries

**THEOREM 1.0.1.** Prove that  $\mathbb{E}[X] = a$  when  $a$  is a point of symmetry of  $X$ .

**PROOF.** We have two method,

(1) Suppose that  $\mathbb{E}[X]$  exists, then

$$\begin{aligned}\mathbb{E}[X] &= \int_{\mathbb{R}} x f(x) dx \\ &= \int_{\mathbb{R}} (a - a + x) f(x) dx \\ &= a + \int_{-\infty}^{\infty} (x - a) f(x) dx \\ &= a + \int_{-\infty}^{\infty} y f(a + y) dy \\ &= a + \int_{-\infty}^0 y f(a + y) dy + \int_0^{\infty} y f(a + y) dy \\ &= a + \int_{\infty}^0 (-1)(-z) f(a - z) dz + \int_0^{\infty} y f(a + y) dy \\ &= a + \int_0^{\infty} (-z) f(a + z) dz + \int_0^{\infty} y f(a + y) dy = a.\end{aligned}$$

(2) Assume that the expectation exists, then by definition of symmetry,  $X - a$  and  $a - X$  have the same distribution, then

$$\mathbb{E}[X - a] = \mathbb{E}[a - X] \implies \mathbb{E}[X] - a = a - \mathbb{E}[X] \implies \mathbb{E}[X] = a.$$

□

**EXAMPLE 1.0.2.** Let  $X_{(1)}, X_{(2)}$  denote the minimum and maximum of  $\{X_1, X_2\}$ . Given that  $F_{X_{(1)}}(x) = 1 - (1 - F(x))^2$  and  $F_{X_{(2)}}(x) = F(x)^2$ . For  $x_1 \geq x_2$ , we have

$$\begin{aligned} P(X_{(1)} \leq x_1, X_{(2)} \leq x_2) &= P(X_{(2)} \leq x_2) - P(X_{(1)} > x_1, X_{(2)} \leq x_2) \\ &= F(x_2)^2. \end{aligned}$$

For  $x_1 < x_2$ , we have

$$\begin{aligned} P(X_{(1)} \leq x_1, X_{(2)} \leq x_2) &= P(X_{(2)} \leq x_2) - P(X_{(1)} > x_1, X_{(2)} \leq x_2) \\ &= F(x_2)^2 - (F(x_2) - F(x_1))^2 \\ &= 2F(x_2)F(x_1) - F(x_1)^2. \end{aligned}$$

Thus

$$P(X_{(1)} \leq x_1, X_{(2)} \leq x_2) = 2F(\min\{x_1, x_2\})F(x_2) - F(\min\{x_1, x_2\})^2.$$

□

**PROPOSITION 1.0.3.** Assume  $F_X$  is continuous and increasing, define  $Y = F_X(x)$  and note that  $Y$  takes values in  $[0, 1]$ . Then  $Y$  follows Uniform  $[0, 1]$ .

**PROOF.** Write

$$F_Y(x') = P(F_X(x) \leq x') = P(x \leq F_X^{-1}(x')) = F_X(F_X^{-1}(x')) = x'.$$

On the other hand, if  $U$  is a uniform random variable that takes values in  $[0, 1]$ ,

$$F_U(x') = \int_{\mathbb{R}} f_U(u) du = \int_0^{x'} 1 du = x',$$

thus,  $F_Y(x') = F_U(x')$  for every  $x \in [0, 1]$ . □

**EXERCISE 1.0.4.** The first-order derivative of joint CDF is

$$\begin{aligned} \frac{\partial}{\partial x} F_{XY}(x, y) &= \frac{\partial}{\partial x} \int_{-\infty}^x \int_{-\infty}^y f_{XY}(t_1, t_2) dt_2 dt_1 \\ &= \int_{-\infty}^y f_{XY}(x, t_2) dt_2 \end{aligned}$$

$$\begin{aligned} &= \int_{-\infty}^y f_{Y|X}(t_2|x) f_X(x) dt_2 \\ &= \int_{-\infty}^y f_{Y|X} f(t_2|x) dt_2 \cdot f_X(x) \\ &= P(Y \leq y | X = x) \cdot f_X(x). \end{aligned}$$

□

## CHAPTER 2

### Exercise

**Exercise 2.1** (1) Let  $H$  be the function defined on  $I^2$  by  $H(x, y) = \max(x, y)$ . Clearly, let  $x_1 \leq x_2, y_1 \leq y_2$ ,

$$H(x_2, y) - H(x_1, y) = \max(x_2, y) - \max(x_1, y) \geq 0,$$

and

$$H(x, y_2) - H(x, y_1) = \max(x, y_2) - \max(x, y_1) \geq 0,$$

Thus  $H$  is non-decreasing in each argument. However,

$$\begin{aligned} V_H(I^2) &= H(1, 1) + H(0, 0) - H(0, 1) - H(1, 0) \\ &= \max(1, 1) + \max(0, 0) - \max(1, 0) - \max(0, 1) \\ &= -1 < 0. \end{aligned}$$

Therefore,  $H$  is not 2-increasing.

(2) Let  $H$  be the function defined on  $I^2$  by  $H(x, y) = (2x - 1)(2y - 1)$ . WLOG, let  $B \in I^2$  with vertices  $(x_1, y_1), (x_2, y_2) \in I^2$  such the  $x_1 \leq x_2, y_1 \leq y_2$ ,

$$\begin{aligned} V_H(B) &= H(x_1, y_1) + H(x_2, y_2) - H(x_1, y_2) - H(x_2, y_1) \\ &= (2x_1 - 1)(2y_1 - 1) + (2x_2 - 1)(2y_2 - 1) - (2x_1 - 1)(2y_2 - 1) - (2x_2 - 1)(2y_1 - 1) \\ &= (2x_1 - 1)(2y_1 - 2y_2) + (2x_2 - 1)(2y_2 - 2y_1) \\ &= (2y_2 - 2y_1)(2x_2 - 2x_1) \geq 0. \end{aligned}$$

Thus  $H$  is 2-increasing, but

$$\frac{\partial H(x, y)}{\partial x} = 4y - 2.$$

It indicates that  $\frac{\partial H(x,y)}{\partial x} < 0$  when  $y \in [0, 1/2)$ . Thus  $H$  is decreasing in  $x$  when  $y \in [0, 1/2)$ . Similarly,  $H$  is decreasing in  $y$  when  $x \in [0, 1/2)$ .  $\square$

**Exercise 2.2** Show that  $M(u, v) = \min(u, v)$ ,  $W(u, v) = \max(u + v - 1, 0)$  and  $\Pi(u, v) = uv$  are indeed copulas.

*Solution.* Let's check the two axioms of a copula.  $\circledast$

$$M(0, v) = \min(0, v) = 0 = \min(u, 0) = M(u, 0),$$

$$M(1, v) = \min(1, v) = v, \quad M(u, 1) = \min(u, 1) = u.$$

and

$$W(0, v) = \max(v - 1, 0) = 0 = \max(u - 1, 0) = W(u, 0),$$

$$W(1, v) = \max(v, 0) = v, \quad W(u, 1) = \max(u, 0) = u.$$

and

$$\Pi(0, v) = 0 \cdot v = 0 = u \cdot 0 = \Pi(u, 0),$$

$$\Pi(1, v) = 1 \cdot v = v, \quad \Pi(u, 1) = u \cdot 1 = u.$$

$\circledast$  Then check the 2-increase of them. For every  $(u_1, v_1), (u_2, v_2) \in I^2$  such that  $u_1 \leq u_2, v_1 \leq v_2$ ,

$$\begin{aligned} & M(u_1, v_1) + M(u_2, v_2) - M(u_1, v_2) - M(u_2, v_1) \\ &= \min(u_1, v_1) + \min(u_2, v_2) - \min(u_1, v_2) - \min(u_2, v_1) = \beta. \end{aligned}$$

If  $u_1 \leq v_1$ ,

$$\beta = u_1 + \min(u_2, v_2) - \min(u_2, v_1) - u_1 \geq 0.$$

If  $u_1 > v_1$ ,

$$\beta = v_1 + \min(u_2, v_2) - \min(u_1, v_2) - v_1 \geq 0.$$

And

$$\alpha = W(u_1, v_1) + W(u_2, v_2) - W(u_1, v_2) - W(u_2, v_1)$$

$$= \max(u_1 + v_1 - 1, 0) + \max(u_2 + v_2 - 1, 0) - \max(u_1 + v_2 - 1, 0) - \max(u_2 + v_1 - 1, 0).$$

If  $u_1 + v_2 \geq 1$ ,

$$\alpha = \max(u_1 + v_1 - 1, 0) + \max(u_2 + v_2 - 1, 0) - \max(u_2 + v_1 - 1, 0) - 0 \geq 0.$$

If  $u_1 + v_2 < 1$ ,

$$\alpha = 0 + \max(u_2 + v_2 - 1, 0) - \max(u_2 + v_1 - 1, 0) + 0 \geq 0.$$

And

$$\begin{aligned} \gamma &= \Pi(u_1, v_1) + \Pi(u_2, v_2) - \Pi(u_1, v_2) - \Pi(u_2, v_1) \\ &= u_1 v_1 + u_2 v_2 - u_1 v_2 - u_2 v_1 \\ &= (u_2 - u_1)(v_2 - v_1) \geq 0. \end{aligned}$$

□

**Exercise 2.3** (a) Let  $C_0$  and  $C_1$  be copulas, and let  $\theta$  be any number in  $I$ . Show that the weighted arithmetic mean  $(1 - \theta)C_0 + \theta C_1$  is also a copula. Hence conclude that *any convex linear combination of copulas is a copula*.

(b) Show that *the geometric mean of two copulas may fail to be a copula*.

*Solution.* (a) Check the two axioms of copula.  $\circledast$  The groundedness and uniform margins are trivial.  $\circledast$  The 2-increase is also trivial as  $(1 - \theta)V_{C_0}(B) + \theta V_{C_1}(B) \geq 0$ , where  $B$  is a rectangle in  $I^2$ .

(b) Let  $C$  be the geometric mean of  $\Pi$  and  $W$ , then the  $C$ -volume of rectangle  $[1/2, 3/4] \times [1/2, 3/4]$  is

$$\begin{aligned} &\sqrt{\Pi(1/2, 1/2)W(1/2, 1/2)} + \sqrt{\Pi(3/4, 3/4)W(3/4, 3/4)} - \sqrt{\Pi(1/2, 3/4)W(1/2, 3/4)} \\ &- \sqrt{\Pi(3/4, 1/2)W(3/4, 1/2)} = \sqrt{1/4 \cdot 0} + \sqrt{9/16 \cdot 1/2} - \sqrt{3/8 \cdot 1/4} - \sqrt{3/8 \cdot 1/4} \\ &= 1/2(\sqrt{9/8} - \sqrt{12/8}) < 0. \end{aligned}$$

Therefore  $C$  is not a copula.

□



**Exercise 2.4** The Fréchet and Mardia families of copulas. (a) Let  $\alpha, \beta$  be in  $I$  with  $\alpha + \beta \leq 1$ . Set

$$C_{\alpha,\beta}(u, v) = \alpha M(u, v) + (1 - \alpha - \beta)\Pi(u, v) + \beta W(u, v).$$

Show that  $C_{\alpha,\beta}$  is a copula (**Fréchet**). A family of copulas that includes  $M, \Pi, W$  is called **comprehensive**.

(b) Let  $\theta$  be in  $[-1, 1]$  and set

$$C_\theta(u, v) = \frac{\theta^2(1 + \theta)}{2}M(u, v) + (1 - \theta^2)\Pi(u, v) + \frac{\theta^2(1 - \theta)}{2}W(u, v).$$

Show that  $C_\theta$  is a copula (**Mardia**).

*Solution.* (a) Since convex combination of copulas is a copula. And  $\alpha, \beta, 1 - \alpha - \beta \in I$  with sum to 1. Thus  $C_{\alpha,\beta}$  is a convex combination of the three copulas, it is indeed a copula.

(b) Similarly to part (a),  $C_\theta$  is a convex combination of copulas. □

**Exercise 2.5** The **Cuadras-Augé** family of copulas. Let  $\theta \in I$ , and set

$$C_\theta(u, v) = [\min(u, v)]^\theta [uv]^{1-\theta} = \begin{cases} uv^{1-\theta}, & u \leq v, \\ u^{1-\theta}v, & u \geq v. \end{cases}$$

Show that  $C_\theta$  is a copula. Note that,  $C_0 = \Pi$  and  $C_1 = M$ . This family is weighted geometric mean of  $M$  and  $\Pi$ .

*Solution.* Check the two axioms of copula.  $\circledast$

$$C_\theta(u, 0) = 0 = C_\theta(0, v), \quad C_\theta(u, 1) = u, \quad C_\theta(1, v) = v.$$

$\circledast$  And for every  $(u_1, v_1), (u_2, v_2) \in I^2$  with  $u_1 \leq u_2, v_1 \leq v_2$ ,

$$\alpha = [\min(u_1, v_1)]^\theta [u_1 v_1]^{1-\theta} + [\min(u_2, v_2)]^\theta [u_2 v_2]^{1-\theta} - [\min(u_1, v_2)]^\theta [u_1 v_2]^{1-\theta} - [\min(u_2, v_1)]^\theta [u_2 v_1]^{1-\theta}.$$

If  $u_2 \leq v_1$ ,

$$\alpha = u_1 v_1^{1-\theta} + u_2 v_2^{1-\theta} - u_1 v_2^{1-\theta} - u_2 v_1^{1-\theta} = (u_2 - u_1)(v_2^{1-\theta} - v_1^{1-\theta}) \geq 0.$$

If  $v_2 > u_2 > v_1 > u_1$ ,

$$\begin{aligned}
\alpha &= u_1 v_1^{1-\theta} + u_2 v_2^{1-\theta} - u_1 v_2^{1-\theta} - u_2^{1-\theta} v_1 \\
&= [u_2 v_2^{1-\theta} - u_1 v_2^{1-\theta}] + v_1 (u_1 v_1^{-\theta} - u_2 u_2^{-\theta}) \\
&\geq [u_2 v_2^{1-\theta} - u_1 v_2^{1-\theta}] + v_1 (u_1 u_2^{-\theta} - u_2 u_2^{-\theta}) \\
&= (u_2 - u_1) (v_2 v_2^{-\theta} - v_1 u_2^{-\theta}) \\
&\geq (u_2 - u_1) (v_2 v_2^{-\theta} - v_1 v_1^{-\theta}) \geq 0.
\end{aligned}$$

If  $v_2 > u_2 > u_1 > v_1$ ,

$$\begin{aligned}
\alpha &= u_1^{1-\theta} v_1 + u_2 v_2^{1-\theta} - u_1 v_2^{1-\theta} - u_2^{1-\theta} v_1 \\
&= [u_2 v_2^{1-\theta} - u_1 v_2^{1-\theta}] + v_1 (u_1 u_1^{-\theta} - u_2 u_2^{-\theta}) \\
&\geq [u_2 v_2^{1-\theta} - u_1 v_2^{1-\theta}] + v_1 (u_1 u_2^{-\theta} - u_2 u_2^{-\theta}) \\
&= (u_2 - u_1) (v_2 v_2^{-\theta} - v_1 u_2^{-\theta}) \\
&\geq (u_2 - u_1) (v_2 v_2^{-\theta} - v_1 v_1^{-\theta}) \geq 0.
\end{aligned}$$

If  $u_2 > v_2 > v_1 > u_1$ ,

$$\begin{aligned}
\alpha &= u_1 v_1^{1-\theta} + u_2^{1-\theta} v_2 - u_1 v_2^{1-\theta} - u_2^{1-\theta} v_1 \\
&= (v_2 - v_1) u_2 u_2^{-\theta} + u_1 (v_1 v_1^{-\theta} - v_2 v_2^{-\theta}) \\
&\geq (v_2 - v_1) u_2 u_2^{-\theta} + u_1 (v_1 v_2^{-\theta} - v_2 v_2^{-\theta}) \\
&= (v_2 - v_1) (u_2 u_2^{-\theta} - u_1 v_2^{-\theta}) \\
&\geq (v_2 - v_1) (u_2 u_2^{-\theta} - u_1 u_1^{-\theta}) \geq 0.
\end{aligned}$$

If  $u_2 > v_2 > u_1 > v_1$ ,

$$\begin{aligned}
\alpha &= u_1^{1-\theta} v_1 + u_2^{1-\theta} v_2 - u_1 v_2^{1-\theta} - u_2^{1-\theta} v_1 \\
&= (v_2 - v_1) u_2 u_2^{-\theta} + u_1 (v_1 u_1^{-\theta} - v_2 v_2^{-\theta}) \\
&\geq (v_2 - v_1) u_2 u_2^{-\theta} + u_1 (v_1 v_2^{-\theta} - v_2 v_2^{-\theta})
\end{aligned}$$

$$\begin{aligned}
&= (v_2 - v_1)(u_2 u_2^{-\theta} - u_1 v_2^{-\theta}) \\
&\geq (v_2 - v_1)(u_2 u_2^{-\theta} - u_1 u_1^{-\theta}) \geq 0.
\end{aligned}$$

□

**Exercise 2.6** Let  $C$  be a copula, and let  $(a, b)$  be any point in  $I^2$ . For  $(u, v)$  in  $I^2$ , define

$$K_{a,b}(u, v) = V_C([a(1-u), u+a(1-u)] \times [b(1-v), v+b(1-v)]).$$

Show that  $K_{a,b}$  is a copula. Note that,

$$K_{0,0}(u, v) = C(u, v),$$

$$K_{0,1}(u, v) = u - C(u, 1-v),$$

$$K_{1,0}(u, v) = v - C(1-u, v),$$

$$K_{1,1}(u, v) = u + v - 1 + C(1-u, 1-v).$$

*Solution.*  $\circledast$  Write

$$\begin{aligned}
K_{a,b}(u, v) &= C(a(1-u), b(1-v)) + C(u+a(1-u), v+b(1-v)) \\
&\quad - C(a(1-u), v+b(1-v)) - C(u+a(1-u), b(1-v)).
\end{aligned}$$

Then

$$\begin{aligned}
K_{a,b}(0, v) &= C(a, b(1-v)) + C(a, v+b(1-v)) - C(a, v+b(1-v)) - C(a, b(1-v)) \\
&= 0 \\
&= C(a(1-u), b) + C(u+a(1-u), b) - C(a(1-u), b) - C(u+a(1-u), b) \\
&= K_{a,b}(u, 0).
\end{aligned}$$

And

$$K_{a,b}(u, 1) = C(a(1-u), 0) + C(u+a(1-u), 1) - C(a(1-u), 1) - C(u+a(1-u), 0)$$

$$= u + a(1 - u) - a(1 - u) = u$$

$$\begin{aligned} K_{a,b}(1, v) &= C(0, b(1 - v)) + C(1, v + b(1 - v)) - C(0, v + b(1 - v)) - C(1, b(1 - v)) \\ &= v + b(1 - v) - b(1 - v) = v. \end{aligned}$$

(\*) Then for every  $(u_1, v_1), (u_2, v_2) \in I^2$  such that  $u_1 \leq u_2, v_1 \leq v_2$ ,

$$\begin{aligned} \alpha &= K_{a,b}(u_1, v_1) + K_{a,b}(u_2, v_2) - K_{a,b}(u_2, v_1) - K_{a,b}(u_1, v_2) \\ &= C(a(1 - u_1), b(1 - v_1)) + C(u_1 + a(1 - u_1), v_1 + b(1 - v_1)) \\ &\quad - C(a(1 - u_1), v_1 + b(1 - v_1)) - C(u_1 + a(1 - u_1), b(1 - v_1)) \\ &\quad + C(a(1 - u_2), b(1 - v_2)) + C(u_2 + a(1 - u_2), v_1 + b(1 - v_2)) \\ &\quad - C(a(1 - u_2), v_2 + b(1 - v_2)) - C(u_2 + a(1 - u_2), b(1 - v_2)) \\ &\quad - C(a(1 - u_1), b(1 - v_2)) - C(u_1 + a(1 - u_1), v_2 + b(1 - v_2)) \\ &\quad + C(a(1 - u_1), v_2 + b(1 - v_2)) + C(u_1 + a(1 - u_1), b(1 - v_2)) \\ &\quad - C(a(1 - u_2), b(1 - v_1)) - C(u_2 + a(1 - u_2), v_1 + b(1 - v_1)) \\ &\quad + C(a(1 - u_2), v_1 + b(1 - v_1)) + C(u_2 + a(1 - u_2), b(1 - v_1)) \\ &= A + B + C + D, \end{aligned}$$

where

$$\begin{aligned} A &= C(a(1 - u_1), b(1 - v_1)) + C(a(1 - u_2), b(1 - v_2)) \\ &\quad - C(a(1 - u_2), b(1 - v_1)) - C(a(1 - u_1), b(1 - v_2)) \\ &= V_C([a(1 - u_2), a(1 - u_1)] \times [b(1 - v_2), b(1 - v_1)]) \geq 0. \end{aligned}$$

Similarly for  $B, C, D$ . Therefore,  $\alpha \geq 0$ . □

**Exercise 2.7** Let  $f$  be a function from  $I^2$  into  $I$  which is non-decreasing in each variable and has margins given by  $f(t, 1) = t = f(1, t)$  for all  $t \in I$ . Prove that  $f$  is grounded.

*Solution.* We need to prove that on  $I^2$  “non-decreasingness + uniform margins  $\implies$  groundedness”. For every  $(x, y) \in I^2$ ,

$$0 \leq f(x, 0) \leq f(1, 0) = 0, \quad 0 \leq f(0, x) \leq f(0, 1) = 0.$$

□

**Exercise 2.8** (a) Show that for any copula  $C$ ,  $\max(2t - 1, 0) \leq \delta_C(t) \leq t$  for all  $t \in I$ .

(b) Show that  $\delta_C(t) = \delta_M(t)$  for all  $t \in I$  implies  $C = M$ .

(c) Show  $\delta_C(t) = \delta_W(t)$  for all  $t \in I$  does not imply that  $C = W$ .

*Solution.* (a) Write

$$W(t, t) \leq C(t, t) \leq M(t, t) \Leftrightarrow \max(2t - 1, 0) \leq \delta_C(t) \leq \min(t, t) = t.$$

(b) Since for all  $t \in I$ ,

$$\delta_M(t) = M(t, t) = \min(t, t) = t \implies \delta_C(t) = C(t, t) = t.$$

Assume that  $C \neq M$ , then there exists  $(u, v) \in I^2$  with  $u \leq v$  such that

$$C(u, v) \neq M(u, v) \implies C(u, v) < M(u, v) = u.$$

Then by non-decreasingness,

$$u > C(u, v) \geq C(u, u) = u,$$

which is a contradiction.

(c) We just need to show that  $\delta_C = \delta_W$  with  $C \neq W$  holds. □

**Exercise 2.9** The **secondary diagonal section** of  $C$  is given by  $C(t, 1 - t)$ . Show that  $C(t, 1 - t) = 0$  for all  $t \in I$  implies  $C = W$ .

*Solution.* Assume that  $C \neq W$ , then for all  $t \in I$ , we have

$$C(t, 1 - t) > W(t, 1 - t) = \max(t + 1 - t, 0) = 0,$$

which contradicts  $C(t, 1 - t) = 0$ . □

**Exercise 2.10** Let  $t$  be in  $[0, 1]$ , and let  $C_t$  be the function from  $I^2$  into  $I$  given by

$$C_t(u, v) = \begin{cases} \max(u + v - 1, t), & (u, v) \in [t, 1]^2, \\ \min(u, v), & \text{o.w.} \end{cases}$$

(a) Show that  $C_t$  is a copula.

(b) Show that the level set  $\{(u, v) \in I^2 | C_t(u, v) = t\}$  is the set of points in the triangle with vertices  $(t, 1)$ ,  $(1, t)$  and  $(t, t)$ .

*Solution.* (a)  $\circledast$  Uniform margins and groundedness are trivial.  $\circledast$  For every  $(u_1, v_1), (u_2, v_2) \in [t, 1]^2$  with  $u_1 \leq u_2, v_1 \leq v_2$ ,

$$\alpha = \max(u_1 + v_1 - 1, t) + \max(u_2 + v_2 - 1, t) - \max(u_1 + v_2 - 1, t) - \max(u_2 + v_1 - 1, t).$$

If  $u_2 + v_1 \geq 1 + t$ , then

$$\alpha = \max(u_1 + v_1 - 1, t) + u_2 + v_2 - 1 - \max(u_1 + v_2 - 1, t) - (u_2 + v_1 - 1) \geq 0.$$

If  $u_2 + v_1 < 1 + t$ , then

$$\alpha = t + \max(u_2 + v_2 - 1, t) - \max(u_1 + v_2 - 1, t) - t \geq 0.$$

Therefore  $C_t$  is a copula.

(b) Using the Fréchet-Hoeffding bounds,

$$W(u, v) \leq C_t(u, v) \leq M(u, v).$$

Since  $C_t$  is non-decreasing in each argument. The level sets  $L = \{(u, v) \in I^2 : C_t(u, v) = t\}$  have bounds

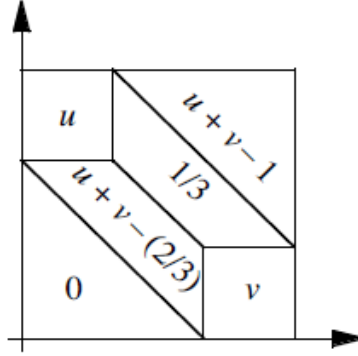
$$\{(u, v) \in I^2 : W(u, v) = t\} \leq L \leq \{(u, v) \in I^2 : M(u, v) = t\}.$$

□

**Exercise 2.11** *This exercise shows that the 2-increasing condition for copulas is not a consequence of simpler properties.* Let  $Q$  be the function from  $I^2$  to  $I$  given by

$$Q(u, v) = \begin{cases} \min(u, v, 1/3, u + v - 2/3), & 2/3 \leq u + v \leq 4/3, \\ \max(u + v - 1, 0), & \text{o.w.} \end{cases}$$

That is,  $Q$  is given as following figure.



(a) Show that for every  $u, v \in I^2$ ,  $Q(u, 0) = 0 = Q(0, v)$ ,  $Q(u, 1) = u$ ,  $Q(1, v) = v$ ;  $W(u, v) \leq Q(u, v) \leq M(u, v)$ ; and that  $Q$  is continuous, satisfies the Lipschitz condition, and is non-decreasing in each variable.

(b) Show that  $Q$  fails to be 2-increasing, and hence is not a copula.

*Solution.* (a) The uniform margins and groundedness are trivial. The upper bounded is clear since  $\min(u, v, 1/3, u + v - 2/3) \leq \min(u, v)$ . The lower bounded is clear from the figure, in the  $1/3$  region,  $1/3$  is the largest value of  $W$  in this region, thus  $W \leq Q$ . The non-decreasingness is obvious. The Lipschitz condition, let  $(u_1, v_1), (u_2, v_2) \in \{2/3 \leq u + v \leq 4/3\}$  with  $u_1 \leq u_2, v_1 \leq v_2$ . If  $(u_2, v_2), (u_1, v_1)$  are in the same region, then  $A = 0$ . Consider different region, if  $(u_1, v_1)$  is in  $u$  region,  $(u_2, v_2)$  is in  $1/3$  region, then

$$Q(u_2, v_2) - Q(u_1, v_1) = 1/3 - u_1 \leq u_2 - u_1 \leq u_2 - u_1 + v_2 - v_1.$$

Similarly, if  $(u_1, v_1)$  is in  $u + v - 2/3$  region,  $(u_2, v_2)$  is in  $1/3$  region, then

$$Q(u_2, v_2) - Q(u_1, v_1) = 1/3 - u_1 - v_1 + 2/3 \leq u_2 - u_1 + v_2 - v_1.$$

If  $(u_1, v_1)$  is in  $u + v - 2/3$  region,  $(u_2, v_2)$  is in  $u$  region, then

$$Q(u_2, v_2) - Q(u_1, v_1) = u_2 - u_1 - v_1 + 2/3 \leq u_2 - u_1 + v_2 - v_1.$$

If  $(u_1, v_1)$  is in  $v$  region,  $(u_2, v_2)$  is in  $1/3$  region, then

$$Q(u_2, v_2) - Q(u_1, v_1) = 1/3 - v_1 \leq v_2 - v_1 \leq u_2 - u_1 + v_2 - v_1.$$

If  $(u_1, v_1)$  is in  $u + v - 2/3$  region,  $(u_2, v_2)$  is in  $v$  region, then

$$Q(u_2, v_2) - Q(u_1, v_1) = v_2 - u_1 - v_1 + 2/3 \leq u_2 - u_1 + v_2 - v_1.$$

Therefore, the Lipschitz condition holds.

(b) Consider the  $Q$ -volume of the rectangle  $[1/3, 2/3]^2$ . Then

$$V_Q([1/3, 2/3]^2) = 0 + 1/3 - 1/3 - 1/3 = -1/3 < 0.$$

$Q$  is not 2-increasing. □

**Exercise 2.12** *Gumbel's bivariate logistic distribution.* Let  $X$  and  $Y$  be random variables with a joint distribution function given by

$$H(x, y) = (1 + e^{-x} + e^{-y})^{-1}$$

for all  $x, y \in \overline{\mathbb{R}}$ . (a) Show that  $X$  and  $Y$  have standard univariate logistic distribution, i.e.,

$$F(x) = (1 + e^{-x})^{-1}, \quad G(y) = (1 + e^{-y})^{-1}.$$

(b) Show that the copula of  $X$  and  $Y$  is the copula given by

$$C(u, v) = \frac{uv}{u + v - uv}.$$

*Solution.* (a) It is obvious.

(b) Since

$$C(u, v) = H(F^{-1}(u), G^{-1}(v)),$$



and

$$F^{-1}(u) = -\ln(u^{-1} - 1), \quad G^{-1}(v) = -\ln(v^{-1} - 1).$$

We have

$$C(u, v) = (1 + u^{-1} - 1 + v^{-1} - 1)^{-1} = \frac{uv}{u + v - uv}.$$

□

**Exercise 2.13** *Type B bivariate extreme value distributions.* Let  $X$  and  $Y$  be random variables with a joint distribution function given by

$$H_\theta(x, y) = \exp \left( - \left( e^{-\theta x} + e^{-\theta y} \right)^{1/\theta} \right)$$

for all  $x, y \in \overline{\mathbb{R}}$ , where  $\theta \geq 1$ . Show that the copula of  $X$  and  $Y$  is given by

$$C_\theta(u, v) = \exp \left( - \left[ -(-\ln u)^\theta + (-\ln v)^\theta \right]^{1/\theta} \right).$$

This parametric family of copulas is known as **Gumbel-Hougaard family**.

*Solution.* The margins are

$$F(x) = H_\theta(x, \infty) = \exp(-e^{-x}), \quad G(y) = H_\theta(\infty, y) = \exp(-e^{-y}).$$

The reverses are

$$F^{-1}(u) = -\ln(-\ln u), \quad G^{-1}(v) = -\ln(-\ln v).$$

Therefore,

$$C_\theta(u, v) = \exp \left( - \left[ -(-\ln u)^\theta + (-\ln v)^\theta \right]^{1/\theta} \right).$$

□

**Exercise 2.14** Note that Gumbel's bivariate logistic distribution suffers from the defect that it lacks a parameter, which limits its usefulness in applications. This can be corrected in a number of ways, one of which is to define  $H_\theta$  as

$$H_\theta(x, y) = (1 + e^{-x} + e^{-y} + (1 - \theta)e^{-x-y})^{-1}$$

for all  $x, y \in \overline{\mathbb{R}}$ , where  $\theta \in [-1, 1]$ . Show that

- (a) the margins are standard logistic distributions;
- (b) when  $\theta = 1$ , we have Gumbel's bivariate logistic distribution;
- (c) when  $\theta = 0$ ,  $X, Y$  are independent;
- (d) the copula of  $X, Y$  is given by

$$C_\theta(u, v) = \frac{uv}{1 - \theta(1 - u)(1 - v)}.$$

This is the **Ali-Mikhail-Haq family** of copulas.

- Solution.* (a) The margins are standard logistic distribution.  
 (b) This is obvious.  
 (c) When  $\theta = 0$ ,

$$H_\theta(x, y) = (1 + e^{-x} + e^{-y} + e^{-x-y})^{-1} = ((1 + e^{-x})(1 + e^{-y}))^{-1}.$$

- (d) The reverses are

$$F^{-1}(u) = -\ln(u^{-1} - 1), \quad G^{-1}(v) = -\ln(v^{-1} - 1).$$

Then

$$\begin{aligned} C_\theta(u, v) &= H_\theta(F^{-1}(u), G^{-1}(v)) = (u^{-1} + v^{-1} - 1 + (1 - \theta)(u^{-1} - 1)(v^{-1} - 1))^{-1} \\ &= \frac{uv}{1 - \theta(1 - u)(1 - v)}. \end{aligned}$$

□

**Exercise 2.15** Let  $X_1, Y_1$  be random variables with continuous distribution functions  $F_1, G_1$ , and copula  $C$ . Let  $F_2, G_2$  be another pair of continuous distribution functions, and set  $X_2 = F_2^{(-1)}(F_1(X_1)), Y_2 = G_2^{(-1)}(G_1(Y_1))$ . Prove that

- (a) the distribution functions of  $X_2, Y_2$  are  $F_2, G_2$ ;
- (b) the copula of  $X_2, Y_2$  is  $C$ .

*Solution.* (a) The CDF of  $X_2$  is

$$P(X_2 \leq x_2) = P(F_2^{(-1)}(F_1(X_1)) \leq x_2)$$

$$\begin{aligned}
&= P(F_1(X_1) \leq F_2(x_2)) \\
&= P(X_1 \leq F_1^{(-1)}(F_2(x_2))) \\
&= F_1(F_1^{(-1)}(F_2(x_2))) = F_2(x_2).
\end{aligned}$$

And similar for  $Y_2$ .

(b) Write

$$\begin{aligned}
C_{X_2, Y_2}(F_2(x), G_2(y)) &= P(X_2 \leq x, Y_2 \leq y) \\
&= P(F_2^{(-1)}(F_1(X_1)) \leq x, G_2^{(-1)}(G_1(Y_1)) \leq y) \\
&= P(X_1 \leq F_1^{(-1)}(F_2(x)), Y_1 \leq G_1^{(-1)}(G_2(y))) \\
&= C_{X_1, Y_1}(F_1[F_1^{(-1)}(F_2(x))], G_1[G_1^{(-1)}(G_2(y))]) \\
&= C_{X_1, Y_1}(F_2(x), G_2(y)).
\end{aligned}$$

□

**Exercise 2.16** (a) Let  $X$  and  $Y$  be continuous random variables with copula  $C$  and univariate distribution functions  $F$  and  $G$ , respectively. The random variables  $\max(X, Y)$  and  $\min(X, Y)$  are the order statistics for  $X, Y$ . Prove that the distribution functions of the order statistics are given by

$$P(\max(X, Y) \leq t) = C(F(t), G(t))$$

and

$$P(\min(X, Y) \leq t) = F(t) + G(t) - C(F(t), G(t)),$$

so that when  $F = G$ ,

$$P(\max(X, Y) \leq t) = \delta_C(F(t)), \quad P(\min(X, Y) \leq t) = 2F(t) - \delta_C(F(t)).$$

(b) Show that bounds on the distribution functions of the order statistics are given by

$$\max(F(t) + G(t) - 1, 0) \leq P(\max(X, Y) \leq t) \leq \min(F(t), G(t))$$

and

$$\max(F(t), G(t)) \leq P(\min(X, Y) \leq t) \leq \min(F(t) + G(t), 1).$$

*Solution.* (a) One have

$$P(\max(X, Y) \leq t) = P(X \leq t, Y \leq t) = H(t, t) = C(F(t), G(t)).$$

And

$$\begin{aligned} P(\min(X, Y) \leq t) &= 1 - P(\min(X, Y) > t) \\ &= 1 - P(X > t, Y > t) \\ &= 1 - [1 - P(X \leq t) - P(Y \leq t) + P(X \leq t, Y \leq t)] \\ &= P(X \leq t) + P(Y \leq t) - P(X \leq t, Y \leq t) \\ &= F(t) + G(t) - C(F(t), G(t)). \end{aligned}$$

(b) We only show the bounds for  $P(\min(X, Y) \leq t)$ . Write

$$\begin{aligned} F(t) + G(t) - \min(F(t), G(t)) &\leq F(t) + G(t) - C(F(t), G(t)) \\ &\leq F(t) + G(t) - \max(F(t) + G(t) - 1, 0), \end{aligned}$$

which is equal to

$$\begin{aligned} \max(F(t), G(t)) &\leq F(t) + G(t) - C(F(t), G(t)) \\ &\leq \min(F(t) + G(t), 1). \end{aligned}$$

□

**Exercise 2.17** [Theorem 2.4.4.] Let  $X$  and  $Y$  be continuous random variables with copula  $C_{XY}$ . Let  $\alpha, \beta$  be strictly monotone on  $\text{ran } X$  and  $\text{ran } Y$ , respectively.

(1) If  $\alpha$  is strictly increasing and  $\beta$  is strictly decreasing, then

$$C_{\alpha(X)\beta(Y)}(u, v) = u - C_{XY}(u, 1 - v).$$

(2) If  $\alpha$  is strictly decreasing and  $\beta$  is strictly increasing, then

$$C_{\alpha(X)\beta(Y)}(u, v) = v - C_{XY}(1 - u, v).$$

(3) If  $\alpha$  and  $\beta$  are both strictly decreasing, then

$$C_{\alpha(X)\beta(Y)}(u, v) = u + v - 1 + C_{XY}(1 - u, 1 - v).$$

PROOF. (1) Write, for any  $x, y \in \overline{\mathbb{R}}$ ,

$$\begin{aligned} C_{\alpha(X)\beta(Y)}(F_{\alpha(X)}(x), G_{\beta(Y)}(y)) &= P(\alpha(X) \leq x, \beta(Y) \leq y) \\ &= P(X \leq \alpha^{-1}(x), Y > \beta^{-1}(y)) \\ &= P(X \leq \alpha^{-1}(x)) - P(X \leq \alpha^{-1}(x), Y \leq \beta^{-1}(y)) \\ &= F_X(\alpha^{-1}(x)) - C_{XY}(F_X(\alpha^{-1}(x)), G_Y(\beta^{-1}(y))) \\ &= F_{\alpha(X)}(x) - C_{XY}(F_{\alpha(X)}(x), 1 - G_{\beta(Y)}(y)). \end{aligned}$$

(2) Similarly, this part is obvious.

(3) Write

$$\begin{aligned} C_{\alpha(X)\beta(Y)}(F_{\alpha(X)}(x), G_{\beta(Y)}(y)) &= P(\alpha(X) \leq x, \beta(Y) \leq y) \\ &= P(X > \alpha^{-1}(x), Y > \beta^{-1}(y)) \\ &= P(X > \alpha^{-1}(x)) - P(X > \alpha^{-1}(x), Y \leq \beta^{-1}(y)) \\ &= P(X > \alpha^{-1}(x)) - P(Y \leq \beta^{-1}(y)) + P(X \leq \alpha^{-1}(x), Y \leq \beta^{-1}(y)) \\ &= F_{\alpha(X)}(x) + G_{\beta(Y)}(y) - 1 + C_{XY}(F_X(\alpha^{-1}(x)), G_Y(\beta^{-1}(y))) \\ &= F_{\alpha(X)}(x) + G_{\beta(Y)}(y) - 1 + C_{XY}(1 - F_{\alpha(X)}(x), 1 - G_{\beta(Y)}(y)). \end{aligned}$$

□

**Exercise 2.18** Let  $S$  be a subset of  $\overline{\mathbb{R}}^2$ . Then  $S$  is non-increasing if and only if for each  $(x, y)$  in  $\overline{\mathbb{R}}^2$ , either

(1) for all  $(u, v)$  in  $S$ ,  $u \leq x$  implies  $v > y$ ; or

(2) for all  $(u, v)$  in  $S$ ,  $v > y$  implies  $u \leq x$ .

*Solution.* “ $\Rightarrow$ ”: Assume  $S$  is non-increasing and neither (1) nor (2) holds. Then there exists points  $(a, b), (c, d)$  in  $S$  such that  $a \leq x, b \leq y$  and  $d > y, c > x$ . Hence  $a \leq c$  and  $b \leq d$ , contradict the non-increasingness.

“ $\Leftarrow$ ”: Assume that  $S$  is not non-increasing. Then there exists points  $(a, b), (c, d)$  in  $S$  with  $a \leq c$  and  $b \geq d$ . For  $(x, y) = ((a + c)/2, (b + d)/2)$ , neither (1) nor (2) holds. □

**Exercise 2.19** Let  $X, Y$  be random variables whose joint distribution function  $H$  is equal to its Fréchet-Hoeffding lower bound. Then for every  $(x, y) \in \overline{\mathbb{R}}^2$ , either  $P(X > x, Y > y) = 0$  or  $P(X \leq x, Y \leq y) = 0$ .

*Solution.* Since

$$P(X > x, Y > y) = 1 - F(x) - G(y) + H(x, y), \quad P(X \leq x, Y \leq y) = H(x, y),$$

$H(x, y) = \max(F(x) + G(y) - 1, 0)$  if and only if either  $P(X > x, Y > y) = 0$  or  $P(X \leq x, Y \leq y) = 0$ . Since if  $P(X \leq x, Y \leq y) = 0$ ,  $F(x) + G(y) - 1 < 0$  and  $H(x, y) = 0$ . □

**Exercise 2.20** [Theorem 2.5.5]

*Solution.* Let  $S$  denote the support of  $H$ , and let  $(x, y)$  be any point in  $\overline{\mathbb{R}}^2$ . Then (1) holds in Exercise 2.18 if and only if

$$\{(u, v) : u \leq x, v \leq y\} \cap S = \emptyset.$$

That is  $P(X \leq x, Y \leq y) = 0$ . By Exercise 2.19, the proof is completed. □

**Exercise 2.21** Let  $X, Y$  be non-negative random variables whose survival function is  $\overline{H}(x, y) = (e^x + e^y - 1)^{-1}$  for  $x, y \geq 0$ .

(a) Show that  $X, Y$  are standard exponential random variables.

(b) Show that the copula of  $X, Y$  is the copula given by

$$C(u, v) = \frac{uv}{u + v + uv}.$$

*Solution.* (a) The univariate survival margins are

$$\bar{F}(x) = \bar{H}(x, -\infty) = (e^x - 1)^{-1}, \quad \bar{G}(y) = \bar{H}(-\infty, y) = (e^y - 1)^{-1}.$$

(b) The inverse functions are

$$\bar{F}^{(-1)}(u) = \ln(u^{-1} + 1), \quad \bar{G}^{(-1)}(v) = \ln(v^{-1} + 1).$$

Then

$$\hat{C}(u, v) = \bar{H}(\bar{F}^{(-1)}(u), \bar{G}^{(-1)}(v)) = \frac{uv}{u + v + uv}.$$

□

**Exercise 2.22** Let  $X, Y$  be continuous random variables whose joint distribution function is given by  $C(F(x), G(y))$ , where  $C$  is the copula of  $X, Y$ , and  $F, G$  are the distribution functions of  $X, Y$  respectively. Verify

$$P(X \leq x \cup Y \leq y) = \tilde{C}(F(x), G(y)), \quad P(X > x \cup Y > y) = C^*(\bar{F}(x), \bar{G}(y)).$$

*Solution.* We have

$$\begin{aligned} P(X \leq x \cup Y \leq y) &= P(X \leq x) + P(Y \leq y) - P(X \leq x, Y \leq y) \\ &= F(x) + G(y) - C(F(x), G(y)) \\ &= \tilde{C}(F(x), G(y)). \end{aligned}$$

And

$$\begin{aligned} P(X > x \cup Y > y) &= 1 - P(X \leq x, Y \leq y) \\ &= 1 - C(F(x), G(y)) \\ &= C^*(\bar{F}(x), \bar{G}(y)). \end{aligned}$$

□

**Exercise 2.23** Let  $X_1, Y_1, F_1, G_1, F_2, G_2$  and  $C$  be as usual. Set  $X_2 = F_2^{(-1)}(1 - F_1(X_1))$  and  $Y_2 = G_2^{(-1)}(1 - G_1(Y_1))$ . Prove that

- (a) The distribution functions of  $X_2, Y_2$  are  $F_2, G_2$ , and
- (b) The copula of  $X_2, Y_2$  is  $\hat{C}$ .

*Solution.* (a) The distribution function of  $X_2$  is

$$\begin{aligned}
 P(X_2 \leq x) &= P(F_2^{(-1)}(1 - F_1(X_1)) \leq x) \\
 &= P((1 - F_1(X_1)) \leq F_2(x)) \\
 &= P(1 - F_2(x) \leq F_1(X_1)) \\
 &= 1 - P(F_1(X_1) \leq 1 - F_2(x)) \\
 &= 1 - P(X_1 \leq F_1^{(-1)}(1 - F_2(x))) \\
 &= 1 - F_1[F_1^{(-1)}(1 - F_2(x))] = F_2(x).
 \end{aligned}$$

Similarly for  $G_2$ .

- (b) Write

$$\begin{aligned}
 P(X_2 \leq x, Y_2 \leq y) &= P(F_2^{(-1)}(1 - F_1(X_1)) \leq x, G_2^{(-1)}(1 - G_1(Y_1)) \leq y) \\
 &= P((1 - F_1(X_1)) \leq F_2(x), (1 - G_1(Y_1)) \leq G_2(y)) \\
 &= P((1 - F_2(x)) \leq F_1(X_1), (1 - G_2(y)) \leq G_1(Y_1)) \\
 &= 1 - [1 - F_2(x) + 1 - G_2(y) - P(F_1(X_1) \leq 1 - F_2(x), G_1(Y_1) \leq 1 - G_2(y))] \\
 &= F_2(x) + G_2(y) - 1 + C(F_1(F_1^{(-1)}(1 - F_2(x))), G_1(G_1^{(-1)}(1 - G_2(y)))) \\
 &= F_2(x) + G_2(y) - 1 + C(1 - F_2(x), 1 - G_2(y)) \\
 &= \hat{C}(F_2(x), G_2(y)).
 \end{aligned}$$

□

**Exercise 2.24** Let  $X, Y$  be continuous random variables with copula  $C$  and a common univariate distribution function  $F$ . Show that the distribution and survival functions of the order statistics are given by



Order statistic	Distribution function	Survival function
$\max(X, Y)$	$\delta(F(t))$	$\delta^*(\bar{F}(t))$
$\min(X, Y)$	$\tilde{\delta}(F(t))$	$\hat{\delta}(\bar{F}(t))$

where  $\delta, \hat{\delta}, \tilde{\delta}$  and  $\delta^*$  denote the diagonal sections of  $C, \hat{C}, \tilde{C}, C^*$ , respectively.

*Solution.* Write

$$\begin{aligned} P(\max(X, Y) \leq t) &= P(X \leq t, Y \leq t) \\ &= C(F(t), F(t)) = \delta(F(t)). \end{aligned}$$

And

$$P(\max(X, Y) > t) = 1 - C(F(t), F(t)) = \delta^*(\bar{F}(t)).$$

Further,

$$\begin{aligned} P(\min(X, Y) \leq t) &= 1 - P(\min(X, Y) > t) \\ &= 1 - P(X > t, Y > t) \\ &= P(X \leq t) + P(Y \leq t) - P(X \leq t, Y \leq t) \\ &= F(t) + F(t) - C(F(t), F(t)) \\ &= \tilde{\delta}(F(t)). \end{aligned}$$

And

$$\begin{aligned} P(\min(X, Y) > t) &= 1 - F(t) - F(t) + C(F(t), F(t)) \\ &= \hat{\delta}(\bar{F}(t)). \end{aligned}$$

□

**Exercise 2.25** Show that under composition, the set of operations of forming the survival copula, the dual of a copula, and the co-copula of a given copula. along with the identity yields the dihedral group:

$\circ$	$i$	$\wedge$	$\sim$	$*$
$i$	$i$	$\wedge$	$\sim$	$*$
$\wedge$	$\wedge$	$i$	$*$	$\sim$
$\sim$	$\sim$	$*$	$i$	$\wedge$
$*$	$*$	$\sim$	$\wedge$	$i$

*Solution.* Clearly,

$$\begin{aligned}\wedge(\wedge(C(u, v))) &= \wedge(u + v - 1 + C(1 - u, 1 - v)) \\ &= u + v - 1 + (1 - u) + (1 - v) - 1 + C(u, v) = C(u, v).\end{aligned}$$

And

$$\begin{aligned}\sim(\sim(C(u, v))) &= \sim(u + v - C(u, v)) \\ &= u + v - u - v + C(u, v) = C(u, v).\end{aligned}$$

And

$$\begin{aligned}\wedge(\sim(C(u, v))) &= \wedge(u + v - C(u, v)) \\ &= u + v - u - v + 1 - C(1 - u, 1 - v) = C^*(u, v).\end{aligned}$$

Others are similar. □

**Exercise 2.26** Prove for any  $(u, v) \in I^2$ ,

$$\hat{C}(u, v) = \overline{H}(\overline{F}^{(-1)}(u), \overline{G}^{(-1)}(v)).$$

*Solution.* One has

$$\begin{aligned}\overline{H}(\overline{F}^{(-1)}(u), \overline{G}^{(-1)}(v)) &= P(X > \overline{F}^{(-1)}(u), Y > \overline{G}^{(-1)}(v)) \\ &= 1 - F(\overline{F}^{(-1)}(u)) - G(\overline{G}^{(-1)}(v)) + H(\overline{F}^{(-1)}(u), \overline{G}^{(-1)}(v)) \\ &= u + v - 1 + C(F(\overline{F}^{(-1)}(u)), G(\overline{G}^{(-1)}(v))) \\ &= u + v - 1 + C(1 - u, 1 - v) = \hat{C}(u, v).\end{aligned}$$



**Exercise 2.27** Let  $X, Y$  be continuous random variables symmetric about  $a$  and  $b$  with marginal distribution function  $F, G$ , and with copula  $C$ . Is  $(X, Y)$  radially symmetric (or jointly symmetric) about  $(a, b)$  if  $C$  is

- (a) a member of the Fréchet family in Exercise 2.4?
- (b) a member of the Cuadras-Augé family in Exercise 2.5?

*Solution.* (a) The Fréchet family is

$$C_{\alpha, \beta}(u, v) = \alpha M(u, v) + (1 - \alpha - \beta)\Pi(u, v) + \beta W(u, v).$$

Then

$$\begin{aligned} \hat{C}_{\alpha, \beta}(u, v) &= u + v - 1 + C_{\alpha, \beta}(1 - u, 1 - v) \\ &= u + v - 1 + \alpha \min(1 - u, 1 - v) + (1 - \alpha - \beta)(1 - u)(1 - v) \\ &\quad + \beta \max(1 - u + 1 - v + 1, 0) \\ &= \alpha[\min(1 - u, 1 - v) - 1 + u + v] + (1 - \alpha - \beta)uv \\ &\quad + \beta[\max(1 - u - v, 0) - 1 + u + v] \\ &= C_{\alpha, \beta}(u, v). \end{aligned}$$

(b) The Cuadras-Augé family is

$$C_{\theta}(u, v) = [\min(u, v)]^{\theta} [uv]^{1-\theta} = \begin{cases} uv^{1-\theta}, & u \leq v, \\ u^{1-\theta}v, & u \geq v. \end{cases}$$

Then when  $u \leq v$ ,

$$\hat{C}_{\theta}(u, v) = u + v - 1 + (1 - u)(1 - v)^{1-\theta}.$$

And similarly for  $u \geq v$ . Clearly,  $\hat{C}_0(u, v) = C_0(u, v)$ . And  $\hat{C}_1(u, v) = C_1(u, v)$ . □

**Exercise 2.28** Suppose  $X, Y$  are identically distributed continuous random variables, each symmetric about  $a$ . Show that “exchangeability” does not imply “radial symmetry”, nor does “radial symmetry” imply “exchangeability”.

*Solution.* Suppose that  $X, Y$  are exchangeable, then

$$C(u, v) = C(v, u).$$

Thus

$$\begin{aligned}\hat{C}(u, v) &= u + v - 1 + C(1 - u, 1 - v) \\ &= u + v - 1 + C(1 - v, 1 - u) \neq C(u, v).\end{aligned}$$

Conversely, assume “radial symmetry”, then

$$\hat{C}(u, v) = C(u, v).$$

Thus

$$C(v, u) = v + u - 1 + C(1 - v, 1 - u) \neq u + v - 1 + C(1 - u, 1 - v) = C(u, v).$$

□

**Exercise 2.29** Let  $X, Y$  be continuous random variables with joint distribution function  $H$  and margins  $F, G$ . Let  $(a, b)$  be a point in  $\mathbb{R}^2$ . Then  $(X, Y)$  is jointly symmetric about  $(a, b)$  if and only if

$$H(a+x, b+y) = F(a+x) - H(a+x, b-y), \quad H(a+x, b+y) = G(b+y) - H(a-x, b+y)$$

for all  $(x, y) \in \overline{\mathbb{R}}^2$ .

*Solution.* According to the definition,

$$\begin{aligned}P(X - a \leq x, Y - b \leq y) &= P(X - a \leq x, b - Y \leq y) \\ &= P(X \leq a + x) - P(X \leq a + x, Y \leq b - y) \\ &= F(x + a) - H(a + x, b - y).\end{aligned}\tag{2.0.1}$$

Similarly,

$$P(X - a \leq x, Y - b \leq y) = P(a - X \leq x, Y - b \leq y) = G(b + y) - H(a - x, b + y).\tag{2.0.2}$$

Let  $y = \infty$ , (2.0.1) and (2.0.2) imply that  $(X, Y)$  is marginally symmetric about  $(a, b)$ ,

$$G(b + y) = \overline{G}(b - y), \quad F(a + x) = \overline{F}(a - x).$$

Then

$$G(b + y) - H(a + x, b + y) = \overline{G}(b - y) - P(X \leq a + x, Y > b - y),$$

which implies

$$P(X > a + x, Y > b - y) = P(X > a + x, Y \leq b + y) = P(X \leq a - x, Y \leq b + y).$$

It follows that (draw a picture)

$$P(X > a + x, Y \leq b + y) = P(X \leq a - x, Y > b - y).$$

It ends the proof. □

**Exercise 2.30** Let  $X, Y$  be continuous random variables with joint distribution function  $H$  and margins  $F, G$  and copula  $C$ . Further suppose that  $X, Y$  are symmetric about  $a$  and  $b$ . Then  $(X, Y)$  is jointly symmetric about  $(a, b)$  if and only if  $C$  satisfies

$$C(u, v) = u - C(u, 1 - v), \quad C(u, v) = v - C(1 - u, v)$$

for all  $(u, v) \in \mathbb{I}^2$ .

*Solution.* Write

$$\begin{aligned} H(a + x, b + y) &= F(x + a) - H(a + x, b - y) \\ &\Leftrightarrow C(F(a + x), G(b + y)) = F(x + a) - C(F(a + x), G(b - y)) \\ &\Leftrightarrow C(F(a + x), G(b + y)) = F(x + a) - C(F(a + x), \overline{G}(b + y)) \\ &\Leftrightarrow C(u, v) = u - C(u, 1 - v). \end{aligned}$$

Similar for the other equation. □

**Exercise 2.31** (a) Show that  $C_1 \prec C_2$  if and only if  $\overline{C}_1 \prec \overline{C}_2$ .

(b) Show that  $C_1 \prec C_2$  if and only if  $\hat{C}_1 \prec \hat{C}_2$ .

*Solution.* (a) We have for all  $(u, v) \in \mathbb{I}^2$ ,

$$\begin{aligned} C_1 \prec C_2 &\Leftrightarrow C_1(u, v) \leq C_2(u, v) \\ &\Leftrightarrow 1 - u - v + C_1(u, v) \leq 1 - u - v + C_2(u, v) \\ &\Leftrightarrow \overline{C}_1 \leq \overline{C}_2 \end{aligned}$$

(b) Similarly, for all  $(u, v) \in \mathbb{I}^2$ ,

$$\begin{aligned} C_1 \prec C_2 &\Leftrightarrow C_1(u, v) \leq C_2(u, v) \\ &\Leftrightarrow C_1(1 - u, 1 - v) \leq C_2(1 - u, 1 - v) \\ &\Leftrightarrow u + v - 1 + C_1(u, v) \leq u + v - 1 + C_2(u, v) \\ &\Leftrightarrow \hat{C}_1 \leq \hat{C}_2 \end{aligned}$$

□

**Exercise 2.32** Show that Ali-Mikhail-Haq family of copulas from Exercise 2.14 is positively ordered.

*Solution.* The copula of  $X, Y$  is given by

$$C_\theta(u, v) = \frac{uv}{1 - \theta(1 - u)(1 - v)},$$

where  $\theta \in [-1, 1]$ . For  $0 \leq \alpha \leq \beta \leq 1$  and  $u, v \in (0, 1)$ ,

$$\frac{C_\alpha(u, v)}{C_\beta(u, v)} = \frac{1 - \beta(1 - u)(1 - v)}{1 - \alpha(1 - u)(1 - v)} \leq 1,$$

thus  $C_\alpha(u, v) \leq C_\beta(u, v)$ . □

**Exercise 2.33** Show that the Mardia family from Exercise 2.4 is neither positively nor negatively ordered.

*Solution.* Let  $\theta$  be in  $[-1, 1]$  and set

$$C_\theta(u, v) = \frac{\theta^2(1 + \theta)}{2}M(u, v) + (1 - \theta^2)\Pi(u, v) + \frac{\theta^2(1 - \theta)}{2}W(u, v).$$

Note that, let  $(u, v) = (3/4, 1/4)$ ,

$$C_0(3/4, 1/4) = \frac{3}{16} = \frac{96}{512}, \quad C_{1/4}(3/4, 1/4) = \frac{95}{512}, \quad C_{1/2}(3/4, 1/4) = \frac{96}{512}.$$

□

**Exercise 2.34** (a) Show that the  $(n-1)$ -margins of an  $n$ -copula are  $(n-1)$ -copulas.

(b) Show that if  $C$  is an  $n$ -copula,  $n \geq 3$ , then for any  $k, 2 \leq k < n$ , all  $\binom{n}{k}$   $k$ -margins of  $C$  are  $k$ -copulas.

*Solution.* (a) The groundedness and uniform margins are directly from the property of this  $n$ -copula. Further, evaluate the  $C^n$ -volume of box

$$[a_1, b_1] \times \cdots \times [a_{k-1}, b_{k-1}] \times [0, 1] \times [a_{k+1}, b_{k+1}] \times \cdots \times [a_n, b_n].$$

Clearly, it is greater than 0, thus the  $C^{n-1}$ -volume of

$$[a_1, b_1] \times \cdots \times [a_{k-1}, b_{k-1}] \times [a_{k+1}, b_{k+1}] \times \cdots \times [a_n, b_n]$$

is also greater than 0.

(b) This is similar to part (a).

□

**Exercise 2.35** Let  $M^n$  and  $\Pi^n$  be multivariate copula, and let  $[\mathbf{a}, \mathbf{b}]$  be an  $n$ -box in  $\mathbb{I}^n$ . Prove that

$$V_{M^n}([\mathbf{a}, \mathbf{b}]) = \max(\min(b_1, b_2, \dots, b_n) - \max(a_1, a_2, \dots, a_n), 0)$$

and

$$V_{\Pi^n}([\mathbf{a}, \mathbf{b}]) = (b_1 - a_1)(b_2 - a_2) \cdots (b_n - a_n),$$

and hence conclude that  $M^n$  and  $\Pi^n$  are  $n$ -copulas for all  $n \geq 2$ .

*Solution.* We only prove for  $M^n$ , for bivariate  $M$ , if  $a_1 \leq a_2$ ,

$$V_M([\mathbf{a}, \mathbf{b}]) = \min(b_1, b_2) - \min(a_1, a_2) \geq 0.$$

If  $a_1 > a_2$ ,

$$V_M([\mathbf{a}, \mathbf{b}]) = \min(b_1, b_2) - \min(a_1, b_2) \geq 0.$$

That is

$$V_M([\mathbf{a}, \mathbf{b}]) = \max(\min(b_1, b_2) - \max(a_1, a_2), 0).$$

Then similarly, for  $M^n$ ,

$$V_{M^n}([\mathbf{a}, \mathbf{b}]) = \max(\min(b_1, b_2, \dots, b_n) - \max(a_1, a_2, \dots, a_n), 0).$$

□

**Exercise 2.36** Show that

$$V_{W^n}([\mathbf{1}/2, \mathbf{1}]) = 1 - n/2,$$

where  $\mathbf{1} = (1, 1, \dots, 1)$  and  $\mathbf{1}/2 = (1/2, 1/2, \dots, 1/2)$ , and hence  $W^n$  fails to be an  $n$ -copula whenever  $n > 2$ .

*Solution.* We have for all vertices  $\mathbf{c}$ , that is  $c_i$  is either  $1/2$  or  $1$ , for all  $i = 1, \dots, n$ ,

$$V_{W^n}([\mathbf{1}/2, \mathbf{1}]) = \sum (-1)^{\#\{c_i=1/2\}} W^n(\mathbf{c}) = -\max(1/2, 0) \cdot n + \max(1, 0) = 1 - n/2.$$

Thus  $W^n$  fails to be an  $n$ -copula whenever  $n > 2$ . □

**Exercise 2.37** Let  $\{X_1, \dots, X_n\}$  be continuous random variables with copula  $C$  and distribution functions  $\{F_1, \dots, F_n\}$ , respectively. Let  $X_{(1)}$  and  $X_{(n)}$  denote the extreme order statistics for  $\{X_1, \dots, X_n\}$ . Prove that the distribution functions  $F_{(1)}, F_{(n)}$  of  $X_{(1)}, X_{(n)}$  satisfy

$$\max(F_1(t), \dots, F_n(t)) \leq F_{(1)}(t) \leq \min\left(\sum_{k=1}^n F_k(t), 1\right)$$

and

$$\max\left(\sum_{k=1}^n F_k(t) - n + 1, 0\right) \leq F_{(n)}(t) \leq \min(F_1(t), F_2(t), \dots, F_n(t)).$$

*Solution.* Note that,

$$\begin{aligned} F_{(1)}(t) &= 1 - P(X_1 > t, \dots, X_n > t) \\ &= (-1)^{1-1} \sum_{i=1}^n P(X_i \leq t) + (-1)^{2-1} \sum_{i_1 \neq i_2}^n P(X_{i_1} \leq t, X_{i_2} \leq t) \end{aligned}$$



$$+ \cdots + (-1)^{n-1} P(X_1 \leq t, \dots, X_n \leq t),$$

which implies that

$$F_{(1)}(t) \leq \min \left( \sum_{k=1}^n F_k(t), 1 \right).$$

Clearly, for  $i = 1, \dots, n$ ,

$$F_{(1)}(t) = 1 - P(X_1 > t, \dots, X_n > t) \geq 1 - P(X_i > t) \implies F_{(1)} \geq \max(F_1, \dots, F_n).$$

Further,

$$\begin{aligned} F_{(n)}(t) &= P(X_1 \leq t, \dots, X_n \leq t) \\ &= 1 + (-1)^1 \sum_{i=1}^n P(X_i > t) + (-1)^2 \sum_{i_1 \neq i_2}^n P(X_{i_1} > t, X_{i_2} > t) \\ &\quad + \cdots + (-1)^n P(X_1 > t, \dots, X_n > t), \end{aligned}$$

which implies that

$$F_{(n)}(t) \geq \max \left( 1 + (-1)^1 \sum_{i=1}^n P(X_i > t), 0 \right) = \max \left( \sum_{k=1}^n F_k(t) - n + 1, 0 \right).$$

Clearly, for  $i = 1, \dots, n$ ,

$$F_{(n)}(t) = P(X_1 \leq t, \dots, X_n \leq t) \leq P(X_i \leq t) \implies F_{(n)} \leq \min(F_1, \dots, F_n).$$

□

**Exercise 3.1** Show that when either of the parameters  $\alpha$  or  $\beta$  is equal to 0 or 1, the function

$$C_{\alpha, \beta} = \min(u^{1-\alpha}v, uv^{1-\beta}) = \begin{cases} u^{1-\alpha}v, & u^\alpha \geq v^\beta, \\ uv^{1-\beta}, & u^\alpha \leq v^\beta. \end{cases}$$

is a copula.

*Solution.* Clearly,

$$C_{0,0} = C_{\alpha,0} = C_{0,\beta} = \Pi, \quad C_{1,1} = M.$$

And

$$C_{\alpha,1} = \min(u^{1-\alpha}v, u), \quad C_{1,\beta} = \min(v, uv^{1-\beta}).$$

The groundedness and uniform margins are obvious. Further, let  $(u_1, v_1), (u_2, v_2) \in I^2$  with  $u_1 \leq u_2, v_1 \leq v_2$ , then

$$\begin{aligned} V_{C_{\alpha,1}}([u_1, u_2] \times [v_1, v_2]) &= \min(u_1^{1-\alpha}v_1, u_1) + \min(u_2^{1-\alpha}v_2, u_2) \\ &\quad - \min(u_2^{1-\alpha}v_1, u_2) - \min(u_1^{1-\alpha}v_2, u_1). \end{aligned}$$

If  $v_2/u_1^\alpha < 1$ ,

$$\begin{aligned} V_{C_{\alpha,1}}([u_1, u_2] \times [v_1, v_2]) &= u_1^{1-\alpha}v_1 + u_2^{1-\alpha}v_2 - u_2^{1-\alpha}v_1 - u_1^{1-\alpha}v_2 \\ &= (v_2 - v_1)(u_2^{1-\alpha} - u_1^{1-\alpha}) \geq 0. \end{aligned}$$

If  $v_2/u_1^\alpha > 1, v_1/u_1^\alpha < 1, v_2/u_2^\alpha < 1$ ,

$$\begin{aligned} V_{C_{\alpha,1}}([u_1, u_2] \times [v_1, v_2]) &= u_1^{1-\alpha}v_1 + u_2^{1-\alpha}v_2 - u_2^{1-\alpha}v_1 - u_1 \\ &\geq u_1^{1-\alpha}v_1 + u_2^{1-\alpha}v_2 - u_2^{1-\alpha}v_1 - u_1^{1-\alpha}v_2 \\ &= (v_2 - v_1)(u_2^{1-\alpha} - u_1^{1-\alpha}) \geq 0. \end{aligned}$$

If  $v_2/u_1^\alpha > 1, v_1/u_1^\alpha < 1, v_2/u_2^\alpha > 1$ ,

$$\begin{aligned} V_{C_{\alpha,1}}([u_1, u_2] \times [v_1, v_2]) &= u_1^{1-\alpha}v_1 + u_2 - u_2^{1-\alpha}v_1 - u_1 \\ &= u_1(v_1/u_1^\alpha - 1) + u_2(1 - v_1/u_2^\alpha) \\ &\geq u_1(v_1/u_1^\alpha - 1) + u_1(1 - v_1/u_2^\alpha) \\ &= v_1/u_1^\alpha - v_1/u_2^\alpha \geq 0. \end{aligned}$$

If  $v_2/u_1^\alpha > 1, v_1/u_1^\alpha > 1, v_2/u_2^\alpha < 1$ ,

$$\begin{aligned} V_{C_{\alpha,1}}([u_1, u_2] \times [v_1, v_2]) &= u_1^{1-\alpha}v_1 + u_2 - u_2 - u_1 \\ &\geq u_1^{1-\alpha}v_1 + u_2 - u_2^{1-\alpha}v_1 - u_1 \geq 0. \end{aligned}$$

If  $v_2/u_1^\alpha > 1, v_1/u_1^\alpha > 1, v_2/u_2^\alpha > 1, v_1/u_2^\alpha > 1$ ,

$$V_{C_{\alpha,1}}([u_1, u_2] \times [v_1, v_2]) = u_1 + u_2 - u_2 - u_1 = 0$$

If  $v_2/u_1^\alpha > 1, v_1/u_1^\alpha > 1, v_2/u_2^\alpha > 1, v_1/u_2^\alpha < 1$ ,

$$V_{C_{\alpha,1}}([u_1, u_2] \times [v_1, v_2]) = u_1 + u_2 - u_2^{1-\alpha}v_1 - u_1 \geq 0.$$

The other  $C_{1,\beta}$  is similar. □

**Exercise 3.2** Show that a version of the Marshall-Olkin bivariate distribution with Pareto margins has joint survival functions given by

$$\overline{H}(x, y) = (1+x)^{-\theta_1}(1+y)^{-\theta_2}[\max(1+x, 1+y)]^{-\theta_{12}},$$

for  $x, y \geq 0$ , where  $\theta_1, \theta_2, \theta_{12}$  are positive parameters.

*Solution.* Note that the Pareto margins are

$$\overline{F}(x) = \begin{cases} (1+x)^{-\theta}, & x \geq 0, \\ 1, & x < 0. \end{cases} \quad \overline{G}(y) = \begin{cases} (1+y)^{-\theta}, & y \geq 0, \\ 1, & y < 0. \end{cases}$$

Then

$$\begin{aligned} \overline{H}(x, y) &= \hat{C}(\overline{F}(x), \overline{G}(y)) \\ &= \begin{cases} (1+x)^{-\theta}(1+y)^{-\theta} \min((1+x)^{\alpha\theta}, (1+y)^{\beta\theta}), & x \geq 0, y \geq 0, \\ (1+x)^{-\theta} \min((1+x)^{\alpha\theta}, 1), & x \geq 0, y < 0, \\ (1+y)^{-\theta} \min(1, (1+y)^{\beta\theta}), & x < 0, y \geq 0, \\ 1, & x < 0, y < 0, \end{cases} \end{aligned}$$

Therefore, for  $x, y \geq 0$ ,

$$\begin{aligned} \overline{H}(x, y) &= (1+x)^{-\theta}(1+y)^{-\theta} \min((1+x)^{\alpha\theta}, (1+y)^{\beta\theta}), \\ &= (1+x)^{-\theta+\alpha\theta}(1+y)^{-\theta} \left[ \min\left(1, \frac{(1+y)^{\beta}}{(1+x)^{\alpha}}\right) \right]^{\theta}, \end{aligned}$$

$$\begin{aligned}
&= (1+x)^{-\theta+\alpha\theta}(1+y)^{-\theta} \left[ \max \left( 1, \frac{(1+x)^\alpha}{(1+y)^\beta} \right) \right]^{-\theta}, \\
&= (1+x)^{-\theta+\alpha\theta}(1+y)^{-\theta-\beta} \left[ \max \left( (1+y)^\beta, (1+x)^\alpha \right) \right]^{-\theta}, \\
&= (1+x)^{-\theta_1}(1+y)^{-\theta_2} [\max(1+x, 1+y)]^{-\theta_{12}}.
\end{aligned}$$

□

**Exercise 3.3** Prove the following generalization of the Marshall-Olkin family of copulas: Suppose that  $a, b$  are increasing functions defined on  $I$  such that  $a(0) = b(0) = 0$  and  $a(1) = b(1) = 1$ . Further suppose that the functions  $u \mapsto a(u)/u$  and  $v \mapsto b(v)/v$  are both increasing on  $(0, 1]$ . Then the function  $C$  defined on  $I^2$  by

$$C(u, v) = \min(va(u), ub(v))$$

is a copula.  $a(u) = u^{1-\alpha}$ ,  $b(v) = v^{1-\beta}$  is a special case.

*Solution.* The groundness and uniform margins are obvious. Let  $(u_1, v_1), (u_2, v_2) \in I^2$  with  $u_1 \leq u_2, v_1 \leq v_2$ , then

$$\alpha = \min(v_1a(u_1), u_1b(v_1)) + \min(v_2a(u_2), u_2b(v_2)) - \min(v_1a(u_2), u_2b(v_1)) - \min(v_2a(u_1), u_1b(v_2)).$$

If  $a(u_1)/u_1 \leq b(v_1)/v_1, a(u_2)/u_2 > b(v_1)/v_1, a(u_2)/u_2 \leq b(v_2)/v_2$ ,

$$\begin{aligned}
\alpha &= v_1a(u_1) + v_2a(u_2) - u_2b(v_1) - v_2a(u_1) \\
&\geq v_1a(u_1) + v_2a(u_2) - v_1a(u_2) - v_2a(u_1) \\
&= (v_2 - v_1)(a(u_2) - a(u_1)) \geq 0.
\end{aligned}$$

Other circumstance are similar.

□

**Exercise 3.4** (a) Show that the following algorithm generates random variates  $(x, y)$  from Marshall-Olkin bivariate-exponential distribution with parameters  $\lambda_1, \lambda_2, \lambda_{12}$ :

1. Generate three independent uniform  $(0, 1)$  variates  $r, s, t$ ;
2. Set  $x = \min \left( \frac{-\ln r}{\lambda_1}, \frac{-\ln t}{\lambda_{12}} \right), y = \min \left( \frac{-\ln s}{\lambda_2}, \frac{-\ln t}{\lambda_{12}} \right)$ ;
3. The desired pair is  $(x, y)$ .

(b) Show that  $u = \exp(-(\lambda_1 + \lambda_{12})x)$  and  $v = \exp(-(\lambda_2 + \lambda_{12})y)$  are uniform  $(0, 1)$  variates whose joint distribution function is a Marshall-Olkin copula.

*Solution.* (a) We have

$$\begin{aligned}
 & P\left(\min\left(\frac{-\ln r}{\lambda_1}, \frac{-\ln t}{\lambda_{12}}\right) > x, \min\left(\frac{-\ln s}{\lambda_2}, \frac{-\ln t}{\lambda_{12}}\right) > y\right) \\
 &= P\left(\frac{-\ln r}{\lambda_1} > x, \frac{-\ln t}{\lambda_{12}} > x, \frac{-\ln s}{\lambda_2} > y, \frac{-\ln t}{\lambda_{12}} > y\right) \\
 &= P(r < \exp(-\lambda_1 x), s < \exp(-\lambda_2 y), t < \exp(-\lambda_{12} \max(x, y))) \\
 &= \exp(-\lambda_1 x) \exp(-\lambda_2 y) \exp(-\lambda_{12} \max(x, y)),
 \end{aligned}$$

which is the Marshall-Olkin bivariate-exponential distribution.

(b) The quasi-inverses are

$$x = \frac{-\ln u}{\lambda_1 + \lambda_{12}}, \quad y = \frac{-\ln v}{\lambda_2 + \lambda_{12}}.$$

Then

$$\begin{aligned}
 \hat{C}(u, v) &= \exp\left(-(\lambda_1 + \lambda_{12})\frac{-\ln u}{\lambda_1 + \lambda_{12}}\right) \exp\left(-(\lambda_2 + \lambda_{12})\frac{-\ln v}{\lambda_2 + \lambda_{12}}\right) \\
 &\quad \cdot \min\left\{\exp\left(\lambda_{12}\frac{-\ln u}{\lambda_1 + \lambda_{12}}\right), \exp\left(\lambda_{12}\frac{-\ln v}{\lambda_2 + \lambda_{12}}\right)\right\} \\
 &= uv \min(u^{-\alpha}, v^{-\beta}),
 \end{aligned}$$

where  $\alpha = \lambda_{12}/(\lambda_1 + \lambda_{12})$ ,  $\beta = \lambda_{12}/(\lambda_2 + \lambda_{12})$ . □

**Exercise 3.5** Let  $(X, Y)$  be random variables with circular uniform distribution. Find the distribution of  $\max(X, Y)$ .

*Solution.* The distribution function of  $\max(X, Y)$  is

$$P(\max(X, Y) \leq t) = H(t, t) = \begin{cases} 3/4, & x^2 + y^2 \leq 1, \\ 1, & x^2 + y^2 > 1, x, y \geq 0, \\ 1 - \frac{\arccos t}{\pi}, & x^2 + y^2 > 1, \min(x, y) < 0 \leq \max(x, y), \\ 0, & x^2 + y^2 > 1, x, y < 0. \end{cases}$$

□

**Exercise 3.6** Let  $Z_1, Z_2, Z_3$  be three mutually independent exponential random variables with parameter  $\lambda > 0$ , and let  $J$  be a Bernoulli random variable, independent with  $Z$ 's, with parameter  $\theta$  in  $(0, 1)$ . Set

$$X = (1 - \theta)Z_1 + JZ_3, \quad Y = (1 - \theta)Z_2 + JZ_3.$$

Show that

(a) for  $x, y \geq 0$ , the joint survival function of  $X$  and  $Y$  is given by

$$\overline{H}(x, y) = \exp[-\lambda(x \vee y)] + \frac{1 - \theta}{1 + \theta} \exp\left[\frac{-\lambda(x + y)}{1 - \theta}\right] \left(1 - \exp\left[\lambda \frac{1 + \theta}{1 - \theta}(x \vee y)\right]\right).$$

(b)  $X, Y$  are exponential with parameter  $\lambda$ ;

(c) the survival copula of  $X, Y$  is given by

$$\hat{C}_\theta(u, v) = M(u, v) + \frac{1 - \theta}{1 + \theta} (uv)^{1/(1-\theta)} (1 - [\max(u, v)]^{-(1+\theta)/(1-\theta)}).$$

(d)  $\hat{C}_\theta$  is absolutely continuous,  $\hat{C}_0 = \Pi, \hat{C}_1 = M$ .

*Solution.* (a)

(b)

□

**Exercise 3.7**

*Solution.*

□

**Exercise 3.35** For Plackett family of copulas, show that

$$(a) C_0(u, v) = \lim_{\theta \rightarrow 0^+} C_\theta(u, v) = \frac{(u+v-1) + |u+v-1|}{2} = W(u, v),$$

$$(b) C_\infty(u, v) = \lim_{\theta \rightarrow \infty} C_\theta(u, v) = \frac{(u+v) - |u-v|}{2} = M(u, v).$$

*Solution.* (a) We have

$$\begin{aligned} & \lim_{\theta \rightarrow 0} \frac{[1 + (\theta - 1)(u + v)] - \sqrt{[1 + (\theta - 1)(u + v)]^2 - 4uv\theta(\theta - 1)}}{2(\theta - 1)} \\ &= \frac{1 - (u + v) - |1 - (u + v)|}{-2} \\ &= \frac{(u + v - 1) + |u + v - 1|}{2} = W(u, v). \end{aligned}$$

(b)

$$\begin{aligned}
& \lim_{\theta \rightarrow \infty} \frac{[1 + (\theta - 1)(u + v)] - \sqrt{[1 + (\theta - 1)(u + v)]^2 - 4uv\theta(\theta - 1)}}{2(\theta - 1)} \\
&= \lim_{\theta \rightarrow \infty} \frac{[1/(\theta - 1) + (u + v)] - \sqrt{[1/(\theta - 1) + (u + v)]^2 - 4uv\theta/(\theta - 1)}}{2} \\
&= \frac{(u + v) - |u - v|}{2} = M(u, v).
\end{aligned}$$

□

**Exercise 3.36** Let  $C_\theta$  be a member of the Plackett family of copulas, where  $\theta$  is in  $(0, \infty)$ .

(a) Show that  $C_{1/\theta}(u, v) = u - C_\theta(u, 1 - v) = v - C_\theta(1 - u, v)$ .

(b) Conclude that  $C_\theta$  satisfies the functional equation  $C = \hat{C}$  for radial symmetry.

*Solution.* (a) Write

$$\begin{aligned}
& C_{1/\theta}(u, v) \\
&= \frac{[1 + (1/\theta - 1)(u + v)] - \sqrt{[1 + (1/\theta - 1)(u + v)]^2 - 4uv(1/\theta)(1/\theta - 1)}}{2(1/\theta - 1)} \\
&= \frac{[\theta + (1 - \theta)(u + v)] - \sqrt{[\theta + (1 - \theta)(u + v)]^2 - 4uv(1 - \theta)}}{2(1 - \theta)} \\
&= \frac{2u(1 - \theta) + 1 - (1 - \theta)(u + 1 - v)}{2(1 - \theta)} \\
&\quad + \frac{\sqrt{[2u(1 - \theta) + 1 - (1 - \theta)(u + 1 - v)]^2 - 4uv(1 - \theta)}}{2(1 - \theta)} \\
&= \frac{2u(1 - \theta) + 1 - (1 - \theta)(u + 1 - v)}{2(1 - \theta)} \\
&\quad + \frac{\sqrt{[1 + (\theta - 1)(u + 1 - v)]^2 + 4u^2(1 - \theta)^2 + 4u(1 - \theta)[1 + (\theta - 1)(u + 1 - v)] - 4uv(1 - \theta)}}{2(1 - \theta)} \\
&= u - \frac{[1 + (\theta - 1)(u + 1 - v)] - \sqrt{[1 + (\theta - 1)(u + 1 - v)]^2 - 4u(1 - v)\theta(\theta - 1)}}{2(\theta - 1)}.
\end{aligned}$$

Also fit for  $\theta = 1$ .

(b) Plackett family is radially symmetric,

$$\begin{aligned}
C(u, v) &= \frac{[1 + (\theta - 1)(u + v)] - \sqrt{[1 + (\theta - 1)(u + v)]^2 - 4uv\theta(\theta - 1)}}{2(\theta - 1)} \\
&= \frac{2(\theta - 1)(u + v - 1) + [1 + (\theta - 1)(2 - u - v)]}{2(\theta - 1)} \\
&\quad - \frac{\sqrt{[2(\theta - 1)(u + v - 1) + 1 + (\theta - 1)(2 - u - v)]^2 - 4uv\theta(\theta - 1)}}{2(\theta - 1)} \\
&= u + v - 1 + \frac{[1 + (\theta - 1)(2 - u - v)] - \sqrt{[1 + (\theta - 1)(2 - u - v)]^2 - 4(1 - u)(1 - v)\theta(\theta - 1)}}{2(\theta - 1)} \\
&= u + v - 1 + C(1 - u, 1 - v) = \hat{C}(u, v).
\end{aligned}$$

Also fit for  $\theta = 1$ . □

**Exercise 3.37** Show that the Plackett family is positively ordered.

*Solution.* Let  $0 < \theta_1 \leq \theta_2 < 1$ ,

$$C_\theta = \frac{[1 + (\theta - 1)(u + v)] - \sqrt{[1 + (\theta - 1)(u + v)]^2 - 4uv\theta(\theta - 1)}}{2(\theta - 1)},$$

then the derivative is

$$\begin{aligned}
\frac{\partial C_\theta}{\partial \theta} &= \frac{\{(u + v) - 1/2A^{-1/2}(2(u + v)[1 + (\theta - 1)(u + v)] - 4uv(2\theta - 1))\}2(\theta - 1)}{[2(\theta - 1)]^2} \\
&\quad - \frac{2\{[1 + (\theta - 1)(u + v)] - \sqrt{[1 + (\theta - 1)(u + v)]^2 - 4uv\theta(\theta - 1)}\}}{[2(\theta - 1)]^2} \\
&\geq 0.
\end{aligned}$$
□

**Exercise 3.38** Show that the following algorithm generates random variates  $(u, v)$  from Plackett distribution with parameter  $\theta$ :

1. Generate two independent uniform  $(0, 1)$  variates  $u, t$ ;
2. Set  $a = t(1 - t)$ ;  $b = \theta + a(\theta - 1)^2$ ;  $c = 2a(u\theta^2 + 1 - u) + \theta(1 - 2a)$ ; and  $d = \sqrt{\theta} \cdot \sqrt{\theta + 4au(1 - u)(1 - \theta)^2}$ ;
3. Set  $v = [c - (1 - 2t)d]/2b$ ;



4. The desired pair is  $(u, v)$ .

*Solution.* We only need to show that

$$[c - (1 - 2t)d]/2b = c_u^{(-1)}(t),$$

where  $c_u(t) = P(V \leq t|U = u) = \frac{\partial C(u, v)}{\partial u}$ . That is

$$\frac{\partial C(u, v)}{\partial u} = \frac{1}{2} - \frac{1}{\sqrt{[1 + (\theta - 1)(u + v)]^2 - 4uv\theta(\theta - 1)}} \cdot [1 + (\theta - 1)(u + v) - 2v\theta],$$

The associated quasi-inverse is

$$c_u^{(-1)}(v) =$$

□

### Exercise 3.39

*Solution.*

□

**Exercise 3.40** Let  $C_\theta$  denote a member of the Ali-Mikhail-Haq family. Show that

$$C_\theta(u, v) = uv \sum_{k=0}^{\infty} [\theta(1 - u)(1 - v)]^k$$

and hence

*Solution.*

□

**Exercise 3.41** (a) Show that the harmonic mean of two Ali-Mikhail-Haq copulas is again an Ali-Mikhail-Haq copula.

(b) Show that each Ali-Mikhail-Haq copula is a weighted harmonic mean of the two extreme members of the family, *i.e.*, for all  $\theta \in [-1, 1]$ ,

$$C_\theta(u, v) = \frac{1}{\frac{1-\theta}{2} \cdot \frac{1}{C_{-1}(u, v)} + \frac{1+\theta}{2} \cdot \frac{1}{C_{+1}(u, v)}}.$$

*Solution.* (a) Let  $C_\alpha, C_\beta$  be Ali-Mikhail-Haq copulas, then

$$\begin{aligned} \frac{2}{\frac{1}{C_\alpha(u, v)} + \frac{1}{C_\beta(u, v)}} &= \frac{2uv}{1 - \alpha(1 - u)(1 - v) + 1 - \beta(1 - u)(1 - v)} \\ &= \frac{2uv}{2 - (\alpha + \beta)(1 - u)(1 - v)} = C_{(\alpha + \beta)/2}. \end{aligned}$$

(b) Write

$$\begin{aligned} \frac{1}{\frac{1-\theta}{2} \cdot \frac{1}{C_{-1}(u,v)} + \frac{1+\theta}{2} \cdot \frac{1}{C_{+1}(u,v)}} &= \frac{2uv}{(1-\theta)[1 + (1-u)(1-v)] + (1+\theta)[1 - (1-u)(1-v)]} \\ &= \frac{uv}{1 - \theta(1-u)(1-v)} = C_\theta(u, v). \end{aligned}$$

□

**Exercise 3.42** Show that the following algorithm generates random variates  $(u, v)$  from an Ali-Mikhail-Haq distribution with parameter  $\theta$ :

1. Generate two independent uniform  $(0, 1)$  variates  $u, t$ ;
2. Set  $a = 1 - u$ ;  $b = -\theta(2at + 1) + 2\theta^2 a^2 t + 1$ ; and  $c = \theta^2(4a^2 t - 4at + 1) - \theta(4at - 4t + 2) + 1$ ;
3. Set  $v = 2t(a\theta - 1)^2 / (b + \sqrt{c})$ ;
4. The desired pair is  $(u, v)$ .

*Solution.* Recall that Ali-Mikhail-Haq copula is

$$C_\theta(u, v) = \frac{uv}{1 - \theta(1-u)(1-v)}$$

for  $\theta \in [-1, 1]$ . Thus

$$c_u(v) = \frac{v - v\theta + v^2\theta}{(1 - \theta(1-u)(1-v))^2}.$$

The associated quasi-inverse is

$$c_u^{(-1)}(v) =$$

□

**Exercise 4.1** [Theorem 4.1.5] Let  $C$  be an Archimedean copula with generator  $\varphi$ . Then:

1.  $C$  is symmetric; i.e.  $C(u, v) = C(v, u)$  for all  $u, v \in I$ ;
2.  $C$  is associative; i.e.,  $C(C(u, v), w) = C(u, C(v, w))$  for all  $u, v, w \in I$ ;
3. If  $c > 0$  is any constant, then  $c\varphi$  is also a generator of  $C$ .

*Solution.* 1. Since

$$C(u, v) = \varphi^{[-1]}(\varphi(u) + \varphi(v)) = \varphi^{[-1]}(\varphi(v) + \varphi(u)) = C(v, u).$$

2. Write

$$\begin{aligned} C(C(u, v), w) &= \varphi^{[-1]}(\varphi[C(u, v)] + \varphi(w)) \\ &= \varphi^{[-1]}(\varphi[\varphi^{[-1]}(\varphi(u) + \varphi(v))] + \varphi(w)) \\ &= \varphi^{[-1]}(\varphi(u) + \varphi(v) + \varphi(w)) \\ &= \varphi^{[-1]}(\varphi(u) + \varphi[\varphi^{[-1]}(\varphi(v) + \varphi(w))]) \\ &= C(u, C(v, w)). \end{aligned}$$

If  $\varphi(u) + \varphi(v) \geq \varphi(0)$  or  $\varphi(v) + \varphi(w) \geq \varphi(0)$ ,  $C(C(u, v), w) = C(u, C(v, w)) = 0$ .

3. Write

$$\begin{aligned} c\varphi^{[-1]}(c\varphi(u) + c\varphi(v)) &= c\varphi^{[-1]}[c(\varphi(u) + \varphi(v))] \\ &= \varphi^{[-1]} \left[ \frac{1}{c} \cdot c(\varphi(u) + \varphi(v)) \right] \\ &= \varphi^{[-1]}(\varphi(u) + \varphi(v)). \end{aligned}$$

If  $\varphi(u) + \varphi(v) \geq \varphi(0)$ , then  $c(\varphi(u) + \varphi(v)) \geq c\varphi(0)$ , it implies that  $c\varphi^{[-1]}[c(\varphi(u) + \varphi(v))] = 0 = C(u, v)$ . □

**Exercise 4.2** The diagonal section of an Archimedean copula  $C$  with generator  $\phi$  in  $\Omega$  is given by  $\delta_C(u) = \varphi^{[-1]}[2\varphi(u)]$ . Prove that if  $C$  is Archimedean, then for  $u \in (0, 1)$ ,  $\delta_C(u) < u$ . Conclude that  $M$  is not Archimedean copula.

*Solution.* If  $2\varphi(u) \geq \varphi(0)$ , then  $\delta_C(u) = 0 < u$ . If  $2\varphi(u) < \varphi(0)$ , since  $\varphi^{[-1]}$  is decreasing,  $\delta_C(u) = \varphi^{[-1]}[2\varphi(u)] < \varphi^{[-1]}[\varphi(u)] = u$ . The diagonal section of  $M$  is

$$\delta_M(u) = \min(u, u) = u,$$

thus  $M$  is not Archimedean. □

**Exercise 4.3** Show that  $\varphi : I \rightarrow [0, \infty]$  is in  $\Omega$  iff  $1 - \varphi^{[-1]}(t)$  is a unimodal distribution function on  $[0, \infty]$  with mode at zero.

*Solution.* The corresponding density function is

$$\frac{d(1 - \varphi^{[-1]}(t))}{dt} = -\frac{1}{\varphi'(\varphi^{[-1]}(t))}. \quad (2.0.3)$$

This distribution is unimodal at zero iff the density function is decreasing on  $[0, \infty]$  and

$$\frac{d(-1/\varphi'(\varphi^{[-1]}(t)))}{dt} = \frac{\varphi''(\varphi^{[-1]}(t))}{\varphi'(\varphi^{[-1]}(t))} < 0.$$

Since  $\varphi'(\varphi^{[-1]}(t)) < 0$  by (2.0.3),  $\varphi''(\varphi^{[-1]}(t)) > 0$ . Thus  $\varphi$  is convex. And we need  $\varphi'(\varphi^{[-1]}(t))$  is a decreasing function of  $t$ . Thus it is iff  $\varphi$  is strictly decreasing and convex with  $\varphi(1) = 0$ . Since if  $\varphi(1) = a > 0$ ,  $1 - \varphi^{[-1]}(0) < 1 - \varphi^{[-1]}(a) = 0$ , which is impossible. □

**Exercise 4.4** Show that non-Archimedean copulas can have

- (a) non-convex level curves;
- (b) convex level curves.

*Solution.* □

**Exercise 4.5** Let  $C$  be an Archimedean copula. Prove that  $C$  is strict if and only if  $C(u, v) > 0$  for  $(u, v) \in (0, 1]^2$ .

*Solution.* We need to prove that  $\varphi(0) = \infty$  iff  $C(u, v) > 0$  for  $(u, v) \in (0, 1]^2$ . That is, for all  $(u, v) \in (0, 1]^2$ ,

$$\begin{aligned} \varphi^{[-1]}(\varphi(u) + \varphi(v)) > 0 &\Leftrightarrow 0 \leq \varphi(u) + \varphi(v) < \varphi(0) \\ &\Leftrightarrow \varphi(0) = \infty. \end{aligned}$$

□

**Exercise 4.6** This exercise shows that different Archimedean copulas can have the same zero set. Let

*Solution.* □

**Exercise 4.7**

*Solution.*



**Exercise 4.8**

*Solution.*



**Exercise 4.9**

*Solution.*



**Exercise 4.10**

*Solution.*



**Exercise 4.11**

*Solution.*



**Exercise 4.12**

*Solution.*



**Exercise 4.13**

*Solution.*



**Exercise 4.14**

*Solution.*



**Exercise 4.15**

*Solution.*



**Exercise 4.16**

*Solution.*



**Exercise 4.17** Show that the following algorithm generates random variates  $(u, v)$  whose joint distribution function is the Clayton copula with parameter  $\theta > 0$ :

- 1.
- 2.
- 3.

*Solution.*



**Exercise 4.18**

*Solution.*



**Exercise 4.19**

*Solution.*



**Exercise 4.20***Solution.*

□

**Exercise 4.21***Solution.*

□

**Exercise 4.22***Solution.*

□

**Exercise 4.23***Solution.*

□

**Exercise 4.24***Solution.*

□

**Exercise 4.25***Solution.*

□

**Exercise 5.1** [Corollary 5.1.2.] 1.  $Q$  is symmetric in its arguments:  $Q(C_1, C_2) = Q(C_2, C_1)$ .

2.  $Q$  is non-decreasing in each argument: if  $C_1 \prec C'_1$  and  $C_2 \prec C'_2$  for all  $(u, v) \in \mathbb{I}^2$ , then  $Q(C_1, C_2) \leq Q(C'_1, C'_2)$ .

3. Copulas can be replaced by survival copulas in  $Q$ , i.e.,  $Q(C_1, C_2) = Q(\hat{C}_1, \hat{C}_2)$ .

*Solution.* 1. We have

$$\begin{aligned}
 Q(C_1, C_2) &= 4 \iint_{\mathbb{I}^2} C_2(u, v) dC_1(u, v) - 1 \\
 &= 2P((X_1 - X_2)(Y_1 - Y_2) > 0) - 1 \\
 &= 2[P(X_1 < X_2, Y_1 < Y_2) + P(X_1 > X_2, Y_1 > Y_2)] - 1 \\
 &= 2 \left[ \iint_{\mathbb{R}^2} P(X_1 < x, Y_1 < y) dC_2(F(x), G(y)) \right. \\
 &\quad \left. + \iint_{\mathbb{R}^2} P(X_1 > x, Y_1 > y) dC_2(F(x), G(y)) \right] - 1 \\
 &= 2 \left[ \iint_{\mathbb{R}^2} P(X_1 < x, Y_1 < y) dC_2(F(x), G(y)) \right. \\
 &\quad \left. + \iint_{\mathbb{R}^2} \{1 - F(x) - G(y) + P(X_1 < x, Y_1 < y)\} dC_2(F(x), G(y)) \right] - 1
 \end{aligned}$$

$$\begin{aligned}
&= 4 \iint_{I^2} C_1(u, v) dC_2(u, v) - 1 \\
&= Q(C_2, C_1).
\end{aligned}$$

2. This is trivial by the definition.

3. For any  $H$ , it is true that

$$\iint H(x, y) dH(x, y) = \iint \bar{H}(x, y) dH(x, y) = \iint \bar{H}(x, y) d\bar{H}(x, y).$$

□

**Exercise 5.2** Let  $X, Y$  be r.v.'s with the Marshall-Olkin bivariate exponential distribution with parameters  $\lambda_1, \lambda_2, \lambda_{12}$ , that is, the survival function is given by

$$\bar{H}(x, y) = \exp(-\lambda_1 x - \lambda_2 y - \lambda_{12} \max(x, y)).$$

(a) Show that the ordinary Pearson product-moment correlation coefficient of  $X, Y$  is given by

$$\frac{\lambda_{12}}{\lambda_1 + \lambda_2 + \lambda_{12}}.$$

(b) Show that Kendall's tau and Pearson's product-moment correlation coefficient are numerically equal for members of this family.

*Solution.* (a) In Marshall's paper, these moments were calculated using MGF.

(b) From Example 5.5.,

$$\begin{aligned}
\tau_{\alpha, \beta} &= \frac{\alpha\beta}{\alpha - \alpha\beta + \beta} \\
&= \frac{\frac{\lambda_{12}}{\lambda_1 + \lambda_{12}} \cdot \frac{\lambda_{12}}{\lambda_2 + \lambda_{12}}}{\frac{\lambda_{12}}{\lambda_1 + \lambda_{12}} - \frac{\lambda_{12}^2}{(\lambda_1 + \lambda_{12})(\lambda_2 + \lambda_{12})} + \frac{\lambda_{12}}{\lambda_2 + \lambda_{12}}} \\
&= \frac{\lambda_{12}^2}{\lambda_{12}(\lambda_2 + \lambda_{12}) - \lambda_{12}^2 + \lambda_{12}(\lambda_1 + \lambda_{12})} \\
&= \frac{\lambda_{12}}{\lambda_1 + \lambda_2 + \lambda_{12}}.
\end{aligned}$$

□

**Exercise 5.3** Prove that an alternate expression for Kendall's tau for an Archimedean copula  $C$  with generator  $\varphi$  is

$$\tau_C = 1 - 4 \int_0^\infty u \left[ \frac{d}{du} \varphi^{[-1]}(u) \right]^2 du.$$

*Solution.* For Archimedean copula  $C$ ,

$$\tau_C = 1 + 4 \int_0^1 \frac{\varphi(t)}{\varphi'(t)} dt.$$

Then let  $\varphi(t) = u$ , one has  $d\varphi^{[-1]}(u)/du = 1/\varphi'(t)$  if  $u \leq \varphi(0)$ , if  $u \geq \varphi(0)$ ,  $d\varphi^{[-1]}(u)/du = 0$ . Thus

$$\begin{aligned} \tau_C &= 1 + 4 \int_{\varphi(0)}^0 u \frac{d}{du} \varphi^{[-1]}(u) d\varphi^{[-1]}(u) \\ &= 1 - 4 \int_0^{\varphi(0)} u \left[ \frac{d}{du} \varphi^{[-1]}(u) \right]^2 du \\ &= 1 - 4 \int_0^\infty u \left[ \frac{d}{du} \varphi^{[-1]}(u) \right]^2 du. \end{aligned}$$

□

**Exercise 5.4** (a) Let  $C_\theta, \theta \in [0, 1]$  be a member of

(b)

*Solution.*

□

**Exercise 5.5** Let  $C$  be a diagonal copula, that is,  $C(u, v) = \min(u, v, (1/2)[\delta(u) + \delta(v)])$ .

(a) Show the Kendall's tau is given by

$$\tau_C = 4 \int_0^1 \delta(t) dt - 1.$$

(b) For diagonal

*Solution.* (a) Write

$$C(u, u) = \min(u, \delta(u)) = \delta(u),$$

since

□

**Exercise 5.6**



*Solution.*

□

### Exercise 5.7

*Solution.*

□

**Exercise 5.8** Let  $C_\theta$  be a member of the Plackett family

$$a$$

for  $\theta > 0$ . Show that Spearman's rho for this  $C_\theta$  is

$$\rho_\theta = \frac{\theta + 1}{\theta - 1} - \frac{2\theta}{(\theta - 1)^2} \ln \theta.$$

There does not appear to be a closed form expression for Kendall's  $\tau$  for members of this family.

*Solution.*

□

**Exercise 5.9** Let  $C_\theta, \theta \in \overline{\mathbb{R}}$  be a member of the Frank family. Show that

$$\tau_\theta = 1 - \frac{4}{\theta}[1 - D_1(\theta)], \quad \rho_\theta = 1 - \frac{12}{\theta}[D_1(\theta) - D_2(\theta)],$$

where  $D_k(x)$  is the Debye function, which is defined for any positive integer  $k$  by

$$D_k(x) = \frac{k}{x^k} \int_0^x \frac{t^k}{e^t - 1} dt.$$

*Solution.*

□

### Exercise 5.10

*Solution.*

□

### Exercise 5.11

*Solution.*

□

### Exercise 5.12

*Solution.*

□

### Exercise 5.13

*Solution.*

□

### Exercise 5.14

*Solution.*

□

**Exercise 5.15***Solution.*

□

**Exercise 5.16** Let  $X, Y$  be continuous random variables with copula  $C$ . Show that an alternate expression for Spearman's rho for  $X, Y$  is

$$\rho = 3 \iint_{\mathbb{I}^2} ([u + v - 1]^2 - [u - v]^2) dC(u, v).$$

*Solution.* Write

$$\begin{aligned} \rho &= 3 \iint_{\mathbb{I}^2} ([u + v - 1]^2 - [u - v]^2) dC(u, v) \\ &= 3 \iint_{\mathbb{I}^2} ((v - 1)^2 + 2uv - 2u - v^2 + 2uv) dC(u, v) \\ &= 3 \iint_{\mathbb{I}^2} (1 - 2v - 2u + 4uv) dC(u, v) \\ &= 12 \iint uv dC(u, v) - 3. \end{aligned}$$

□

**Exercise 5.17** Let  $X$  and  $Y$  be continuous random variables with copula  $C$ . Establish the following inequalities between Blomqvist's  $\beta$  and Kendall's  $\tau$ , Spearman's  $\rho$ , and Gini's  $\gamma$ :

$$\begin{aligned} \frac{1}{4}(1 + \beta)^2 - 1 &\leq \tau \leq 1 - \frac{1}{4}(1 - \beta)^2, \\ \frac{3}{16}(1 + \beta)^3 - 1 &\leq \rho \leq 1 - \frac{3}{16}(1 - \beta)^3, \\ \frac{3}{8}(1 + \beta)^2 - 1 &\leq \gamma \leq 1 - \frac{3}{8}(1 - \beta)^2. \end{aligned}$$

*Solution.*

□

**Exercise 5.18***Solution.*

□

**Exercise 5.19***Solution.*

□

**Exercise 5.20** Let  $X, Y$  be continuous random variables whose copula  $C$  satisfies one (or both) of the functional equations

$$C(u, v) = u - C(u, 1 - v), \quad C(u, v) = v - C(1 - u, v).$$

for joint symmetry. Show that

$$\tau_{X,Y} = \rho_{X,Y} = \gamma_{X,Y} = \beta_{X,Y} = 0.$$

*Solution.* Clearly,  $C(1/2, 1/2) = 1/2 - C(1/2, 1/2) \implies C(1/2, 1/2) = 1/4$ ,

$$\beta_{X,Y} = 4C(1/2, 1/2) - 1 = 1 - 1 = 0.$$

Since these four are measures of concordance,

$$\kappa_{-X,Y} = \kappa_{X,-Y} = -\kappa_{X,Y}.$$

From the monotonic transformation,

$$C(u, v) = u - C(u, 1 - v), \quad C(u, v) = v - C(1 - u, v) \implies C_{X,Y} = C_{X,-Y}, \quad C_{X,Y} = C_{-X,Y}.$$

Thus

$$\kappa_{X,Y} = \kappa_{X,-Y} = -\kappa_{X,Y}, \quad \kappa_{X,Y} = \kappa_{-X,Y} = -\kappa_{X,Y}.$$

Therefore,  $\kappa_{X,Y} = 0$  as its range is  $[-1, 1]$ . □

**Exercise 5.21** Another measure of association between two variates is **Spearman's foot-rule**, for which the sample version is

$$f = 1 - \frac{3}{n^2 - 1} \sum_{i=1}^n |p_i - q_i|,$$

where  $p_i, q_i$  denote the ranks of a sample of size  $n$  of two continuous random variables  $X, Y$ .

(a) Show that the population version of the foot-rule, which is

$$\phi = 1 - 3 \iint_{\mathbf{I}^2} |u - v| dC(u, v) = \frac{1}{2} [3Q(C, M) - 1]$$

(b) Show that  $\phi$  fails to satisfy

$$-1 \leq \kappa_{X,Y} \leq 1, \kappa_{X,X} = 1, \kappa_{X,-X} = -1,$$

and

$$\kappa_{-X,Y} = \kappa_{X,-Y} = -\kappa_{X,Y}.$$

Hence it is not a “measure of concordance”.

*Solution.* (a) Rewrite  $f$  as

$$\begin{aligned} f &= 1 - \frac{3n^2}{n^2 - 1} \left[ \sum_{i=1}^n \left| \frac{p_i}{n} - \frac{q_i}{n} \right| \right] \cdot \frac{1}{n} \\ &= 1 - 3\mathbb{E}[|U - V|] \\ &= 1 - 3 \iint_{\mathbb{I}^2} |u - v| dC(u, v). \end{aligned}$$

Recall that

$$\begin{aligned} Q(C, M) &= 4 \iint_{\mathbb{I}^2} M(u, v) dC(u, v) - 1 \\ &= 2 \iint_{\mathbb{I}^2} [u + v - |u - v|] dC(u, v) - 1 \\ &= 1 - 2 \iint_{\mathbb{I}^2} |u - v| dC(u, v). \end{aligned}$$

Hence

$$f = 1 - \frac{3}{2} (1 - Q(C, M)) = \frac{1}{2} [3Q(C, M) - 1].$$

(b) Since  $Q(C, M) \in [0, 1]$ ,  $f \in [-1/2, 1]$ . We can never attain  $-1$  for  $f$ . Besides,

$$\begin{aligned} \frac{3}{2} Q(C_{X,-Y}, M) - \frac{1}{2} &= 6 \int_{\mathbb{I}} u - C(u, 1 - u) du - \frac{3}{2} - \frac{1}{2} \\ &= 1 - 6 \int_{\mathbb{I}} C(u, 1 - u) du \\ &\neq \frac{1}{2} - 6 \int_{\mathbb{I}} C(u, u) du + \frac{3}{2} \\ &= \frac{1}{2} - \frac{3}{2} Q(C_{X,Y}, M). \end{aligned}$$



**Exercise 5.22** (a) Show that

$$P(X \leq x, Y \leq y) \geq P(X \leq x)P(Y \leq y)$$

and

$$P(X > x, Y > y) \geq P(X > x)P(Y > y)$$

are equivalent.

(b) Show that

$$H(x, y) \geq F(x)G(y), \quad \forall (x, y) \in \mathbb{R}^2$$

is equivalent to

$$\overline{H}(x, y) \geq \overline{F}(x)\overline{G}(y), \quad \forall (x, y) \in \mathbb{R}^2.$$

*Solution.* (a) Write

$$P(X \leq x, Y \leq y) \geq P(X \leq x)P(Y \leq y)$$

$$\Leftrightarrow 1 - P(X > x) - P(Y > y) + P(X > x, Y > y) \geq (1 - P(X > x))(1 - P(Y > y))$$

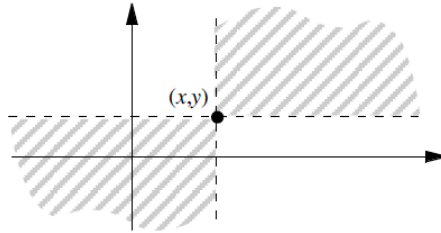
$$\Leftrightarrow P(X > x, Y > y) \geq P(X > x)P(Y > y).$$

(b) It is directly from part (a). □

**Exercise 5.23** (a) Let  $X, Y$  be random variables with joint distribution function  $H$  and margins  $F$  and  $G$ . Show that PQD( $X, Y$ ) iff for any  $(x, y) \in \mathbb{R}^2$ ,

$$H(x, y)[1 - F(x) - G(y) + H(x, y)] \geq [F(x) - H(x, y)][G(y) - H(x, y)],$$

that is the product of two probabilities corresponding to the two shaded quadrants is at least as great as the unshaded quadrant in the following figure:



(b) Give an interpretation of quadrant dependence in terms of the cross product ratio

$$\theta = \frac{H(x, y)[1 - F(x) - G(y) + H(x, y)]}{[F(x) - H(x, y)][G(y) - H(x, y)]}$$

for continuous random variables.

(c) In copula notation,

$$C(u, v)[1 - u - v + C(u, v)] \geq [u - C(u, v)][v - C(u, v)].$$

*Solution.* (a) From Exercise 5.22,

$$\begin{aligned} H(x, y)[1 - F(x) - G(y) + H(x, y)] &\geq [1 - F(x)][1 - G(y)]F(x)G(y) \\ &= [G(y) - F(x)G(y)][F(x) - F(x)G(y)] \\ &\geq [G(y) - H(x, y)][F(x) - H(x, y)]. \end{aligned}$$

(b) If  $\theta \geq 1$ ,  $X, Y$  are positive quadrant dependent, if  $\theta \leq 1$ ,  $X, Y$  are negative quadrant dependent.

(c) The interpretation is the product of two probabilities for  $(U, V)$  corresponding to the two shaded quadrants is at least as great as the unshaded quadrant.  $\square$

**Exercise 5.24** (a) Show that if  $X, Y$  are PQD, then  $-X, Y$  are NQD,  $X$  and  $-Y$  are NQD, and  $-X, -Y$  are PQD.

(b) Show that if  $C$  is the copula of PQD random variables, then so is  $\hat{C}$ .

*Solution.* (a) If  $X, Y$  are PQD,

$$P(X \leq x, Y \leq y) \geq P(X \leq x)P(Y \leq y).$$

One has

$$\begin{aligned} P(-X \leq x, Y \leq y) &= P(X \geq -x, Y \leq y) \\ &= P(Y \leq y) - P(X \leq -x, Y \leq y) \\ &= G(Y) - H(-x, y). \end{aligned}$$

Since PQD,

$$H(-x, y) \geq F(-x)G(y),$$

then

$$\begin{aligned} P(-X \leq x, Y \leq y) &= G(y) - H(-x, y) \\ &\leq G(y) - F(-x)G(y) \\ &= (1 - F(-x))G(y) \\ &= P(-X \leq x)P(Y \leq y). \end{aligned}$$

The others are in the same fashion.

(b) The PQD is

$$C(u, v) \geq uv.$$

From part (a),  $-X, -Y$  are also PQD,

$$C_{-X, -Y}(u, v) \geq uv \Leftrightarrow \hat{C}_{X, Y}(u, v) \geq uv.$$

□

**Exercise 5.25** Consider the random variable  $Z = H(X, Y) - F(X)G(Y)$ .

(a) Show that  $\mathbb{E}[Z] = (3\tau_C - \rho_C)/12$ .

(b) Show that  $\omega_C = 6\mathbb{E}[Z] = (3\tau_C - \rho_C)/2$  can be interpreted as a measure of “expected” quadrant dependence for which  $\omega_M = 1$ ,  $\omega_\Pi = 0$ ,  $\omega_W = -1$ .

(c) Show that  $\omega_C$  fails to be a measure of concordance.

*Solution.* (a) Write

$$\begin{aligned} \mathbb{E}[Z] &= \mathbb{E}[C(U, V)] - E[UV] \\ &= \frac{\tau_C + 1}{4} - \frac{\rho_C + 3}{12} \\ &= \frac{3\tau_C - \rho_C}{12}. \end{aligned}$$

(b) Write

$$6\mathbb{E}[Z] = 6 \iint_{\mathbb{I}^2} [C(u, v) - uv] dC(u, v).$$

(c)

□

**Exercise 5.26** *Hoeffding's lemma.* Let  $X, Y$  be random variables with joint distribution function  $H$  and margins  $F, G$ , such that  $\mathbb{E}[|X|], \mathbb{E}[|Y|]$  and  $\mathbb{E}[|XY|]$  are finite. Prove that

$$\mathbb{Cov}(X, Y) = \iint_{\mathbb{R}^2} [H(x, y) - F(x)G(y)] dx dy.$$

*Solution.*

□

**Exercise 5.27** Let  $X, Y$  be random variables. Show that if  $\text{PQD}(X, Y)$ , then  $\mathbb{Cov}(X, Y) \geq 0$ , and hence Pearson's product-moment correlation coefficient is non-negative for positively quadrant dependent random variables.

*Solution.* This is directly from Exercise 5.26.

□

**Exercise 5.28** Show that  $X, Y$  are PQD iff  $\mathbb{Cov}[f(X), g(Y)] \geq 0$  for all functions  $f, g$  that are non-decreasing in each place and for which expectations  $\mathbb{E}[f(X)], \mathbb{E}[g(Y)], \mathbb{E}[f(X)g(Y)]$  exist.

*Solution.*

□

**Exercise 5.29** Prove that if the copula of  $X, Y$  are max-stable, then  $\text{PQD}(X, Y)$ .

*Solution.*

□

**Exercise 5.30**

*Solution.*

□

**Exercise 5.31** Let  $X$  and  $Y$  be continuous random variables whose copula is  $C$ .

(a) Show that if  $C = \hat{C}$ , then  $\text{LTD}(Y|X)$  iff  $\text{RTI}(Y|X)$ , and  $\text{LTD}(X|Y)$  iff  $\text{RTI}(X|Y)$ .

(b) Show that if  $C$  is symmetric, then  $\text{LTD}(Y|X)$  iff  $\text{LTD}(X|Y)$ , and  $\text{RTI}(Y|X)$  iff  $\text{RTI}(X|Y)$ .

*Solution.* (a)  $\text{LTD}(Y|X)$ , if and only if  $C(u, v)/u$  is non-increasing in  $u$ . That is

$$\frac{\frac{\partial C(u, v)}{\partial u} u - C(u, v)}{u^2} \leq 0 \implies \frac{\partial C(u, v)}{\partial u} u \leq C(u, v). \quad (2.0.4)$$



$\text{RTI}(Y|X)$  iff  $\hat{C}(1-u, 1-v)/(1-u)$  is non-decreasing in  $u$ . Assume  $\hat{C} = C$ ,  $\hat{C}(1-u, 1-v)/(1-u) = C(1-u, 1-v)/(1-u)$ , then

$$\frac{\frac{\partial C(1-u, 1-v)}{\partial(1-u)}(1-u)(-1) + C(1-u, 1-v)}{(1-u)^2} \geq 0$$

by (2.0.4).

(b) Assume  $C(u, v) = C(v, u)$ ,  $C(u, v)/v = C(v, u)/v$ , the derivative

$$\frac{\frac{\partial C(v, u)}{\partial v}v - C(v, u)}{v^2} \leq 0 \Leftrightarrow \frac{\partial C(v, u)}{\partial v}v \leq C(v, u)$$

iff

$$\frac{\partial C(u, v)}{\partial u}u \leq C(v, u) \Leftrightarrow \frac{\frac{\partial C(u, v)}{\partial u}u - C(u, v)}{u^2} \leq 0.$$

□

### Exercise 5.32

*Solution.*

□

### Exercise 5.33

*Solution.*

□

### Exercise 5.34

*Solution.*

□

### Exercise 5.35

*Solution.*

□

**Exercise 5.36** Show that (a) if the function  $u - C(u, v)$  is  $\text{TP}_2$ , then  $\text{LTD}(Y|X)$  and  $\text{RTI}(X|Y)$ ;

(b) if the function  $v - C(u, v)$  is  $\text{TP}_2$ , then  $\text{LTD}(X|Y)$  and  $\text{RTI}(Y|X)$ ;

(c) the function  $1 - u - v + C(u, v)$  is  $\text{TP}_2$  iff  $\hat{C}$  is  $\text{TP}_2$ .

*Solution.* (a)  $u - C(u, v)$  is  $\text{TP}_2$ , that is

$$F(x) - H(x, y) = P(X \leq x) - P(X \leq x, Y \leq y) = P(X \leq x, Y > y).$$

(b)

(c)

□

**Exercise 5.37***Solution.*

□

**Exercise 5.38***Solution.*

□

**Exercise 5.39***Solution.*

□

**Exercise 5.40** Let  $X, Y$  be continuous random variables whose copula  $C$  is a member of a totally ordered family that include  $\Pi$ . Show that  $\sigma_{X,Y} = |\rho_{X,Y}|$ .

*Solution.*

□

**Exercise 5.41***Solution.*

□

**Exercise 5.42***Solution.*

□

**Exercise 5.43***Solution.*

□

**Exercise 5.44** Show that  $k_p$  is given by

$$k_p = \frac{\Gamma(2p+3)}{2\Gamma^2(p+1)}.$$

*Solution.*

□

**Exercise 5.45** Show that the “ $\ell_p$ ” generalization of  $\gamma_C, \rho_C$  leads to measures of association given by

*Solution.*

□

**Exercise 5.46** Show that the “ $L_p$ ” generalization of  $\gamma_C$  leads to measures of association given by

*Solution.*

□

**Exercise 5.47** Verify the entries for  $\lambda_U, \lambda_L$ .

*Solution.*

□

**Exercise 5.48** Write  $\lambda_U(C), \lambda_L(C)$  to specify the copula under consideration. Prove that  $\lambda_U(\hat{C}) = \lambda_L(C)$  and  $\lambda_L(\hat{C}) = \lambda_U(C)$ .

*Solution.* Write

$$\begin{aligned}\lambda_U(\hat{C}) &= \lim_{t \rightarrow 1^-} \frac{2 - 2t - 1 + [2t - 1 + C(1 - t, 1 - t)]}{1 - t} \\ &= \lim_{t \rightarrow 1^-} \frac{C(1 - t, 1 - t)}{1 - t} \\ &= \lim_{t \rightarrow 0^+} \frac{C(t, t)}{t} = \lambda_L(C).\end{aligned}$$

The other one is similar. □

**Exercise 5.49**

*Solution.* □

**Exercise 5.50**

*Solution.* □