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RANDOM NOISE AND PERTURBATION OF COPULAS

RADKO MESIAR, AYYUB SHEIKHI AND MAGDA KOMORNÍKOVÁ

For a random vector (X, Y) characterized by a copula $C_{X,Y}$ we study its perturbation $C_{X+Z,Y}$ characterizing the random vector $(X + Z, Y)$ affected by a noise Z independent of both X and Y . Several examples are added, including a new comprehensive parametric copula family $(\mathcal{C}_k)_{k \in [-\infty, \infty]}$.

Keywords: copula, noise, perturbation of copula, random vector

Classification: 60E05, 62H20

1. INTRODUCTION AND PRELIMINARIES

Observation of random variables is often polluted by a random noise. On the other hand, the stochastic dependence of a couple of random variables is described by means of copulas, e. g., Sklar (1959) [6], Joe (1997) [4], Nelsen (2006) [5] and Durante & Sempi (2016) [2].

The main aim of this contribution is the study of perturbation of a copula $C_{X,Y}$ related to a random vector (X, Y) in the case when the first coordinate X is polluted by some noise Z , i. e., we are interested in copula $C_{X+Z,Y}$ when Z is random variable independent of both X and Y .

Recall that due to Sklar theorem [6] for any random vector (X, Y) such that

$$F_{X,Y} : \mathbb{R}^2 \rightarrow [0, 1]$$

is its joint distribution function and

$$F_X, F_Y : \mathbb{R} \rightarrow [0, 1]$$

are distribution functions of random variables X, Y respectively, there exists a copula

$$C : [0, 1]^2 \rightarrow [0, 1]$$

such that

$$F_{X,Y}(x, y) = C(F_X(x), F_Y(y)). \tag{1}$$

Copula C is unique if both X and Y are continuous random variables.

We denote the copula coupling X and Y as $C_{X,Y}$. Note also that we have an axiomatic characterization of copulas, see [2, 4, 5]:

A function $C : [0, 1]^2 \rightarrow [0, 1]$ is a copula if and only if it satisfies the next three axioms:

- C1** $C(x, 0) = C(0, x) = 0$ for each $x \in [0, 1]$ (C is grounded);
- C2** $C(x, 1) = C(1, x) = x$ for each $x \in [0, 1]$ (C has a neutral element $e = 1$);
- C3** for all $x_1, x_2, y_1, y_2 \in [0, 1]$, $x_1 \leq x_2$, $y_1 \leq y_2$ it holds $C(x_1, y_1) + C(x_2, y_2) - C(x_2, y_1) - C(x_1, y_2) \geq 0$ (C is 2-increasing).

A similar axiomatization describe n -ary copulas characterizing the stochastic dependence structure of n -ary random vectors (X_1, \dots, X_n) . In this paper, we consider only binary copulas, and for readers interested in n -ary copulas, $n \geq 2$ we recommend books [2, 4, 5].

There are many copula families available in the literature, in which, the upper Fréchet-Hoeffding, the lower Fréchet-Hoeffding and the product copula are known as the three basic copulas and denoted by M , $M(u, v) = \min(u, v)$, W , $W(u, v) = \max(0, u + v - 1)$ and Π , $\Pi(u, v) = uv$, respectively. It is well known that for any copula C it holds

$$W \leq C \leq M.$$

More, it is known that when X and Y are linked by a copula $C_{X,Y}$, the distribution function of $X + Y$ of their sum is given by

$$F_{X+Y}(s) = \int \int_{x+y \leq s} dH_{X,Y}(x, y) \quad s \in \mathbb{R}. \tag{2}$$

Where $H_{X,Y}(x, y) = C_{X,Y}(F_X(x), F_Y(y))$ and $dH_{X,Y}(x, y)$ is its derivative. When $C = \Pi$, the formula (2) is referred as convolution of two independent random variables X and Y . Williamson and Downs (1990) [8] discussed the distribution of some arithmetic convolution function of X, Y where their marginal distributions is fixed. One may refer to Cherubini et al. (2016) [1], for more details of distributions of convolution of variables.

Also, the distribution of linear combination of two (or more) random variables in the case when their marginal distributions as well as their copula are available, has been investigated in the literature. Durante and Sempi (2016) [2] have obtained the distribution of $X + Y$ as well as the copula of $X + Y$ and Y in this situation. Gijbels and Herrmann (2014) [3] have found the distribution of sums of random variables with copula-induced dependence. See also Williamson and Downs (1990) [8], Wang (2014) [7] and Durante and Sempi (2016) [2].

As already mentioned, the main aim of this paper is related to the description of copula $C_{X+Z,Y}$ which can be seen as a perturbation of the copula $C_{X,Y}$. Here we suppose the independence of Z on X and Y , and the main results are given in the next section. In Section 3, we study a particular case when X and Y are uniformly distributed on $[0, 1]$, and Z_ε is uniformly distributed on $[0, \varepsilon]$, where $\varepsilon \in]0, \infty[$ is a positive parameter charactering the considered noise. Our approach allows to construct a new parametric comprehensive family of copulas [5], i. e., family containing all three basic copulas M , W and Π . Finally, some concluding remarks are added.

2. PERTURBATION OF COPULAS UNDER NOISE

Consider two random variables X and Y linked by a copula $C_{X,Y}$. Let Z be a random variable Z , independent of both X and Y . Evidently, the knowledge of $C_{X,Y}$ and of distribution function F_Z is not sufficient to determination of the copula $C_{X+Z,Y}$. This fact follows from the well known facts that $C_{X,Y} = C_{\varphi(X),\eta(Y)}$ for any transformations $\varphi, \eta : \mathbb{R} \rightarrow \mathbb{R}$ which are strictly increasing on $Ran X$ and $Ran Y$, respectively. Therefore, to describe the perturbed copula $C_{X+Z,Y}$, we need to know the original copula $C_{X,Y}$ and distribution functions F_X , F_Y and F_Z (possible modifications of values of F_Y which do not influence the form of copula $C_{X,Y}$ do not play a role in the next considerations). More, we will consider in the rest of the paper continuous random variables X, Y and Z . This is caused not only by the fact that then the copulas $C_{X,Y}$ and $C_{X+Z,Y}$ are unique, but also by the formula inverse to (1) valid in this case, namely that then

$$C_{X,Y}(u, v) = F_{X,Y}(F_X^{-1}(u), F_Y^{-1}(v)), \tag{3}$$

where $F_X^{-1}, F_Y^{-1} :]0, 1[\rightarrow \mathbb{R}$ are the related quantile functions for some bijective functions F_X and F_Y respectively (for more details see [2, 4, 5]).

Theorem 1. If $C_{X,Y}$ is the copula function of X and Y and the random variable Z is independent of X and Y , then the copula function $C_{X+Z,Y}$ is given by

$$C_{X+Z,Y}(u, v) = \int \int_0^v D_2 C_{X,Y}(F_X(F_{X+Z}^{-1}(u) - z), r) \text{drd}F_Z(z) \tag{4}$$

where $D_2 C_{X,Y}(u, v) = \frac{\partial F_{X,Y}(u,v)}{\partial v}$ if this derivative exists, and $D_2 C_{X,Y}(u, v) = 0$ otherwise. More, the bijective function F_{X+Z} is the standard convolution of F_X and F_Z .

Proof. Using the theorem of total probability we have

$$\begin{aligned} F_{X+Z,Y}(s, t) &= P(X + Z < s, Y < t) = \int P(X + Z < s, Y < t | Z = z) \text{d}F_Z(z) \\ &= \int P(X + z < s, Y < t | Z = z) \text{d}F_Z(z) = \int P(X < s - z, Y < t) \text{d}F_Z(z) \\ &= \int \int_{-\infty}^t P(X < s - z | Y = y) \text{d}F_Y(y) \text{d}F_Z(z) \\ &= \int \int_{-\infty}^t D_2 C_{X,Y}(F_X(s - z), F_Y(y)) \text{d}F_Y(y) \text{d}F_Z(z). \end{aligned}$$

Defining $r = F_Y(y)$, it becomes

$$F_{X+Z,Y}(s, t) = \int \int_0^{F_Y(t)} D_2 C_{X,Y}(F_X(s - z), r) \text{dr} \text{d}F_Z(z).$$

So, the perturbed copula $C_{X+Z,Y}$ is given by

$$C_{X+Z,Y}(u, v) = \int \int_0^v D_2 C_{X,Y}(F_X(F_{X+Z}^{-1}(u) - z), r) \text{drd}F_Z(z).$$

□

Example 1. Let X and Y be two random variables uniformly distributed on $[0, 1]$ linked with minimum copula, $C_{X,Y} = M$, and let the random variable Z independent of X and Y has uniform distribution on $[0, 1]$. Clearly, then $X = Y$, and

$$C_{X+Z,Y}(u, v) = \begin{cases} u & \text{if } 0 \leq u \leq \frac{1}{2}, \sqrt{2u} \leq v \\ v\sqrt{2u} - \frac{v^2}{2} & \text{if } 0 \leq u \leq \frac{1}{2}, \sqrt{2u} \geq v \\ v - \frac{(\sqrt{2(1-u)}+v-1)^2}{2} & \text{if } \frac{1}{2} \leq u \leq 1, 1 - \sqrt{2(1-u)} \leq v \\ v & \text{if } \frac{1}{2} \leq u \leq 1, 1 - \sqrt{2(1-u)} \geq v. \end{cases} \tag{5}$$

Proof. Let $F_{X+Z}(s)$ is the CDF of Irwin-Hall distribution for $n = 2$, Since the support of the joint density of X, Z is $[0, 1]^2$ and is finite so $F_{X+Z}(s)$ is a strictly increasing function which yields the bijectivity property of $F_{X+Z}(s)$. By inverting the $F_{X+Z}(s)$ we obtain

$$F_{X+Z}^{-1}(s) = \begin{cases} 0 & \text{if } s \leq 0 \\ \sqrt{2s} & \text{if } 0 \leq s \leq \frac{1}{2} \\ 2 - \sqrt{2(1-s)} & \text{if } \frac{1}{2} \leq s \leq 1 \\ 2 & \text{if } 1 \leq s. \end{cases}$$

Also, since $D_2C_{X,Y}(u, v) = \frac{\partial M(u,v)}{\partial v} = \mathbf{1}_{(u < v)}$, where $\mathbf{1}_{(u < v)}$ is an indicator function which equals 1 if $u < v$ and zero otherwise, we have

$$D_2C_{X,Y}(F_X(F_{X+Y}^{-1}(u) - z), r) = \mathbf{1}_{(r < F_{X+Y}^{-1}(u) - z)}$$

and based on (4) we have the following four parts:

1. if $0 \leq u \leq \frac{1}{2}, \sqrt{2u} \leq v$

$$C_{X+Z,X}(u, v) = \int_0^{\sqrt{2u}} \int_0^{\sqrt{2u}-z} dr dz = u,$$

2. if $0 \leq u \leq \frac{1}{2}, \sqrt{2u} \geq v$

$$C_{X+Z,X}(u, v) = \int_0^{\sqrt{2u}-v} \int_0^v dr dz + \int_{\sqrt{2u}-v}^{\sqrt{2u}} \int_0^{\sqrt{2u}-z} dr dz = v\sqrt{2u} - \frac{v^2}{2},$$

3. for the sake of notational simplicity, define $g(u) = 2 - \sqrt{2(1-u)}$, if $\frac{1}{2} \leq u \leq 1, g(u) - 1 \leq v$,

$$C_{X+Z,X}(u, v) = \int_0^{g(u)-v} \int_0^v dr dz + \int_{g(u)-v}^1 \int_0^{g(u)-z} dr dz = v - \frac{(\sqrt{2(1-u)}+v-1)^2}{2},$$

4. if $\frac{1}{2} \leq u \leq 1, g(u) - 1 \geq v$

$$C_{X+Z,X}(u, v) = \int_0^1 \int_0^v dr dz = v,$$

which completes the proof. □

Figure 1 depicts this copula function.

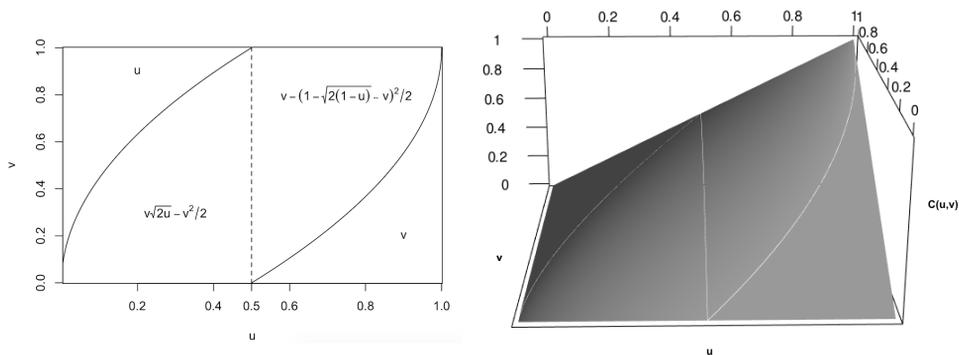


Fig. 1. 2D and 3D view of Copula function $C_{X+Z,X}$.

It should be noticed the L_∞ norm of the difference between $C_{X,X}$ and $C_{X+Z,X}$ can be obtained as follows.

Corollary 1. Under the assumption of Example 1,

$$L_\infty(C_{X,X}, C_{X+Z,X}) = \frac{1}{8}.$$

Proof. We have two cases: $0 \leq u \leq \frac{1}{2}$ and $\frac{1}{2} \leq u \leq 1$.
 If $0 \leq u \leq \frac{1}{2}$, using (5) and $C_{X,Y}(u, v) = M(u, v)$ we have

$$\begin{aligned} L_\infty(C_{X,X}, C_{X+Z,X}) &= \max |C_{X,X}(u, v) - C_{X+Z,X}(u, v)| \\ &= \begin{cases} \max |u - u| & \text{if } u \leq v \text{ \& } v \leq v\sqrt{2u} - \frac{v^2}{2} \\ \max \left| u - v\sqrt{2u} + \frac{v^2}{2} \right| & \text{if } u \leq v \text{ \& } v \geq v\sqrt{2u} - \frac{v^2}{2} \\ \max \left| v - v + \frac{(\sqrt{2(1-u)} + v - 1)^2}{2} \right| & \text{if } u \geq v \text{ \& } v \leq v\sqrt{2u} - \frac{v^2}{2} \\ \max |v - v| & \text{if } u \geq v \text{ \& } v \geq v\sqrt{2u} - \frac{v^2}{2} \end{cases} \\ &= \begin{cases} 0 & \text{if } u \leq v \text{ \& } v \leq v\sqrt{2u} - \frac{v^2}{2} \\ \max \left| u - \frac{1}{2}\sqrt{2u} + \frac{1}{8} \right| = \frac{1}{8} & \text{if } u \leq v \text{ \& } v \geq v\sqrt{2u} - \frac{v^2}{2} \\ \max \left| \frac{v^2}{2} \right| = \frac{1}{8} & \text{if } u \geq v \text{ \& } v \leq v\sqrt{2u} - \frac{v^2}{2} \\ 0 & \text{if } u \geq v \text{ \& } v \geq v\sqrt{2u} - \frac{v^2}{2} \end{cases} \end{aligned}$$

and similarly, if $\frac{1}{2} \leq u \leq 1$, we have four parts; so

$$\begin{aligned} L_\infty(C_{X,X}, C_{X+Z,X}) &= \max |C_{X,X}(u, v) - C_{X+Z,X}(u, v)| \\ &= \begin{cases} 0 & \text{if } u \leq v \text{ \& } v \geq v(2 - \sqrt{2u}) - \frac{v^2 - (1 - \sqrt{2(1-u)})^2}{2} \\ \max \left| u - \sqrt{2u} - \frac{1}{2} \right| = \frac{1}{8} & \text{if } u \leq v \text{ \& } v \leq v(2 - \sqrt{2u}) - \frac{v^2 - (1 - \sqrt{2(1-u)})^2}{2} \\ \max \left| v - (v - \frac{v^2}{2}) \right| = \frac{1}{8} & \text{if } u \geq v \text{ \& } v \leq v(2 - \sqrt{2u}) - \frac{v^2 - (1 - \sqrt{2(1-u)})^2}{2} \\ 0 & \text{if } u \geq v \text{ \& } v \geq v(2 - \sqrt{2u}) - \frac{v^2 - (1 - \sqrt{2(1-u)})^2}{2} \end{cases}. \end{aligned}$$

□

Example 2. Let X and Y be two random variables uniformly distributed on $[0, 1]$ with the Fréchet copula W , $C_{X,Y}(u, v) = \max(0, u + v - 1)$, (i.e., $Y = 1 - X$) and random variable Z , independent of X and Y , has uniform distribution on $[0, 1]$, then

$$C_{X+Z,Y}(u, v) = \begin{cases} 0 & \text{if } 0 \leq u \leq \frac{1}{2}, \sqrt{2u} \leq 1 - v \\ \frac{(\sqrt{2u}+v-1)^2}{2} & \text{if } 0 \leq u \leq \frac{1}{2}, \sqrt{2u} \geq 1 - v \\ u + v - 1 & \text{if } \frac{1}{2} \leq u \leq 1, \sqrt{2(1-u)} \leq v \\ v(1 - \sqrt{2(1-u)}) + \frac{v^2}{2} & \text{if } \frac{1}{2} \leq u \leq 1, \sqrt{2(1-u)} \geq v. \end{cases} \tag{6}$$

Proof. Similar to Example 1 we have the same $F_{X+Z}^{-1}(s)$ and since $D_2C_{X,Y}(u, v) = \frac{\partial W(u,v)}{\partial v} = \mathbf{1}_{(u>1-v)}$, then

1. if $0 \leq u \leq \frac{1}{2}, \sqrt{2u} \geq 1 - v$

$$C_{X+Z,1-X}(u, v) = \int_{1-v}^{\sqrt{2u}} \int_0^{\sqrt{2u}-z} drdz = \frac{(\sqrt{2u} + v - 1)^2}{2}$$

2. if $0 \leq u \leq \frac{1}{2}, \sqrt{2u} \leq 1 - v$

$$C_{X+Z,1-X}(u, v) = 0,$$

3. if $\frac{1}{2} \leq u \leq 1, v \geq \sqrt{2(1-u)}$, by defining $g(u) = 2 - \sqrt{2(1-u)}$,

$$C_{X+Z,1-X}(u, v) = v - \int_{g(u)-1}^1 \int_{g(u)-z}^1 drdz = v + u - 1,$$

4. if $\frac{1}{2} \leq u \leq 1, v \leq \sqrt{2(1-u)}$

$$\begin{aligned} C_{X+Z,1-X}(u, v) &= \int_0^{g(u)-1} \int_{1-v}^1 drdz + \int_{g(u)-1}^{g(u)-v+1} \int_{1-v}^{g(u)-z} drdz \\ &= v(1 - \sqrt{2(1-u)}) + \frac{v^2}{2}. \end{aligned}$$

Hence,

$$C_{X+Z,Y}(u, v) = \begin{cases} 0 & \text{if } 0 \leq u \leq \frac{1}{2}, \sqrt{2u} \leq 1 - v \\ \frac{(\sqrt{2u}+v-1)^2}{2} & \text{if } 0 \leq u \leq \frac{1}{2}, \sqrt{2u} \geq 1 - v \\ u + v - 1 & \text{if } \frac{1}{2} \leq u \leq 1, \sqrt{2(1-u)} \leq v \\ v(1 - \sqrt{2(1-u)}) + \frac{v^2}{2} & \text{if } \frac{1}{2} \leq u \leq 1, \sqrt{2(1-u)} \geq v. \end{cases}$$

□

Figure 2 presents this copula function and similar to Corollary 1, after a simple computation, we can obtain that

$$L_\infty(C_{X,1-X}, C_{X+Z,1-X}) = \max |C_{X,1-X}(u, v) - C_{X+Z,1-X}(u, v)| = \frac{1}{8}.$$

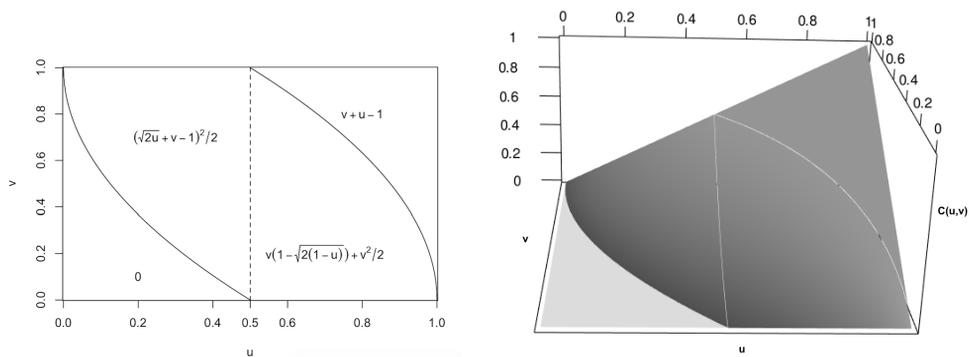


Fig. 2. 2D and 3D view of copula function $C_{X+Z,1-X}$.

The proof of the following example is trivial.

Example 3. If X, Y and Z are independent random variables with uniform distribution on $[0, 1]$ then

$$C_{X+Z,Y}(u, v) = uv.$$

Remark 1. (i) Note that Theorem 1 can be applied also in case when Z is not independent of X and Y . Obviously, then the distribution function F_{X+Z} should be computed by means of formula

$$F_{X+Y}(s) = \int \int_{x+y \leq s} dH_{X,Y}(x, y) \quad s \in R.$$

In particular, if $Z = Y$, then the copula $C_{X+Y,Y}$ is given by

$$C_{X+Y,Y}(u, v) = \int_0^v D_2 C_{X,Y}(F_X(F_{X+Y}^{-1}(u) - F_Y^{-1}(r)), r) dr.$$

Recall that this result was obtained as Theorem (3.4.2) already in book [2].

(ii) Observe that copulas $C_{X+Z,X}$ and $C_{X+Z,1-X}$ discussed in Example 1 and Example 2, respectively, are related by the flipping relations,

$$C_{X+Z,1-X}(u, v) = u - C_{X+Z,X}(u, 1 - v) = v - C_{X+Z,X}(1 - u, v).$$

Consequently, both these copulas are survival invariant, i. e.,

$$C_{X+Z,X}(u, v) = 1 - u - v + C_{X+Z,X}(1 - u, 1 - v)$$

and

$$C_{X+Z,1-X}(u, v) = 1 - u - v + C_{X+Z,1-X}(1 - u, 1 - v).$$

3. A PARAMETRIC COMPREHENSIVE FAMILY OF COPULAS

In this section, a new parametric comprehensive family of copulas is introduced. It is based on the subsequent Examples 4 and 5.

Example 4. Let X and Y be two random variables with uniform distribution on $[0, 1]$ with minimum copula, $C_{X,Y}(u, v) = \min(u, v)$, and random variable Z , independent of X and Y , has uniform distribution on $[0, \varepsilon]$, where $\varepsilon \in]0, \infty[$. Then

a) if $\varepsilon \leq 1$,

$$C_{X+Z,Y}(u, v) = \begin{cases} u & \text{if } 0 \leq u \leq \frac{\varepsilon}{2} \ \& \ \sqrt{2\varepsilon u} \leq v \\ & \text{or } \frac{\varepsilon}{2} \leq u \leq 1 - \frac{\varepsilon}{2} \ \& \ u + \frac{\varepsilon}{2} \leq v \leq 1 \\ \frac{1}{\varepsilon} [v\sqrt{2\varepsilon u} - \frac{v^2}{2}] & \text{if } 0 \leq u \leq \frac{\varepsilon}{2} \ \& \ v \leq \sqrt{2\varepsilon u} \\ u - \frac{1}{\varepsilon} \frac{(u + \frac{\varepsilon}{2} - v)^2}{2} & \text{if } \frac{\varepsilon}{2} \leq u \leq 1 - \frac{\varepsilon}{2} \ \& \ u - \frac{\varepsilon}{2} \leq v \leq u + \frac{\varepsilon}{2} \\ v - \frac{1}{\varepsilon} \frac{(v - g(u))^2}{2} & \text{if } 1 - \frac{\varepsilon}{2} \leq u \leq 1 \ \& \ g(u) \leq v \\ v & \text{if } \frac{\varepsilon}{2} \leq u \leq 1 - \frac{\varepsilon}{2} \ \& \ 0 \leq v \leq u - \frac{\varepsilon}{2} \\ & \text{or } 1 - \frac{\varepsilon}{2} \leq u \leq 1 \ \& \ v \leq g(u) \end{cases} \tag{7}$$

b) if $\varepsilon > 1$,

$$C_{X+Z,Y}(u, v) = \begin{cases} u & \text{if } 0 \leq u \leq \frac{1}{2\varepsilon} \ \& \ \sqrt{2\varepsilon u} \leq v \\ \frac{1}{\varepsilon} [v\sqrt{2\varepsilon u} - \frac{v^2}{2}] & \text{if } 0 \leq u \leq \frac{1}{2\varepsilon} \ \& \ v \leq \sqrt{2\varepsilon u} \\ \frac{1}{\varepsilon} [v(\varepsilon u + \frac{1}{2} - v) + \frac{v^2}{2}] & \text{if } \frac{1}{2\varepsilon} \leq u \leq 1 - \frac{1}{2\varepsilon} \ \& \ 0 \leq v \leq 1 \\ v - \frac{(v - g(u))^2}{2\varepsilon} & \text{if } 1 - \frac{1}{2\varepsilon} \leq u \leq 1 \ \& \ v \geq g(u) \\ v & \text{if } 1 - \frac{1}{2\varepsilon} \leq u \leq 1 \ \& \ v \leq g(u) \end{cases} \tag{8}$$

where $g(u) = 1 - \sqrt{2\varepsilon(1 - u)}$.

Proof.

a) Similar to Example 1 we have the bijective function

$$F_{X+Z}(s) = \begin{cases} \frac{s^2}{2\varepsilon} & \text{if } 0 \leq s \leq \varepsilon \\ s - \frac{\varepsilon}{2} & \text{if } \varepsilon \leq s \leq 1 \\ 1 - \frac{(1 + \varepsilon - s)^2}{2\varepsilon} & \text{if } 1 \leq s \leq \varepsilon + 1 \end{cases}$$

and equal zero for $s < 0$ and 1 for $s > \varepsilon + 1$. So, its inverse in the corresponding intervals is

$$F_{X+Z}^{-1}(s) = \begin{cases} \sqrt{2\varepsilon s} & \text{if } 0 \leq s \leq \frac{\varepsilon}{2} \\ s + \frac{\varepsilon}{2} & \text{if } \frac{\varepsilon}{2} \leq s \leq 1 - \frac{\varepsilon}{2} \\ \varepsilon + 1 - \sqrt{2\varepsilon(1 - s)} & \text{if } 1 - \frac{\varepsilon}{2} \leq s \leq 1 \end{cases}$$

and similar to the Example 1, $D_2C_{X,Y}(u, v) = \mathbf{1}_{(r < F_{X+Z}^{-1}(s) - z)}$. So, we have the following cases:

1. if $0 \leq u \leq \frac{\varepsilon}{2}, \sqrt{2\varepsilon u} \leq v$

$$C_{X+Z,X}(u, v) = \int_0^{\sqrt{2\varepsilon u}} \int_0^{\sqrt{2\varepsilon u}-z} \frac{1}{\varepsilon} drdz = u,$$
2. if $0 \leq u \leq \frac{\varepsilon}{2}, \sqrt{2\varepsilon u} \geq v$

$$C_{X+Z,X}(u, v) = \int_0^{\sqrt{2\varepsilon u}-v} \int_0^v \frac{1}{\varepsilon} drdz + \int_{\sqrt{2\varepsilon u}-v}^{\sqrt{2\varepsilon u}} \int_0^{\sqrt{2\varepsilon u}-z} \frac{1}{\varepsilon} drdz = \frac{1}{\varepsilon}(v\sqrt{2\varepsilon u} - \frac{v^2}{2}),$$
3. if $\frac{\varepsilon}{2} \leq u \leq 1 - \frac{\varepsilon}{2}$ and
 - i) if $0 \leq v \leq u - \frac{\varepsilon}{2}$

$$C_{X+Z,X}(u, v) = \int_0^{\varepsilon} \int_0^v \frac{1}{\varepsilon} drdz = v$$
 - ii) if $u - \frac{\varepsilon}{2} \leq v \leq u + \frac{\varepsilon}{2}$

$$C_{X+Z,X}(u, v) = \int_0^{u+\frac{\varepsilon}{2}-v} \int_0^v \frac{1}{\varepsilon} drdz + \int_{u+\frac{\varepsilon}{2}-v}^{\varepsilon} \int_0^{u+\frac{\varepsilon}{2}-z} \frac{1}{\varepsilon} drdz = u - \frac{1}{\varepsilon} \frac{(u+\frac{\varepsilon}{2}-v)^2}{2}$$
 - iii) if $u + \frac{\varepsilon}{2} \leq v \leq 1$

$$C_{X+Z,X}(u, v) = \int_0^{\varepsilon} \int_0^{u-\frac{\varepsilon}{2}} \frac{1}{\varepsilon} drdz + \int_0^{\varepsilon} \int_{u-\frac{\varepsilon}{2}}^{u+\frac{\varepsilon}{2}-z} \frac{1}{\varepsilon} drdz = u$$
4. if $1 - \frac{\varepsilon}{2} \leq u \leq 1, v \leq g(u),$

$$C_{X+Z,X}(u, v) = \int_0^{\varepsilon} \int_0^v \frac{1}{\varepsilon} drdz = v$$
5. if $1 - \frac{\varepsilon}{2} \leq u \leq 1, g(u) \leq v$

$$C_{X+Z,X}(u, v) = \int_0^{\varepsilon} \int_0^v \frac{1}{\varepsilon} drdz - \int_{g(u)+\varepsilon-v}^v \int_{g(u)+\varepsilon-z}^v \frac{1}{\varepsilon} dzdr = v - \frac{1}{\varepsilon} \frac{(v-g(u))^2}{2}.$$

which is (7). Figure 3 presents this copula function for $\varepsilon = \frac{1}{2}$.

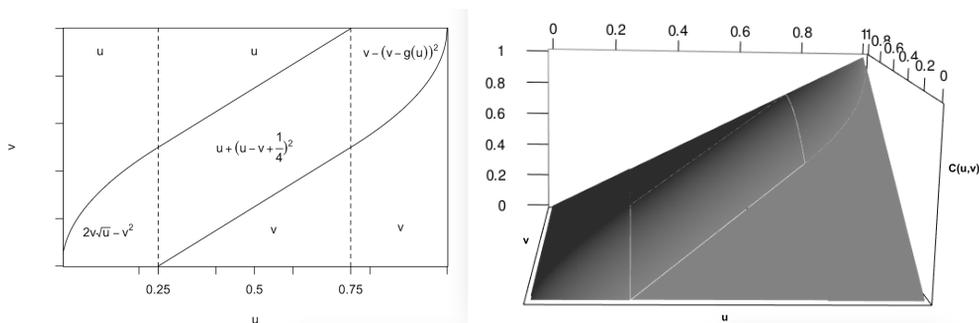


Fig. 3. 2D and 3D view of copula function $C_{X+Z,X}$ from Example 4 for $\varepsilon = \frac{1}{2}$.

b) Again, similar to to Example 1 we have the bijective function

$$F_{X+Z}(s) = \begin{cases} \frac{s^2}{2\varepsilon} & \text{if } 0 \leq s \leq 1 \\ \frac{2s-1}{2\varepsilon} & \text{if } 1 \leq s \leq \varepsilon \\ 1 - \frac{(1+\varepsilon-s)^2}{2\varepsilon} & \text{if } \varepsilon \leq s \leq \varepsilon + 1 \end{cases}$$

and equals zero for $s < 0$ and 1 for $s > \varepsilon + 1$. So, its inverse in the corresponding intervals is

$$F_{X+Z}^{-1}(s) = \begin{cases} \sqrt{2\varepsilon s} & \text{if } 0 \leq s \leq \frac{1}{2\varepsilon} \\ \varepsilon s + \frac{1}{2} & \text{if } \frac{1}{2\varepsilon} \leq s \leq 1 - \frac{1}{2\varepsilon} \\ \varepsilon + 1 - \sqrt{2\varepsilon(1-s)} & \text{if } 1 - \frac{1}{2\varepsilon} \leq s \leq 1 \end{cases}$$

and similar to the Example 1, $D_2C_{X,Y}(u, v) = \mathbf{1}_{(v < F_{X+Z}^{-1}(s) - z)}$. So, we have the following five cases:

1. if $0 \leq u \leq \frac{1}{2\varepsilon}$, $\sqrt{2\varepsilon u} \leq v$
 $C_{X+Z,X}(u, v) = \int_0^{\sqrt{2\varepsilon u}} \int_0^{\sqrt{2\varepsilon u} - z} \frac{1}{\varepsilon} \, drdz = u,$
2. if $0 \leq u \leq \frac{1}{2\varepsilon}$, $\sqrt{2\varepsilon u} \geq v$
 $C_{X+Z,X}(u, v) = \int_0^{\sqrt{2\varepsilon u} - v} \int_0^v \frac{1}{\varepsilon} \, drdz + \int_{\sqrt{2\varepsilon u} - v}^{\sqrt{2\varepsilon u}} \int_0^{\sqrt{2\varepsilon u} - z} \frac{1}{\varepsilon} \, drdz = \frac{1}{\varepsilon}(v\sqrt{2\varepsilon u} - \frac{v^2}{2}),$
3. if $\frac{1}{2\varepsilon} \leq u \leq 1 - \frac{1}{2\varepsilon}$,
 $C_{X+Z,X}(u, v) = \int_0^{\varepsilon u + \frac{1}{2} - v} \int_0^v \frac{1}{\varepsilon} \, drdz + \int_{\varepsilon u + \frac{1}{2} - v}^{\varepsilon} \int_0^{\varepsilon u + \frac{1}{2} - z} \frac{1}{\varepsilon} \, drdz = \frac{1}{\varepsilon}(v(\varepsilon u + \frac{1}{2} - v) + \frac{v^2}{2}),$
4. if $1 - \frac{1}{2\varepsilon} \leq u \leq 1$, $v \geq g(u)$,
 $C_{X+Z,X}(u, v) = \int_0^{\varepsilon + g(u) - v} \int_0^v \frac{1}{\varepsilon} \, drdz + \int_{\varepsilon + g(u) - v}^{\varepsilon} \int_0^{\varepsilon + g(u) - z} \frac{1}{\varepsilon} \, drdz = v - \frac{(v - g(u))^2}{2\varepsilon},$
5. if $1 - \frac{1}{2\varepsilon} \leq u \leq 1$, $v \leq g(u)$
 $C_{X+Z,X}(u, v) = \int_0^{\varepsilon} \int_0^v \frac{1}{\varepsilon} \, drdz = v,$

which is (8).

□

Figure 4 presents this copula function for $\varepsilon = 2$.

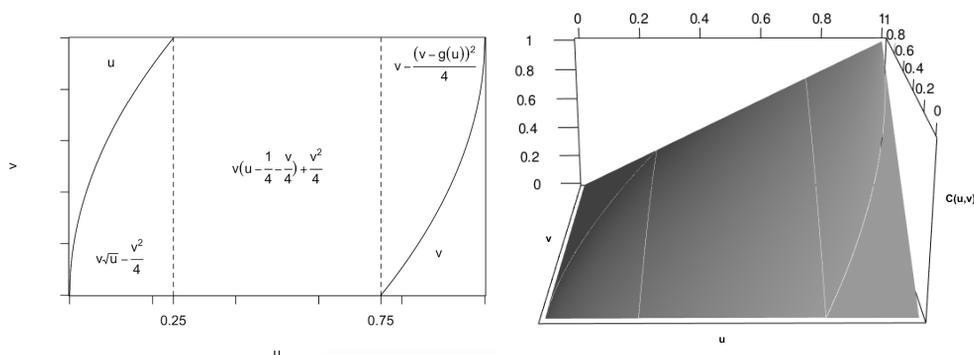


Fig. 4. 2D and 3D view of copula function $C_{X+Z,X}$ from Example 4 for $\varepsilon = 2$.

Using a similar computation, we will obtain the following results when $Y = 1 - X$.

Example 5. Let X and Y be two random variables uniformly distributed on $[0, 1]$ linked with copula W , $C_{X,Y}(u, v) = \max(0, u + v - 1)$ (i. e., $Y = 1 - X$), and let the random variable Z independent of X and Y has uniform distribution on $[0, \varepsilon]$, then

a) if $\varepsilon \leq 1$,

$$C_{X+Z,Y}(u, v) = \begin{cases} 0 & \text{if } 0 \leq u \leq \frac{\varepsilon}{2} \ \& \ \sqrt{2\varepsilon u} \leq 1 - v \\ & \text{or} \\ & \frac{\varepsilon}{2} \leq u \leq 1 - \frac{\varepsilon}{2} \ \& \ u + \frac{\varepsilon}{2} \leq 1 - v \\ \frac{1}{2\varepsilon}(\sqrt{2\varepsilon u} + v - 1)^2 & \text{if } 0 \leq u \leq \frac{\varepsilon}{2} \ \& \ \sqrt{2\varepsilon u} \geq 1 - v \\ \frac{1}{\varepsilon} \frac{(u + \frac{\varepsilon}{2} - v - 1)^2}{2} & \text{if } \frac{\varepsilon}{2} \leq u \leq 1 - \frac{\varepsilon}{2} \ \& \ u - \frac{\varepsilon}{2} \leq 1 - v \leq u + \frac{\varepsilon}{2} \\ \frac{1}{\varepsilon}[v(g(u) + \varepsilon - 1) + \frac{v^2}{2}] & \text{if } 1 - \frac{\varepsilon}{2} \leq u \leq 1 \ \& \ g(u) \leq 1 - v \\ u + v - 1 & \text{if } \frac{\varepsilon}{2} \leq u \leq 1 - \frac{\varepsilon}{2} \ \& \ 1 - v \leq u - \frac{\varepsilon}{2} \\ & \text{or} \\ & 1 - \frac{\varepsilon}{2} \leq u \leq 1 \ \& \ g(u) \geq 1 - v \end{cases} \tag{9}$$

b) if $\varepsilon > 1$,

$$C_{X+Z,Y}(u, v) = \begin{cases} 0 & \text{if } 0 \leq u \leq \frac{1}{2\varepsilon} \ \& \ \sqrt{2\varepsilon u} \leq 1 - v \\ \frac{1}{2\varepsilon}(\sqrt{2\varepsilon u} + v - 1)^2 & \text{if } 0 \leq u \leq \frac{1}{2\varepsilon} \ \& \ \sqrt{2\varepsilon u} \geq 1 - v \\ \frac{(\varepsilon u + v - \frac{1}{2})^2 - (\varepsilon u - \frac{1}{2})^2}{2\varepsilon} & \text{if } \frac{1}{2\varepsilon} \leq u \leq 1 - \frac{1}{2\varepsilon} \\ \frac{(h(u) + v)^2 - h^2(u)}{2\varepsilon} & \text{if } 1 - \frac{1}{2\varepsilon} \leq u \leq 1 \ \& \ h(u) \leq \varepsilon - v \\ u + v - 1 & \text{if } 1 - \frac{1}{2\varepsilon} \leq u \leq 1 \ \& \ h(u) \geq \varepsilon - v \end{cases} \tag{10}$$

where $g(u)$ is defined as in Example 4 and $h(u) = \varepsilon - \sqrt{2\varepsilon(1 - u)}$.

It should be noted that the copula functions (7) and (9) can be reduced to (5) as well as (8) and (10) reduce to (6) when $\varepsilon = 1$. The following corollaries describe the asymptotic behaviors of these copula functions.

Corollary 2. Under the assumptions of Example 4 we have the following results

a) $C_{X+Z,X}(u, v) \rightarrow M(u, v)$, as $\varepsilon \rightarrow 0$

b) $C_{X+Z,X}(u, v) \rightarrow \Pi(u, v)$, as $\varepsilon \rightarrow \infty$.

Proof.

a) When $\varepsilon \rightarrow 0$, the third and the fifth part of (7) respectively will be

$$\begin{aligned} \lim_{\varepsilon \rightarrow 0} C_{X+Z,X}(u, v) I_{(\frac{\varepsilon}{2}, 1 - \frac{\varepsilon}{2})}(u) I_{(0, u - \frac{\varepsilon}{2})}(v) &= v \lim_{\varepsilon \rightarrow 0} I_{(\frac{\varepsilon}{2}, 1 - \frac{\varepsilon}{2})}(u) I_{(0, u - \frac{\varepsilon}{2})}(v) \\ &= v I_{(0,1)}(u) I_{(0,u)}(v) \end{aligned}$$

and

$$\begin{aligned} \lim_{\varepsilon \rightarrow 0} C_{X+Z,X}(u, v) I_{(\frac{\varepsilon}{2}, 1-\frac{\varepsilon}{2})}(u) I_{(u+\frac{\varepsilon}{2}, 1)}(v) &= u \lim_{\varepsilon \rightarrow 0} I_{(\frac{\varepsilon}{2}, 1-\frac{\varepsilon}{2})}(u) I_{(u+\frac{\varepsilon}{2}, 1)}(v) \\ &= u I_{(0,1)}(u) I_{(u,1)}(v). \end{aligned}$$

Hence $C_{X+Z,X}(u, v) \rightarrow \min(u, v) = M(u, v)$ as $\varepsilon \rightarrow 0$.

b) When $\varepsilon \rightarrow \infty$, the third part of (8) will be

$$\begin{aligned} &\lim_{\varepsilon \rightarrow \infty} C_{X+Z,X}(u, v) I_{(\frac{1}{2\varepsilon}, 1-\frac{1}{2\varepsilon})}(u) I_{(0,1)}(v) \\ &= \lim_{\varepsilon \rightarrow \infty} \frac{1}{\varepsilon} [v(\varepsilon u + \frac{1}{2} - v) + \frac{v^2}{2}] I_{(\frac{1}{2\varepsilon}, 1-\frac{1}{2\varepsilon})}(u) I_{(0,1)}(v) uv I_{(0,1)}(u) I_{(0,1)}(v) \\ &= \Pi(u, v). \end{aligned}$$

□

Similar to Corollary 2 we readily can prove the following corollary.

Corollary 3. Under the assumptions of Example 5 we have the following results

- a) $C_{X+Z,1-X}(u, v) \rightarrow W(u, v)$, as $\varepsilon \rightarrow 0$,
- b) $C_{X+Z,1-X}(u, v) \rightarrow \Pi(u, v)$, as $\varepsilon \rightarrow \infty$.

For $k \in]0, \infty[$, let us denote by \mathcal{C}_k the copula $C_{X+Z,X}$ from Example 4 in the case when $k = \frac{1}{\varepsilon}$. Similarly for $k \in]-\infty, 0[$, we denote by \mathcal{C}_k the copula $C_{X+Z,1-X}$ from Example 5 in the case when $k = -\frac{1}{\varepsilon}$. Moreover, let $\mathcal{C}_0 = \Pi$, $\mathcal{C}_\infty = M$ and $\mathcal{C}_{-\infty} = W$.

Theorem 2. The parametric copula family $(\mathcal{C}_k)_{k \in [-\infty, \infty]}$ is a comprehensive family of copulas continuous and strictly increasing in parameter k .

The proof of the above theorem follows from Examples 4, 5 and Corollaries 2, 3. Evidently, for any dependence parameter like Kendal’s tau, Spearman’s rho, Gini’s gamma, Blomquist beta, see [2, 4, 5], the range of our comprehensive family $(\mathcal{C}_k)_{k \in [-\infty, \infty]}$ is full, i. e., it is $[-1, 1]$, and any of these dependence parameters is increasing in parameter k .

More, any copula \mathcal{C}_k is survival invariant,

$$\mathcal{C}_k(u, v) = 1 - u - v + \mathcal{C}_k(1 - u, 1 - v),$$

and \mathcal{C}_k and \mathcal{C}_{-k} are linked by the flipping relations,

$$\mathcal{C}_{-k}(u, v) = u - \mathcal{C}_k(u, 1 - v) = v - \mathcal{C}_k(1 - u, v).$$

4. CONCLUDING REMARKS

We have discussed the form of copula $C_{X+Z,Y}$ when the first coordinate X of a random vector (X, Y) is affected by a noise Z (independent of both X and Y). Obviously, similar results for the case when the second coordinate is affected, i. e., for $C_{X,Y+Z}$, can be obtained in a straight way. Repeating our approach twice, one can obtain the copula $C_{X+Z_1,Y+Z_2}$ dealing with affection of both coordinates. A deeper study of this case, including the dropping of the independence of Z (Z_1, Z_2) and X, Y , is a topic for the further study. More, in the next study the general case of n -dimensional random vectors affected by noise should be considered.

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