

## End-point linear functions

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## Abstract

Positive homogeneity is represented as a constraint  $\mathbf{0}$ -homogeneity and generalized into  $\mathbf{z}$ -homogeneity, called also  $\mathbf{z}$ -end point linearity. Several special  $\mathbf{z}$ -homogeneous aggregation functions are studied, in particular semicopulas, quasi-copulas, copulas, overlap functions, etc.

*Keywords:* Aggregation function, copula, end-point linear function.

## 1 Introduction

Linearity of functions defined on a segment is a most applied approach in modelling of dependences between variables in physics or engineering. Recall that a function  $f : D \rightarrow \mathbb{R}$  defined on a convex subset  $D$  of a vector space  $(\mathbb{X}, +, \cdot)$  is linear if

$$f(a\mathbf{x} + b\mathbf{y}) = af(\mathbf{x}) + bf(\mathbf{y}),$$

for any real constants  $a, b \in \mathbb{R}$  and  $\mathbf{x}, \mathbf{y} \in D$  such that  $a\mathbf{x} + b\mathbf{y} \in D$ . Equivalently,  $f$  is linear if and only if for any  $\lambda \in [0, 1]$  and  $\mathbf{x}, \mathbf{y} \in D$  it holds

$$f(\lambda\mathbf{x} + (1 - \lambda)\mathbf{y}) = \lambda f(\mathbf{x}) + (1 - \lambda)f(\mathbf{y}).$$

Note that in this case,  $\lambda\mathbf{x} + (1 - \lambda)\mathbf{y} \in D$  due to the convexity of  $D$ .

A weaker form of linearity is related to a fixed point  $\mathbf{z} \in D$ .

We tell that  $f$  is *end-point linear in  $\mathbf{z}$*  whenever for any  $\mathbf{x} \in D$  and  $\lambda \in [0, 1]$  it holds

$$f(\lambda\mathbf{z} + (1 - \lambda)\mathbf{x}) = \lambda f(\mathbf{z}) + (1 - \lambda)f(\mathbf{x}).$$

Clearly, then, for any  $\mathbf{x} \in D$ ,  $f$  is linear on the segment  $\langle \mathbf{z}, \mathbf{x} \rangle$ . Also,  $f$  is linear if and only if it is  $\mathbf{z}$ -end point linear for any  $\mathbf{z} \in D$ .

In particular, if  $\mathbf{0} \in D$  then  $f$  is  $\mathbf{0}$ -end point linear whenever

$$f(\lambda\mathbf{x}) = \lambda f(\mathbf{x}) + (1 - \lambda)f(\mathbf{0}),$$

for any  $\mathbf{x} \in D$  and any  $\lambda \in [0, 1]$ .

If  $\mathbf{0} \in D$ , then  $f : D \rightarrow \mathbb{R}$  is called positively homogeneous if

$$f(a\mathbf{x}) = af(\mathbf{x}) \text{ for any } a \in [0, \infty[ \text{ and any } \mathbf{x} \in D \text{ such that } a\mathbf{x} \in D.$$

Obviously, then  $f(\mathbf{0}) = 0$  and, equivalently, for any  $\mathbf{x} \in D$  and  $\lambda \in [0, 1]$ ,  $f(\lambda\mathbf{x}) = \lambda f(\mathbf{x})$ . Observe that the positive homogeneity of the function  $f$  means that  $f(\mathbf{0}) = 0$  and, for any  $\mathbf{x} \in D$ , the restriction of  $f$  to the segment  $\langle \mathbf{0}, \mathbf{x} \rangle$  is a linear function, i.e.,  $f$  is  $\mathbf{0}$ -end point linear. The aim of this paper is a generalization of the positive homogeneity of functions vanishing in  $\mathbf{0}$  seen as linearity on  $\langle \mathbf{0}, \mathbf{x} \rangle$  for any  $\mathbf{x} \in D$  by  $\mathbf{z}$ -homogeneity, where  $\mathbf{z}$  is a fixed point in  $D$ .

**Definition 1.1.** Let  $f : D \rightarrow \mathbb{R}$  and  $\mathbf{z} \in D$ . Then  $f$  is  $\mathbf{z}$ -homogeneous if, for any  $\mathbf{x} \in D$ ,  $f$  is linear on the segment  $\langle \mathbf{z}, \mathbf{x} \rangle$ .

Clearly, the standard positive homogeneity implies the  $\mathbf{0}$ -homogeneity, but not vice-versa. Moreover,  $\mathbf{z}$ -homogeneity can be seen as an end-point linearity, as the  $\mathbf{z}$ -homogeneous  $f$  is linear on any segment in  $D$  with an end point  $\mathbf{z}$ . Our main aim is the study of end-point linear functions acting on the domain  $[0, 1]^n \subseteq \mathbb{R}^n$  (i.e., we consider the vector space  $(\mathbb{R}^n, +, \cdot)$ ), with a special stress on the case  $[0, 1]^2$ . Considering some additional properties, we study and characterise some particular subclasses of end-point linear functions.

The paper is organized as follows. In the next section, we discuss and exemplify some particular  $\mathbf{z}$ -homogeneous functions  $f : [0, 1]^n \rightarrow [0, 1]$ . In Section 3, we focus on functions  $f : [0, 1]^2 \rightarrow [0, 1]$  and characterize end-point linear semicopulas (quasi-copulas, copulas, t-norms, t-conorms), overlap and grouping functions, quasi-arithmetic means, etc. Finally, some concluding remarks are added.

## 2 End-point linear fusion functions, general case

For  $n \geq 2$ , we will consider functions  $f : [0, 1]^n \rightarrow [0, 1]$ . Note that these functions assign to  $n$ -tuples of inputs from the unit interval  $[0, 1]$  an output value from the same interval  $[0, 1]$ . To stress this property, they are also called *fusion functions*, see, e.g., [3]. Some particular subclasses of fusion functions we will deal with are:

- *semi-aggregation functions*, i.e., fusion functions satisfying two boundary conditions  $f(\mathbf{0}) = 0$  and  $f(\mathbf{1}) = 1$ ;
- *aggregation functions*, i.e., semi-aggregation functions which are directionally increasing for any direction  $\vec{r} \in [0, 1]^n \setminus \{\mathbf{0}\}$ ;

Recall that  $f : [0, 1]^n \rightarrow [0, 1]$  is  $\vec{r}$ -directionally increasing whenever  $f(\mathbf{x} + c\vec{r}) \geq f(\mathbf{x})$  for any  $\mathbf{x} \in [0, 1]^n$  and  $c > 0$  such that  $\mathbf{x} + c\vec{r} \in [0, 1]^n$ . For more details we recommend [4, 8]. Note also that aggregation functions can be equivalently characterized by two boundary conditions and increasingness in each coordinate, i.e., by  $\mathbf{e}_i$ -directional increasingness for any unit vector  $\mathbf{e}_i = (0, \dots, \underbrace{1}_{i\text{-th coordinate}}, \dots, 0)$ ,  $i = 1, 2, \dots, n$ .

As already mentioned, positively homogeneous functions are a particular case of end-point linear functions, namely, they are  $\mathbf{0}$ -homogeneous and vanishing in  $\mathbf{0}$ . As an example of  $\mathbf{0}$ -homogeneous fusion function which is not positively homogeneous consider, for example  $f(\mathbf{x}) = \frac{2 - \min(\mathbf{x})}{3}$ . It is not difficult to check that from any fusion function  $g : [0, 1]^n \rightarrow [0, 1]$  one can construct a  $\mathbf{0}$ -homogeneous function.

**Proposition 2.1.** Let  $g : [0, 1]^n \rightarrow [0, 1]$  be a fusion function. Let  $f = g_{\mathbf{0}} : [0, 1]^n \rightarrow [0, 1]$  be given by

$$f(\mathbf{x}) = \begin{cases} 0 & \text{if } \mathbf{x} = \mathbf{0}, \\ \max(\mathbf{x}) \cdot g\left(\frac{\mathbf{x}}{\max(\mathbf{x})}\right) & \text{otherwise.} \end{cases} \quad (1)$$

Then  $f$  is a  $\mathbf{0}$ -homogeneous fusion function. If  $g$  is also semi-aggregation function, then also  $f$  is a semi-aggregation function.

The proof is obvious and therefore omitted.

Note that we can replace  $\max$  in formula (1) by any other  $\mathbf{0}$ -homogeneous fusion function  $h : [0, 1]^n \rightarrow [0, 1]$ , and then  $f : [0, 1]^n \rightarrow [0, 1]$  is given by

$$f(\mathbf{x}) = \begin{cases} 0 & \text{if } h(\mathbf{x}) = 0, \\ h(\mathbf{x}) \cdot g\left(\frac{\mathbf{x}}{h(\mathbf{x})}\right) & \text{otherwise.} \end{cases} \quad (2)$$

Then again  $f$  is a  $\mathbf{0}$ -homogeneous fusion function. To see the correctness of (2), observe that for any  $\mathbf{0}$ -homogeneous fusion function  $h$  it holds, for  $\mathbf{x} \neq \mathbf{0}$ ,  $h(\mathbf{x}) = \max(\mathbf{x}) \cdot h\left(\frac{\mathbf{x}}{\max(\mathbf{x})}\right) \leq \max(\mathbf{x})$  and thus  $\frac{\mathbf{x}}{h(\mathbf{x})} \in [0, 1]^n$  whenever  $h(\mathbf{x}) > 0$ .

For constructing the  $\mathbf{0}$ -homogeneous aggregation function, some stronger constraints are necessary. The next result can be found in [7], see also [10].

**Theorem 2.2.** Let  $g : [0, 1]^n \rightarrow [0, 1]$  be an aggregation function such that, for any  $\mathbf{x}, \mathbf{y} \in [0, 1]^n$ ,  $\mathbf{x} \leq \mathbf{y}$ , it holds

$$\frac{g(\mathbf{x})}{g(\mathbf{y})} \geq \min\left(\frac{x_1}{y_1}, \dots, \frac{x_n}{y_n}\right), \quad (3)$$

with the convention  $\frac{0}{0} = 1$ . Then the function  $f = g_{\mathbf{0}}$  given by (1) is a  $\mathbf{0}$ -homogeneous aggregation function.

Observe also that for any  $\mathbf{0}$ -homogenous aggregation function  $f : [0, 1]^n \rightarrow [0, 1]$ , necessarily  $\frac{f(\mathbf{x})}{f(\mathbf{y})} \geq \min\left(\frac{x_1}{y_1}, \dots, \frac{x_n}{y_n}\right)$  for any  $\mathbf{0} \leq \mathbf{x} \leq \mathbf{y} \leq \mathbf{1}$ .

Recall that for any fusion function (semi-aggregation function, aggregation function)  $f : [0, 1]^n \rightarrow [0, 1]$  its dual  $f^d : [0, 1]^n \rightarrow [0, 1]$  given by

$$f^d(\mathbf{x}) = 1 - f(\mathbf{1} - \mathbf{x}),$$

is a fusion function (semi-aggregation function, aggregation function).

**Proposition 2.3.** *Let a fusion function (semi-aggregation function, aggregation function)  $f : [0, 1]^n \rightarrow [0, 1]$  be  $\mathbf{z}$ -homogeneous for some  $\mathbf{z} \in [0, 1]^n$ . Then its dual  $f^d$  is  $(\mathbf{1} - \mathbf{z})$ -homogeneous.*

*Proof.* For any  $\mathbf{x} \in [0, 1]$  and  $\lambda \in [0, 1]$ ,  $\mathbf{y} = \lambda\mathbf{x} + (1 - \lambda)(\mathbf{1} - \mathbf{z}) \in \langle \mathbf{1} - \mathbf{z}, \mathbf{x} \rangle$  and thus

$$\mathbf{1} - \mathbf{y} = \mathbf{1} - (\lambda\mathbf{x} + (1 - \lambda)\mathbf{1} - (1 - \lambda)\mathbf{z}) = \lambda(\mathbf{1} - \mathbf{x}) + (1 - \lambda)\mathbf{z}.$$

Due to the  $\mathbf{z}$ -homogeneity of  $f$  it holds

$$\begin{aligned} f^d(\mathbf{y}) &= 1 - f(\mathbf{1} - \mathbf{y}) = 1 - f(\lambda(\mathbf{1} - \mathbf{x}) + (1 - \lambda)\mathbf{z}) \\ &= 1 - \lambda f(\mathbf{1} - \mathbf{x}) - (1 - \lambda)f(\mathbf{z}) \\ &= 1 - \lambda(1 - f^d(\mathbf{x})) - (1 - \lambda)(1 - f^d(\mathbf{1} - \mathbf{z})) \\ &= \lambda f^d(\mathbf{x}) + (1 - \lambda)f^d(\mathbf{1} - \mathbf{z}), \end{aligned}$$

proving the linearity of  $f^d$  on the segment  $\langle \mathbf{1} - \mathbf{z}, \mathbf{x} \rangle$ . Thus  $f^d$  is  $(\mathbf{1} - \mathbf{z})$ -homogeneous.  $\square$

Due to Proposition 2.3, one can consider  $\mathbf{1}$ -homogeneity of functions satisfying  $f(\mathbf{1}) = 1$  as a dual positive homogeneity, and this implies also the next results.

**Corollary 2.4.** *Let  $g : [0, 1]^n \rightarrow [0, 1]$  be a fusion function (semi-aggregation function). Then the function  $f = g_{\mathbf{1}} : [0, 1]^n \rightarrow [0, 1]$  given by*

$$f(\mathbf{x}) = \begin{cases} 1 & \text{if } \mathbf{x} = \mathbf{1}, \\ \min(\mathbf{x}) + (1 - \min(\mathbf{x})) \cdot g\left(\frac{\mathbf{x} - \min(\mathbf{x}) \cdot \mathbf{1}}{1 - \min(\mathbf{x})}\right) & \text{otherwise.} \end{cases} \quad (4)$$

*is a  $\mathbf{1}$ -homogeneous fusion function (semi-aggregation function). The function  $f$  is an aggregation function whenever  $g$  is an aggregation function satisfying, for all  $\mathbf{x}$  and  $\mathbf{y}$  with  $\mathbf{0} \leq \mathbf{y} \leq \mathbf{x} \leq \mathbf{1}$ ,*

$$\frac{1 - g(\mathbf{x})}{1 - g(\mathbf{y})} \geq \min\left(\frac{1 - x_1}{1 - y_1}, \dots, \frac{1 - x_n}{1 - y_n}\right), \quad (5)$$

*with the convention  $\frac{0}{0} = 1$ .*

Similarly as in Proposition 2.3, one can show the next result.

**Proposition 2.5.** *Let  $f : [0, 1]^n \rightarrow [0, 1]$  be  $\mathbf{z}$ -homogeneous fusion function (semi-aggregation function, aggregation function) and let  $\sigma : \{1, \dots, n\} \rightarrow \{1, \dots, n\}$  be a permutation. Then the function  $f_\sigma : [0, 1]^n \rightarrow [0, 1]$ ,*

$$f_\sigma(\mathbf{x}) = f(\mathbf{x}_\sigma) = f(x_{\sigma(1)}, \dots, x_{\sigma(n)}),$$

*is a  $\mathbf{z}_\sigma$ -homogenous fusion function (semi-aggregation function, aggregation function).*

Due to Proposition 2.5, each symmetric fusion function  $f$  which is  $\mathbf{z}$ -homogeneous is also  $\mathbf{z}_\sigma$ -homogeneous for any permutation  $\sigma$ . Moreover, due to the symmetry of  $f$ ,  $f(\mathbf{z}_\sigma) = f(\mathbf{z}) = c$  for any permutation  $\sigma$ . Therefore, for all permutations  $\sigma$  and  $\tau$  such that  $\mathbf{z}_\sigma \neq \mathbf{z}_\tau$ , the function  $f$  is constant (attaining the value  $c$ ) on the segment which is the intersection of the domain  $[0, 1]^n$  and the straight line determined by points  $\mathbf{z}_\sigma$  and  $\mathbf{z}_\tau$ .

Total end-point linearity can be seen as  $\mathbf{z}$ -homogeneity valid for any end-point  $\mathbf{z} \in [0, 1]^n$ .

**Theorem 2.6.** *A fusion function  $f : [0, 1]^n \rightarrow [0, 1]$  is total end-point linear if and only if it is linear, i.e.,  $f(\mathbf{x}) = a + \sum_{i=1}^n b_i x_i$  for some real constants  $a, b_1, \dots, b_n$  such that  $a + \sum_{b_i < 0} b_i \geq 0$  and  $a + \sum_{b_i > 0} b_i \leq 1$ . This  $f$  is a semi-aggregation function if and only if  $a = 0$  and  $\sum_{i=1}^n b_i = 1$  and then  $f$  is also an aggregation function.*

*Proof.* The sufficiency of both claims is obvious.

To see the necessity, denote  $a = f(\mathbf{0})$ ,  $b = f(1, 0, \dots, 0)$ ,  $c = f(0, 1, 0, \dots, 0)$ , and  $d = f(1, 1, 0, \dots, 0)$ . Observe that for any  $x_1 \in [0, 1]$ ,  $(x_1, 0, \dots, 0) = (1 - x_1)\mathbf{0} + x_1(1, 0, \dots, 0)$ .

- Due to  $\mathbf{0}$ -homogeneity of  $f$ , it holds

$$f(x_1, 0, \dots, 0) = (1 - x_1)f(\mathbf{0}) + x_1f(1, 0, \dots, 0) = (1 - x_1)a + x_1b = a + (b - a)x_1.$$

Similarly  $f(0, x_2, 0, \dots, 0) = a + (c - a)x_2$ . Next

- $(1, 0, \dots, 0)$ -homogeneity ensures  $f(1, x_2, 0, \dots, 0) = b + (d - b)x_2$ , and
- $(0, 1, 0, \dots, 0)$ -homogeneity forces  $f(x_1, 1, 0, \dots, 0) = c + (d - c)x_1$ .
- Also,  $(0, x_2, 0, \dots, 0)$ -homogeneity and the equality

$$\begin{aligned} (x_1, x_2, 0, \dots, 0) &= x_1(1, x_2, 0, \dots, 0) + (1 - x_1)(0, x_2, 0, \dots, 0) \text{ implies,} \\ f(x_1, x_2, 0, \dots, 0) &= a + (b - a)x_1 + (c - a)x_2 + (d - c - b + a)x_1x_2. \end{aligned}$$

- Due to

$$\left(\frac{1}{2}, \frac{1}{2}, 0, \dots, 0\right) = \frac{1}{2}(\mathbf{0} + (1, 1, 0, \dots, 0)) = \frac{1}{2}((1, 0, \dots, 0) + (0, 1, 0, \dots, 0)),$$

it holds  $f\left(\frac{1}{2}, \frac{1}{2}, 0, \dots, 0\right) = \frac{a+d}{2} = \frac{b+c}{2}$ , and, finally

$$f(x_1, x_2, 0, \dots, 0) = a + (b - a)x_1 + (c - a)x_2.$$

Denoting  $b - a = b_1$  and  $c - a = b_2$ , we see that  $f(x_1, x_2, 0, \dots, 0) = a + b_1x_1 + b_2x_2$ .

In similar way, one can prove that  $f(x_1, 0, x_3, 0, \dots, 0) = a + b_1x_1 + b_3x_3$ . Now, denote

$$\begin{aligned} \alpha &= f(x_1, 0, \dots, 0) = a + b_1x_1 \\ \beta &= f(x_1, 1, 0, \dots, 0) = a + b_1x_1 + b_2 \\ \gamma &= f(x_1, 0, 1, 0, \dots, 0) = a + b_1x_1 + b_3 \quad \text{and} \\ \delta &= f(x_1, 1, 1, 0, \dots, 0) \end{aligned}$$

Similarly as in the case of  $f(x_1, x_2, 0, \dots, 0)$ , we can show that

$$f(x_1, x_2, x_3, 0, \dots, 0) = \alpha + (\beta - \alpha)x_2 + (\gamma - \alpha)x_3 + (\delta - \gamma - \beta + \alpha)x_2x_3,$$

and  $f\left(x_1, \frac{1}{2}, \frac{1}{2}, 0, \dots, 0\right) = \frac{\alpha+\delta}{2} = \frac{\beta+\gamma}{2}$ . Then

$$f(x_1, x_2, x_3, 0, \dots, 0) = a + b_1x_1 + b_2x_2 + b_3x_3.$$

By induction we get  $f(x_1, \dots, x_n) = a + \sum_{i=1}^n b_i x_i$ , i.e.,  $f$  is a linear function. Its extremal values on  $[0, 1]^n$  are  $a + \sum_{b_i < 0} b_i$  and  $a + \sum_{b_i > 0} b_i$ , and thus  $f$  is a fusion function only if  $a + \sum_{b_i < 0} b_i \geq 0$  and  $a + \sum_{b_i > 0} b_i \leq 1$ .

Clearly, if  $f$  is semi-aggregation function then  $f(\mathbf{0}) = a = 0$  and  $f(\mathbf{1}) = a + \sum_{i=1}^n b_i = \sum_{i=1}^n b_i = 1$  and  $a + \sum_{b_i < 0} b_i = \sum_{b_i < 0} b_i \geq 0$  ensures there is no negative  $b_i$ , i.e., each  $b_i \geq 0$ . Obviously, then  $f$  is an aggregation function.  $\square$

Based on Theorem 2.6 we see that the only total end-point linear (semi-) aggregation functions are just weighted arithmetic means.

**Example 2.7.** The functions max and min are  $\mathbf{c}$ -homogeneous aggregation functions for any  $\mathbf{c} = (c, \dots, c)$ ,  $c \in [0, 1]$ , but not  $\mathbf{z}$ -homogeneous whenever  $\mathbf{z}$  is non-constant. Consider, e.g.,  $n = 2$ ,  $\mathbf{z} = (1, 0)$  and suppose max (min) is  $\mathbf{z}$ -homogeneous. Then, knowing that  $\max(1, 0) = \max(0, 1) = 1$  ( $\min(1, 0) = \min(0, 1) = 0$ ), from the  $\mathbf{z}$ -homogeneity it follows  $\max(x, 1 - x) = 1$  ( $\min(x, 1 - x) = 0$ ) for each  $x \in [0, 1]$ , which is a contradiction.

Observe that  $\mathbf{c}$ -homogeneity,  $c \in [0, 1]$ , holds for any Choquet integral [5, 6, 9].

**Remark 2.8.** Observe that if  $\mathbf{z} \in ]0, 1[^n$ , then any  $\mathbf{z}$ -homogeneous fusion function  $f$  is determined by the value  $f(\mathbf{z})$  and values of  $f$  on the boundary points, i.e., on  $[0, 1]^n \setminus ]0, 1[^n$ . If  $\mathbf{z} \notin ]0, 1[^n$ , then even a proper subset of  $[0, 1]^n \setminus ]0, 1[^n$  is enough to be considered. For example, consider  $\mathbf{z} = (\frac{1}{2}, \dots, \frac{1}{2})$  and  $f(\mathbf{z}) = \alpha$ ,  $f(\mathbf{u}) = \beta$  for any  $\mathbf{u} \in [0, 1]^n \setminus ]0, 1[^n$ ,  $\alpha, \beta \in [0, 1]$ . Then the related  $\mathbf{z}$ -homogeneous fusion function  $f : [0, 1]^n \rightarrow [0, 1]$  is given by

$$f(\mathbf{x}) = \alpha + (\beta - \alpha) \max(|2x_1 - 1|, \dots, |2x_n - 1|).$$

For  $n = 2$ , we continue in Remark 2.8 and show the link between the values of a  $\mathbf{z}$ -homogeneous fusion functions and the values of  $f$  on the boundary  $[0, 1]^2 \setminus ]0, 1[^2$  and in  $\mathbf{z}$ . For  $\mathbf{z} \in ]0, 1[^2$ ,  $\mathbf{z}$ -homogeneous binary functions  $f$  are fully determined by  $f(\mathbf{z})$  and its four boundaries that is,  $f(0, \cdot)$ ,  $f(\cdot, 0)$ ,  $f(1, \cdot)$  and  $f(\cdot, 1)$ .

**Proposition 2.9.** Let  $\mathbf{z} = (z_1, z_2) \in ]0, 1[^2$  and  $f : [0, 1]^2 \rightarrow [0, 1]$  be  $\mathbf{z}$ -homogeneous. Then, if  $(x_1, x_2) \neq \mathbf{z}$ ,  $f$  is such that, it holds

(i) for  $(x_1, x_2)$  from the triangle  $\langle \mathbf{0}, \mathbf{z}, (1, 0) \rangle$ ,

$$f(x_1, x_2) = \frac{x_2}{z_2} f(\mathbf{z}) + \frac{z_2 - x_2}{z_2} f\left(\frac{x_1 z_2 - x_2 z_1}{z_2 - x_2}, 0\right);$$

(ii) for  $(x_1, x_2)$  from the triangle  $\langle \mathbf{0}, \mathbf{z}, (0, 1) \rangle$ ,

$$f(x_1, x_2) = \frac{x_1}{z_1} f(\mathbf{z}) + \frac{z_1 - x_1}{z_1} f\left(0, \frac{x_2 z_1 - x_1 z_2}{z_1 - x_1}\right);$$

(iii) for  $(x_1, x_2)$  from the triangle  $\langle (0, 1), \mathbf{z}, \mathbf{1} \rangle$ ,

$$f(x_1, x_2) = \frac{1 - x_2}{1 - z_2} f(\mathbf{z}) + \frac{x_2 - z_2}{1 - z_2} f\left(\frac{x_1 - z_1 - x_1 z_2 + x_2 z_1}{x_2 - z_2}, 1\right);$$

(iv) for  $(x_1, x_2)$  from the triangle  $\langle (1, 0), \mathbf{z}, \mathbf{1} \rangle$ ,

$$f(x_1, x_2) = \frac{1 - x_1}{1 - z_1} f(\mathbf{z}) + \frac{x_1 - z_1}{1 - z_1} f\left(1, \frac{x_2 - z_2 - x_2 z_1 + x_1 z_2}{x_1 - z_1}\right).$$

The proof is a matter of simple linear interpolation and therefore omitted.

### 3 2-dimensional end-point linear aggregation functions

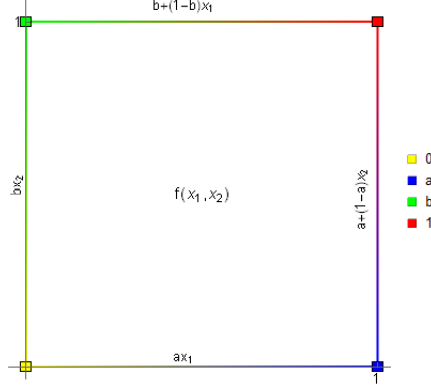
Now, we focus on binary semi-aggregation functions with linear boundaries, i.e., functions  $f : [0, 1]^2 \rightarrow [0, 1]$  which are linear on segments  $\langle \mathbf{0}, (0, 1) \rangle$ ,  $\langle \mathbf{0}, (1, 0) \rangle$ ,  $\langle \mathbf{1}, (0, 1) \rangle$  and  $\langle \mathbf{1}, (1, 0) \rangle$ . Clearly, then  $f(\mathbf{0}) = 0$  and  $f(\mathbf{1}) = 1$ , and  $f(1, 0) = a$ ,  $f(0, 1) = b$  for some constants  $a, b \in [0, 1]$ . Due to the linearity on boundaries,  $f(x_1, 0) = ax_1$ ,  $f(0, x_2) = bx_2$ ,  $f(x_1, 1) = b + (1 - b)x_1$  and  $f(1, x_2) = a + (1 - a)x_2$ , see Figure 1. For the sake of brevity,  $f$  will be then called  $\langle a, b \rangle$ -boundary linear function.

As a particular  $\langle a, b \rangle$ -boundary linear function we recall the Choquet integral [5]. Note that the Choquet integral  $\mathcal{C}_m$  with respect to a capacity  $m$  such that  $m(\{1\}) = a$  and  $m(\{2\}) = b$ , is a function  $\mathcal{C}_m : [0, 1]^2 \rightarrow [0, 1]$  given by

$$\mathcal{C}_m(x_1, x_2) = \begin{cases} ax_1 + (1 - a)x_2 & \text{if } x_1 \geq x_2, \\ (1 - b)x_1 + bx_2 & \text{otherwise.} \end{cases}$$

The next result relates the Choquet integrals and  $\mathbf{0}$ -homogeneous ( $\mathbf{1}$ -homogeneous)  $\langle a, b \rangle$ -boundary linear functions.

**Proposition 3.1.** The semi-aggregation function  $f : [0, 1]^2 \rightarrow [0, 1]$  is  $\langle a, b \rangle$ -boundary linear for some  $a, b \in [0, 1]$  and  $\mathbf{0}$ -homogeneous (or  $\mathbf{1}$ -homogeneous) if and only if  $f = \mathcal{C}_m$  is the Choquet integral with respect to a capacity  $m$  such that  $m(\{1\}) = a$  and  $m(\{2\}) = b$ .

Figure 1: The  $\langle a, b \rangle$ -boundary linear function

*Proof.* Obviously,  $\mathcal{C}_m$  is linear on triangles determined by points  $(0, 0), (1, 1), (1, 0)$ , and  $(0, 0), (1, 1), (0, 1)$ , which ensures the  $\langle a, b \rangle$ -boundary linearity,  $\mathbf{0}$ -homogeneity and  $\mathbf{1}$ -homogeneity of  $f = \mathcal{C}_m$ .

On the other hand, suppose that  $f$  is  $\langle a, b \rangle$ -boundary linear function which is  $\mathbf{0}$ -homogeneous (i.e., positively homogeneous).

Consider  $(x_1, x_2) \in [0, 1]^2$  such that  $x_1 \geq x_2 > 0$ .

Then  $(x_1, x_2) \in \langle (0, 0), (1, \frac{x_2}{x_1}) \rangle$ ,  $(x_1, x_2) = (1 - x_1)(0, 0) + x_1(1, \frac{x_2}{x_1})$ , and thus

$$f(x_1, x_2) = x_1 f\left(1, \frac{x_2}{x_1}\right) = x_1 \left(a + (1 - a) \frac{x_2}{x_1}\right) = ax_1 + (1 - a)x_2 = \mathcal{C}_m(x_1, x_2).$$

Similarly, if  $0 < x_1 \leq x_2$  then

$$f(x_1, x_2) = (1 - b)x_1 + bx_2 = \mathcal{C}_m(x_1, x_2).$$

Summarizing, we have  $f = \mathcal{C}_m$ . □

Proposition 3.1 can be modified replacing the  $\langle a, b \rangle$ -boundary linearity and  $\mathbf{0}$ -homogeneity by  $\mathbf{z}$ -homogeneity for any  $\mathbf{z} = (z, z)$ ,  $z \in [0, 1]$ .

When considering  $\mathbf{z}$ -homogeneous  $\langle a, b \rangle$ -boundary linear functions on  $[0, 1]^2$ , then each such  $f$  is linear on (possibly degenerated) triangles determined by point  $\mathbf{z}$  and two neighbouring vertices of  $[0, 1]^2$  square. In general, denote  $f(\mathbf{z}) = c$  (clearly,  $f(0, 1) = b$ , similarly  $f(1, 0) = a$ .) Then  $f$  is univocally determined by  $\mathbf{z}$  and parameters  $a, b, c$  but it need not be, in general, increasing and thus not aggregation function.

**Theorem 3.2.** *Let  $f : [0, 1]^2 \rightarrow [0, 1]$  be an  $\langle a, b \rangle$ -boundary linear function which is  $\mathbf{z}$ -homogeneous and  $f(\mathbf{z}) = c$ . Then  $f$  is a semi-aggregation function such that*

(i) for  $(x_1, x_2)$  from the triangle  $\langle \mathbf{0}, \mathbf{z}, (1, 0) \rangle$ ,

$$f(x_1, x_2) = ax_1 + \frac{x_2(c - az_1)}{z_2},$$

(if  $z_2 = 0$ , the degenerated triangle  $\langle \mathbf{0}, \mathbf{z}, (1, 0) \rangle$  coincides with the segment  $\langle \mathbf{0}, (1, 0) \rangle$  and then  $f(x_1, 0) = ax_1$ );

(ii) for  $(x_1, x_2)$  from the triangle  $\langle \mathbf{0}, \mathbf{z}, (0, 1) \rangle$ ,

$$f(x_1, x_2) = \frac{x_1(c - bz_2)}{z_1} + bx_2,$$

(if  $z_1 = 0$ , the degenerated triangle  $\langle \mathbf{0}, \mathbf{z}, (0, 1) \rangle \equiv \langle \mathbf{0}, (0, 1) \rangle$  and then  $f(0, x_2) = bx_2$ );

(iii) for  $(x_1, x_2)$  from the triangle  $\langle (0, 1), \mathbf{z}, \mathbf{1} \rangle$ ,

$$f(x_1, x_2) = (1 - b)x_1 + \frac{(x_2 - 1)(b + (1 - b)z_1 - c)}{1 - z_2} + b,$$

(if  $z_2 = 1$ , the degenerated triangle  $\langle (0, 1), \mathbf{z}, \mathbf{1} \rangle \equiv \langle (0, 1), \mathbf{1} \rangle$  and then  $f(x_1, 1) = b + (1 - b)x_1$ );

(iv) for  $(x_1, x_2)$  from the triangle  $\langle(1, 0), \mathbf{z}, \mathbf{1}\rangle$ ,

$$f(x_1, x_2) = \frac{(x_1 - 1)(a + (1 - a)z_2 - c)}{1 - z_1} + (1 - a)x_2 + a,$$

(if  $z_1 = 1$ , the degenerated triangle  $\langle(1, 0), \mathbf{z}, \mathbf{1}\rangle \equiv \langle(1, 0), \mathbf{1}\rangle$  and then  $f(1, x_2) = a + (1 - a)x_2$ ).

See, Figure 2.

*Proof.* The function  $f$  is linear on each of 4 discussed triangles and thus to determine  $f(x_1, x_2)$  it is enough to find the triangle containing  $(x_1, x_2)$ . Then  $f(x_1, x_2)$  is a convex combination of values of  $f$  on the related triangle vertices corresponding to the convex combination of related vertices resulting into  $(x_1, x_2)$ . As all 4 vertices of the square  $[0, 1]^2$  have value from the set  $\{0, 1, a, b\} \subset [0, 1]$ ,  $f$  is a fusion function. More, due to  $f(\mathbf{0}) = 0$  and  $f(\mathbf{1}) = 1$  we see that  $f$  is a semi-aggregation function.  $\square$

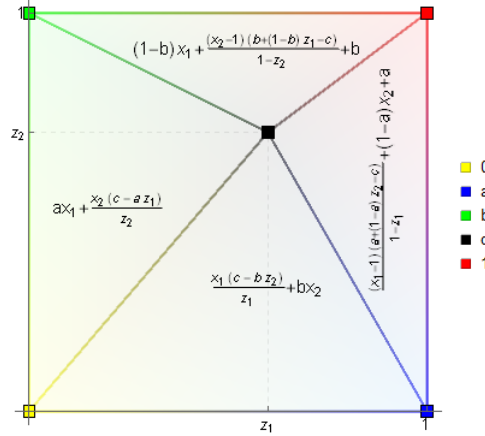


Figure 2: The semi-aggregation function

Based on Theorem 3.2, it is not difficult to see when the discussed  $f$  is an aggregation function.

**Corollary 3.3.** *Under the constraints of Theorem 3.2,  $f$  is an aggregation function if and only if*

$$c \geq az_1 \text{ and } c \geq bz_2 \text{ and } c \leq a + z_2 - az_2 \text{ and } c \leq b + z_1 - bz_1.$$

Considering some particular subclasses of aggregation functions, note that the only boundary linear (weighted) quasi-arithmetic means are weighted arithmetic means  $W_{(w,1-w)}$  given by

$$W_{(w,1-w)}(x_1, x_2) = wx_1 + (1 - w)x_2, \quad w \in [0, 1],$$

which are  $\langle w, 1 - w \rangle$ -boundary linear and  $\mathbf{z}$ -homogeneous for an arbitrary  $\mathbf{z} \in [0, 1]^2$  and  $c = f(\mathbf{z}) = wz_1 + (1 - w)z_2$ .

Next we focus on semicopulas and their subclasses quasi-copulas, copulas and triangular norms. Recall that an aggregation function  $f : [0, 1]^2 \rightarrow [0, 1]$  is a *semicopula* whenever 1 is its neutral element,  $f(x, 1) = f(1, x)$  for all  $x \in [0, 1]$ . Obviously, semicopulas are just  $\langle 0, 0 \rangle$ -boundary linear aggregation functions. Next, a semicopula  $f : [0, 1]^2 \rightarrow [0, 1]$  is a *quasi-copula* if it is 1-Lipschitz, i.e.,

$$|f(x_1, x_2) - f(y_1, y_2)| \leq |x_1 - y_1| + |x_2 - y_2| \quad \text{for all } (x_1, x_2), (y_1, y_2) \in [0, 1]^2.$$

A quasi-copula  $f : [0, 1]^2 \rightarrow [0, 1]$  is a *copula* if it is supermodular, i.e.

$$f((x_1, x_2) \vee (y_1, y_2)) + f((x_1, x_2) \wedge (y_1, y_2)) \geq f(x_1, x_2) + f(y_1, y_2),$$

for all  $(x_1, x_2), (y_1, y_2) \in [0, 1]^2$ . Finally, a semicopula  $f$  is a *triangular norm* if it is symmetric (commutative) and associative. For more details we recommend [7].

**Proposition 3.4.** A  $\langle 0, 0 \rangle$ -boundary linear  $\mathbf{z}$ -homogeneous function  $f : [0, 1]^2 \rightarrow [0, 1]$ ,  $f(\mathbf{z}) = c$ , is a semicopula if and only if  $c \leq \min(z_1, z_2)$ .

*Proof.* For any semicopula  $f$  and  $(x_1, x_2) \in [0, 1]^2$ , it holds  $f(x_1, x_2) \leq \min(x_1, x_2)$ , and thus  $f(\mathbf{z}) = c \leq \min(z_1, z_2)$ , showing the necessity. To see the sufficiency, note that due to Corollary 3.3, the above considered  $f$  is an aggregation function if and only if  $c \leq \min(z_1, z_2)$ . Also, due to  $a = 0$  we have  $f(1, y) = y$  for all  $y \in [0, 1]$ . Similarly, due to  $b = 0$ , it holds  $f(x, 1) = x$  for all  $x \in [0, 1]$ , and thus 1 is a neutral element of  $f$ . Hence,  $f$  is a semicopula.  $\square$

**Example 3.5.** Consider  $\langle 0, 0 \rangle$ -boundary linear  $(k, k)$ -homogeneous semicopula  $f_{k,c}$ , where  $c \in [0, k]$ . Then, for the greatest  $c = k$ ,

$$f_{k,k}(x_1, x_2) = \min(x_1, x_2),$$

i.e.,  $f_{k,k}$  is also a quasi-copula, a copula and a  $t$ -norm. On the other hand, for the smallest  $c = 0$ , it holds

$$f_{k,0}(x_1, x_2) = \begin{cases} 0 & \text{if } x_2 \leq \min\left(1 + \frac{k-1}{k}x_1, \frac{k}{k-1}x_1 - \frac{k}{k-1}\right), \\ \min(x_1, x_2) + \frac{k}{1-k} \max(x_1, x_2) - \frac{k}{1-k} & \text{otherwise.} \end{cases}$$

Then  $f_{k,0}$  is a semicopula for each  $k \in [0, 1[$ , it is quasicopula and copula for each  $k \in [0, \frac{1}{2}]$  and it is a  $t$ -norm only for  $k \in \{0, \frac{1}{2}\}$ . Note that  $f_{0,0} = \min$  and  $f_{\frac{1}{2},0} = T_L$  is the Łukasiewicz  $t$ -norm given by

$$T_L(x_1, x_2) = \max(x_1 + x_2 - 1, 0).$$

**Theorem 3.6.** For a  $\langle 0, 0 \rangle$ -boundary linear  $\mathbf{z}$ -homogeneous function  $f : [0, 1]^2 \rightarrow [0, 1]$ ,  $f(\mathbf{z}) = c$ , the following are equivalent:

- (i)  $f$  is a quasi-copula;
- (ii)  $f$  is a copula;
- (iii)  $\max(z_1 + z_2 - 1, 0) \leq c \leq \min(z_1, z_2)$ .

*Proof.* Recall that  $f$  is a semicopula if and only if  $c \leq \min(z_1, z_2)$ , see Proposition 3.4. Then  $f$  is a quasi-copula only if it is 1-Lipschitz on each of 4 triangles determined by  $\mathbf{z}$  considered in Theorem 3.2. As on each of these triangles  $f$  is linear (and increasing in both coordinates), it is 1-Lipschitz if and only if the coefficients by  $z_1$  and by  $z_2$  are bounded from above by 1. Hence,  $f$  is a quasi-copula if and only if

$$c \leq \min(z_1, z_2) \quad \text{and} \quad \frac{z_2 - c}{1 - z_1} \leq 1, \text{ (i.e., } c \geq z_1 + z_2 - 1) \quad \text{and} \quad \frac{z_1 - c}{1 - z_2} \leq 1, \text{ (i.e., } c \geq z_1 + z_2 - 1).$$

Summarizing, we see that (i) and (iii) are equivalent. Next, each copula is also a quasi-copula, hence (ii) implies (i) and (iii).

On the other hand suppose that (iii) holds. Recall that the supermodularity of a function  $f : [0, 1]^2 \rightarrow [0, 1]$  trivially holds if  $(x_1, x_2)$  and  $(y_1, y_2)$  are comparable, and hence we need to discuss the case when  $(x_1, x_2)$  and  $(y_1, y_2)$  are incomparable only. Suppose  $x_1 < y_1$  and  $x_2 > y_2$ . Then  $f$  is supermodular only if

$$f(y_1, x_2) + f(x_1, y_2) \geq f(x_1, x_2) + f(y_1, y_2),$$

i.e., if  $V_f(R) \geq 0$ , where  $R$  is the rectangle  $[x_1, y_2] \times [y_1, x_2]$  and  $V_f(R)$  is its volume given by

$$V_f(R) = f(y_1, x_2) + f(x_1, y_2) - f(x_1, x_2) - f(y_1, y_2).$$

Note that if  $f$  is linear on some domain  $D$ , then  $V_f(R) = 0$  for any  $R \subseteq D$ . Also, if  $R = R_1 \cup R_2$  for some non-overlapping rectangles  $R_1$  and  $R_2$  (i.e.,  $R_1 \cap R_2$  has Lebesgue measure 0), then  $V_f(R) = V_f(R_1) + V_f(R_2)$ . These facts allow to restrict our considerations to special rectangles only, namely to  $R = [x_1, y_2] \times [y_1, x_2]$  with diagonal segment  $\langle (x_1, y_2), (y_1, x_2) \rangle$  which is a subset of segment  $\langle \mathbf{z}, \mathbf{u} \rangle$ , where  $\mathbf{u} \in \{(0, 0), (1, 0), (0, 1), (1, 1)\}$ . It is not difficult to check the next 4 cases:

- i) if  $\langle (x_1, y_2), (y_1, x_2) \rangle \subseteq \langle \mathbf{z}, \mathbf{0} \rangle$  then

$$V_f(R) = \frac{c(x_2 - y_2)}{z_2} = \frac{c(y_1 - x_1)}{z_1} \geq 0;$$



ii) if  $\langle (x_1, y_2), (y_1, x_2) \rangle \subseteq \langle \mathbf{z}, (1, 0) \rangle$  then

$$V_f(R) = (z_2 - c) \frac{x_2 - y_2}{z_2} \geq 0, \quad (\text{note that } c \leq z_2).$$

iii) if  $\langle (x_1, y_2), (y_1, x_2) \rangle \subseteq \langle \mathbf{z}, (0, 1) \rangle$  then

$$V_f(R) = (z_1 - c) \frac{y_1 - x_1}{z_1} \geq 0, \quad (\text{note that } c \leq z_1).$$

iv) if  $\langle (x_1, y_2), (y_1, x_2) \rangle \subseteq \langle \mathbf{z}, \mathbf{1} \rangle$  then

$$V_f(R) = (1 - z_1 - z_2 + c) \frac{x_2 - y_2}{1 - z_2} = (1 - z_1 - z_2 + c) \frac{y_1 - x_1}{1 - z_1} \geq 0,$$

(note that  $c \geq \max(z_1 + z_2 - 1, 0)$  and thus  $1 - z_1 - z_2 + c \geq 0$ ).

Summarizing, we see that  $V_f(R) \geq 0$  for any rectangle  $R \subseteq [0, 1]^2$ , and thus  $f$  is a copula.  $\square$

**Remark 3.7.** Due to Theorem 3.6, each  $\langle 0, 0 \rangle$ -boundary linear  $\mathbf{z}$ -homogeneous function  $f : [0, 1]^2 \rightarrow [0, 1]$ ,  $f(\mathbf{z}) = c \in [T_L(z_1, z_2), \min(z_1, z_2)]$  is a copula. This copula is singular and its support with the corresponding masses is formed by 4 segments (possibly degenerated):

- $\langle \mathbf{0}, \mathbf{z} \rangle$  with mass  $c$  uniformly distributed over this segment;
- $\langle \mathbf{z}, (0, 1) \rangle$  with mass  $z_1 - c$  uniformly distributed over this segment;
- $\langle \mathbf{z}, (1, 0) \rangle$  with mass  $z_2 - c$  uniformly distributed over this segment;
- $\langle \mathbf{z}, \mathbf{1} \rangle$  with mass  $1 - z_1 - z_2 + c$  uniformly distributed over this segment.

As we have seen, the class of  $\langle a, b \rangle$ -boundary linear  $\mathbf{z}$ -homogeneous copulas is quite rich. This is not the case of triangular norms as shown in the next theorem.

**Theorem 3.8.** An  $\langle a, b \rangle$ -boundary linear function  $f$  which is  $\mathbf{z}$ -homogeneous for some  $\mathbf{z} \in [0, 1]^2$  is a t-norm if and only if  $f \in \{T_L, \min\}$ .

*Proof.* The sufficiency is obvious. Indeed, the Łukasiewicz t-norm  $T_L$  is  $\langle 0, 0 \rangle$ -boundary linear and  $\mathbf{z} = (z, 1 - z)$ -homogeneous for any  $z \in [0, 1]$  and  $f(\mathbf{z}) = 0$ . Similarly, the strongest t-norm,  $\min$ , is  $\langle 0, 0 \rangle$ -boundary linear and  $\mathbf{z} = (z, z)$ -homogeneous with  $c = z, z \in [0, 1]$ .

The necessity is more tricky. Obviously, each t-norm  $f$  is  $\langle 0, 0 \rangle$ -boundary linear. Suppose it is  $\mathbf{z}$ -homogeneous and  $f(\mathbf{z}) = c$ . Clearly,  $c \in [0, \min(z_1, z_2)]$ . Suppose first  $z_1 \leq z_2$ .

If  $z_2 = 1$  then necessarily  $f(\mathbf{z}) = z_1 = c$  and  $f$  is symmetric and t-norm only if  $z_1 = c = 1$  (then  $f = \min$ ).

If  $z_2 < 1$ , we will consider  $x$  sufficiently large, close to 1. Then

$$x^{(2)} = f(x, x) = u = x \frac{1 - z_1 + z_2 - c}{1 - z_1} - \frac{z_2 - c}{1 - z_1},$$

and

$$x^{(4)} = u^{(2)} = f(u, u) = u \frac{1 - z_1 + z_2 - c}{1 - z_1} - \frac{z_2 - c}{1 - z_1} = x \left( \frac{1 - z_1 + z_2 - c}{1 - z_1} \right)^2 - \left( \frac{z_2 - c}{1 - z_1} \right) \left( 2 + \frac{z_2 - c}{1 - z_1} \right).$$

On the other hand,

$$x^{(3)} = f(x, x^{(2)}) = f(x, u) = u + \frac{(x - 1)(z_2 - c)}{1 - z_1} = x \left( 1 + 2 \frac{z_2 - c}{1 - z_1} \right) - 2 \frac{z_2 - c}{1 - z_1}$$

and, due to the associativity of t-norms,

$$x^{(4)} = f(x, x^{(3)}) = x^{(3)} + \frac{(x - 1)(z_2 - c)}{1 - z_1} = x \left( 1 + 3 \frac{z_2 - c}{1 - z_1} \right) - 3 \frac{z_2 - c}{1 - z_1}.$$

This two expressions for  $x^{(4)}$  result into the next two equalities:

$$\left(\frac{1 - z_1 + z_2 - c}{1 - z_1}\right)^2 = 1 + 3 \frac{z_2 - c}{1 - z_1} \quad \text{and} \quad 3 \frac{z_2 - c}{1 - z_1} = 2 \frac{z_2 - c}{1 - z_1} = \left(\frac{z_2 - c}{1 - z_1}\right)^2.$$

Hence  $\left(\frac{z_2 - c}{1 - z_1}\right)^2 = \frac{z_2 - c}{1 - z_1}$  which implies either  $c = z_2$  (and then necessarily  $z_1 = z_2 = c$ ) or  $c = z_1 + z_2 - 1$ .

On the other hand, if  $z_1 = 0$ , then necessarily  $c = 0$  and  $f = \min$  if  $z_2 = 0$ . If  $z_2 > 0$  in this case, then  $f$  is not symmetric and thus not a t-norm.

Suppose  $z_1 > 0$ . For any  $x \in ]0, z_1]$  it holds

$$x^{(2)} = u = \frac{xc}{z_2}, \quad x^{(4)} = u^{(2)} = \frac{uc}{z_2} = x \left(\frac{c}{z_2}\right)^2.$$

More,

$$x^{(3)} = f(x, u) = \frac{uc}{z_2} = x \left(\frac{c}{z_2}\right)^2 \quad \text{and} \quad x^{(4)} = f(x, x^{(3)}) = \frac{x^{(3)}c}{z_2} = x \left(\frac{c}{z_2}\right)^3.$$

Thus  $\left(\frac{c}{z_2}\right)^2 = \left(\frac{c}{z_2}\right)^3$ , which implies either  $c = 0$  or  $c = z_2$  (and hence  $z_1 = z_2 = c$ ).

Summarizing, we see that

$$c = z_1 = z_2 \quad \text{or} \quad c = z_1 + z_2 - 1, \quad \text{and} \quad c = 0 \quad \text{or} \quad c = z_2 = z_1.$$

Then either  $c = z_1 = z_2$  and the resulting  $f = \min$ , or  $c = 0 = z_1 + z_2 - 1$ , and then  $\mathbf{z} = (z_1, 1 - z_1)$ ,  $z_1 \in ]0, \frac{1}{2}]$ ,  $f(\mathbf{z}) = 0$  and  $f = T_L$ .

Similar results are obtained if we suppose  $z_1 \geq z_2$  (then we have the case  $\mathbf{z} = (z_1, 1 - z_1)$  for  $z_1 \in ]\frac{1}{2}, 1]$ ). In all possible situations, the only possibilities are  $f = \min$  or  $f = T_L$ .  $\square$

**Remark 3.9.** Recall that if  $f$  is a t-norm then its dual  $f^d$  is a t-conorm. Due to Theorem 3.8 we see that an  $\langle a, b \rangle$ -boundary linear  $\mathbf{z}$ -homogeneous function  $f : [0, 1]^2 \rightarrow [0, 1]$  is a t-conorm if and only if  $f \in \{S_L, \max\}$ , ( $S_L$  is the Lukasiewicz t-conorm given by  $S_L(x_1, x_2) = \min(x_1 + x_2, 1)$ ). Note that the t-conorm  $\max$  is  $\langle 1, 1 \rangle$ -boundary linear and  $(z, z)$ -homogeneous with  $c = z, z \in [0, 1]$ . Similarly,  $S_L$  is  $\langle 1, 1 \rangle$ -boundary linear and  $(z, 1 - z)$ -homogeneous with  $c = 1, z \in [0, 1]$ .

Recently, overlap and grouping functions were introduced, especially for applications in image processing [2].

**Definition 3.10.** A fusion function  $f : [0, 1]^2 \rightarrow [0, 1]$  is called an overlap function whenever

- (i)  $f$  is continuous and symmetric;
- (ii)  $f$  is increasing in both coordinates;
- (iii)  $f(x, y) = 0$  if and only if  $x = 0$  or  $y = 0$ ;
- (iv)  $f(x, y) = 1$  if and only if  $x = y = 1$ .

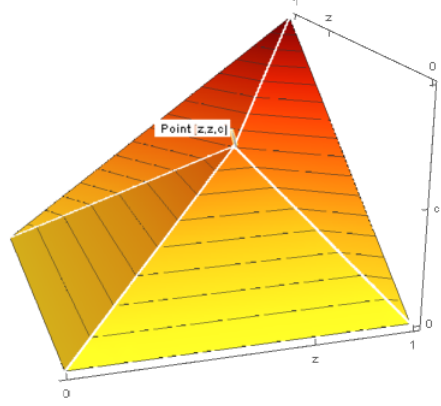
Dual function  $f^d : [0, 1]^2 \rightarrow [0, 1]$  to an overlap function  $f$  is called a grouping function.

Obviously, for any overlap function  $f$  it holds  $f(1, 0) = f(0, 1) = 0$ , and thus the particular subclass of overlap functions belong to the class of  $\langle 0, 0 \rangle$ -boundary linear aggregation functions. Clearly, these overlap functions are symmetric semicopulas, too. Thus, based on Proposition 3.4, we have the next result.

**Corollary 3.11.** A  $\langle 0, 0 \rangle$ -boundary linear  $\mathbf{z}$ -homogeneous function  $f : [0, 1]^2 \rightarrow [0, 1]$ , is an overlap function if and only if  $\mathbf{z} = (z, z), z \in [0, 1]$  and  $c = z$  if  $z \in \{0, 1\}$ , and  $c \in ]0, z[$  if  $z \in ]0, 1[$ . Then  $f = f_z$ , where

$$f_z(x, y) = \begin{cases} \min(x, y) & \text{if } c = z \\ \max\left(\frac{c}{z} \cdot \min(x, y), \min(x, y) + \frac{z-c}{1-z} \cdot \max(y-1, x-1)\right) & \text{else,} \end{cases}$$

see Figure 3.

Figure 3: The  $\mathbf{z}$ -homogeneous overlap function

*Proof.* The result that  $f = \min$  in the case when  $\mathbf{z} = (z, z)$  and  $c = z$  is trivial. Suppose that  $f$  is  $(z_1, z_2)$ -homogeneous,  $f(\mathbf{z}) = c$ . Due to the symmetry,  $f$  is also  $(z_2, z_1)$ -homogeneous, and  $f(z_2, z_1) = c$ . Due to properties of overlap functions,  $\mathbf{z} \in ]0, 1[^2$  or  $\mathbf{z} \in \{\mathbf{0}, \mathbf{1}\}$ . Consider  $\mathbf{z} \in ]0, 1[^2$ . Clearly,  $f(\mathbf{z}) = c > 0$ .

As our  $f$  is  $\langle 0, 0 \rangle$ -boundary linear semicopula, we have 4 possibilities how to evaluate  $f(z_2, z_1)$ , depending on the triangle with vertex  $\mathbf{z}$  considered in Theorem 3.2 where  $(z_2, z_1)$  belongs. Thus either

- i)  $\frac{z_2 c}{z_1} = c$  implying  $z_1 = z_2 = z$ ;
- ii)  $\frac{z_1 c}{z_2} = c$  implying  $z_1 = z_2$ ;
- iii)  $z_2 + \frac{(z_1 - 1)(z_1 - c)}{1 - z_2} = c$  implying  $z_1 = z_2$  or  $c = z_1 + z_2 - 1$ , and thus also  $z_1 + z_2 > 1$  (else the axiom (iii) from Definition 3.10 of overlap functions will be violated);
- iv)  $z_1 + \frac{(z_2 - 1)(z_2 - c)}{1 - z_1} = c$ , leading to the same conclusions as in the case iii).

Suppose  $z_1 + z_2 > 1$  and  $f(z_1, z_2) = f(z_2, z_1) = c = z_1 + z_2 - 1 > 0$ . Due to  $\mathbf{z}$ -homogeneity of  $f$ ,  $f\left(\frac{z_1}{2}, \frac{z_2}{2}\right) = \frac{c}{2}$ . On the other hand,  $f$  is also  $(z_2, z_1)$ -homogeneous and then  $f\left(\frac{z_1}{2}, \frac{z_2}{2}\right) = c \frac{z_1}{2z_2}$ , implying  $z_1 = z_2$ . Summarizing necessity  $\mathbf{z} = (z, z)$  for  $z \in [0, 1]$  and  $c = 0$  if  $z = 0$  (then  $f = \min$ ),  $c = 1$  if  $z = 1$  (then also  $f = \min$ ) and  $c \in ]0, z[$  if  $z \in ]0, 1[$ .  $\square$

Corollary 3.11 has introduced a 2-parametric family  $(f_z, c)$  of overlap functions, where  $(z, c) \in \{(0, 0), (1, 1), (u, v) | u \in ]0, 1[, v \in ]0, u]\}$ . Due to duality of overlap and grouping functions, one can characterize all  $\langle 1, 1 \rangle$ -boundary linear  $\mathbf{z}$ -homogenous grouping functions, with  $\mathbf{z} = (z, z)$ ,  $z \in [0, 1]$ , where  $z = c$  if  $z \in \{0, 1\}$ , and else  $c \in [z, 1[$ .

## 4 Concluding remarks

We have introduced and discussed  $\mathbf{z}$ -homogeneous fusion functions, in particular aggregation functions and semi-aggregation functions. Our approach generalizes the positive homogeneity of functions, which in our terminology, for functions vanishing in  $\mathbf{0}$  (i.e.,  $f(\mathbf{0}) = 0$ ), is just the  $\mathbf{0}$ -homogeneity. We expect applications of our results in several engineering domains and physics where the end-point linearity related to one fixed end-point  $\mathbf{z}$  is considered. Then the  $\mathbf{z}$ -homogeneity, possibly with some other given properties, allows to build consistent models of real world dependencies, requiring few accurate measurements only. This fact was exemplified in the case of binary (semi-)aggregation functions  $f : [0, 1]^2 \rightarrow [0, 1]$  supposing the boundary linearity. We have completely characterized several particular classes of binary aggregation functions which are boundary linear and  $\mathbf{z}$ -homogenous, such as semicopulas, quasi-copulas, copulas, triangular norms, overlap functions. Then also dual aggregation functions are completely characterized what was exemplified on t-conorms and grouping functions.

Our approach can be helpful also when constructing some other types of fusion functions, such as fuzzy implications, co-implications, restricted dissimilarity functions, etc. For example, consider boundary linear  $\mathbf{z}$ -homogenous fuzzy implication functions  $I : [0, 1]^2 \rightarrow [0, 1]$  (for more details see [1]). Recall that  $I$  is a fuzzy implication if it extends the classical Boolean implication, i.e.,  $I(0, 0) = I(0, 1) = I(1, 1) = 0$  and  $I(1, 0) = 0$ , and it is decreasing in the first coordinate and increasing in the second coordinate. Clearly, if  $I$  is boundary linear, then  $I(0, y) = I(x, 1) = 1$ ,  $I(x, 0) = 1 - x$  and  $I(1, y) = y$  for all  $x, y \in [0, 1]$ . Then the function  $f : [0, 1]^2 \rightarrow [0, 1]$  given by

$$f(x, y) = 1 - I(x, 1 - y),$$

is a semicopula. If  $I$  is also  $\mathbf{z}$ -homogeneous,  $\mathbf{z} = (z_1, z_2)$ ,  $I(\mathbf{z}) = c$ , then  $f$  is  $(z_1, 1 - z_2)$ -homogeneous and  $f(z_1, 1 - z_2) = 1 - c$ . Based on Proposition 3.4,  $1 - c \leq \min(z_1, 1 - z_2)$ , i.e.,  $c \geq \max(1 - z_1, z_2)$ .

For example, consider  $z_1 = z_2 = c = \frac{1}{2}$ . Then the corresponding fuzzy implication  $I$  which is  $\mathbf{z}$ -homogeneous and  $I(\mathbf{z}) = \frac{1}{2}$  is just the Kleene-Dienes implication  $I_{KD}$  given by  $I_{KD}(x, y) = \max(1 - x, y)$ .

Similarly, for any  $\mathbf{z} = (z, z)$ ,  $z \in [0, 1]$ , and  $I(\mathbf{z}) = 1$ , the corresponding  $\mathbf{z}$ -homogeneous implication  $I_L$  is given by  $I_L(x, y) = \min(1, 1 - x + y)$ .

## Acknowledgement

The support of the grants APVV-17-0066 and VEGA 1/0468/20 is kindly announced.

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## End-point linear functions

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### توابع خطی نقطه-پایان

چکیده. همگنی مثبت به عنوان یک  $+$  همگنی محدود نشان داده شده است و به  $Z$ -همگنی تعمیم داده می شود که خطی بودن نقطه  $Z$ -پایان نیز نامیده می شود. چندین تابع خاص تجمع  $Z$ -همگن مورد مطالعه قرار گرفته است، به طور خاص نیم‌رابطها، رابطها، توابع همپوشانی و غیره.