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# Aggregation Function Constructed from Semi-copula

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**Abstract** In this work, we show that almost all aggregation functions can be constructed by composing semi-copulas and tuples of non-decreasing univariate functions. In particular, this construction method works for all strictly increasing aggregation functions. This method also resembles Sklar's construction of multivariate distribution functions. As a demonstration, construction examples via this method are also given.

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# **1. INTRODUCTION**

One important tool for analysis and interpretation of data is a representative function. Aggregation functions are functions with special properties. They are usually used for data representation. In addition, they interpret the data contained in a k-tuple into a single-value representation. For a few decade, aggregation functions are extensively used in many fields such as applied mathematics, computer and engineering sciences, economics, and finance. One can refer to [1-3] for related literature. Each aggregation is used depending on a suitable situation. For example, we may use the mean or the median to aggregate some data, but it is not suitable for the set of data containing extremely different values. There are a lot of aggregation functions, then we should pick the appropriate functions for given data. Hence, the study of a classification is very important, the more different class, the better we get. Although, aggregation functions have used a long time ago, but construction methods are widely studied nowadays.

There are several methods to construct aggregation functions, for example, extensions of functions to aggregation functions [4-6], transformations of functions to aggregation functions [7-14], transformations of aggregation functions to other aggregation functions [15-17], and penalty-based constructions [18-23].

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We are interested in a method, which constructs aggregation functions based on copulas, Sklars copula-based method [24].

In 2008, Durante et al. [15] introduced new methods to construct aggregation functions. They applied Sklar's method, considered a multivariate distribution as a 2-increasing aggregation function and obtained similar results.

In this work, we show that almost all aggregation functions can be constructed by composing semi-copulas and tuples of non-decreasing univariate functions. This method also resembles Sklars construction of multivariate distribution functions. On the other hand, we obtain the necessary and sufficient conditions for representing aggregation functions in a form of our method. Moreover, we show that all continuous aggregation functions can be approximated by aggregation functions in this form.

In the next section, we will provide some preliminary concepts about aggregation functions, copulas, and semi-copulas. In Section 3, we will present our construction method with its properties. The proof of main theorems will be presented in Section 4. Finally, some conclusions are given.

#### 2. Preliminaries

For reasons of completeness, we will assume that the data is lying in the interval  $\mathbb{I} = [0, 1]$ . The argument  $\vec{u} = (u_1, \ldots, u_k)$  denotes an element of  $\mathbb{I}^k$  with special cases for the zero vector  $\vec{0} = (0, \ldots, 0)$  and the vector  $\vec{1} = (1, \ldots, 1)$ . The vector which has 1 at *i*-th coordinate and 0 otherwise is denoted by  $\vec{e_i}$ .

**Definition 2.1.** A function  $A : \mathbb{I}^k \to \mathbb{I}$  is called an *aggregation function* if A is nondecreasing (in each variable) with the boundary conditions,  $A(\vec{0}) = 0$  and  $A(\vec{1}) = 1$ .

Note that the notation  $\vec{u} < \vec{v}$  stand for  $u_i < v_i$  for all *i*. Similarly,  $\vec{u} \leq \vec{v}$  we mean  $u_i \leq v_i$ .

**Definition 2.2.** For any aggregation function  $A : \mathbb{I}^k \to \mathbb{I}$  and all  $a_i, b_i \in \mathbb{I}$  in which  $a_i \leq b_i$ . The volume of A, denoted by  $V_A$ , is defined by

$$V_A\left(\prod_{i=1}^{k} (a_i, b_i]\right) := \sum_{\vec{x} \in \prod_{i=1}^{k} \{a_i, b_i\}} (-1)^{N(\vec{x}, \vec{a})} A(\vec{x}),$$

where  $N(\vec{x}, \vec{a})$  is the number of *i* such that  $x_i = a_i$ .

A k-dimensional aggregation function is called a k-increasing if the volume of A is non-negative.

**Example 2.3.** The arithmetic mean, Mean :  $\mathbb{I}^k \to \mathbb{I}$  defined by  $\text{Mean}(\vec{u}) = \frac{1}{k} \sum_{i=1}^k u_i$ , is an appropriate function

an aggregation function.

**Definition 2.4.** A function  $S : \mathbb{I}^k \to \mathbb{I}$  is called a *k-semi-copula* (or a *semi-copula*) if

- (i) S is non-decreasing (in each variable),
- (ii)  $S(\vec{u}) = u_i$  if  $u_j = 1$  for all  $j \neq i$ .

Note that  $S(\vec{u}) = 0$  when  $u_i = 0$  for some *i*.

Clearly, every semi-copula is an aggregation function, but aggregation functions are not semi-copulas by Example 2.3.

**Example 2.5.** The function  $L : \mathbb{I}^k \to \mathbb{I}$  defined by

$$L(\vec{u}) = \begin{cases} u_i & \text{if } u_j = 1 \text{ for all } i \neq j, \\ 0 & \text{otherwise,} \end{cases}$$

and the minimum function M in k-dimensional defined by

$$M(\vec{u}) = \min\{u_1, \dots, u_k\}$$

are semi-copulas. Moreover,  $L \leq S \leq M$  for all semi-copula S.

**Definition 2.6.** A *k*-copula (or a copula) is a function  $C : \mathbb{I}^k \to \mathbb{I}$  satisfying the following properties:

- (i) C is a k-increasing,
- (ii)  $C(\vec{u}) = 0$  whenever  $u_i = 0$  for some  $i = 1, \dots, k$ ,
- (iii)  $C(\vec{u}) = u_i$  whenever  $u_j = 1$  for all  $j \neq i$ .

It can be shown that a copula is a semi-copula satisfying a k-increasing condition. The function L is a semi-copula but is not a copula.

One important theorem in copula theory is the Sklar's theorem [24]. The theorem is demonstated that every multivariate distribution function can be constructed by a copula and marginal distribution functions. In 2008, Durante et al.[15] shown that a 2-increasing aggregation function can be represented by a copula and univariate non-decreasing functions using Sklar's method.

In this work, we are motivated by Sklar's method and results of Durante et al. We will show that this method also works for almost all aggregation functions if we choose semi-copulas instead of a copula and then compose them with univariate non-decreasing functions.

The aim of the next section is to present our construction method together with some construction examples.

### 3. Main Results

In this section, we will present our construction method and some examples. The construction part is actually simple and natural as stated in the following theorem.

**Theorem 3.1.** Let S be a semi-copula and  $F_i$  be a non-decreasing function on  $\mathbb{I}$  with  $F_i(0) = 0$  and  $F_i(1) = 1$  for all i = 1, 2, ..., k. Then the function A defined by

$$A(\vec{u}) = S(F_1(u_1), F_2(u_2), \dots, F_k(u_k)),$$
(3.1)

for all  $\vec{u} \in \mathbb{I}^k$ , is an aggregation function. Moreover, the aggregation function A satisfies the following condition for all i = 1, 2, ..., k.

$$F_i(u) = F_i(v) \text{ implies } A(\vec{u} + (u - u_i)\vec{e}_i) = A(\vec{u} + (v - u_i)\vec{e}_i) \text{ for all } \vec{u} \in \mathbb{I}^k, u, v \in \mathbb{I}.$$

$$(3.2)$$

**Example 3.2.** Let  $S : \mathbb{I}^2 \to \mathbb{I}$  be given by  $S(u, v) = u^2 v + uv^2 - u^2 v^2$ .

We known that the function S is a semi-copula [14]. Let  $F_1, F_2$  be functions on I given by  $F_1(u) = u$  and  $F_2(v) = v^2$  for all  $u, v \in \mathbb{I}$ . Then the function A constructed via

$$A(u, v) = S(u, v^{2}) = u^{2}v^{2} + uv^{4} - u^{2}v^{4}$$

is an aggregation function.

The proof of Theorem 3.1 is obvious and simple that follows from a non-decreasing property and extreme-value conditions. A more complicated question is whether this construction is widely adaptable. In other words, whether we can use this method to construct many aggregation functions. Obviously, not all aggregation functions can be constructed via (3.1) because such aggregation functions also satisfy (3.2), it turns out that (3.2) is also sufficient for an aggregation function represented in the form of (3.1). Therefore, (3.2) can be considered as a necessary and sufficient condition yielding (3.1).

**Theorem 3.3.** Let A be an aggregation function,  $F_j(u) = A(\vec{1} - (1 - u)\vec{e_j})$  for all  $u \in \mathbb{I}, j \in \{1, 2, ..., k\}$ . If (3.2) holds, then there is a semi-copula S such that

$$A(\vec{u}) = S(F_1(u_1), \dots, F_k(u_k)).$$

Even though the above condition seems natural, but the proof is not easy. Thus, we will postpone the proof to the next section. It should be mentioned that condition (3.2) fails for disjunctive aggregation functions, e.g., those that have  $F_j(u) = 1$  for all  $u \in \mathbb{I}$ . This can also happen for an aggregation function A such that  $F_j$  is not constant.

**Example 3.4.** Let  $A : \mathbb{I}^2 \to \mathbb{I}$  be defined by

$$A(u,v) = \begin{cases} \frac{u+v}{2} & \text{if } (u,v) \in \mathbb{I} \times [0,\frac{1}{2}]\\ \max\{u,v\} & \text{if } (u,v) \in \mathbb{I} \times (\frac{1}{2},1] \end{cases}$$

for all  $u, v \in \mathbb{I}$ .

Let  $F_1(u) = A(u, 1)$  for all u and  $F_2(v) = A(1, v)$  for all v.

First, we will show that A is an aggregation function.

Let us look at the boundary conditions of A. It can be easily seen that A(0,0) = 0 and A(1,1) = 1. Now, we will show a non-decreasing property in the first coordinate. Let  $u_1, u_2 \in \mathbb{I}$  be such that  $u_1 \leq u_2$ .

**Case 1**  $v \in [0, \frac{1}{2}].$ 

$$A(u_1, v) = \frac{u_1 + v}{2} \le \frac{u_2 + v}{2} = A(u_2, v).$$

**Case 2**  $v \in (\frac{1}{2}, 1]$ .

$$A(u_1, v) = \max\{u_1, v\} \le \max\{u_2, v\} = A(u_2, v).$$

Next, we will show a non-decreasing property in the second coordinate. Let  $v_1, v_2 \in \mathbb{I}$  be such that  $v_1 \leq v_2$ . It is obvious in the case of  $v_1, v_2 \in [0, \frac{1}{2}]$  and  $v_1, v_2 \in (\frac{1}{2}, 1]$ .

Consider in the case of  $v_1 \in [0, \frac{1}{2}]$  and  $v_2 \in (\frac{1}{2}, 1]$ . If  $u \leq v_1 < v_2$ , then

$$A(u, v_1) = \frac{u + v_1}{2} < \frac{2v_1}{2} = v_1 < v_2 = \max\{u, v_2\} = A(u, v_2).$$

Therefore,  $A(u, v_1) \leq A(u, v_2)$ . If  $u > v_1$ , then we consider 2 cases. **Case 1**  $v_1 < u \leq v_2$ .

$$A(u, v_1) = \frac{u + v_1}{2} < \frac{2u}{2} = u \le v_2 = \max\{u, v_2\} = A(u, v_2).$$

Therefore,  $A(u, v_1) \le A(u, v_2)$ . Case 2  $v_1 < v_2 < u$ .

$$A(u, v_1) = \frac{u + v_1}{2} < \frac{2u}{2} = u = \max\{u, v_2\} = A(u, v_2).$$

Therefore,  $A(u, v_1) \leq A(u, v_2)$ . Now, we can conclude that A is an aggregation function. Moreover,  $F_1$  and  $F_2$  are non-decreasing. Consider  $F_1(\frac{1}{3}) = A(\frac{1}{3}, 1) = \max\{\frac{1}{3}, 1\} = 1$  and  $F_1(\frac{2}{3}) = A(\frac{2}{3}, 1) = \max\{\frac{2}{3}, 1\} = 1$ , but  $A(\frac{1}{3}, \frac{1}{2}) = \frac{1}{2}(\frac{1}{3} + \frac{1}{2}) = \frac{5}{12} \neq \frac{7}{12} = \frac{1}{2}(\frac{2}{3} + \frac{1}{2}) = A(\frac{2}{3}, \frac{1}{2}).$ 

In most cases, aggregation functions satisfy (3.2). Thus, they can be constructed via (3.1). Also, this construction works for strictly increasing aggregation functions as the following result.

**Corollary 3.5.** Let A be a strictly increasing aggregation function. Then there are semicopula S and strictly increasing  $F_j$  where j = 1, 2, ..., k such that

$$A(\vec{u}) = S(F_1(u_1), \dots, F_k(u_k)).$$

*Proof.* Let  $A : \mathbb{I}^k \to \mathbb{I}$  be a strictly increasing aggregation function.

Define  $F_j : \mathbb{I} \to \mathbb{I}$  by  $F_j(u) = A(\vec{1} - (1 - u)\vec{e_j})$  for each  $j \in \{1, 2, \dots, k\}$ .

We can see that  $F_j$  are non-decreasing for all j. Next, we will show that the above statement satisfies the sufficient condition of Theorem 3.3. Let  $u, v, u_j \in \mathbb{I}$  for all  $j = 1, 2, \ldots, k$ .

For each *i*, assume that  $F_i(u) = F_i(v)$ . We have

$$A(\vec{1} - (1 - u)\vec{e}_i) = F_i(u) = F_i(v) = A(\vec{1} - (1 - v)\vec{e}_i).$$

Since A is strictly increasing, we get A is injective, that is, u = v. It implies that  $A(u_1, \ldots, u, \ldots, u_k) = A(u_1, \ldots, v, \ldots, u_k)$ . By Theorem 3.3, there is a semi-copula S such that

$$A(\vec{u}) = S(F_1(u_1), \dots, F_k(u_k)).$$

Observe that a strictly increasing property is stronger than (3.2). However, we can show that the class of strictly increasing aggrgation functions that represented in the form (3.1) is almost all aggregation functions via approximation method.

**Theorem 3.6.** For any an aggregation function A, there is a sequence  $\{A_n\}$  of (strictly) aggregation functions that can be written in the form (3.1) and that the sequence  $\{A_n\}$  converges uniformly to A.

*Proof.* Let A be an aggregation function. For each  $n \in \mathbb{N}$ , define

$$A_n = (1 - \frac{1}{n})A + \frac{1}{n}M.$$

For each n, let  $F_{i,n}(u) = A_n(\vec{1} - (1 - u)\vec{e_i})$  for all *i*. We can see that  $A_n$  is strictly increasing. It implies that  $F_{i,n}$  are also strictly increasing for all *i*. By Corollary 3.5, we obtain that for each n, there is a strictly semi-copula  $S_n$  such that

$$A_n = S_n(F_{1,n},\ldots,F_{k,n}).$$

Next, we prove that  $A_n$  converges uniformly to an aggregation function A. Let  $u_j \in \mathbb{I}$  for all j = 1, 2, ..., k.

$$\begin{aligned} |A_n(\vec{u}) - A(\vec{u})| &= |(1 - \frac{1}{n})A(\vec{u}) + \frac{1}{n}M(\vec{u}) - A(\vec{u})| \\ &= |-\frac{1}{n}A(\vec{u}) + \frac{1}{n}M(\vec{u})| \\ &\leq \frac{1}{n}(|A(\vec{u})| + |M(\vec{u})|) \\ &\leq \frac{2}{n} \to 0 \text{ as } n \to \infty. \end{aligned}$$

Therefore,  $\{A_n\}$  converges uniformly to an aggregation function A.

# 4. PROOF OF THEOREM 3.3

To prove Theorem 3.3, we study some following results. Recall the standard conventions,  $\sup \emptyset = 0$  and  $\inf \emptyset = 1$ .

**Proposition 4.1.** Let  $X \subseteq \mathbb{I}$  be closed,  $f : X \to \mathbb{I}$  be non-decreasing, let  $u, v \in \mathbb{I}$  with  $u \leq v$ . Then

(i)  $u_X^- \le u \le u_X^+$  for all  $u \in \mathbb{I}$ ,

- (ii) If  $u \in \mathbb{I} \setminus X$ , then  $u_X^-, u_X^+ \in X$ ,
- (iii) If  $u \in X, v \in \mathbb{I} \setminus X$ , then  $f(u) \le f(v_X^-)$ ,
- (iv) If  $u \in \mathbb{I} \setminus X, v \in X$ , then  $f(u_X^+) \leq f(v)$

where  $u_X^- := \sup\{t \in X : t < u\}, \ u_X^+ := \inf\{t \in X : t > u\}.$ 

*Proof.* (i) Let  $u \in \mathbb{I}$ . If  $u \in X$ , then  $u_X^- = u = u_X^+$ . Suppose that  $u \in \mathbb{I} \setminus X$ . If there are  $a_1, a_2 \in X$  such that  $a_1 \leq u \leq a_2$ , we have  $u_X^- \leq u \leq u_X^+$ . If  $u \leq a$  for all  $a \in X$ , then  $u_X^- = \sup \emptyset = 0 \leq u$  and  $u \leq \inf X = u_X^+$ . In the case of  $u \geq a$  for all  $a \in X$ , we have  $u_X^- = \sup X \leq u$  and  $u \leq 1 = \inf \emptyset = u_X^+$ .

(*ii*) Let  $u \in \mathbb{I} \setminus X$ . The proof is similar to (*i*).

(*iii*) Let  $u \in X, v \in \mathbb{I} \setminus X$ . Also, we get u < v. Then we have  $u \leq v_X^-$ . Since f is non-decreasing and  $v_X^- \in X$ , we obtain that  $f(u) \leq f(v_X^-)$ .

(iv) Let  $u \in \mathbb{I} \setminus X, v \in X$ . Also, we get u < v and hence  $u_X^+ \leq v$ . Since f is non-decreasing and  $u_X^+ \in X$ , we get  $f(u_X^+) \leq f(v)$ .

**Proposition 4.2.** Let  $X \subseteq \mathbb{I}$ . Define  $X^- = \{u \in \mathbb{I} : u_X^- = u\}$  and  $X^+ = \{u \in \mathbb{I} : u_X^+ = u\}$ . Then the closure of X is a subset of  $X^* = X^- \cup X^+$ .

*Proof.* Denote the closure of X by  $\overline{X}$ . Let  $a \in \overline{X}$ . Then there is a sequence  $\{a_n\} \subseteq X$  which converges to a. If  $a \in X$ , then  $a = a_X^- = a_X^+$ . Thus,  $a \in \overline{X} \subseteq X^*$ . Consider in the case of  $a \in \mathbb{I} \setminus X$ , we have  $a_n \neq a$  for all  $n \in \mathbb{N}$ . **Case 1**  $a_n > a$  for all  $n \in \mathbb{N}$ .

There is a monotone subsequence  $\{a_{n_k}\}$  of  $\{a_n\}$  such that  $a_{n_k} > a$  for all  $k \in \mathbb{N}$ . Moreover, we have  $a_{n_k} \to a$  as  $k \to \infty$ . Thus,  $a = \inf_{k \in \mathbb{N}} a_{n_k}$ .

Since  $\{a_{n_k} \in X : k \in \mathbb{N}\} \subseteq \{t \in X : t > a\}$ , we get  $a = \inf_{k \in \mathbb{N}} a_{n_k} \ge \inf\{t \in X : t > a\}$ . We know that  $a \le \inf\{t \in X : t > a\}$ . Then we have  $a = \inf\{t \in X : t > a\} = a_X^-$ , that is,  $\bar{X} \subseteq X^- \subseteq X^*$ . **Case 2**  $a_n < a$  for all  $n \in \mathbb{N}$ . The proof of this case is similar to Case 1.

**Case 3** There is a subsequence  $\{a_{n_k}\}$  of  $\{a_n\}$  such that for each  $k \in \mathbb{N}$ , either  $a_{n_k} > a$ or  $a_{n_k} < a$ . There is a subsequence  $\{a_{n_{k_l}}\}$  of  $\{a_n\}$  such that either  $a_{n_{k_l}} > a$  or  $a_{n_{k_l}} < a$ for all  $l \in \mathbb{N}$ . By the proof of Case 1 and Case 2, we have  $\overline{X} \subseteq X^*$ .

Observe that if we have a non-decreasing function F on I, we can define  $F^{\downarrow}(u) =$  $\inf\{x: F(x) \ge u\}$  and  $F^{\uparrow}(u) = \sup\{x: F(x) \le u\}$ . It is easy to prove that  $F^{\downarrow}$  and  $F^{\uparrow}$ are non-decreasing. Moreover,  $F^* := \frac{1}{2}F^{\downarrow} + \frac{1}{2}F^{\uparrow}$  is also non-decreasing.

**Lemma 4.3.** Let  $F : \mathbb{I} \to \mathbb{I}$  be non-decreasing. Then  $FF^*(u) = u$  for all  $u \in Range(F)$ . Moreover,  $FF^*F(u) = F(u)$  for all  $u \in \mathbb{I}$ .

*Proof.* Suppose F is non-decreasing. Let  $u \in Range(F)$ . Then there is  $u' \in \mathbb{I}$  such that F(u') = u.

**Case 1** u' is the only one point such that F(u') = u. Then  $F^{\uparrow}(u) = F^{\downarrow}(u) = u'$ . We have  $F^*(u) = \frac{1}{2}F^{\downarrow}(u) + \frac{1}{2}F^{\uparrow}(u) = F^{\downarrow}(u) = u'$ . Hence,  $FF^*(u) = F(u') = u$ . **Case 2** u' is not only one point such that F(u') = u.

Let  $X = \{z \in \mathbb{I} : F(z) = u\}$ . We can see that X is a bounded set. We set  $l = \inf X$  and  $m = \sup X$ . Thus,  $F^{\uparrow}(u) = m$ . Similarly,  $F^{\downarrow}(u) = l$ . In this case,  $F^{\downarrow}(u) = l < m = F^{\uparrow}(u)$ . We have

$$F^*(u) = \frac{1}{2}F^{\downarrow}(u) + \frac{1}{2}F^{\uparrow}(u) = \frac{1}{2}l + \frac{1}{2}m > \frac{1}{2}l + \frac{1}{2}l = l.$$

Similarly,  $F^*(u) < m$ . Hence, there are  $z_1, z_2 \in \mathbb{I}$  such that  $F(z_1) = u = F(z_2)$  and l < l $z_1 \leq F^*(u) \leq z_2 < m$ . Therefore,  $FF^*(u) = F(z_1) = F(z_2) = u$ . Moreover,  $FF^*F(u) = F(z_1) = F(z_2) = u$ . F(u) for all  $u \in \mathbb{I}$ .

**Theorem 4.4.** Let A be an aggregation function, S be a semi-copula,  $F_i$  be a monotone function for all  $i = 1, 2, \ldots, k$  such that

$$A(\vec{u}) = A(u_1, \dots, u_k) = S(F_1(u_1), \dots, F_k(u_k))$$

where  $u_i \in \mathbb{I}$  for all *i*. Then the following conditions are satisfied:

- (i)  $F_i(1) = 1$  for all i = 1, 2, ..., k,
- (*ii*) For each  $i \in \{1, 2, ..., k\}$ ,  $F_i(u) = A(\vec{1} (1 u)\vec{e_i})$  for all u,
- (*iii*)  $S(\vec{u}) = A(F_1^*(u_1), F_2^*(u_2), \dots, F_k^*(u_k))$  for all  $u_i \in Range(F_i)$ ,
- $\begin{array}{l} (iv) \quad A(F_1^*(1),F_2^*(1),\ldots,F_k^*(1)) = 1, \\ (v) \quad For \ fixed \ i, \ A(\sum_{j=1}^k F_j^*(1)\vec{e_j} + (F_i^*(u_i) F_i^*(1))\vec{e_i}) = u_i \ for \ all \ u_i \in Range(F_i). \end{array}$

*Proof.* (i) By assumption,  $1 = A(\vec{1}) = S(F_1(1), F_2(1), \dots, F_k(1)).$ Since S is non-decreasing, for each  $i, 1 \leq S(\vec{1} - (1 - F_i(1))\vec{e}_i) = F_i(1)$ . Thus,  $F_i(1) = 1$  for each  $i \in \{1, 2, ..., k\}$ .

(*ii*) For each *i*, we have  $A(\vec{1} - (1 - u)\vec{e}_i) = S(\vec{1} - (1 - F_i(u))\vec{e}_i) = F_i(u)$ .

(iii) Let  $u_i \in Range(F_i)$  for all i = 1, 2, ..., k. By Lemma 4.3, we have  $F_i F_i^*(u_i) = u_i$ . Thus,  $A(F_1^*(u_1), F_2^*(u_2), \dots, F_k^*(u_k)) = S(F_1F_1^*(u_1), F_2F_2^*(u_2), \dots, F_kF_k^*(u_k)) = S(\vec{u}).$ 

(iv) From (iii) holds and  $\hat{S}$  is a semi-copula, we get  $A(F_1^*(1), F_2^*(1), \ldots, F_k^*(1)) =$  $S(\vec{1}) = 1.$ 

(v) For fixed i, let  $u_i \in Range(F_i)$ . From (iii) holds and S is a semi-copula, we get  $u_i = S(\vec{1} - (1 - u_i)\vec{e}_i) = A(\sum_{i=1}^k F_i^*(1)\vec{e}_j + (F_i^*(u_i) - F_i^*(1))\vec{e}_i).$ 

**Lemma 4.5.** Let A be an aggregation function, let  $u \in \mathbb{I}$ . For each  $j \in \{1, 2, ..., k\}$ , define  $F_j(u) = A(\vec{1} - (1 - u)\vec{e_j})$ . Assume that

$$A(\sum_{j=1}^{k} F_{j}^{*}(1)\vec{e_{j}} + (u - F_{i}^{*}(1))\vec{e_{i}}) = F_{i}(u).$$

Let  $u_i \in Range(F_i)$  for all *i*. Then

$$A(\sum_{j=1}^{n} F_{j}^{*}(1)\vec{e_{j}} + (F_{i}^{*}(u_{i}) - F_{i}^{*}(1))\vec{e_{i}}) = u_{i} \text{ for all } i.$$

*Proof.* Let  $u_i \in Range(F_i)$  for fixed  $i \in \{1, 2, ..., k\}$ . By assumption, we have  $A(\sum_{j=1}^k F_j^*(1)\vec{e_j} + (u - F_i^*(1))\vec{e_i}) = F_i(u)$  for all u. By Lemma 4.3, we get  $A(\sum_{j=1}^k F_j^*(1)\vec{e_j} + (F_i^*(u_i) - F_i^*(1))\vec{e_i}) = F_i(F_i^*(u_i)) = u_i$ . ■

**Lemma 4.6.** Let A be an aggregation function, let  $u \in \mathbb{I}$ . For each  $j \in \{1, 2, ..., k\}$ , assume that  $F_j(u) = A(\vec{1} - (1 - u)\vec{e_j})$  for all j. For fixed  $i \in \{1, 2, ..., k\}$  and  $u_i \in Range(F_i)$ , suppose that

$$A(\sum_{j=1}^{k} F_{j}^{*}(1)\vec{e_{j}} + (F_{i}^{*}(u_{i}) - F_{i}^{*}(1))\vec{e_{i}}) = u_{i}$$

Define

$$S(\vec{u}) = S(u_1, \dots, u_k) = A(F_1^*(u_1), \dots, F_k^*(u_k))$$

for all  $u_j \in Range(F_j)$ . Then S is non-decreasing with  $S(u_1, \ldots, u_k) = u_i$  for all  $u_i \in Range(F_i)$  where  $u_j = 1$  for all  $j \neq i$ .

Proof. Define  $S : \prod_{j=1}^{k} Range(F_j) \to \mathbb{I}$  by  $S(\vec{u}) = A(F_1^*(u_1), \dots, F_k^*(u_k))$  for all  $u_j \in Range(F_j)$ .

We will show that S is non-decreasing with  $S(\vec{u}) = u_i$  for all  $u_i \in Range(F_i)$  where  $u_j = 1$  for all  $j \neq i$ . Let  $u_i, v_i \in Range(F_i)$  be such that  $u_i \leq v_i$  for all i. It implies that

$$S(\vec{u}) = A(F_1^*(u_1), \dots, F_k^*(u_k)) \le A(F_1^*(v_1), \dots, F_k^*(v_k)) = S(\vec{v}).$$

Therefore, S is non-decreasing. Moreover, for  $u_i \in Range(F_i)$ ,

$$S(\vec{1} - (1 - u_i)\vec{e}_i) = A(F_1^*(1), \dots, F_i^*(u_i), \dots, F_k^*(1)) = u_i.$$

Henceforth, for each  $u \in \mathbb{I}$ , we put  $S(\vec{u}+(u-u_i)\vec{e_i})$  instead of  $S(u_1, ..., u_{i-1}, u, u_{i+1}, ..., u_k)$ . For now, Lemmas 4.7-4.10 are the tools that help us to prove Theorem 4.11 which shows that a non-decreasing function S can be extended to a semi-copula under some conditions.

**Lemma 4.7.** Let  $S: \prod_{j=1}^{k} X_j \to \mathbb{I}$  be non-decreasing where  $X_j \subseteq \mathbb{I}$  for all j = 1, 2, ..., k. For each j, define  $\hat{X}_j = \{u \in \mathbb{I} : u_{X_j}^- = u\}$ . For fixed  $i \in \{1, 2, ..., k\}$ , define

$$\hat{S}(\vec{u}) = \begin{cases} S(\vec{u}) & \text{if } \vec{u} \in \prod_{j=1}^{k} X_j, \\ \sup_{t < u_i} S(\vec{u} + (t - u_i)\vec{e}_i) & \text{if } \vec{u} \in \hat{X}_i \backslash X_i \times \prod_{\substack{j=1\\ j \neq i}}^{k} X_j. \end{cases}$$

Then  $\hat{S}: \hat{X}_i \times \prod_{\substack{j=1 \ j \neq i}}^k X_j \to \mathbb{I}$  is also non-decreasing. Moreover, if  $S(\vec{1} - (1 - u_i)\vec{e}_i) = u_i$  for all  $u_i \in X_i$ , then  $\hat{S}(\vec{1} - (1 - u_i)\vec{e}_i) = u_i$  for all  $u_i \in \hat{X}_i$  also.

*Proof.* Let  $\vec{u}, \vec{v} \in \hat{X}_i \times \prod_{\substack{j=1\\j \neq i}}^k X_j$  with  $\vec{u} \le \vec{v}$ . It can be shown that  $X_j \subseteq \hat{X}_j$  for all j. To prove  $\hat{S}$  is non-decreasing on its domain, we can divide the situation into four cases. **Case 1**  $\vec{u}, \vec{v} \in \prod_{\substack{k\\j=1}}^k X_j$ . It is obvious.

**Case 2** 
$$\vec{u} \in \prod_{j=1}^{k} X_j$$
 and  $\vec{v} \in \hat{X}_i \setminus X_i \times \prod_{\substack{j \neq i \\ j \neq i}}^{k} X_j$ . Then  
 $\hat{S}(\vec{u}) = S(\vec{u}) \leq \sup_{u_i \leq t < v_i} S(\vec{u} + (t - u_i)\vec{e}_i) \leq \sup_{t < v_i} S(\vec{v} + (t - v_i)\vec{e}_i) = \hat{S}(\vec{v}).$ 

**Case 3**  $\vec{u} \in \hat{X}_i \setminus X_i \times \prod_{\substack{j=1 \ j \neq i}}^k X_j$  and  $\vec{v} \in \prod_{\substack{j=1 \ j \neq i}}^k X_j$ . We have

$$\hat{S}(\vec{u}) = \sup_{t < u} S(\vec{u} + (t - u_i)\vec{e}_i) \le S(\vec{u} + (v_i - u_i)\vec{e}_i) \le \hat{S}(\vec{v}).$$

**Case 4**  $\vec{u}, \vec{v} \in \hat{X}_i \setminus X_i \times \prod_{\substack{j=1\\ j \neq i}}^k X_j$ . We have  $\hat{S}(\vec{u}) = \sup S(\vec{u} + (t - u_i)\vec{e}_i) \le \sup$ 

$$\hat{S}(\vec{u}) = \sup_{t < u_i} S(\vec{u} + (t - u_i)\vec{e}_i) \le \sup_{t < u_i} S(\vec{v} + (t - v_i)\vec{e}_i) \le \sup_{t < v_i} S(\vec{v} + (t - v_i)\vec{e}_i) = \hat{S}(\vec{v})$$

Now, S is non-decreasing.

Suppose  $S(\vec{1} - (1 - u_i)\vec{e}_i) = u_i$  for all  $u_i \in X_i$ .  $(1 \in X_j$  for all  $j \neq i$ ) We will show that  $\hat{S}(\vec{1} - (1 - u_i)\vec{e}_i) = u_i$  for all  $u_i \in \hat{X}_i$ . Let  $u \in \hat{X}_i$ .

If  $u_i \in X_i$ , then we are done. Consider in case  $u_i \in \hat{X}_i \setminus X_i$ . Thus, we obtain that  $\sup\{t \in X_i : t < u_i\} = u_{X_i}^- = u_i$ . It implies that

$$\hat{S}(\vec{1} - (1 - u_i)\vec{e}_i) = \sup_{t < u_i} S(\vec{1} - (1 - u_i)\vec{e}_i) = \sup_{t < u_i} t = u_{X_i}^- = u_i.$$

From Lemma 4.7, we can extend  $X_j$  to  $\hat{X}_j$  for all j = 1, 2, ..., k by the same method.

**Lemma 4.8.** Let  $S : \prod_{j=1}^{k} X_j \to \mathbb{I}$  be non-decreasing where  $X_j \subseteq \mathbb{I}$  for all j = 1, 2, ..., k. For each j, define  $\hat{X}_j = \{u \in \mathbb{I} : u_{X_j}^+ = u\}$ . For fixed  $i \in \{1, 2, ..., k\}$ , define

$$\hat{S}(\vec{u}) = \begin{cases} S(\vec{u}) & \text{if } \vec{u} \in \prod_{j=1}^{k} X_{j}, \\ \inf_{t > u_{i}} S(\vec{u} + (t - u_{i})\vec{e_{i}}) & \text{if } \vec{u} \in \hat{X}_{i} \backslash X_{i} \times \prod_{\substack{j \neq i \\ j \neq i}}^{k} X_{j}. \end{cases}$$

Then  $\hat{S}: \hat{X}_i \times \prod_{\substack{j=1 \ j \neq i}}^k X_j \to \mathbb{I}$  is also non-decreasing. If  $S(\vec{1} - (1 - u_i)\vec{e}_i) = u_i$  for all  $u_i \in X_i$ , then  $\hat{S}(\vec{1} - (1 - u_i)\vec{e}_i) = u_i$  for all  $u_i \in \hat{X}_i$  also.

*Proof.* Let  $\vec{u}, \vec{v} \in \hat{X}_i \times \prod_{\substack{j=1\\j \neq i}}^k X_j$  with  $\vec{u} \le \vec{v}$ . It can be shown that  $X_j \subseteq \hat{X}_j$  for all j. To prove  $\hat{S}$  is non-decreasing on its domain, we can divide the situation into four cases. **Case 1**  $\vec{u}, \vec{v} \in \prod_{j=1}^k X_j$ . It is obvious.

**Case 2**  $\vec{u} \in \prod_{j=1}^{k} X_j$  and  $\vec{v} \in \hat{X}_i \setminus X_i \times \prod_{\substack{j=1 \ j \neq i}}^{k} X_j$ . Then

$$\hat{S}(\vec{u}) = S(\vec{u}) \le \inf_{t \ge u_i} S(\vec{u} + (t - u_i)\vec{e}_i) \le \inf_{t > v_i} S(\vec{u} + (t - u_i)\vec{e}_i) = \hat{S}(\vec{u} + (v - u_i)\vec{e}_i) \le \hat{S}(\vec{v}).$$

**Case 3**  $\vec{u} \in \hat{X}_i \setminus X_i \times \prod_{\substack{j=1 \ j \neq i}}^k X_j$  and  $\vec{v} \in \prod_{j=1}^k X_j$ . We have

$$\hat{S}(\vec{u}) = \inf_{t > u_i} S(\vec{u} + (t - u_i)\vec{e}_i) \le S(\vec{u} + (v_i - u_i)\vec{e}_i) \le S(\vec{v}) = \hat{S}(\vec{v}).$$

**Case 4**  $\vec{u}, \vec{v} \in \hat{X}_i \setminus X_i \times \prod_{\substack{j=1\\j\neq i}}^k X_j$ . We have  $\hat{S}(\vec{u}) = \inf S(\vec{u} + (t-u_i)\vec{e}_i) < \inf S(\vec{u})$ 

$$\hat{S}(\vec{u}) = \inf_{t > u_i} S(\vec{u} + (t - u_i)\vec{e}_i) \le \inf_{t > v_i} S(\vec{u} + (v_i - u_i)\vec{e}_i) \le \inf_{t > v_i} S(\vec{v} + (t - v_i)\vec{e}_i) = \hat{S}(\vec{v}).$$

Similarly to Lemma 4.7, we can prove that  $\hat{S}(\hat{1} - (1 - u_i)\vec{e}_i) = u_i$ .

**Lemma 4.9.** Let  $S : \prod_{j=1}^{k} X_j \to \mathbb{I}$  be non-decreasing where  $X_j \subseteq \mathbb{I}$  for all j = 1, 2, ..., k. Then S can be extended to a non-decreasing function  $\hat{S}$  on  $\prod_{j=1}^{k} \bar{X}_j$ . Moreover, if  $S(\vec{1} - (1 - u_i)\vec{e}_i) = u_i$  for all  $u_i \in X_i$ , then  $\hat{S}(\vec{1} - (1 - u_i)\vec{e}_i) = u_i$  for all  $u_i \in \bar{X}_i$  also.

Proof. Let  $S : \prod_{j=1}^{k} X_j \to \mathbb{I}$  be non-decreasing where  $X_j \subseteq \mathbb{I}$  for all j = 1, 2, ..., k. Fixed  $i \in \{1, 2, ..., k\}$ , suppose  $S(\vec{1} - (1 - u_i)\vec{e_i}) = u_i$  for all  $u_i \in X_i$ . By Lemma 4.7, we can extend S to S' on  $\prod_{j=1}^{k} X'_j$  where  $X'_j = \{u \in \mathbb{I} : u_{X_j} = u\}$ . Moreover,  $S'(\vec{1} - (1 - u_i)\vec{e_i}) = u_i$  for all  $u_i \in X_i$ . By Lemma 4.8, S' can be extended to a non-decreasing function S'' on  $\prod_{j=1}^{k} X''_j$  where

 $X_j'' = \{u \in \mathbb{I} : u_{X_j'}^+ = u\}.$  Furthermore,  $S''(\vec{1} - (1 - u_i)\vec{e}_i) = u_i \text{ for all } u_i \in X_i.$ 

By Proposition 4.2, we obtain that  $\bar{X}_j \subseteq X_j''$  for all j.

Thus, we can restrict the domain of S'' into  $\prod_{j=1}^{k} \bar{X}_j$ . Hence, we get a non-decreasing function  $\hat{S} : \prod_{j=1}^{k} \bar{X}_j \to \mathbb{I}$ . In addition,  $\hat{S}(\vec{1} - (1 - u_i)\vec{e}_i) = u_i$  for all  $u_i \in \bar{X}_i$ .

**Lemma 4.10.** Let  $X_j$  be a closed subset of  $\mathbb{I}$  for all j = 1, 2, ..., k. Let  $S : \prod_{j=1}^k X_j \to \mathbb{I}$  be non-decreasing. For fixed  $i \in \{1, 2, ..., k\}$ , define

$$\hat{S}(\vec{u}) = \begin{cases} S(\vec{u}) & \text{if } \vec{u} \in \prod_{j=1}^{k} X_j, \\ \frac{1}{u_{X_i}^+ - u_{X_i}^-} \Big[ (u_{X_i}^+ - u_i) S(\vec{u} + (u_{X_i}^- - u_i) \vec{e}_i) \\ + (u_i - u_{X_i}^-) S(\vec{u} + (u_{X_i}^+ - u_i) \vec{e}_i) \Big] & \text{if } \vec{u} \in \mathbb{I} \setminus X_i \times \prod_{\substack{j=1\\ j \neq i}}^{k} X_j \end{cases}$$

where  $u_{X_i}^- = \sup\{t \in X_i : t < u_i\}$  and  $u_{X_i}^+ = \inf\{t \in X_i : t > u_i\}$ . Then  $\hat{S} : \mathbb{I} \times \prod_{\substack{j=1 \ j \neq i}}^k X_j \to \mathbb{I}$  is also non-decreasing. Moreover, if  $S(\vec{1} - (1 - u_i)\vec{e_i}) = u_i$  for all  $u_i \in X_i$ , then  $\hat{S}(\vec{1} - (1 - u_i)\vec{e_i}) = u_i$  for all  $u_i \in \mathbb{I}$  also.

*Proof.* Let  $\vec{u}, \vec{v} \in \mathbb{I} \times \prod_{\substack{j=1\\j\neq i}}^{k} X_j$  with  $\vec{u} \leq \vec{v}$ . To prove  $\hat{S}$  is non-decreasing on its domain, we divide the situation into four cases. **Case 1**  $\vec{u}, \vec{v} \in \prod_{j=1}^{k} X_j$ . It is obvious. **Case 2**  $\vec{u} \in \prod_{i=1}^{k} X_i$  and  $\vec{v} \in \mathbb{I} \setminus X_i \times \prod_{j=1}^{k} X_j$ . Then

$$\begin{split} \hat{S}(\vec{u}) &= S(\vec{u}) \\ &\leq S(\vec{u} + (v_{X_i}^- - u_i)\vec{e}_i) \\ &\leq S(\vec{v} + (v_{X_i}^- - v_i)\vec{e}_i) \\ &\leq \left(\frac{v_{X_i}^+ - v}{v_{X_i}^+ - v_{X_i}^-}\right) S(\vec{u} + (v_{X_i}^- - v_i)\vec{e}_i) + \left(\frac{v - v_{X_i}^-}{v_{X_i}^+ - v_{X_i}^-}\right) S(\vec{u} + (v_{X_i}^+ - v_i)\vec{e}_i) = \hat{S}(\vec{v}). \end{split}$$

 $\begin{aligned} \mathbf{Case \ 3} \ \vec{u} \in \mathbb{I} \setminus X_i \times \prod_{\substack{j=1 \ j \neq i}}^k X_j \ \text{and} \ \vec{v} \in \prod_{j=1}^k X_j. \ \text{Then} \\ \hat{S}(\vec{u}) &= \left(\frac{u_{X_i}^+ - u_i}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{u} + (u_{X_i}^- - u_i)\vec{e}_i) + \left(\frac{u_i - u_{X_i}^-}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{u} + (u_{X_i}^+ - u_i)\vec{e}_i) \\ &\leq S(\vec{u} + (u_{X_i}^+ - u_i)\vec{e}_i) \\ &\leq S(\vec{u} + (v_i - u_i)\vec{e}_i) \\ &< \hat{S}(\vec{v}). \end{aligned}$ 

Case 4  $\vec{u}, \vec{v} \in \mathbb{I} \setminus X_i \times \prod_{\substack{j=1 \ j \neq i}}^k X_j.$ 

**Case 4.1** There is  $a \in X_i^{j \neq i}$  such that  $u_i \leq a \leq v_i$ . Thus,  $u_{X_i}^+ \leq v_{X_i}^-$ . It implies that

$$\begin{split} \hat{S}(\vec{u}) &= \left(\frac{u_{X_i}^+ - u_i}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{u} + (u_{X_i}^- - u_i)\vec{e}_i) + \left(\frac{u_i - u_{X_i}^-}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{u} + (u_{X_i}^+ - u_i)\vec{e}_i) \\ &\leq \left(\frac{u_{X_i}^+ - u_i}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{u} + (u_{X_i}^+ - u_i)\vec{e}_i) + \left(\frac{u_i - u_{X_i}^-}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{u} + (u_{X_i}^+ - u_i)\vec{e}_i) \\ &= \left(\frac{u_{X_i}^+ - u_i - u_{X_i}^-}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{u} + (u_{X_i}^+ - u_i)\vec{e}_i) \\ &= \left(\frac{u_{X_i}^+ - u_{X_i}^-}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{u} + (u_{X_i}^+ - u_i)\vec{e}_i) \\ &= S(\vec{u} + (u_{X_i}^+ - u_i)\vec{e}_i) \\ &\leq S(\vec{u} + (v_{X_i}^- - u_i)\vec{e}_i) \\ &\leq \left(\frac{v_{X_i}^+ - v_{X_i}^-}{v_{X_i}^+ - v_{X_i}^-}\right) S(\vec{u} + (v_{X_i}^- - u_i)\vec{e}_i) + \left(\frac{v - v_{X_i}^-}{v_{X_i}^+ - v_{X_i}^-}\right) S(\vec{u} + (v_{X_i}^+ - u_i)\vec{e}_i) \\ &\leq \hat{S}(\vec{v}). \end{split}$$

**Case 4.2** For each  $a \in X_i$ , either  $a \leq u_i \leq v_i$  or  $u_i \leq v_i \leq a$ . Thus,  $u_{X_i}^- = v_{X_i}^-$  and  $u_{X_i}^+ = v_{X_i}^+$ . For convenience, for each  $u \in X_i$ , we put  $S(\vec{u} + (u - u_i)\vec{e_i})$  instead of  $S(u_1, \ldots, u_{i-1}, u, u_{i+1}, \ldots, u_k)$ . Then

$$\begin{split} \hat{S}(\vec{v}) - \hat{S}(\vec{u}) &= \left(\frac{v_{X_i}^+ - v_i}{v_{X_i}^+ - v_{\overline{X}_i}^-}\right) S(\vec{v} + (v_{\overline{X}_i}^- - v_i)\vec{e}_i) + \left(\frac{v_i - v_{\overline{X}_i}^-}{v_{X_i}^+ - v_{\overline{X}_i}^-}\right) S(\vec{v} + (v_{X_i}^+ - v_i)\vec{e}_i) \\ &- \left(\frac{u_{X_i}^+ - u_i}{u_{X_i}^+ - u_{\overline{X}_i}^-}\right) S(\vec{u} + (u_{\overline{X}_i}^- - u_i)\vec{e}_i) - \left(\frac{u_i - u_{\overline{X}_i}^-}{u_{X_i}^+ - u_{\overline{X}_i}^-}\right) S(\vec{u} + (u_{X_i}^+ - u_i)\vec{e}_i) \\ &= \left(\frac{v_{X_i}^+ - v_i}{v_{X_i}^+ - v_{\overline{X}_i}^-}\right) S(\vec{v} + (v_{\overline{X}_i}^- - v_i)\vec{e}_i) + \left(\frac{v_i - v_{\overline{X}_i}^-}{v_{X_i}^+ - v_{\overline{X}_i}^-}\right) S(\vec{v} + (v_{X_i}^+ - u_i)\vec{e}_i) \\ &- \left(\frac{v_{X_i}^+ - u_i}{v_{X_i}^+ - v_{\overline{X}_i}^-}\right) S(\vec{u} + (v_{\overline{X}_i}^- - u_i)\vec{e}_i) - \left(\frac{u_i - v_{\overline{X}_i}^-}{v_{X_i}^+ - v_{\overline{X}_i}^-}\right) S(\vec{u} + (v_{X_i}^+ - u_i)\vec{e}_i) \end{split}$$

$$\begin{split} &= \left(\frac{v_{X_i}^+ - v_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{\overline{X_i}}^- - v_i)\vec{e_i}) - \left(\frac{v_{X_i}^+ - u_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{u} + (v_{\overline{X_i}}^- - u_i)\vec{e_i}) \\ &+ \left(\frac{v_i - v_{\overline{X_i}}}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{X_i}^+ - v_i)\vec{e_i}) - \left(\frac{u_i - v_{\overline{X_i}}}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{u} + (v_{X_i}^+ - u_i)\vec{e_i}) \\ &\geq \left(\frac{v_{X_i}^+ - v_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{\overline{X_i}}^- - v_i)\vec{e_i}) - \left(\frac{v_{X_i}^+ - u_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{\overline{X_i}}^- - v_i)\vec{e_i}) \\ &+ \left(\frac{v_i - v_{\overline{X_i}}}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{X_i}^+ - v_i)\vec{e_i}) - \left(\frac{u_i - v_{\overline{X_i}}}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{X_i}^+ - v_i)\vec{e_i}) \\ &= \left(\frac{v_i^+ - v_i - v_{X_i}^+ + u_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{\overline{X_i}}^- - v_i)\vec{e_i}) \\ &+ \left(\frac{v_i - v_{\overline{X_i}}^- - u_i + v_{\overline{X_i}}^-}{v_{\overline{X_i}}^+}\right) S(\vec{v} + (v_{\overline{X_i}}^+ - v_i)\vec{e_i}) \\ &= \left(\frac{u_i - v_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{\overline{X_i}}^- - v_i)\vec{e_i}) - \left(\frac{v_i - u_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{\overline{X_i}}^- - v_i)\vec{e_i}) \\ &= \left(\frac{v_i - u_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{\overline{X_i}}^+ - v_i)\vec{e_i}) - \left(\frac{v_i - u_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) S(\vec{v} + (v_{\overline{X_i}^- - v_i)\vec{e_i}) \\ &= \left(\frac{v_i - u_i}{v_{X_i}^+ - v_{\overline{X_i}}^-}\right) \left[S(\vec{v} + (v_{X_i}^+ - v_i)\vec{e_i}) - S(\vec{v} + (v_{\overline{X_i}^- - v_i)\vec{e_i})\right] \\ &\geq 0. \end{split}$$

Therefore,  $\hat{S}$  is non-decreasing. Suppose  $S(\vec{1} - (1 - u_i)\vec{e_i}) = u_i$  for all  $u_i \in X_i$ . We will prove that  $\hat{S}(\vec{1} - (1 - u_i)\vec{e_i}) = u_i$  for all  $u_i \in \mathbb{I}$ . Let  $u_i \in \mathbb{I}$ . If  $u_i \in X_i$ , then we are done. In the case of  $u_i \in \mathbb{I} \setminus X_i$ , consider

$$\begin{split} \hat{S}(\vec{1} - (1 - u_i)\vec{e_i}) &= \left(\frac{u_{X_i}^+ - u_i}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{1} - (1 - u_{X_i}^-)\vec{e_i}) \\ &+ \left(\frac{u_i - u_{X_i}^-}{u_{X_i}^+ - u_{X_i}^-}\right) S(\vec{1} - (1 - u_{X_i}^+)\vec{e_i}) \\ &= \left(\frac{u_{X_i}^+ - u_i}{u_{X_i}^+ - u_{X_i}^-}\right) u_{X_i}^- + \left(\frac{u_i - u_{X_i}^-}{u_{X_i}^+ - u_{X_i}^-}\right) u_{X_i}^+ \\ &= \left(\frac{u_{X_i}^+ u_{X_i}^- - u_i u_{X_i}^-}{u_{X_i}^+ - u_{X_i}^-}\right) + \left(\frac{u_i u_{X_i}^+ - u_{X_i}^- u_{X_i}^+}{u_{X_i}^+ - u_{X_i}^-}\right) \\ &= \frac{u_i (u_{X_i}^+ - u_{X_i}^-)}{u_{X_i}^+ - u_{X_i}^-} = u_i. \end{split}$$

From Lemma 4.10, we can extend  $X_j$  to  $\mathbb{I}$  for all  $j = 1, 2, \ldots, k$  by the same method.

**Theorem 4.11.** Let S be non-decreasing such that  $S(\vec{1} - (1 - u_i)\vec{e_i}) = u_i$  where  $u_i \in A_i$ and  $A_i \subseteq \mathbb{I}$  for all i = 1, 2, ..., k. Then S can be extended to a semi-copula.

*Proof.* Direct proof by using Lemma 4.9 and 4.10.

From previous results, we are now able to show the proof of Theorem 3.3.

Proof of Theorem 3.3. Fixed  $i \in \{1, 2, ..., k\}$ . Since  $1 \in Range(F_j)$  for all j, we get  $F_jF_j^* = 1 = F_j(1)$  by Lemma 4.3. Let  $u \in \mathbb{I}$ . By assumption,

$$A(F_1^*(1),\ldots,F_{i-1}^*(1),u,F_{i+1}^*(1),\ldots,F_k^*(1)) = A(1 - (1 - F_i^*(1))\vec{e}_i) = F_i(u).$$

By Lemma 4.5, we have  $A(\sum_{j=1}^{k} F_{j}^{*}(1)\vec{e_{j}} + (F_{i}^{*}(u_{i}) - F_{i}^{*}(1))\vec{e_{i}}) = u_{i}$  for all  $u_{i} \in Range(F_{i})$ . Define  $\hat{S} : \prod_{j=1}^{k} A_{j} \to \mathbb{I}$  by  $\hat{S}(u_{1}, \ldots, u_{k}) = A(F_{1}^{*}(u_{1}), \ldots, F_{k}^{*}(u_{k}))$  for all  $u_{j} \in Range(F_{j})$ . By Lemma 4.6, we get  $\hat{S}$  is non-decreasing with  $\hat{S}(\vec{1} - (1 - u)\vec{e_{j}}) = u_{j}$  for all  $u_{j} \in Range(F_{j})$ . By Theorem 4.11,  $\hat{S}$  can be extended to a semi-copula S. Furthermore, by Lemma 4.3, we get  $F_{j}F_{j}^{*}F_{j}(u_{j}) = F_{j}(u_{j})$  for all j. Therefore,

$$S(F_1(u_1), \dots, F_k(u_k)) = A(F_1F_1^*(u_1), \dots, F_kF_k^*(u_k))$$
  
=  $A(u_1, F_2F_2^*(u_2), \dots, F_kF_k^*(u_k))$   
=  $A(u_1, u_2, F_3F_3^*(u_3), \dots, F_kF_k^*(u_k))$   
:  
=  $A(u_1, \dots, u_k).$ 

## 5. Conclusions

In this work, we resemble Sklars construction of multivariate distribution functions by replacing copulas with semi-copulas. We find necessary and sufficient conditions of aggregation functions that can be constructed by composing between semi-copulas and tuples of non-decreasing univariate functions. Furthermore, we prove that almost all aggregation functions can be constructed by this way. This construction method can be used to construct all strictly increasing aggregation functions. Also, all aggregation functions can be approximated by aggregation functions in our form.

During the revision process, we also found that one of our results (Theorem 3.3) is similar to [Scarsini [25], Theorem 9] (see also [26]) where the latter is proved for distribution functions of supermodular fuzzy measures instead of general aggregation functions. One interesting question is whether an aggregation function satisfied (3.2) can be used to construct a supermodular fuzzy measure. Hopefully, this can be answered in the future work.

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#### References

 G. Beliakov, A. Pradera, T. Calvo, Aggregation Functions: A Guide for Practitioners, Springer, Berlin, Germany, 2007.

- [2] M. Grabisch, J.-L. Marichal, R. Mesiar, E. Pap, Aggregation Functions, volume 127. Cambridge University Press, 2009.
- [3] V. Torra, Y. Narukawa, Modeling Decisions: Information Fusion and Aggregation Operators, Springer Science & Business Media, 2007.
- [4] J.J. Arias-García, R. Mesiar, E.P. Klement, S. Saminger-Platz, B. De Baets, Extremal Lipschitz continuous aggregation functions with a given diagonal section, Fuzzy Set Syst. 346 (2018) 147–167.
- [5] T. Jwaid, B. De Baets, H. De Meyer, Biconic aggregation functions, Inform. Sci. 187 (2012) 129–150.
- [6] T. Jwaid, B. De Baets, J. Kalická, R. Mesiar, Conic aggregation functions, Fuzzy Set Syst. 167 (1) (2011) 3–20.
- [7] G. Beliakov, Construction of aggregation functions from data using linear programming, Fuzzy Set Syst. 160 (1) (2009) 65–75.
- [8] P. Boonmee, P. Chanthorn, Quadratic transformations of multivariate semi-copulas, Thai J. Math. 18 (4) (2020) 1917–1931.
- [9] O. Csiszár, J. Fodor, Threshold constructions of aggregation functions, In 2013 IEEE 9th International Conference on Computational Cybernetics (ICCC) IEEE (2013), 191–194.
- [10] A.F. Roldán López de Hierro, C. Roldán, H. Bustince, J. Fernández, I. Rodriguez, H. Fardoun, J. Lafuente, Affine construction methodology of aggregation functions, Fuzzy Set Syst. (2020) 146–164.
- [11] F. Kouchakinejad, A. Sipošová, J. Siráň, Aggregation functions with given superadditive and sub-additive transformations, Int J Gen Syst. 46 (3) (2017) 225–234.
- [12] S. Tasena, Characterization of quadratic aggregation functions, IEEE Trans Fuzzy Syst. 27 (4) (2019) 824–829.
- [13] S. Tasena, Polynomial copula transformations, International Journal of Approximate Reasoning 107 (2019) 65–78.
- [14] P. Boonmee, S. Tasena, Quadratic transformation of multivariate aggregation functions, Depend Model. 8 (1) (2020) 254–261.
- [15] F. Durante, S. Saminger-Platz, P. Sarkoci, On representations of 2-increasing binary aggregation functions, Information Sciences 178 (23) (2008) 4534–4541.
- [16] M. Grabisch, J.-L.Marichal, R. Mesiar, E. Pap, Aggregation functions: Construction methods, conjunctive, disjunctive and mixed classes, Inform Sci. 181 (1) (2011) 23 -43.
- [17] A. Kolesároá, R. Mesiar, J. Kalická, On a new construction of 1-lipschitz aggregation functions, quasi-copulas and copulas, Fuzzy Set Syst. 226 (2013) 19–31.
- [18] A.H. Altalhi, J.I. Forcén, M. Pagola, E. Barrenechea, H. Bustince, Z. Takáč, Moderate deviation and restricted equivalence functions for measuring similarity between data, Inform Sci. 501 (2019) 19–29.
- [19] H. Bustince, G. Beliakov, G.P. Dimuro, B. Bedregal, R. Mesiar, On the definition of penalty functions in data aggregation, Fuzzy Set Syst. 323 (2017) 1–18.
- [20] M. Decký, R. Mesiar, A. Stupňanová, Aggregation functions based on deviations, In International Conference on Information Processing and Management of Uncertainty

in Knowledge-Based Systems, Springer, Cham. (2018), 151–159.

- [21] M. Decký, R. Mesiar, A. Stupňanová, Deviation-based aggregation functions, Fuzzy Set Syst. 332 (2018) 29–36.
- [22] M. Gagolewski, Penalty-based aggregation of multidimensional data, Fuzzy Set Syst. 325 (2017) 4–20.
- [23] P. Tongjundee, S. Tasena, Aggregation functions based on quadratic deviations, In the 24th Annual Meeting in Mathematics (AMM 2019) (2020) 67–77.
- [24] A. Sklar, Fonctions de rèpartition á n dimensions et leurs marges. Publ. Inst. Statist. Univ. Paris 8 (1959) 229–231.
- [25] M. Scarsini, Copula of capacities on product spaces, In: Distribution Functions with Fixed Marginals and Related Topics (L. Ruschendorf, B. Schweizer, and M. D. Taylor, eds.), Institute of Mathematical Statistics (Lecture Notes – Monograph Series, 28 (1996), 307–318.
- [26] F. Durante, J.J. Quesada-Molina, C. Sempi, Semicopulas: characterizations and applicability, Kybernetika 42 (3) (2006) 287–302.