## k-intolerant capacities and Choquet integrals

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### Abstract

We define an aggregation function to be (at most) k-intolerant if it is bounded from above by its kth lowest input value. Applying this definition to the discrete Choquet integral and its underlying capacity, we introduce the concept of k-intolerant capacities which, when varying kfrom 1 to n, cover all the possible capacities on n objects. as the concepts of k-additive capacities and p-symmetric capacities have been previously introduced essentially to overcome the problem of computational complexity of capacities, k-intolerant capacities are proposed here for the same purpose but also for dealing with intolerant or tolerant behaviors of aggregation. We also introduce axiomatically indices to appraise the extent to which a given capacity is k-intolerant.

**Keywords:** multi-criteria analysis, interacting criteria; capacities; Choquet integral.

### 1 Introduction

In a previous work [9] the author investigated the intolerant behavior of the discrete Choquet integral when used to aggregate interacting criteria. Roughly speaking, the Choquet integral  $C_v$ , or equivalently its associated capacity v, has an intolerant behavior if its output (aggregated) value is often close to the lowest of its input values. More precisely, consider the domain  $[0,1]^n$  of  $\mathcal{C}_v$  as a probability space, with uniform distribution, and the mathematical expectation of  $C_v$ , which expresses the typical position of  $C_v$  within the unit interval. A low expectation then means that the Choquet integral is rather intolerant and behaves nearly like the minimum on average. Similarly, a high expectation means that the Choquet integral is rather tolerant and behaves nearly like the maximum on average. Note that such an analysis is meaningless when criteria are independent since, in that case, the Choquet integral boils down to a weighted arithmetic mean whose expectation is always one half (neither tolerant nor intolerant.)

In this paper we pursue this idea by defining k-intolerant Choquet integrals. The case k=1 corresponds to the unique most intolerant Choquet integral, namely the minimum. The case k=2 corresponds to the subclass of n-variable Choquet integrals that are bounded from above by their second lowest input values. Those Choquet integrals are more or less intolerant but not as much as the minimum.

More generally, denoting by  $x_{(1)}, \ldots, x_{(n)}$  the order statistics resulting from reordering  $x_1, \ldots, x_n$  in the nondecreasing order, we say that an n-variable Choquet integral  $C_v$ , or equivalently its underlying capacity v, is at most k-intolerant if

$$C_v(x) \le x_{(k)} \qquad (x \in [0,1]^n) \tag{1}$$

and it is exactly k-intolerant if, in addition,

there is  $x^* \in [0,1]^n$  such that  $C_v(x^*) > x^*_{(k-1)}$ , with convention that  $x_{(0)} := 0$ .

Interestingly, condition (1) clearly implies that the output value of  $C_v$  is zero whenever at least k input values are zeros. We will see in Section 3 that the converse holds true as well.

At first glance, defining k-intolerant aggregation functions may appear as a pure mathematical exercise without any real application behind. In fact, in many real-life decision problems, experts or decision-makers are or must be intolerant. This is often the case when, in a given selection problem, we search for most qualified candidates among a wide population of potential alternatives. It is then sensible to reject every candidate which fails at least k criteria.

**Example 1.1.** Consider a (simplified) problem of selecting candidates applying for a university permanent position and suppose that the evaluation procedure is handled by appointed expert-consultants on the basis of the following academic selection criteria:

- 1. Scientific value of curriculum vitae,
- 2. Teaching effectiveness,
- 3. Ability to supervise staff and work in a team environment,
- 4. Ability to easily communicate in English,
- 5. Work experience in the industry,
- 6. Recommendations by faculty and other individuals.

Assume also that one of the rules of the evaluation procedure states that the complete failure of any two of these criteria results in automatic rejection of the applicant. This quite reasonable rule forces the Choquet integral, when used for the aggregation procedure, to be 2-intolerant, thus restricting the class of possible Choquet integrals for such a selection problem.

On the other hand, there are real-life situations where it is recommended to be tolerant, especially if the criteria are hard to meet simultaneously and if the potential alternatives are not numerous. To deal with such situations, we introduce k-tolerant aggregation functions and we will say that an n-variable Choquet integral  $C_v$ , or equivalently its underlying capacity v, is at most k-tolerant if

$$C_v(x) \ge x_{(n-k+1)}$$
  $(x \in [0,1]^n).$ 

In that case, the output value of  $C_v$  is one whenever at least k input values are ones.

**Example 1.2.** Consider a family who consults a Real Estate agent to buy a house. The parents propose the following house buying criteria:

- 1. Close to a school,
- 2. With parks for their children to play in,
- 3. With safe neighborhood for children to grow up in,
- 4. At least 100 meters from the closest major road,
- 5. At a fair distance from the nearest shopping mall,
- 6. Within reasonable distance of the airport.

Feeling that it is likely unrealistic to satisfy all six criteria simultaneously, the parents are ready to accept a house that would fully succeed any five over the six criteria. If a 6variable Choquet integral is used in this selection problem, it must be 5-tolerant.

Considering k-intolerant and k-tolerant capacities can also be viewed as a way to make real applications easier to model from a computational viewpoint. Those "simplified" capacities indeed require less parameters than classical capacities (actually  $O(n^{k-1})$  parameters instead of  $O(2^n)$ ; see Section 3). Moreover, when varying k from 1 to n, we clearly recover all the possible capacities on n objects.

Notice however that this idea of partitioning capacities into subclasses is not new. Grabisch [4] proposed the k-additive capacities,

which gradually cover all the possible capacities starting from additive capacities (k = 1). Later, Miranda et al. [10] introduced the psymmetric capacities, also covering the possible capacities but starting from symmetric capacities (p = 1). Note also that other approaches to overcome the exponential complexity of capacities have also been previously proposed in the literature: Sugeno  $\lambda$ measures [12], \perp -decomposable measures (see e.g. [6]), hierarchically decomposable measures [13], distorted probabilities (see e.g. [11]) to name a few.

It is also noteworthy that, in a given multicriteria sorting or ordering procedure, when the capacity must be learnt from a set of examples, it is sometimes interesting or even recommended to restrict the admissible capacities to k-intolerant capacities <sup>1</sup>, starting from k = 1 and incrementing this value until a solution is found. This makes it possible to simplify the aggregation model as much as possible while keeping an interpretation of the solution.

The outline of the paper is as follows. In Section 2 we introduce and formalize the concepts of k-intolerance and k-tolerance for arbitrary aggregation functions. In Section 3 we apply these concepts to the Choquet integral, thus introducing the k-intolerant and k-tolerant capacities. In Section 4 we investigate the veto and favor indices when used with those particular capacities. Finally, in Section 5 we axiomatically introduce new indices measuring the extent to which the Choquet integral is at most k-intolerant or k-tolerant.

#### 2 Basic definitions

Let  $F:[0,1]^n \to [0,1]$  be an aggregation function. By considering the cube  $[0,1]^n$  as a probability space with uniform distribution, we can compute the mathematical expectation of F, that is,

$$E(F) := \int_{[0,1]^n} F(x) \, dx. \tag{2}$$

This value gives the average position of Fwithin the interval [0, 1].

When F is internal (i.e.,  $\min \leq F \leq \max$ ) then it is convenient to rescale E(F) within the interval  $[E(\min), E(\max)]$ . This leads to the following normalized and mutually complementary values [1, 9]:

and 
$$\operatorname{ness}(F) := \frac{E(\max) - E(F)}{E(\max) - E(\min)}$$
 (3)  
 $\operatorname{orness}(F) := \frac{E(F) - E(\min)}{E(\max) - E(\min)}$  (4)

orness(F) := 
$$\frac{E(F) - E(\min)}{E(\max) - E(\min)}$$
(4)

Thus defined, the degree of andness (resp. orness) of F represents the degree or intensity (between 0 and 1) to which the average value of F is close to that of "min" (resp. "max"). In some sense, it also reflects the extent to which F behaves like the minimum (resp. the maximum) on average.

Define the kth order statistic function  $OS_k$ :  $[0,1]^n \to [0,1]$  as

$$OS_k(x) = x_{(k)}$$
  $(x \in [0, 1]^n),$ 

where  $x_{(k)}$  is the kth lowest coordinate of x.

**Definition 2.1.** Let  $k \in \{1,\ldots,n\}$ . An aggregation function  $F: [0,1]^n \to [0,1]$  is at most k-intolerant if  $F \leq OS_k$ . It is kintolerant if, in addition,  $F \nleq \mathrm{OS}_{k-1}$ , where  $OS_0 := 0$  by convention.

It follows immediately from this definition that, for any k-intolerant function F, we have  $E(F) \leq E(OS_k)$  and, if F is internal, we have  $andness(F) \ge andness(OS_k)$  and  $orness(F) \le$ orness( $OS_k$ ).

By duality, we can also introduce k-tolerant functions as follows:

**Definition 2.2.** Let  $k \in \{1, \ldots, n\}$ . An aggregation function  $F: [0,1]^n \rightarrow [0,1]$  is at most k-tolerant if  $F \geq OS_{n-k+1}$ . It is ktolerant if, in addition,  $F \ngeq OS_{n-k+2}$ , where  $OS_{n+1} := 1$  by convention.

It is immediate to see that when a function  $F: [0,1]^n \rightarrow [0,1]$  is k-intolerant, its dual  $F^*: [0,1]^n \to [0,1]$ , defined by

$$F^*(x_1, \dots, x_n) := 1 - F(1 - x_1, \dots, 1 - x_n)$$
(5)

<sup>&</sup>lt;sup>1</sup>or k-tolerant, or k-additive, etc., according to the feeling of the decision maker.

is k-tolerant and vice versa.

In the next section we investigate the particular case where F is the Choquet integral and we define the concepts of k-intolerant and k-tolerant capacities.

# 3 Case of Choquet integrals and capacities

The use of the Choquet integral has been proposed by many authors as an adequate substitute to the weighted arithmetic mean to aggregate interacting criteria; see e.g. [2, 7]. In the weighted arithmetic mean model, each criterion is given a weight representing the importance of this criterion in the decision. In the Choquet integral model, where criteria can be dependent, a capacity is used to define a weight on each combination of criteria, thus making it possible to model the interaction existing among criteria.

Let us first recall the formal definitions of these concepts. Throughout, we will use the notation  $N := \{1, \ldots, n\}$  for the set of criteria.

**Definition 3.1.** A capacity on N is a set function  $v: 2^N \to [0,1]$ , that is nondecreasing with respect to set inclusion and such that  $v(\emptyset) = 0$  and v(N) = 1.

**Definition 3.2.** Let v be a capacity on N. The *Choquet integral* of  $x: N \to \mathbb{R}$  with respect to v is defined by

$$C_v(x) := \sum_{i=1}^n x_{(i)} \left[ v(A_{(i)}) - v(A_{(i+1)}) \right],$$

permutation on N such that  $x_{(1)} \leq \ldots \leq x_{(n)}$ . Furthermore  $A_{(i)} := \{(i), \ldots, (n)\}$  and  $A_{(n+1)} := \varnothing$ .

In this section we apply the ideas of k-intolerance and k-tolerance to the Choquet integral. Since this integral is internal, it can be seen as a function from  $[0,1]^n$  to [0,1].

It is useful here to remember the following identity (see e.g. [8])

$$v(T) = \mathcal{C}_v(\mathbf{1}_T) \qquad (T \subseteq N),$$

where  $\mathbf{1}_T$  denotes the characteristic vector of subset T in  $\{0,1\}^n$ .

Let us denote by  $\mathcal{F}_N$  the set of all capacities on N. The following proposition gives equivalent conditions for a Choquet integral to be at most k-intolerant.

**Proposition 3.1.** Let  $k \in \{1, ..., n\}$  and  $v \in \mathcal{F}_N$ . Then the following assertions are equivalent:

- i)  $C_v(x) \le x_{(k)} \ \forall x \in [0,1]^n$ ,
- ii)  $v(T) = 0 \ \forall T \subseteq N \ such \ that |T| \le n k,$
- iii)  $C_v(x) = 0 \ \forall x \in [0,1]^n \ such \ that$  $x_{(k)} = 0,$
- iv)  $C_v(x)$  is independent of  $x_{(k+1)}, \ldots, x_{(n)}$ ,
- v)  $\exists \lambda \in [0,1) \text{ such that } \forall x \in [0,1]^n \text{ we}$ have  $x_{(k)} \leq \lambda \Rightarrow C_v(x) \leq \lambda.$

As we can see, some assertions of Proposition 3.1 are natural and can be interpreted easily. Some others are more surprising and show that the Choquet integral may have an unexpected behavior.

First, assertion (ii) enables us to define k-intolerant capacities as follows:

**Definition 3.3.** Let  $k \in \{1, ..., n\}$ . A capacity  $v \in \mathcal{F}_N$  is k-intolerant if v(T) = 0 for all  $T \subseteq N$  such that  $|T| \le n - k$  and there is  $T^* \subseteq N$ , with  $|T^*| = n - k + 1$ , such that  $v(T^*) \ne 0$ .

Assertion (iii) says that the output value of the Choquet integral is zero whenever at least k input values are zeros. This is actually a straightforward consequence of k-intolerance.

Assertion (iv) is more surprising. It says that the output value of the Choquet integral does not take into account the values of  $x_{(k+1)}, \ldots, x_{(n)}$ . Back to Example 1.1, only the two lowest scores are taken into account to provide a global evaluation, regardless of the other scores.

Assertion (v) is also of interest. By imposing that  $C_v(x) \leq \lambda$  whenever  $x_{(k)} \leq \lambda$  for a given threshold  $\lambda \in [0,1)$ , we necessarily force  $C_v$  to be at most k-intolerant. For instance, consider the problem of evaluating students with respect to different courses and suppose that it is decided that if the lowest k marks obtained by a student are less than 18/20 then his/her global mark must be less than 18/20. In this case, the Choquet integral utilized is

at most k-intolerant.

Proposition 3.1 can be easily rewritten for ktolerance by considering the dual  $C_v^*$  of the
Choquet integral  $C_v$  as defined in Eq. (5). On
this issue, Grabisch et al. [5, §4] showed that
the dual  $C_v^*$  of  $C_v$  is the Choquet integral  $C_{v^*}$ defined from the dual capacity  $v^*$ , which is
constructed from v by

$$v(T) = 1 - v(N \setminus T)$$
  $(T \subseteq N).$ 

We then have

$$C_v \ge OS_{n-k+1} \Leftrightarrow C_{v^*} \le OS_k.$$

**Proposition 3.2.** Let  $k \in \{1, ..., n\}$  and  $v \in \mathcal{F}_N$ . Then the following assertions are equivalent:

- i)  $C_v(x) \ge x_{(n-k+1)} \ \forall x \in [0,1]^n,$
- ii)  $v(T) = 1 \ \forall T \subseteq N \ such \ that \ |T| \ge k$ ,
- iii)  $C_v(x) = 1 \ \forall x \in [0,1]^n \ such \ that$  $x_{(n-k+1)} = 1,$
- iv)  $C_v(x)$  is independent of  $x_{(1)}, \ldots, x_{(n-k)}$ ,
- v)  $\exists \lambda \in (0,1] \text{ such that } \forall x \in [0,1]^n \text{ we}$  $have \ x_{(n-k+1)} \ge \lambda \implies C_v(x) \ge \lambda.$

Here again, some assertions are of interest. First, assertion (ii) enables us to define k-tolerant capacities as follows:

**Definition 3.4.** Let  $k \in \{1, ..., n\}$ . A capacity  $v \in \mathcal{F}_N$  is k-tolerant if v(T) = 1 for all  $T \subseteq N$  such that  $|T| \ge k$  and there is  $T^* \subseteq N$ , with  $|T^*| = k - 1$ , such that  $v(T^*) \ne 1$ .

Assertion (iii) says that the output value of the Choquet integral is one whenever at least k input values are ones.

Assertion (iv) says that the output value of the Choquet integral does not take into account the values of  $x_{(1)}, \ldots, x_{(n-k)}$ . As an application, consider students who are evaluated according to n homework assignments and assume that the evaluation procedure states that the two lowest homework scores of each student are dropped, which implies that each student can miss two homework assignments without affecting his/her final grade. If a n-variable Choquet integral is used to aggregate the homework scores, it should not take  $x_{(1)}$  and  $x_{(2)}$  into consideration and hence it is at most (n-2)-tolerant.

### 4 Links with veto and favor indices

Definition of k-intolerant aggregation functions (cf. Definition 2.1) is actually inspired from the following concept of veto criterion, which was introduced in multi-criteria decision-making by Grabisch [3]:

Let  $F: [0,1]^n \to [0,1]$  be an arbitrary aggregation function. A criterion  $j \in N$  is said to be a *veto* for F if

$$F(x) \le x_i$$
  $(x \in [0,1]^n).$ 

Even though this definition resembles that of k-intolerance, it involves only one criterion. Clearly, the failure of this criterion necessarily entails a low global score.

Similarly, a criterion  $j \in N$  is a favor for F if

$$F(x) \ge x_i$$
  $(x \in [0,1]^n).$ 

Here the satisfaction of a favor criterion entails a high global score.

When F is the Choquet integral, analog versions of Propositions 3.1 and 3.2 can be easily obtained (see also [9,  $\S4$ ]):

**Proposition 4.1.** Let  $j \in N$  and  $v \in \mathcal{F}_N$ . Then the following assertions are equivalent:

- i)  $C_v(x) \leq x_i \ \forall x \in [0,1]^n$ ,
- ii)  $v(T) = 0 \ \forall T \subseteq N \ such \ that \ T \not\ni j,$
- iii)  $C_v(x) = 0 \ \forall x \in [0,1]^n \ such \ that \ x_j = 0,$
- iv)  $C_v(x)$  is independent of  $x_i$   $(i \in N \setminus \{j\})$ whenever  $x_i \ge x_j$ ,
- v)  $\exists \lambda \in [0,1) \text{ such that } \forall x \in [0,1]^n \text{ we}$ have  $x_j \leq \lambda \Rightarrow \mathcal{C}_v(x) \leq \lambda$ .

**Proposition 4.2.** Let  $j \in N$  and  $v \in \mathcal{F}_N$ . Then the following assertions are equivalent:

- i)  $C_v(x) \ge x_i \ \forall x \in [0,1]^n$ ,
- ii)  $v(T) = 1 \ \forall T \subseteq N \ such \ that \ T \ni j,$
- iii)  $C_v(x) = 1 \ \forall x \in [0,1]^n \ such \ that \ x_j = 1,$
- iv)  $C_v(x)$  is independent of  $x_i$   $(i \in N \setminus \{j\})$ whenever  $x_i \leq x_j$ ,
- v)  $\exists \lambda \in (0,1] \text{ such that } \forall x \in [0,1]^n \text{ we}$  $have \ x_i \geq \lambda \implies \mathcal{C}_v(x) \geq \lambda.$

Since they present rather extreme behaviors, veto and favor criteria rarely occur in practical applications. It is then natural to wonder if one can define indices measuring the intensity (between 0 and 1) to which a given criterion  $j \in N$  behaves like a veto or a favor for the Choquet integral  $C_v$ .

Considering again  $x \in [0,1]^n$  as a multidimensional random variable uniformly distributed, we could propose to define such indices as

$$veto(\mathcal{C}_v, j) := \Pr[\mathcal{C}_v(x) \le x_j \mid x \in [0, 1]^n]$$
  
$$favor(\mathcal{C}_v, j) := \Pr[\mathcal{C}_v(x) \ge x_j \mid x \in [0, 1]^n]$$

Unfortunately, as pointed out in [9, §4], these definitions lead to rather intricate formulas, which are not even continuous with respect to the capacity v.

Alternative indices have been proposed axiomatically by the author [9, §4] as follows

$$veto(C_{v}, j) := 1 - \frac{1}{n-1} \sum_{t=0}^{n-1} \frac{1}{\binom{n-1}{t}} \sum_{\substack{T \subseteq N \setminus \{j\} \\ |T| = t}} v(T)$$

favor
$$(C_v, j) := \frac{1}{n-1} \sum_{t=0}^{n-1} \frac{1}{\binom{n-1}{t}} \sum_{T \subseteq N \setminus \{j\}} v(T \cup \{j\}) - \frac{1}{n-1}$$

Besides the advantage of being linear in terms of v, these indices can have a straightforward interpretation, as we will now show.

For any aggregation function  $F:[0,1]^n \to [0,1]$ , we define the *conditional expectation* of F given  $x_j = \lambda \in \{0,1\}$  as

$$E(F \mid x_j = \lambda) := \int_{[0,1]^n} F(x \mid x_j = \lambda) \, dx.$$
(6)

Now, by rewriting Eqs. (3) and (4) by means of the conditional expectation (6) instead of the classical expectation (2), we naturally define the *conditional* andness and orness degrees given  $x_j = \lambda$ . These new definitions then enable us to easily rewrite the veto and favor indices as in Eqs. (7) and (8) below, which shows that these indices somehow represent the intensity to which assertions (iii) in Propositions 4.1 and 4.2 are true.

**Proposition 4.3.** For any  $v \in \mathcal{F}_N$  and any  $j \in N$ , we have

$$\operatorname{veto}(C_v, j) = \operatorname{andness}(C_v \mid x_j = 0), (7)$$
  
 $\operatorname{favor}(C_v, j) = \operatorname{orness}(C_v \mid x_j = 1). (8)$ 

Now, let us investigate the behavior of veto and favor indices when used with k-intolerant and k-tolerant capacities.

If  $C_v$  is k-intolerant then, from Eqs. (7) and (8), it follows immediately that, for any  $j \in N$ ,

$$\text{veto}(C_v, j) \ge \text{veto}(OS_k, j)$$

and

$$favor(C_v, j) \leq favor(OS_k, j),$$

which shows that any criterion is more a veto for  $C_v$  than for  $OS_k$  and less a favor for  $C_v$  than for  $OS_k$ .

Similarly, if  $C_v$  is k-tolerant then

$$\operatorname{veto}(\mathcal{C}_v, j) \le \operatorname{veto}(\operatorname{OS}_{n-k+1}, j)$$

and

$$favor(C_v, j) \ge favor(OS_{n-k+1}, j),$$

with similar interpretations.

Of course, these latter four inequalities are tight for  $C_v = OS_k$  and  $C_v = OS_{n-k+1}$ , respectively.

### 5 Intolerance and tolerance indices

Exactly as for veto and favor phenomena, it is legitimate to wonder how we could define an index measuring the degree to which a given Choquet integral or its capacity is at most k-intolerant or at most k-tolerant. Again, we can think of the probabilities

$$\operatorname{intol}_{\leq k}(\mathcal{C}_{v}) := \Pr[\mathcal{C}_{v}(x) \leq x_{(k)} \mid x \in [0, 1]^{n}],$$
  
 $\operatorname{tol}_{\leq k}(\mathcal{C}_{v}) := \Pr[\mathcal{C}_{v}(x) \geq x_{(n-k+1)} \mid x \in [0, 1]^{n}]$   
 $= \operatorname{intol}_{\leq k}(\mathcal{C}_{v^{*}}),$ 

which lead to nonlinear formulas.

Alternatively, we can proceed as in Proposition 4.3 and hence focus on assertions (*iii*) of Propositions 3.1 and 3.2.

First, define the following conditional expectations:

$$E(\mathcal{C}_{v} \mid x_{(k)} = 0) := \frac{1}{\binom{n}{k}} \sum_{\substack{K \subseteq N \\ |K| = k}} E(\mathcal{C}_{v} \mid x \mathbf{1}_{K} = 0 \mathbf{1}_{K}), \quad (9)$$

$$E(\mathcal{C}_{v} \mid x_{(n-k+1)} = 1) := \frac{1}{\binom{n}{k}} \sum_{\substack{K \subseteq N \\ |K| = k}} E(\mathcal{C}_{v} \mid x \mathbf{1}_{K} = \mathbf{1}_{K}) \quad (10)$$

$$= 1 - E(\mathcal{C}_{v^{*}} \mid x_{(k)} = 0).$$

These definition are based on the idea that condition  $x_{(k)} = 0$  (resp.  $x_{(n-k+1)} = 1$ ) means that at least k coordinates of x are zeros (resp. ones).

Next, one can show that

$$E(C_v \mid x_{(k)} = 0) = \frac{1}{n - k + 1} \sum_{t=0}^{n-k} \frac{1}{\binom{n}{t}} \sum_{\substack{T \subseteq N \\ |T| = t}} v(T),$$

$$E(C_v \mid x_{(n-k+1)} = 1) = \frac{1}{n - k + 1} \sum_{t=k}^{n} \frac{1}{\binom{n}{t}} \sum_{\substack{T \subseteq N \\ |T| = t}} v(T),$$

with particular cases (when k = n)

$$E(C_v \mid x_{(n)} = 0) = 0$$
 and  $E(C_v \mid x_{(1)} = 1) = 1$ .

Finally, by rewriting Eqs. (3) and (4) by means of the conditional expectations (9) and (10), we obtain the following conditional andness and orness degrees (for  $k \neq n$ ):

andness(
$$C_v \mid x_{(k)} = 0$$
) =
$$1 - \frac{1}{n-k} \sum_{t=0}^{n-k} \frac{1}{\binom{n}{t}} \sum_{\substack{T \subseteq N \\ |T| = t}} v(T)$$
orness( $C_v \mid x_{(n-k+1)} = 1$ ) =
$$\frac{1}{n-k} \sum_{t=k}^{n} \frac{1}{\binom{n}{t}} \sum_{\substack{T \subseteq N \\ |T| = t}} v(T) - \frac{1}{n-k}$$

Hence the following definition of intolerance and tolerance indices, constructed in the spirit of assertions (*iii*) of Propositions 3.1 and 3.2:

**Definition 5.1.** For any  $k \in \{1, ..., n-1\}$ , we define the k-intolerance and k-tolerance indices, respectively, as

$$intol_{\leq k}(\mathcal{C}_v) := andness(\mathcal{C}_v \mid x_{(k)} = 0)$$

$$tol_{\leq k}(\mathcal{C}_v) := orness(\mathcal{C}_v \mid x_{(n-k+1)} = 1)$$

$$= andness(\mathcal{C}_{v^*} \mid x_{(k)} = 0).$$

It is clear that  $C_v$  is k-intolerant (resp. k-tolerant) if and only if  $\operatorname{intol}_{\leq k}(C_v) = 1$  (resp.  $\operatorname{tol}_{\leq k}(C_v) = 1$ ). For example, for any  $k \in \{1, \ldots, n-1\}$  and any  $l \in \{1, \ldots, n\}$ , we have

$$\operatorname{intol}_{\leq k}(OS_{l}) = \operatorname{tol}_{\leq k}(OS_{n-l+1}) = 1 - \frac{(l-k)^{+}}{n-k}$$

where  $r^+ := \max(r, 0)$  means the positive part of  $r \in \mathbb{R}$ .

The intolerance and tolerance indices proposed in Definition 5.1 have been defined in a constructive way. To fully justify their use, we need to propose an axiomatic characterization of them. The next result, inspired from [9, Theorem 4.1], deals with this issue.

For any capacity  $v \in \mathcal{F}_N$  and any permutation  $\pi$  on N,  $\pi v$  will denote the capacity of  $\mathcal{F}_N$  defined by  $\pi v(\pi(S)) = v(S)$  for all  $S \subseteq N$ , where  $\pi(S) = \{\pi(i) \mid i \in S\}$ .

**Theorem 5.1.** Let  $k \in \{1, ..., n-1\}$  and consider a family of real numbers  $\{\psi_k(\mathcal{C}_v) \mid v \in \mathcal{F}_N\}$ . These numbers

• are linear w.r.t. the capacity, that is, there exist real constants  $p_T^k$   $(T \subseteq N)$ such that

$$\psi_k(\mathcal{C}_v) = \sum_{T \subseteq N} p_T^k v(T) \qquad (v \in \mathcal{F}_N)$$

• fulfill the "symmetry" axiom, that is, for any permutation  $\pi$  on N, we have

$$\psi_k(\mathcal{C}_v) = \psi_k(\mathcal{C}_{\pi v}) \qquad (v \in \mathcal{F}_N)$$

• fulfill the "boundary" axiom, that is, for any  $l \in \{1, ..., n\}$ , we have

$$\psi_k(OS_l) = 1 - \frac{(l-k)^+}{n-k}$$

(resp. 
$$\psi_k(OS_{n-l+1}) = 1 - \frac{(l-k)^+}{n-k}$$
)

if and only if  $\psi_k(\mathcal{C}_v) = \operatorname{intol}_{\leq k}(\mathcal{C}_v)$  (resp.  $\psi_k(\mathcal{C}_v) = \operatorname{tol}_{\leq k}(\mathcal{C}_v)$ ) for all  $v \in \mathcal{F}_N$ .

The axioms of Theorem 5.1 can be interpreted as follows. As for veto and favor indices, we ask the intolerance and tolerance indices to be linear with respect to the capacity. We also require that these indices be independent of the numbering of criteria. The third axiom is motivated by the following observation: For a fixed  $k \in \{1, \ldots, n-1\}$  the expression  $\psi_k(\mathrm{OS}_l)$  (resp.  $\psi_k(\mathrm{OS}_{n-l+1})$ ) must be

- one, whenever  $1 \le l \le k$ ,
- zero, when l = n (limit condition),
- a decreasing linear expression of l, when  $k \leq l \leq n$ .

### 6 Conclusion

In this paper we have proposed the concepts of k-intolerant and k-tolerant Choquet integrals and capacities. Besides the obvious computational advantage of these concepts (comparable to that of k-additive and p-symmetric capacities), they can be easily interpreted in practical decision problems where the decision makers must be intolerant or tolerant.

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