

# AGGREGATION OPERATORS FOR MULTICRITERIA DECISION AID

by

Jean-Luc Marichal  
University of Liège

# 1. Aggregation in MCDM

Set of alternatives  $A = \{a, b, c, \dots\}$

Set of criteria  $N = \{1, \dots, n\}$ .

For all  $i \in N$ ,  $\omega_i$  = weight associated to criterion  $i$ .

Profile:  $a \in A \rightarrow (x_1^a, \dots, x_n^a) \in E^n$ ,  $E$  = real interval.

$x_i^a$  = partial score of  $a$  w.r.t. criterion  $i$ .

Aggregation operator  $M : E^n \rightarrow F$

Example:  $\text{WAM}_\omega(x) = \sum_{i=1}^n \omega_i x_i$  with  $\sum_{i=1}^n \omega_i = 1$ ,  $\omega_i \geq 0$ .

	criterion 1	...	criterion $n$	global score
alternative $a$	$x_1^a$	...	$x_n^a$	$M(x_1^a, \dots, x_n^a)$
alternative $b$	$x_1^b$	...	$x_n^b$	$M(x_1^b, \dots, x_n^b)$
:	:		:	:

## Phases of multicriteria decision making procedures:

1. Modelling phase: How to construct  $x_i^a$  and  $\omega_i$ ?
2. Aggregation phase: How to build  $M(x_1^a, \dots, x_n^a)$ ?
3. Exploitation phase: Which are the best alternatives?

## Hypotheses:

- The weights  $\omega_i$  are defined according to a cardinal scale
- All the partial scores  $x_i^a$  are commensurable.

## 2. Some aggregation operators

Continuity (Co)

Increasing monotonicity (In):

$$x_i \leq x'_i \quad \forall i \Rightarrow M(x_1, \dots, x_n) \leq M(x'_1, \dots, x'_n)$$

Idempotence (Id):

$$M(x, \dots, x) = x$$

Associativity (A):

$$M(x_1, M(x_2, x_3)) = M(M(x_1, x_2), x_3)$$

$$M(x_1, \dots, x_n, x_{n+1}) = M(M(x_1, \dots, x_n), x_{n+1})$$

An *extended aggregation operator* is a sequence  $M = (M^{(n)})_{n \in \mathbb{N}_0}$  of aggregation operators  $M^{(n)} : E^n \rightarrow F$ . The set of all those sequences is denoted by  $A(E, F)$ .

### Theorem

$M \in A(E, \mathbb{R})$  fulfills (Co, In, Id, A) if and only if there exist  $\alpha, \beta \in E$  such that

$$M^{(n)}(x) = (\alpha \wedge x_1) \vee (\bigvee_{i=2}^{n-1} (\alpha \wedge \beta \wedge x_i)) \vee (\beta \wedge x_n) \vee (\bigwedge_{i=1}^n x_i) \quad \forall n \in \mathbb{N}_0$$

+ Symmetry (Sy)

### Theorem

$M \in A(E, \mathbb{R})$  fulfills (Sy, Co, In, Id, A) if and only if there exists  $\alpha \in E$  such that

$$M^{(n)}(x) = \text{median}(\bigwedge_{i=1}^n x_i, \bigvee_{i=1}^n x_i, \alpha) \quad \forall n \in \mathbb{N}_0.$$

Decomposability (D):

for all  $k \leq n$ ,

$$\begin{aligned} M^{(k)}(x_1, \dots, x_k) &= M^{(k)}(x'_1, \dots, x'_k) \\ &\Downarrow \\ M^{(n)}(x_1, \dots, x_k, x_{k+1}, \dots, x_n) &= M^{(n)}(x'_1, \dots, x'_k, x_{k+1}, \dots, x_n) \end{aligned}$$

Strict increasing monotonicity (SIn)

**Theorem** (Kolmogoroff-Nagumo, 1930)

$M \in A(E, \mathbb{R})$  fulfills (Sy, Co, SIn, Id, D) if and only if there exists a continuous strictly monotonic function  $f : E \rightarrow \mathbb{R}$  such that

$$M^{(n)}(x) = f^{-1}\left[\frac{1}{n} \sum_{i=1}^n f(x_i)\right], \quad n \in \mathbb{N}_0.$$

## Remarks

1. The family of  $M \in A(E, \mathbb{R})$  that satisfy (Sy, Co, In, Id, D) has a rather intricate structure (see §3.2.2).
2. (Sy, Co, SIn, Id, D)  $\Leftrightarrow$  (Co, SIn, Id, SD) (see §3.2.1).

The quasi-linear means (Aczél, 1948):

$$M(x) = f^{-1}\left[\sum_{i=1}^n \omega_i f(x_i)\right], \quad \text{with } \sum_{i=1}^n \omega_i = 1, \quad \omega_i \geq 0.$$

The weighted arithmetic means:

$$\text{WAM}_\omega(x) = \sum_{i=1}^n \omega_i x_i, \quad \text{with } \sum_{i=1}^n \omega_i = 1, \quad \omega_i \geq 0.$$

General bisymmetry (GB):

$M^{(1)}(x) = x$  for all  $x \in E$ , and

$$\begin{aligned} & M^{(p)}(M^{(n)}(x_{11}, \dots, x_{1n}), \dots, M^{(n)}(x_{p1}, \dots, x_{pn})) \\ = & M^{(n)}(M^{(p)}(x_{11}, \dots, x_{p1}), \dots, M^{(p)}(x_{1n}, \dots, x_{pn})) \end{aligned}$$

for all matrices

$$X = \begin{pmatrix} x_{11} & \cdots & x_{1n} \\ \vdots & & \vdots \\ x_{p1} & \cdots & x_{pn} \end{pmatrix} \in E^{p \times n}.$$

Stability for the admissible positive linear transformations (SPL):

$$M(r x_1 + s, \dots, r x_n + s) = r M(x_1, \dots, x_n) + s$$

for all  $x \in E^n$  and all  $r > 0, s \in \mathbb{R}$  such that  $r x_i + s \in E$  for all  $i \in N$ .

We can assume w.l.o.g. that  $E = [0, 1]$

### Theorem

$M \in A([0, 1], \mathbb{R})$  fulfills (In, SPL, GB) if and only if

- either:  $\forall n \in \mathbb{N}_0, \exists S \subseteq \{1, \dots, n\}$  such that  $M^{(n)} = \min_S$ ,
- or:  $\forall n \in \mathbb{N}_0, \exists S \subseteq \{1, \dots, n\}$  such that  $M^{(n)} = \max_S$ ,
- or:  $\forall n \in \mathbb{N}_0, \exists \omega \in [0, 1]^n$  such that  $M^{(n)} = \text{WAM}_\omega$ .

$$\min_S(x) := \bigwedge_{i \in S} x_i \quad \max_S(x) := \bigvee_{i \in S} x_i$$

### Theorem

$M \in A([0, 1], \mathbb{R})$  fulfills (SIn, SPL, GB) if and only if for all  $n \in \mathbb{N}_0$ , there exists  $\omega \in ]0, 1[^n$  such that  $M^{(n)} = \text{WAM}_\omega$ .

### 3. The weighted arithmetic means

$$\text{WAM}_\omega(x) = \sum_{i \in N} \omega_i x_i, \quad \text{with } \sum_{i \in N} \omega_i = 1, \quad \omega_i \geq 0.$$

#### Definition

For any  $S \subseteq N$ , we define  $e_S \in \{0, 1\}^n$  as the binary profile whose  $i$ -th component is 1 iff  $i \in S$ .

We observe that

$$\text{WAM}_\omega(e_{\{i\}}) = \omega_i$$

The weight  $\omega_i$  can be viewed as the global score obtained with the profile  $e_i$

**Additivity (Add):**

$$M(x_1 + x'_1, \dots, x_n + x'_n) = M(x_1, \dots, x_n) + M(x'_1, \dots, x'_n)$$

#### Theorem

$M : [0, 1]^n \rightarrow \mathbb{R}$  fulfills (In, SPL, Add) if and only if there exists  $\omega \in [0, 1]^n$  such that  $M = \text{WAM}_\omega$ .

#### Remark

Weighted arithmetic means can be used only when criteria are “independent” !!!

Example of correlated criteria:

Statistics	Probability	Algebra
0.3	0.3	0.4

# Preferential independence

Let  $x, x'$  be two profiles in  $[0, 1]^n$ .

The profile  $x$  is said to be preferred to the profile  $x'$  ( $x \succeq x'$ ) if  $M(x) \geq M(x')$ .

## Definition

The subset  $S$  of criteria is said to be *preferentially independent* of  $N \setminus S$  if, for all  $x, x' \in [0, 1]_S$  and all  $y, z \in [0, 1]_{N \setminus S}$ , we have

$$(x, y) \succeq (x', y) \Leftrightarrow (x, z) \succeq (x', z).$$

**Theorem** (Scott and Suppes, 1958)

*If a weighted arithmetic mean is used as an aggregation operator then every subset  $S$  of criteria is preferentially independent of  $N \setminus S$ .*

## Example:

	price	consumption	comfort
car 1	10.000 Euro	10 $\ell/100$ km	very good
car 2	10.000 Euro	9 $\ell/100$ km	good
car 3	30.000 Euro	10 $\ell/100$ km	very good
car 4	30.000 Euro	9 $\ell/100$ km	good

No weighted arithmetic mean can model the following preferences:

$$\text{car 2} \succeq \text{car 1} \quad \text{and} \quad \text{car 3} \succeq \text{car 4}.$$

## 4. The Choquet integral

**Definition** (Choquet, 1953; Sugeno, 1974)

A (discrete) fuzzy measure on  $N$  is a set function  $\mu : 2^N \rightarrow [0, 1]$  satisfying

- i)  $\mu_\emptyset = 0, \mu_N = 1,$
- ii)  $S \subseteq T \Rightarrow \mu_S \leq \mu_T.$

$\mu_S$  is regarded as the weight of importance of the combination  $S$  of criteria.

A fuzzy measure is additive if  $\mu_{S \cup T} = \mu_S + \mu_T$  whenever  $S \cap T = \emptyset$ .

When the fuzzy measure is not additive then some criteria interact. For example, we should have

$$\mu_{\{\text{St}, \text{Pr}\}} < \mu_{\{\text{St}\}} + \mu_{\{\text{Pr}\}}.$$

We search for a suitable aggregation operator  $M_\mu : [0, 1]^n \rightarrow \mathbb{R}$ , which generalizes the weighted arithmetic mean.

As for the weighted arithmetic means, we assume that the weight  $\mu_S$  is defined as the global score of the profile  $e_S$ :

$$\mu_S = M_\mu(e_S) \quad (S \subseteq N).$$

We observe that  $\mu$  can be expressed in a unique way as:

$$\mu_S = \sum_{T \subseteq S} a_T \quad (S \subseteq N)$$

where  $a_T \in \mathbb{R}$ .

$a$  viewed as a set function on  $N$  is called the Möbius transform of  $\mu$ , which is given by:

$$a_S = \sum_{T \subseteq S} (-1)^{|T|-|S|} \mu_T \quad (S \subseteq N).$$

For example,

$$\begin{aligned} a_\emptyset &= 0, \\ a_{\{i\}} &= \mu_{\{i\}}, \\ a_{\{i,j\}} &= \mu_{\{i,j\}} - [\mu_{\{i\}} + \mu_{\{j\}}] \\ &\leq 0 \quad (\text{overlap effect}) \\ &\geq 0 \quad (\text{positive synergy}) \\ &= 0 \quad (\text{no interaction}) \end{aligned}$$

If  $\mu$  is additive then we have  $a_S = 0$  for all  $S \subseteq N$ ,  $|S| \geq 2$ .

$$M_\mu(x) = \sum_{i \in N} a_{\{i\}} x_i \quad (\text{weighted arithmetic mean}).$$

When  $\mu$  is not additive, we can introduce

$$\begin{aligned} M_\mu(x) &= \sum_{i \in N} a_{\{i\}} x_i + \sum_{\{i,j\} \subseteq N} a_{\{i,j\}} [x_i \wedge x_j] + \dots \\ &= \sum_{T \subseteq N} a_T \bigwedge_{i \in T} x_i. \end{aligned}$$

(Choquet integral)

Such a function satisfies (Co), (In), (Id), and (SPL).  
It violates (Add).

## Definition (Choquet, 1953)

Let  $\mu$  be a fuzzy measure on  $N$ . The (discrete) Choquet integral of the profile  $x : N \rightarrow [0, 1]$  w.r.t.  $\mu$  is defined by

$$\mathcal{C}_\mu(x) = \sum_{i=1}^n x_{(i)} [\mu_{\{(i), \dots, (n)\}} - \mu_{\{(i+1), \dots, (n)\}}]$$

with the convention that  $x_{(1)} \leq \dots \leq x_{(n)}$ .

## Particular cases:

- When  $\mu$  is additive,  $\mathcal{C}_\mu$  identifies with the weighted arithmetic mean (Lebesgue integral):

$$\mathcal{C}_\mu(x) = \sum_{i=1}^n x_i \mu_{\{i\}} = \sum_{i=1}^n \omega_i x_i$$

- $\mathcal{C}_\mu$  is symmetric (Sy) iff  $\mu$  depends only on the cardinality of subsets (Grabisch, 1995). Setting

$$\omega_i := \mu_{\{(i), \dots, (n)\}} - \mu_{\{(i+1), \dots, (n)\}},$$

we see that  $\mathcal{C}_\mu$  identifies with an *ordered weighted averaging* operator (OWA):

$$\text{OWA}_\omega(x) = \sum_{i=1}^n \omega_i x_{(i)} \quad \text{with } \sum_{i=1}^n \omega_i = 1, \quad \omega_i \geq 0$$

(Yager, 1988)

Two profiles  $x, x' \in [0, 1]^n$  are said to be *comonotonic* if there exists a permutation  $\pi$  of  $N$  such that

$$x_{\pi(1)} \leq \cdots \leq x_{\pi(n)} \quad \text{and} \quad x'_{\pi(1)} \leq \cdots \leq x'_{\pi(n)}.$$

**Comonotonic additivity (CoAdd):**

$$M(x_1 + x'_1, \dots, x_n + x'_n) = M(x_1, \dots, x_n) + M(x'_1, \dots, x'_n)$$

for any two comonotonic profiles  $x, x' \in [0, 1]^n$ .

**Theorem** (Schmeidler, 1986)

$M : [0, 1]^n \rightarrow \mathbb{R}$  fulfills (In, SPL, CoAdd) if and only if there exists a fuzzy measure  $\mu$  on  $N$  such that  $M = \mathcal{C}_\mu$ .

**Theorem**

The aggregation operator  $M_\mu : [0, 1]^n \rightarrow \mathbb{R}$

- is linear w.r.t. the fuzzy measure  $\mu$ :

there exist  $2^n$  functions  $f_T : [0, 1]^n \rightarrow \mathbb{R}$ ,  $T \subseteq N$ , such that

$$M_\mu = \sum_{T \subseteq N} a_T f_T \quad \forall \mu,$$

- satisfies (In),
- satisfies (SPL),
- and is such that

$$M_\mu(e_S) = \mu_S, \quad (S \subseteq N),$$

if and only if  $M_\mu = \mathcal{C}_\mu$ .

## 5. Behavioral analysis of aggregation

### 5.1 Shapley power index

Given  $i \in N$ , it may happen that

- $\mu_{\{i\}} = 0$ ,
- $\mu_{T \cup \{i\}} \gg \mu_T$  for many  $T \not\ni i$

The overall importance of  $i \in N$  should not be solely determined by  $\mu_{\{i\}}$ , but also by all  $\mu_{T \cup \{i\}}$  such that  $T \not\ni i$ .

The marginal contribution of  $i$  in combination  $T \subseteq N$  is defined by

$$\mu_{T \cup \{i\}} - \mu_T$$

The *Shapley power index* for  $i$  is defined as an average value of the marginal contributions of  $i$  alone in all combinations:

$$\begin{aligned} \phi_\mu(i) &:= \frac{1}{n} \sum_{t=0}^{n-1} \frac{1}{\binom{n-1}{t}} \sum_{\substack{T \not\ni i \\ |T|=t}} [\mu_{T \cup \{i\}} - \mu_T] \\ &= \sum_{T \not\ni i} \frac{(n-t-1)! t!}{n!} [\mu_{T \cup \{i\}} - \mu_T] \\ &= \sum_{T \ni i} \frac{1}{t} a_T \end{aligned}$$

This index has been introduced axiomatically by Shapley (1953) in game theory.

## 5.2 Interaction index

Consider a pair  $\{i, j\}$  of criteria. If

$$\underbrace{\mu_{T \cup \{i, j\}} - \mu_{T \cup \{i\}}}_{\substack{\text{contribution of } j \text{ in} \\ \text{the presence of } i}} < \underbrace{\mu_{T \cup \{j\}} - \mu_T}_{\substack{\text{contribution of } j \text{ in} \\ \text{the absence of } i}} \quad \forall T \not\ni i, j$$

then there is an overlap effect between  $i$  and  $j$ .

Criteria  $i$  and  $j$  interfere in a positive way in case of  $>$  and are independent of each other in case of  $=$ .

An *interaction index* for the pair  $\{i, j\} \subseteq N$  is given by an average value of the marginal interaction between  $i$  and  $j$ , conditioned to the presence of elements of the subset  $T \not\ni i, j$ :

$$\begin{aligned} I_\mu(ij) &= \sum_{T \not\ni i, j} \frac{(n-t-2)!t!}{(n-1)!} [\mu_{T \cup \{i, j\}} - \mu_{T \cup \{i\}} - \mu_{T \cup \{j\}} + \mu_T] \\ &= \sum_{T \ni i, j} \frac{1}{t-1} a_T \end{aligned}$$

This interaction index has been proposed by Murofushi and Soneda (1993).

## Notes

1. Interaction indices among a combination  $S$  of criteria have been introduced and characterized by Grabisch and Roubens (1998).
2. Another definition has also been introduced and investigated by Marichal and Roubens (1998) (see §5.4)

### 5.3 Degree of disjunction (cf. Dujmovic, 1974)

We observe that

$$\min x_i \leq \mathcal{C}_\mu(x) \leq \max x_i \quad \forall x \in [0, 1]^n.$$

Define the *average value* of  $\mathcal{C}_\mu$  as

$$m(\mathcal{C}_\mu) := \int_{[0,1]^n} \mathcal{C}_\mu(x) dx$$

We then have

$$\frac{1}{n+1} = m(\min) \leq m(\mathcal{C}_\mu) \leq m(\max) = \frac{n}{n+1}$$

A *degree of disjunction* of  $\mathcal{C}_\mu$  corresponds to

$$\text{orness}(\mathcal{C}_\mu) := \frac{m(\mathcal{C}_\mu) - m(\min)}{m(\max) - m(\min)} \in [0, 1].$$

#### Theorem

For any Choquet integral  $\mathcal{C}_\mu$ , we have

$$\text{orness}(\mathcal{C}_\mu) = \frac{1}{n-1} \sum_{T \subseteq N} \frac{n-t}{t+1} a_T$$

Moreover, we have

$$\text{orness}(\mathcal{C}_\mu) = 1 \iff \mathcal{C}_\mu = \max$$

$$\text{orness}(\mathcal{C}_\mu) = 0 \iff \mathcal{C}_\mu = \min$$

$\mathcal{C}_\mu$	$\text{orness}(\mathcal{C}_\mu)$
$\text{WAM}_\omega$	$1/2$
$\text{OWA}_\omega$	$\frac{1}{n-1} \sum_{i=1}^n (i-1) \omega_i$

## 5.4 Veto and favor effects

Let  $M : [0, 1]^n \rightarrow [0, 1]$  be an aggregation operator. A criterion  $i \in N$  is a

- *veto* for  $M$  if

$$M(x_1, \dots, x_n) \leq x_i \quad \forall x \in [0, 1]^n$$

- *favor* for  $M$  if

$$M(x_1, \dots, x_n) \geq x_i \quad \forall x \in [0, 1]^n$$

(Dubois and Koning, 1991; Grabisch, 1997)

Given a criterion  $i \in N$  and a fuzzy measure  $\mu$  on  $N$ , how can we define a degree of veto of  $i$  for  $\mathcal{C}_\mu$ ?

**First attempt:** Let  $x \in [0, 1]^n$  be a random variable uniformly distributed. A degree of veto of  $i$  is given by

$$\Pr[\mathcal{C}_\mu(x) \leq x_i].$$

However,

$$\Pr[\text{WAM}_\omega(x) \leq x_i] = \begin{cases} 1, & \text{if } \omega_i = 1 \\ 1/2, & \text{otherwise} \end{cases}$$

is non-continuous w.r.t. the fuzzy measure !!!

**Second attempt:** Axiomatic characterization.

$$\text{veto}(\mathcal{C}_\mu; i) := 1 - \frac{n}{n-1} \sum_{T \not\ni i} \frac{1}{t+1} a_T$$

(Similar definition for  $\text{favor}(\mathcal{C}_\mu; i)$ )

## Theorem

The real-valued function  $\psi(\mathcal{C}_\mu; i)$  satisfies the

- *linearity axiom*:

there exist real numbers  $p_T^i$ ,  $T \subseteq N$ ,  $i \in N$ , such that

$$\psi(\mathcal{C}_\mu; i) = \sum_{T \subseteq N} \mu_T p_T^i \quad \forall i \forall \mu,$$

- *symmetry axiom*:

for any permutation  $\pi$  of  $N$ ,

$$\psi(\mathcal{C}_\mu; i) = \psi(\mathcal{C}_{\pi\mu}; \pi(i)) \quad \forall i \forall \mu,$$

where  $\pi\mu$  is defined by  $\pi\mu_{\{\pi(i)\}} = \mu_{\{i\}}$  for all  $i$ .

- *boundary axiom*:

for all  $S \subseteq N$  and all  $i \in S$ ,

$$\psi(\min_S; i) = 1, \quad (\text{cf. } \min_S(x) \leq x_i \forall i \in S)$$

- *normalization axiom*:

$$\psi(\mathcal{C}_\mu; i) = \psi(\mathcal{C}_\mu; j) \quad \forall i, j \in N$$

$\Downarrow$

$$\psi(\mathcal{C}_\mu; i) = \text{andness}(\mathcal{C}_\mu) \quad \forall i \in N.$$

if and only if  $\psi(\mathcal{C}_\mu; i) = \text{veto}(\mathcal{C}_\mu; i)$ .

## 5.5 Measure of dispersion

Consider a symmetric Choquet integral (OWA):

$$\text{OWA}_\omega(x) = \sum_{i=1}^n \omega_i x_{(i)}.$$

Yager (1988) proposed to use the *entropy* of  $\omega$  as degree of the use of the partial scores  $x$ :

$$\text{disp}(\omega) = -\frac{1}{\ln n} \sum_{i=1}^n \omega_i \ln \omega_i \in [0, 1]$$

Examples:

$\text{OWA}_\omega$	$\omega$	$\text{orness}(\text{OWA}_\omega)$	$\text{disp}(\omega)$
AM	$(1/n, \dots, 1/n)$	$1/2$	$1$
median	$(0, \dots, 1, \dots, 0)$	$1/2$	$0$

Measure of dispersion of a fuzzy measure:

$$\text{disp}(\mu) := -\frac{1}{\ln n} \sum_{i=1}^n \sum_{T \not\ni i} \frac{(n-t-1)! t!}{n!} [\mu_{T \cup \{i\}} - \mu_T] \ln [\mu_{T \cup \{i\}} - \mu_T]$$

**Theorem** The following properties hold:

- i)  $\text{disp}(\mu_{\text{WAM}_\omega}) = \text{disp}(\mu_{\text{OWA}_\omega}) = -\frac{1}{\ln n} \sum_{i=1}^n \omega_i \ln \omega_i$
- ii)  $0 \leq \text{disp}(\mu) \leq 1$
- iii)  $\text{disp}(\mu) = 1 \Leftrightarrow \mu = \mu_{\text{AM}}$
- iv)  $\text{disp}(\mu) = 0 \Leftrightarrow \mu_S \in \{0, 1\} \forall S \subseteq N$   
 $\Rightarrow \mathcal{C}_\mu(x) \in \{x_1, \dots, x_n\}.$