

Interaction between criteria

In classical MCDM procedures (ELECTRE & PROMETHEE ...) we consider additive weights : $p(i)$ related to criterion i such that

$$\sum_{i=1,\dots,n} p(i) = 1$$

in order to introduce compensation via weighted sums in aggregation phase :

$$C(a, b) = \sum_{i=1,\dots,n} p(i) r_{ab}^{(i)} = p_{ab}^+ + p_{ab}^-$$

$$\begin{aligned} r_{ab}^{(i)} &= 1 \text{ if } a \geq_i b \\ &= 0 \text{ otherwise.} \end{aligned}$$

Additive model : $\left\{ \begin{array}{l} \mu(1, \dots, n) = 1 \\ \mu(i, j) = \mu(i) + \mu(j) \\ \mu(i) = p(i). \end{array} \right.$

This model can be extended in a Choquet capacity (fuzzy measure, Sugeno measure) :

X : set of criteria $\mathcal{P}(X)$: power set of X

$$\mu : \mathcal{P}(X) \rightarrow [0, 1]$$

$$\left\{ \begin{array}{l} \mu(P) \geq 0, \quad P \in \mathcal{P}(X) \\ \mu(\emptyset) = 0 \\ \mu(X) = 1 \text{ (not compulsory)} \\ \mu(Q) \leq \mu(P) \text{ if } Q \subseteq P \end{array} \right.$$

Pseudo-Boolean functions and Choquet capacity

μ can be rewritten under the form of a *unique* pseudo-Boolean function

$$f(x_1, \dots, x_n) = \sum_{L \subset X} \left[a(L) \prod_{i \in L} x_i \right], \quad x_i \in \{0, 1\}$$

$$\begin{aligned} \mu(P) &= f(\dots \underset{i \in P}{\overset{i}{1}} \dots \underset{j \notin P}{\overset{j}{0}} \dots) \\ &= f(x_1, \dots, x_n \mid x_i = 1 \text{ if } i \in P, \\ &\qquad\qquad\qquad x_i = 0 \text{ otherwise}) \\ &= \sum_{L \subset P} a(L) \end{aligned}$$

$$\left\{ \begin{array}{l} \mu(\emptyset) = 0 \\ \mu(i) = a(i) \\ \mu(i, j) = a(i) + a(j) + a(i, j) \\ \vdots \\ \mu(X) = 1 \end{array} \right.$$

↓

$$\left\{ \begin{array}{l} a(i, j) = 0 : \text{additivity} \\ a(i, j) > 0 : \text{synergy effect (superadditivity)} \\ a(i, j) < 0 : \text{redundancy effect (subadditivity)} \end{array} \right.$$

Möbius transform

$$\mu(P) = \sum_{L \subset P} a(L)$$

$$\begin{cases} a(\emptyset) = 0 \\ \mu(i) = a(i) \\ \mu(i, j) = a(i) + a(j) + a(ij) \\ \vdots \end{cases}$$

$$a(P) = \sum_{L \subset P} (-1)^{|P|-|L|} \mu(L)$$

$$\begin{cases} a(i) = \mu(i) \\ a(i, j) = \mu(\emptyset) - \mu(i) - \mu(j) + \mu(i, j) \\ \vdots \end{cases}$$

From $\mu \rightarrow a$: Möbius transform

$$\underline{n=1} \quad \begin{pmatrix} a(\emptyset) \\ a(1) \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} \mu(\emptyset) \\ \mu(1) \end{pmatrix}$$

$$\tilde{a}_{(1)} = M_{(1)} \tilde{\mu}_{(1)}.$$

$n > 1$: $\mu(P)$ and $a(P)$ are represented as vectors where the elements of $\mathcal{P}(X)$ are represented in the following order :

$$\begin{aligned}\tilde{\mu}_{(n)} &= \left\{ \frac{\underline{\mu(\emptyset)} \underline{\mu(1)}_{n=1} \underline{\mu(2)}_{n=2} \underline{\mu(1, 2)}}{\underline{\mu(3)\mu(1, 3)\mu(2, 3)}_{n=3} \underline{\mu(1, 2, 3)}_{n=3} \dots} \right\} \\ \tilde{a}_{(n)} &= \{a(\emptyset)a(1)a(2)a(1, 2) \dots\}\end{aligned}$$

$n = 2$

$$\begin{pmatrix} a(\emptyset) \\ a(1) \\ a(2) \\ a(1, 2) \end{pmatrix} = \left(\begin{array}{cc|cc} +1 & 0 & 0 & 0 \\ -1 & +1 & 0 & 0 \\ \hline -1 & 0 & +1 & 0 \\ +1 & -1 & -1 & +1 \end{array} \right) \begin{pmatrix} \mu(\emptyset) \\ \mu(1) \\ \mu(2) \\ \mu(1, 2) \end{pmatrix}$$

$$\tilde{a}_{(a)} = M_{(2)} \tilde{\mu}_{(2)}$$

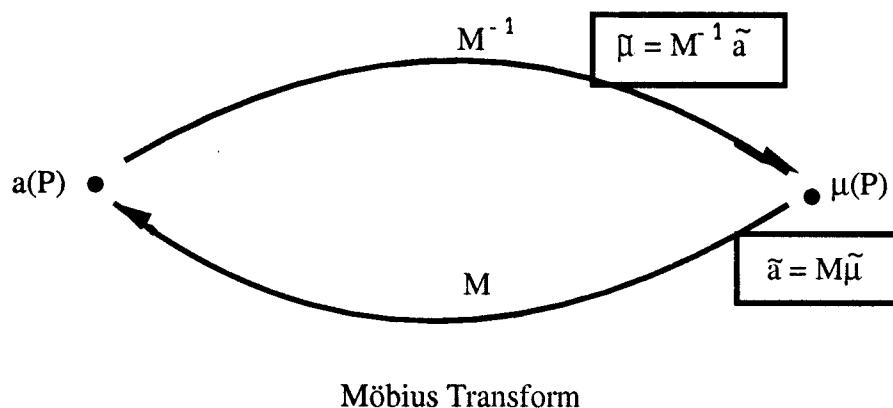
Two equivalent representations

$$\begin{aligned}
 \tilde{a}_{(1)} &= \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix} \tilde{\mu}_{(1)} = M_{(1)} \tilde{\mu}_{(1)} \text{ dim. 2} \\
 \tilde{a}_{(2)} &= \left[\begin{array}{c|c} 1M_{(1)} & 0M_{(1)} \\ \hline -1M_{(1)} & 1M_{(1)} \end{array} \right] \tilde{\mu}_{(2)} = M_{(2)} \tilde{\mu}_{(2)} \text{ (dim. } 2^2\text{)} \\
 &\vdots \\
 \tilde{a}_{(n)} &= \underbrace{\left[\begin{array}{c|c} 1M_{(n-1)} & 0M_{(n-1)} \\ \hline -1M_{(n-1)} & 1M_{(n-1)} \end{array} \right]}_{M_{(n)}} \tilde{\mu}_{(n)} \text{ (dim. } 2^n\text{)}
 \end{aligned}$$

$M_{(n)}$ is a triangular fractal matrix :

$M_{(1)} \otimes \cdots \otimes M_{(1)}$

(tensorial product of matrix $M_{(1)}$)



Weights and power indices

Back to cooperative game theory.

$\mu(P)$: weight of coalition

$$\begin{aligned} \mu(P) &= \begin{cases} 1 & \text{if } P \text{ has the majority} \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

swing related to $\{i\}$:

$$\mu(P) = 0 \nearrow \mu(P \cup \{i\}) = 1$$

Banzhaf power index : “probability” of a swing related to $\{i\}$.

$$b_i(X) = \frac{1}{2^{n-1}} \sum_{P \subset X - \{i\}} [\mu(P \cup \{i\}) - \mu(P)]$$

$$2^{n-1} = |\mathcal{P}(X - \{i\})|.$$

b_i has some nice properties :

1) If $\{i\}$ is a dummy :

$$\mu(P \cup \{i\}) = \mu(P), \forall P \subset X - \{i\} \Rightarrow b_i(X) = 0,$$

the real weight of $\{i\}$ is zero (even if $\mu(i) > 0$).

2) If $\{i\}$ and $\{j\}$ are symmetric players :

$$\mu(P \cup \{i\}) = \mu(P \cup \{j\}), \forall P \subset X - \{i, j\} \Rightarrow b_i(X) = b_j(X)$$

they have the same real power (even if $\mu(i) \neq \mu(j)$).

3) Hammer and Holzman proved in 1992

$$\min_{p_i} \sum_{R \subset X} [\mu(R) - \hat{\mu}(R)]^2$$

with $\hat{\mu}(R) = \sum_{i \in R} p_i$ (related additive measure)

is such that $p_i = b_i(X)$ but

$$\hat{\mu}(X) = \sum_{i \in X} b_i(X) \neq 1 \quad !!!$$

Example related to Owen (1968)

4 stock holders in a company :

- (1) : 10 shares of stock
- (2) : 30 shares
- (3) : 30 shares
- (4) : 40 shares

weight is : $\frac{10}{110}$; $\frac{30}{110}$; $\frac{30}{110}$; $\frac{40}{110}$ (9%; 27%; 27%; 36%)

Consider now :

$$\begin{aligned}\mu(1) &= 0 & \mu(1, 2) &= 0 & \mu(i, j, k) &= 1, \quad \forall i, j, k \\ \mu(2) &= 0 & \mu(1, 3) &= 0 \\ \mu(3) &= 0 & \mu(1, 4) &= 0 \\ \mu(4) &= 0 & \mu(2, 3) &= 1 \\ && \mu(2, 4) &= 1 \\ && \mu(3, 4) &= 1\end{aligned}$$

$$\left\{ \begin{array}{lcl} b_1(X) & = & 0 \\ \\ b_2(X) & = & \frac{1}{8} \overbrace{\{\mu(23) - \mu(3)}^{\text{swing}} + \mu(24) - \mu(4) \\ & & + \mu(124) - \mu(14) + \mu(123) - \mu(13)\} \\ & = & \frac{4}{8} \\ \\ b_3(X) & = & \frac{4}{8} \\ \\ b_4(X) & = & \frac{1}{8} \{\mu(24) - \mu(2) + \mu(34) - \mu(4) \\ & & + \mu(124) - \mu(12) + \mu(134) - \mu(13)\} \\ & = & \frac{4}{8} \end{array} \right.$$

We switch from the weights (9%; 27%; 27%; 36%) to the power indices (0, $\underbrace{.50, .50, .50}_{\text{same weight}}$)

Back to MCDM

- Practically, it is useless to think on a general Choquet capacity to determine the weights of criteria including interaction between criteria.

For n criteria, $2^{n \text{**}}$ weights should be determined !!

$$\mu(1), \dots, \mu(n), \mu(1, 2), \mu(1, 3), \dots, \mu(1, 2, 3), \dots$$

We might restrict ourselves to a second order model :

$$a(P) = 0, \forall P : |P| > 2$$

$$(\Leftrightarrow J(P) = 0, \forall P : |P| > 2)$$

$$2^{n \text{**}} \rightarrow \binom{n+1}{2} = \frac{n(n+1)}{2} \text{ parameters.}$$

- Extension of ELECTRE II using interaction.

We have seen that

$$C(a, b) = \sum_{\substack{i=1, \dots, n \\ a \geq_i b}} p(i) r_{ab}^{(i)} = \sum_i p_i$$

The weighted sum can be extended to a Choquet integral.

Def. : If $x_{(1)} \leq x_{(2)} \leq \dots \leq x_{(n)}$, the Choquet integral for (x_1, \dots, x_n) :

$$\begin{aligned} C[x_1, \dots, x_n] &= x_{(1)}[1 - \mu(2, \dots, n)] \\ &\quad + x_{(2)}[\mu(2, \dots, n) - \mu(3, \dots, n)] \\ &\quad + x_{(3)}[\mu(3, \dots, n) - \mu(4, \dots, n)] \\ &\quad \vdots \\ &\quad + x_{(n)}[\mu(n)]. \end{aligned}$$

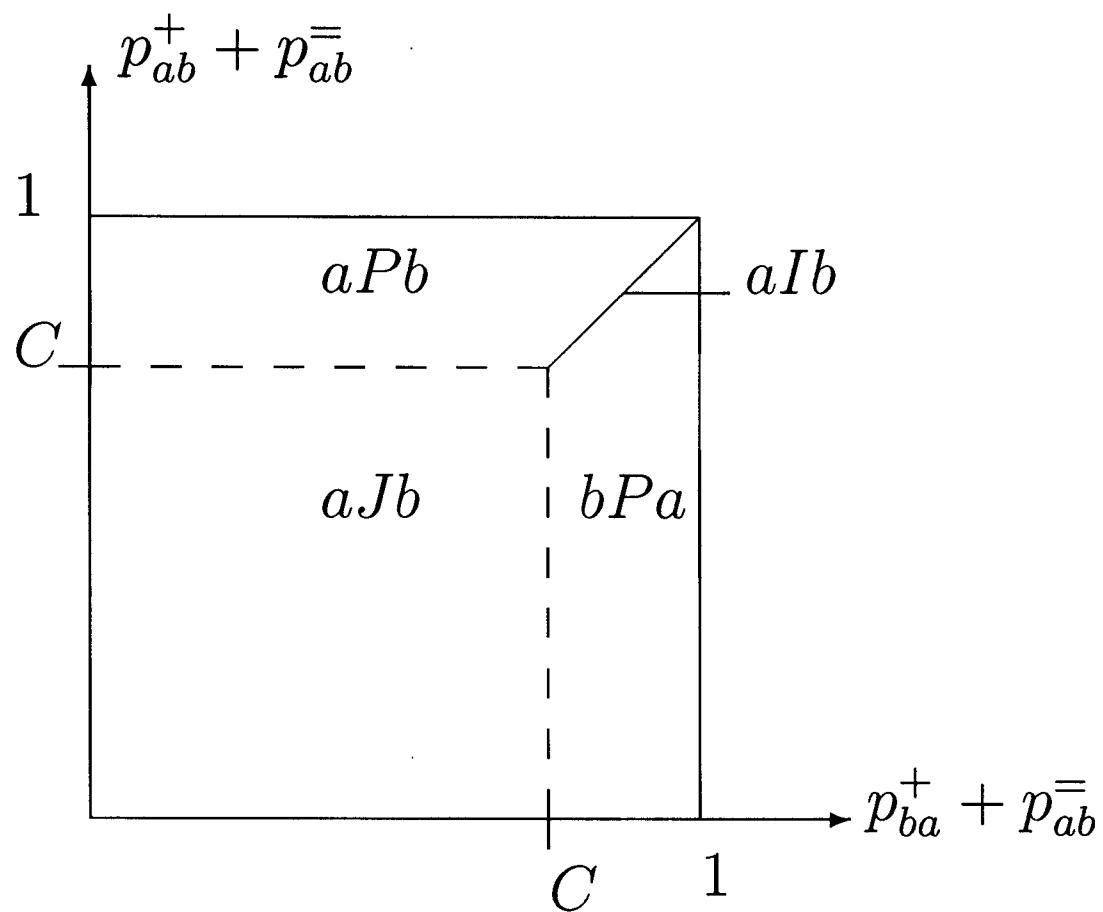
$$(x_1, \dots, x_n) \rightarrow (r_{ab}^{(1)} \dots r_{ab}^{(n)}) \in \{0, 1\}^n$$

$$C[r_{ab}^{(1)} \dots r_{ab}^{(n)}] = \mu(i_1, \dots, i_k \mid a \geq_{i_1} b, \dots, a \geq_{i_k} b)$$

$$\sum_{i=1,\dots,n} p(i) r^{(i)}_{ab} = p^+_{ab} + p^{\equiv}_{ab}$$

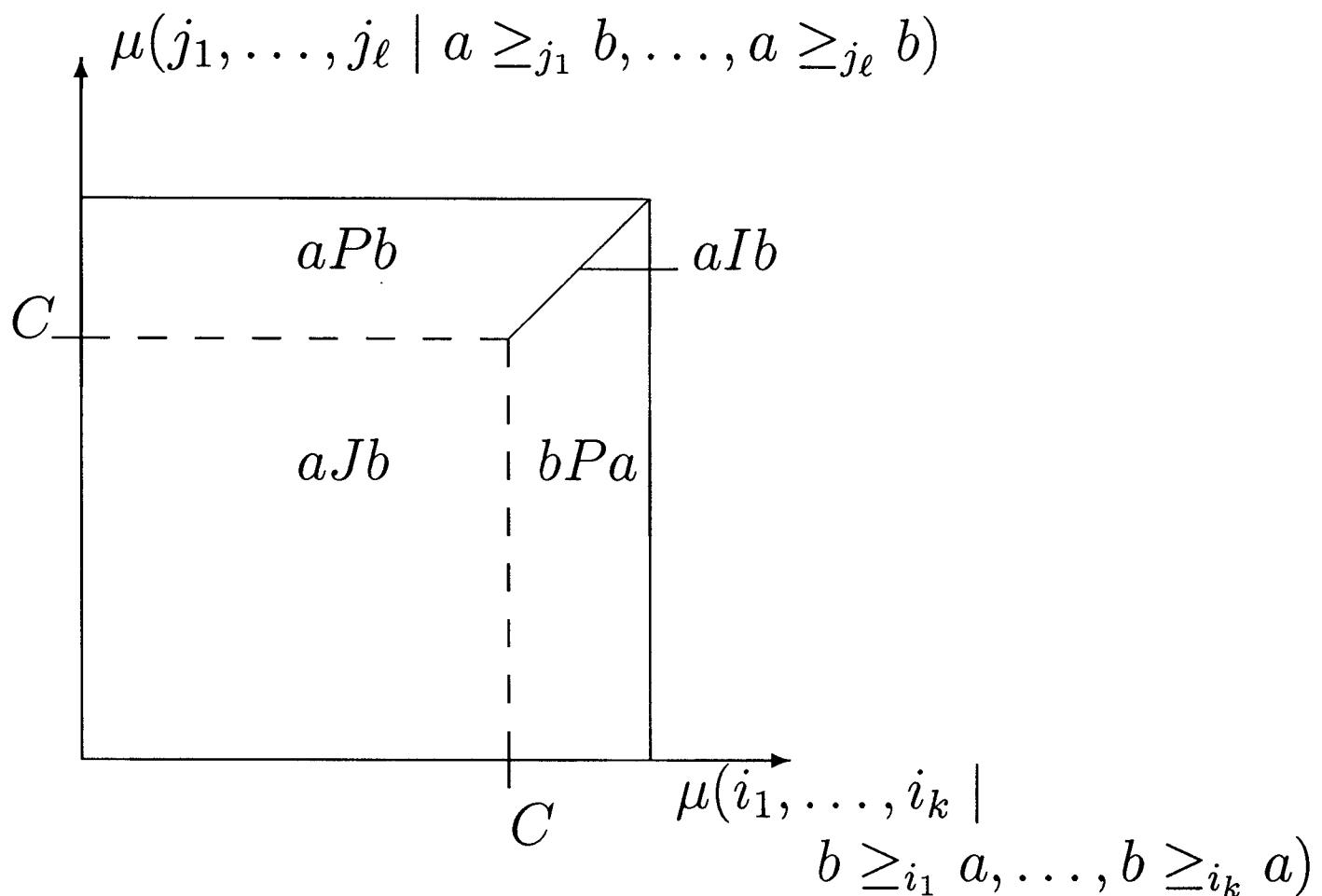
$$\Rightarrow \mu(i_1, \dots, i_k \mid a \geq_{i_1} b, \dots, a \geq_{i_k} b)$$

Classical ELECTRE II (Concordance test)



(concordance level)

ELECTRE II with interaction between criteria



Interaction in a formal way

$$f(\underline{x}) = f(x_1, \dots, x_n) = \sum_{L \subset X} \left[a(L) \prod_{i \in L} x_i \right];$$

$$\mu(P) = \sum_{L \subset P} a(L)$$

$$\begin{aligned} \delta_i f(\underline{x}) &= f(x_1, \dots, x_i = 1, \dots, x_n) \\ &\quad - f(x_1, \dots, x_i = 0, \dots, x_n) \end{aligned}$$

$$\delta_{ij} f(\underline{x}) = \delta_i [\delta_j f(\underline{x})]$$

⋮

$$\delta_P f(\underline{x}) = \delta_{i_1} [\delta_{i_2 \dots i_p} f(\underline{x})] \text{ if } P = \{i_1, \dots, i_p\}$$

$$\begin{aligned} f(\underline{0}) &= f(0, \dots, 0) = 0 \\ f(\underline{1}) &= f(1, \dots, 1) = \sum_{L \subset X} a(L) \\ &= \mu(1, \dots, n) = 1 \end{aligned}$$

$$J(P) = \delta_P f \left(\frac{1}{2} \right)$$

with two particular cases :

$$\begin{aligned}
J(i) &= \delta_i f\left(\frac{1}{2}\right) = b_i(X) \\
&= \frac{1}{2^{n-1}} \sum_{L \subset X - \{i\}} [\mu(L \cup \{i\}) - \mu(L)] \\
J(i, j) &= \delta_{ij} f\left(\frac{1}{2}\right) \\
&= \frac{1}{2^{n-2}} \sum_{L \subset X - \{i, j\}} [\mu(L \cup \{i, j\}) \\
&\quad - \mu(L \cup \{i\}) - \mu(L \cup \{j\}) + \mu(L)] \\
\tilde{J}_{(2)} &= \begin{pmatrix} J(\emptyset) \\ J(1) \\ J(2) \\ J(1, 2) \end{pmatrix} \\
&= \left(\begin{array}{cc|cc} 1 & \frac{1}{2} & \frac{1}{2} & \frac{1}{4} \\ 0 & 1 & 0 & \frac{1}{2} \\ \hline 0 & 0 & 1 & \frac{1}{2} \\ 0 & 0 & 0 & 1 \end{array} \right) \begin{pmatrix} a(\emptyset) \\ a(1) \\ a(2) \\ a(1, 2) \end{pmatrix} \\
&= R_{(2)} \tilde{a}_{(2)}
\end{aligned}$$

but

$$R_{(2)} = \begin{bmatrix} 1R_{(1)} & \frac{1}{2}R_{(1)} \\ 0R_{(1)} & 1R_{(1)} \end{bmatrix}$$

and we obtain a fractal representation :

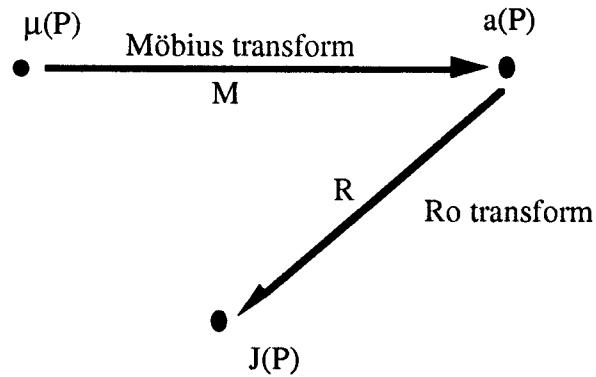
$$\tilde{J}_{(1)} = R_{(1)}\tilde{a}_{(1)}$$

$$\tilde{J}_{(2)} = R_{(2)}\tilde{a}_{(2)}$$

⋮

$$\tilde{J}_{(n)} = \begin{bmatrix} 1R_{(n-1)} & \frac{1}{2}R_{(n-1)} \\ 0 & 1R_{(n-1)} \end{bmatrix} \tilde{a}_{(n)}$$

A third equivalent representation



$$J(P) = \delta_P f\left(\frac{1}{2}\right) = \sum_{L \subset X - P} \frac{1}{2^{|L|}} a(L \cup P)$$

$$J(\emptyset) = \sum_{L \subset X} \frac{1}{2^{|L|}} a(L)$$

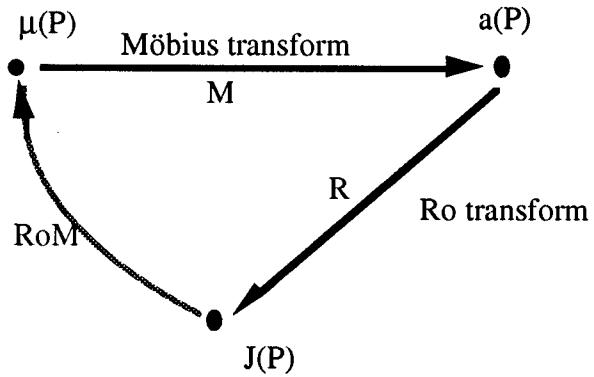
$$J(i) = \sum_{L \subset X - \{i\}} \frac{1}{2^{|L|}} a(L \cup \{i\})$$

$$J(i, j) = \sum_{L \subset X - \{i, j\}} \frac{1}{2^{|L|}} a(L \cup \{i, j\})$$

⋮

$$\begin{aligned} \tilde{J}_{(1)} &= \begin{pmatrix} J(\emptyset) \\ J(1) \end{pmatrix} = \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix} \begin{pmatrix} a(\emptyset) \\ a(1) \end{pmatrix} \\ &= R_{(1)} \tilde{a}_{(1)} \end{aligned}$$

Three equivalent presentations of Choquet capacities of fractal style



$$M_{(1)} = \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix} \quad R_{(1)} = \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix}$$

$$R_{(1)} \circ M_{(1)} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ -1 & 1 \end{bmatrix}$$

$$\tilde{J}_{(n)} = \begin{bmatrix} \frac{1}{2} R_{(n-1)} \circ M_{(n-1)} & \frac{1}{2} R_{(n-1)} \circ M_{(n-1)} \\ -1 R_{(n-1)} \circ M_{(n-1)} & 1 R_{(n-1)} \circ M_{(n-1)} \end{bmatrix}$$

of course inversion is easy and is also fractal

$$M_{(1)}^{-1} = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \quad R_{(1)}^{-1} = \begin{bmatrix} 1 & -\frac{1}{2} \\ 0 & 1 \end{bmatrix}$$

$$M_{(1)}^{-1} \circ R_{(1)}^{-1} = \begin{bmatrix} 1 & -\frac{1}{2} \\ 1 & \frac{1}{2} \end{bmatrix}$$