

Faculté Polytechnique



Comparison of Decisional Maps

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PLAN

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- Motivation
- Example

2 Models

- Case of one decisional map
- Case of one decisional map and one “attribute” map
- Case of one decisional map and several “attribute” maps

3 Conclusion and Perspectives

Introduction

There are many interesting decision problems related with GIS

- Suitability for housing (*Joerin, 1997*)
- Railway corridor (*Mousseau and Chakhar, 2008*)
- Risk of degradation of a region (*Metchebon, 2010*)

⇒ map partitioned in geographic units (g.u.)
each g.u. assessed on an ordinal scale

= decisional map

After some time, the map has evolved

Example : Loulouka

Study of Loulouka's basin in Burkina Faso

Map representing the response to the risk of degradation

Geographic units : 229 squares 500m x 500m

Scale for response : 4 categories :
adequate, moderately adequate, weakly adequate, not adequate

Example of evolution

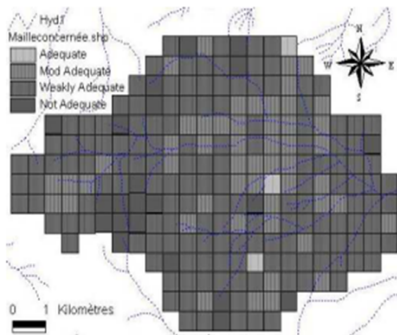


FIGURE 1 – Previous state

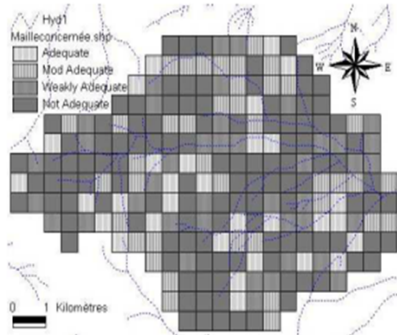


FIGURE 2 – New state

MODEL 1 :

Case of one decisional map

Model 1

Hypothesis

We assume that a map A is equivalent to the distribution of g.u. in categories :

$$x(A) = (x_1(A), x_2(A), \dots, x_n(A))$$

Comparing maps is equivalent to comparing probability distributions

Model : EU (expected utility)

$$A \succsim B \iff \sum_i x_i(A) u_i \geq \sum_i x_i(B) u_i$$

where u_i is a value associated with the map in which all g.u. belong to category i

Model 1

Characterization : Adapting Jensen's axioms (Fishburn 82)

- The set of maps is a mixture set : for $\lambda \in [0, 1]$,

$$\lambda A \oplus (1 - \lambda)B$$

is a map in which a portion λ has the same distribution as A and a portion $1 - \lambda$ has the same distribution as B

- $A = x_1(A)e_1 \oplus \dots \oplus x_i(A)e_i \oplus \dots \oplus x_n(A)e_n$ where e_i is a map with all g.u. in category i
- \succsim is the DM's preference on the set of maps
- if this preference satisfies some axioms then it has a linear EU representation

Model 1

Axioms

A_1 : \succsim is a weak order on the set of maps

A_2 : for any maps A, B, C and $\forall \lambda \in]0, 1]$

$$A \succ B \Rightarrow \lambda A \oplus (1 - \lambda)C \succ \lambda B \oplus (1 - \lambda)C$$

A_3 : [Continuity] $A \succ B$ and $B \succ C$ imply

there exists a number $\alpha \in]0, 1[$ such that $\alpha A \oplus (1 - \alpha)C \succ B$

there exists a number $\beta \in]0, 1[$ such that $B \succ \beta A \oplus (1 - \beta)C$

Model 1

Representation theorem

There is a unique linear utility function u representing \succsim ; i.e.

$$A \succsim B \iff u(A) \geq u(B)$$

Linear : $u(\lambda A \oplus (1 - \lambda)B) = \lambda u(A) + (1 - \lambda)u(B)$

Unique : if u' represents \succsim then there are $a > 0$ and b such that $u' = au + b$

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Unique : if u' represents \succsim then there are $a > 0$ and b such that $u' = au + b$

Benefits

$$A = x_1(A)e_1 \oplus \cdots \oplus x_i(A)e_i \oplus \cdots \oplus x_n(A)e_n$$

$$u(A) = x_1(A)u(e_1) + \cdots + x_i(A)u(e_i) + \cdots + x_n(A)u(e_n)$$

We only have to determine $u(e_i) := u_i$ for $i = 1, \dots, n$

Model 1

Additional axiom

Due to the ordering of the categories, the DM's preference surely satisfies :

$$e_1 \succ e_2 \succ \cdots \succ e_n$$

which results in :

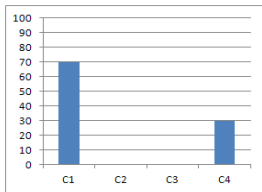
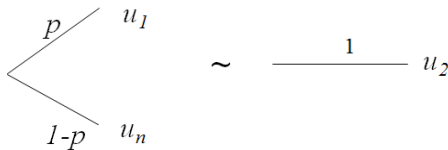
$$u_1 > u_2 > \cdots > u_n$$

Using the two degrees of freedom, we set

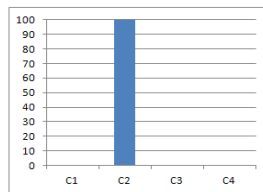
$$u_1 = 100 \text{ and } u_n = 0$$

Elicitation

Inspired by comparison of lotteries



\sim



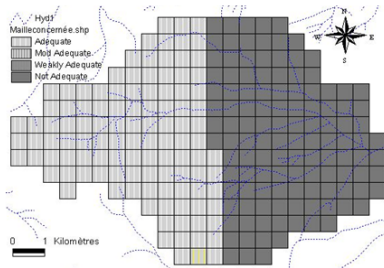
Remarks

Model 1 only takes into account the proportion of g.u. in the categories
→ a bit disappointing : geographic aspects neglected

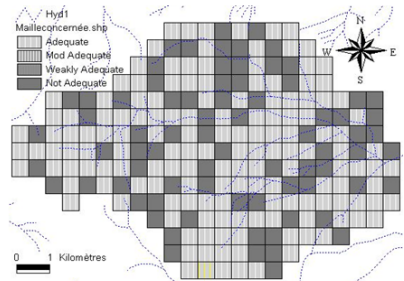
One geographic aspect : contiguity

Same distribution

Grouped



Scattered



Remarks

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One geographic aspect : contiguity

→ model based on the Choquet's integral w.r.t. a 2-additive capacity

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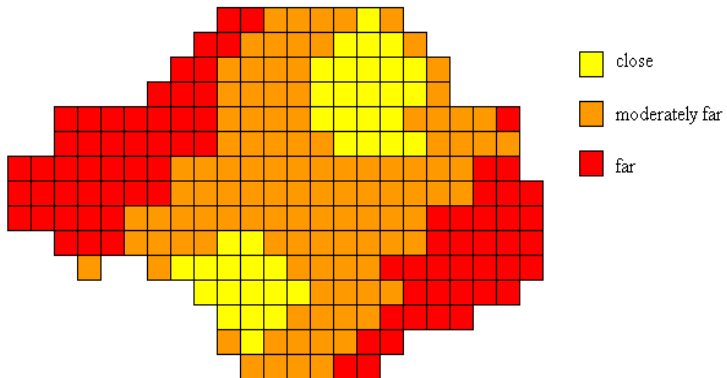
Other geographic aspects : proximity of a village, of a road, of a watercourse, ...

→ attribute map(s)

MODEL 2 :

Case of one decisional map
and
one attribute map

Fictitious attribute map



Model 2

We consider :

- a decisional map A
- a fixed attribute map G partitionned as A

We note (A, G) :

- a map partitionned in g.u.
- each g.u. assessed on 2 scales : the same as A and the same as G

Model 2

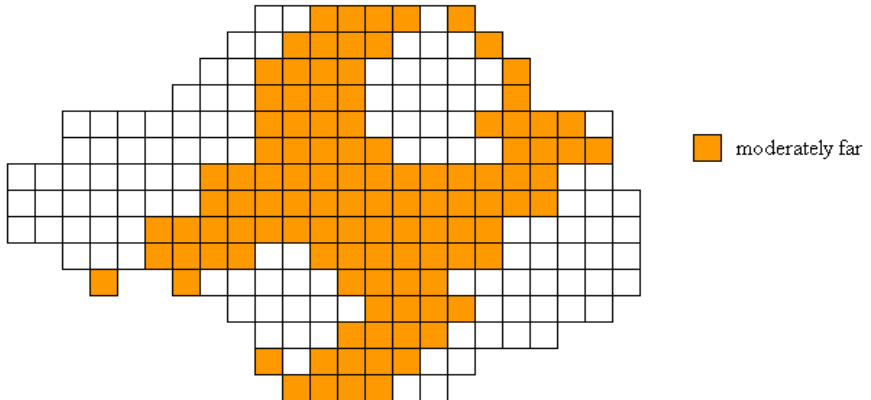
Hypothesis

We assume that a map (A, G) is equivalent to the distribution of g.u. in categories :

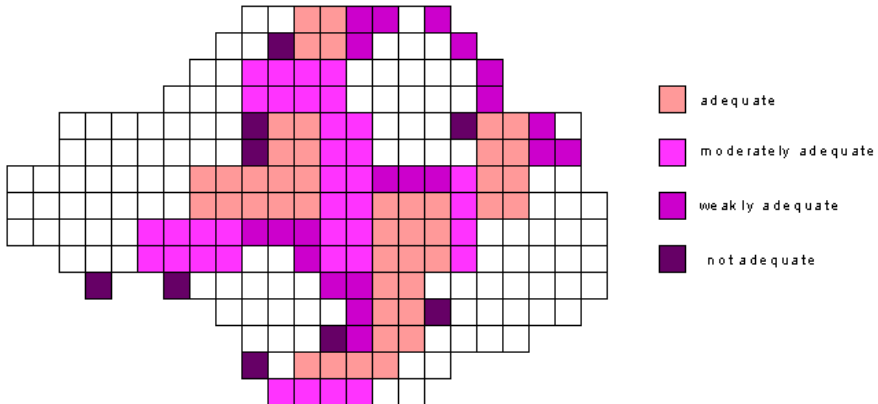
$$\begin{aligned}x(A, G) &= ((x_{11}(A, G), \dots, x_{n1}(A, G)), \dots, (x_{1m}(A, G), \dots, x_{nm}(A, G))) \\ &:= (x_1(A, G), \dots, x_m(A, G))\end{aligned}$$

Comparing maps is equivalent to comparing vectors of probability distributions

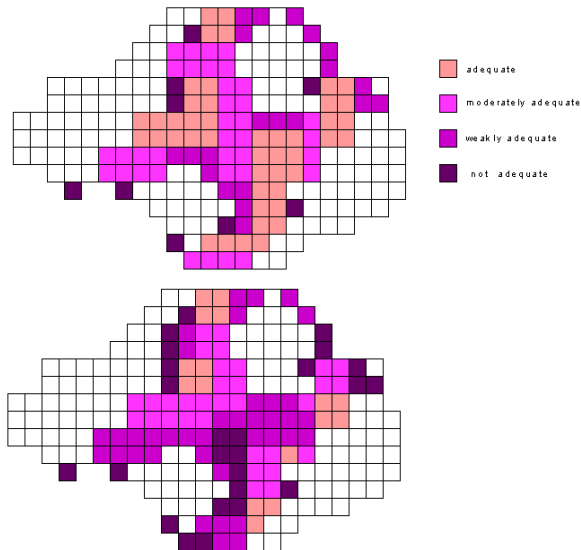
Example of partial map



Example of distribution in categories for a partial map



Preferences on partial maps



Model 2

Notation and definitions

- For $j \in \{1, \dots, m\}$ we write X_j the set of all possible distributions in categories of a partial map

- For $j \in \{1, \dots, m\}$ we write $X_{-j} = \prod_{\substack{k=1 \\ k \neq j}}^m X_k$

- For $x_j \in X_j$ and $a_{-j} \in X_{-j}$, we write

$$(x_j, a_{-j}) = (a_1, \dots, a_{j-1}, x_j, a_{j+1}, \dots, a_m)$$

- We define a preference relation \succsim_j on X_j as follows :

$$\forall x_j, y_j \in X_j, \quad x_j \succsim_j y_j \iff \forall a_{-j} \in X_{-j} \quad (x_j, a_{-j}) \succsim (y_j, a_{-j})$$

Model 2

Notations and definitions (following)

- Each set X_j is a mixture set : for $\lambda \in [0, 1]$

$$\lambda x_j \oplus (1 - \lambda)y_j$$

is a partial map in which a portion λ has the same distribution as x_j ,
the other portion $1 - \lambda$ having the same distribution as y_j

Model 2

Axioms

B_1 : \succsim is a weak order

B_2 : $\forall j \in \{1, \dots, m\} \quad \forall x_j, y_j, z_j \in X_j \quad \forall a_{-j} \in X_{-j} \text{ and } \forall \lambda \in]0, 1[$

$$[(x_j, a_{-j}) \succ (y_j, a_{-j})] \Rightarrow$$

$$[\lambda(x_j, a_{-j}) \oplus (1 - \lambda)(z_j, a_{-j}) \succ \lambda(y_j, a_{-j}) \oplus (1 - \lambda)(z_j, a_{-j})]$$

B_3 : $\forall j \in \{1, \dots, m\} \quad \forall x_j, y_j, z_j \in X_j \quad \forall a_{-j} \in X_{-j}$

$$[(x_j, a_{-j}) \succ (y_j, a_{-j}) \succ (z_j, a_{-j})] \Rightarrow \exists \alpha, \beta \in]0, 1[:$$

$$\alpha(x_j, a_{-j}) \oplus (1 - \alpha)(z_j, a_{-j}) \succ (y_j, a_{-j})$$

and

$$(y_j, a_{-j}) \succ \beta(x_j, a_{-j}) \oplus (1 - \beta)(z_j, a_{-j})$$

Model 2

Axioms (following)

B_4 [Essentiality] : $\forall j \in \{1, \dots, m\} \exists x_j, y_j \in X_j \exists a_{-j} \in X_{-j} :$

$$(x_j, a_{-j}) \succ (y_j, a_{-j})$$

B_5 [Independence] : $\forall j \in \{1, \dots, m\} \forall x_j, y_j \in X_j \forall a_{-j}, b_{-j} \in X_{-j}$

$$(x_j, a_{-j}) \succsim (y_j, a_{-j}) \Rightarrow (x_j, b_{-j}) \succsim (y_j, b_{-j})$$

B_6 [Restricted solvability] :

$$\forall j \in \{1, \dots, m\} \forall x_j, z_j \in X_j \forall a_{-j} \in X_{-j} \forall y \in X$$

$$(x_j, a_{-j}) \succsim y \succsim (z_j, a_{-j}) \Rightarrow \exists w_j \in X_j : y \sim (w_j, a_{-j})$$

B_7 [Archimedean] : $\forall j \in \{1, \dots, m\}$, if a standard sequence on X_j is bounded, then it is finite

Model 2

Representation theorem

- \succsim is represented by an additive value function

$$(A, G) \succsim (B, G) \iff \sum_{j=1}^m u_j(x_j(A, G)) \geq \sum_{j=1}^m u_j(x_j(B, G))$$

- \succsim_j is represented by a linear EU function

$$x_j(A, G) = x_{1j}(A, G)e_{1j} \oplus \cdots \oplus x_{nj}(A, G)e_{nj}$$

where e_{ij} is a partial map where all g.u. are in i and j

$$u_j(x_j(A, G)) = \sum_{i=1}^n x_{ij}(A, G)u_j(e_{ij})$$

Model 2

Representation theorem

$$(A, G) \succsim (B, G) \iff \sum_j \sum_i x_{ij}(A, G) u_j(e_{ij}) \geq \sum_j \sum_i x_{ij}(B, G) u_j(e_{ij})$$

We have to determine $u_j(e_{ij})$ for $i = 1, \dots, n$ and $j = 1, \dots, m$

Model 2

Additional axiom

Due to the ordering of the categories, the DM's preference surely satisfies, for any j :

$$e_{1j} \succ e_{2j} \succ \cdots \succ e_{nj}$$

which results in :

$$u_j(e_{1j}) \succ u_j(e_{2j}) \succ \cdots \succ u_j(e_{nj})$$

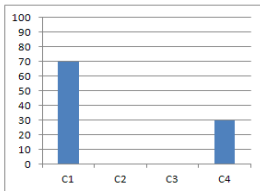
Using the two degrees of freedom, we set

$$u_j(e_{1j}) = 100 \text{ and } u_j(e_{nj}) = 0$$

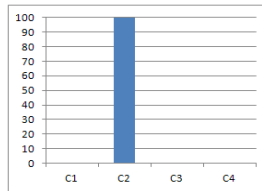
Elicitation

Inspired by comparison of lotteries

$$\begin{array}{l} p \quad u_j(e_{1j}) \\ \swarrow \quad \searrow \\ \sim \quad \text{---} 1 \text{---} u_j(e_{2j}) \\ \nwarrow \quad \nearrow \\ 1-p \quad u_j(e_{nj}) \end{array}$$



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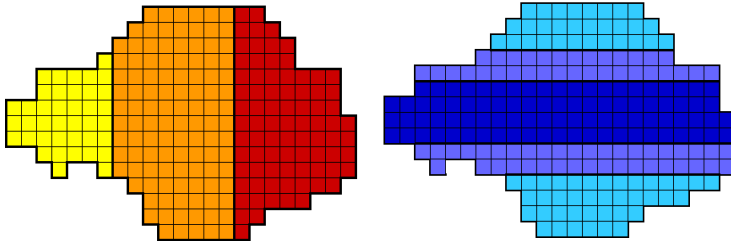


MODEL 3 :

Case of one decisional map
and
several attribute maps

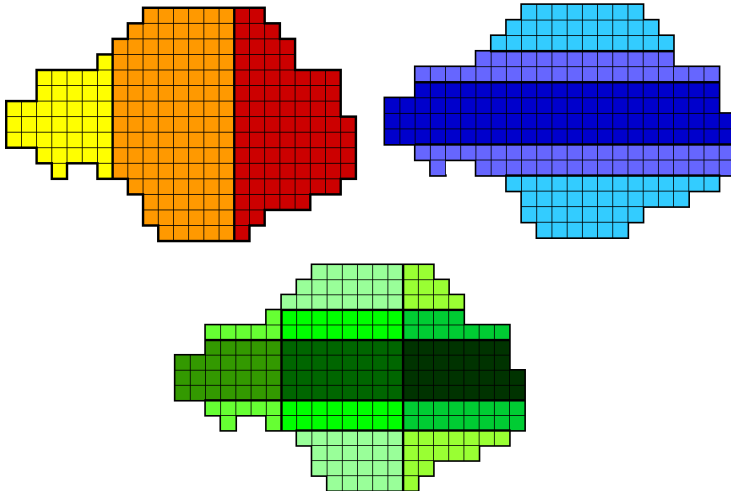
Idea : create a unique attribute map

- each g.u. assessed on $I + 1$ scales (if I attribute maps)



Idea : create a unique attribute map

- each g.u. assessed on $I + 1$ scales (if I attribute maps)



Model 3

Model : EU (expected utility)

$$\begin{aligned}
 (A, G_1, \dots, G_l) \succsim (B, G_1, \dots, G_l) &\iff \\
 \sum_{j_1} \cdots \sum_{j_l} \sum_i x_{i,j_1, \dots, j_l}(A, G_1, \dots, G_l) &u_{j_1, \dots, j_l}(e_{i,j_1, \dots, j_l}) \\
 \geq \sum_{j_1} \cdots \sum_{j_l} \sum_i x_{i,j_1, \dots, j_l}(B, G_1, \dots, G_l) &u_{j_1, \dots, j_l}(e_{i,j_1, \dots, j_l})
 \end{aligned}$$

where $u_{j_1, \dots, j_l}(e_{i,j_1, \dots, j_l})$ is a value associated with the map in which all g.u. belong to category i and characteristic j_k on G_k , $k = 1, \dots, l$

Conclusion

Emphasize the usefulness of characterizations → elicitation

Much work still to be done

For instance :

- implement a decision deck tool for questioning the DM in terms of histograms comparisons or maps comparisons
- include contiguity in models 2 and 3
- reduce the number of parameters to be elicited → other characterization
- outranking methods