"Social" Multicriteria Evaluation: An Introduction

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Structure of the talk

•What is multi-criteria evaluation
•Why Social Multi-criteria Evaluation (SMCE)?
•How such an approach should be developed?
•Conclusions



Figure 1. A lexicographic Decision Process

The Lexicographic Model

- LEARNING
- PROCESS
- WEIGHTS
- COMPENSABILITY







Example of an impact matrix

	Alterr	natives		
Criteria	\mathbf{a}_1	\mathbf{a}_2	a ₃	a ₄
g ₁	$g_1(a_1)$	$g_1(a_2)$	•	$g_1(a_4)$
g ₂	•	•	•	•
g ₃	•	•	•	•
g ₄	•	•	•	•
g ₅	•	•	•	•
g ₆	$g_6(a_1)$	$g_6(a_2)$	•	$g_6(a_4)$



addressed

gnored

Complexity is an inherent property of natural and social systems



COMPLEX SYSTEMS CANNOT BE CAPTURED BY A SINGLE DIMENTION/PERSPECTIVE

Complexity: the ontological dimension

the existence of *different levels and scales* at which a hierarchical system can be analyzed implies the unavoidable existence of non-equivalent descriptions of it

Orientation of the coastal line of Maine



Complexity: the epistemological dimension



1 measurable property: MONETARY 1 particular perspective: EFFICIENCY hard topology only RCC KED) statistical value of human life



A co-evolutionary interpretation of a city



"The issue is not whether it is only the marketplace that can determine value, for economists have long debated other means of valuation; our concern is with the assumption that in any dialogue, all valuations or "numeraires" should be reducible to a single one-dimension standard".

(Funtowicz and Ravetz, 1994, p. 198)

Strong comparability



Weak comparability

incommensurability

GOVERNANCE in a COMPLEX world

Contradictory scientific findings and lay opinions must be integrated into the policy.

•Who has the power to impose a language of valuation?

•Who has the power to privilege one analytical level or time-space scale?

•Who has the power to simplify the complexity?

Multi-, inter-, trans-disciplinarity?

- Multi-: each expert takes his part
- Inter-: methodological choices are discussed across the disciplines
 - Informing the others about object matter
 - Criticism, reflexivity
- Trans-: What is it?

Consequences: 1) MULTIDISCIPLINARITY





Consequence: 2) PARTICIPATIVE TECHNIQUES

- In-Depth Interviews
- Focus Groups
- Questionnaires
- Institutional Analysis
- e_democracy

Objectives and Methodology of DIAFANIS



Consequences: 3) ETHICS MATTERS



Weights in a social framework

Political Democracy

Economic Democracy

Sustainability

Precautionary Principle



Consequence: 4)THE AXIOMATIZATION ISSUE

The Plurality Rule



The Plurality Rule!

Number of criteria	3	5	7	6
	a	a	b	С
	b	С	d	b
	С	b	С	d
	d	d	a	a

An Original Condorcet's Numerical Example

Number of criteria	23	17	2	10	8
	a	b	b	С	С
	b	С	a	a	b
	С	a	С	b	a

The Borda Solution

	Alternatives	a	b	С	
Ranking					Points
1-st		23	19	18	2
2-nd		12	31	17	1
3-rd		25	10	25	0

a = 58, b = 69, c = 53

The Condorcet Solution

$$\begin{bmatrix} a & b & c \\ a & 0 & 33 & 25 \\ b & 27 & 0 & 42 \\ c & 35 & 18 & 0 \end{bmatrix}$$

It is: aPb, bPc and cPa, thus due to the transitive property a cycle exists and no alternative can be selected!

Fishburn Numerical Example on Borda Rule

Number of criteria	3	2	2
	С	b	a
	b	a	d
	a	d	С
	d	С	b

Fishburn Numerical Example on Borda Rule

	Alternatives	a	b	С	d	
Ranking						Points
1-st		2	2	3	0	3
2-nd		2	3	0	2	2
3-rd		3	0	2	2	1
4-th		0	2	2	3	0

$$a = 13, b = 12, c = 11, d = 6$$

Fishburn Numerical Example on Borda Rule

• Let's now suppose that alternative *d* is removed from the analysis. Since *d* was at the bottom of the ranking, nobody should have any reasonable doubt that alternative *a* is still the best alternative. Let's check if this assumption is correct.

Fishburn Numerical Example on Borda Rule Alternatives b a С **Points** Ranking 2 3 2 1-st 2 3 2 2-nd 2 1 3 2 2 3-rd 0

Frequency Matrix Without *d*

a = 6, b = 7, c = 8

thus alternative *c* is now chosen!

• Both social choice literature and multi-criteria decision theory agree that whenever the majority rule can be operationalized, it should be applied. However, the majority rule often produces undesirable intransitivities, thus *"more limited ambitions are compulsory. The* next highest ambition for an aggregation algorithm is to be Condorcet" (Arrow and Raynaud, 1986, p. 77).

Applying the maximum likelihhod ranking procedure to the original Condorcet Example

a	b	С	100
b	С	a	104
С	a	b	86
b	a	С	94
C	b	a	80
a	С	b	76

Sustainal	oility	v Ind	licator
Oustania		y 111 0	icator

	Indic.	GDP	Unemp. Rate	Solid wastes	Inc. disp.	Crime rate
Country						
Α		25,000	0.15	0.4	9.2	40
В		45,000	0.10	0.7	13.2	52
С		20,000	0.08	0.35	5.3	80
weights		0.165	0.165	0.333	0.165	0.165

	Α	В	С
Α	0	0.666	0.333
В	0.333	0	0.333
С	0.666	0.666	0

ABC = 0.666 + 0.333 + 0.333 = 1.333BCA = 0.333 + 0.666 + 0.333 = 1.333**CAB** = 0.666 + 0.666 + 0.666 = 2ACB = 0.333 + 0.666 + 0.666 = 1.666BAC = 0.333 + 0.333 + 0.333 = 1CBA = 0.666 + 0.333 + 0.666 = 1.666

The Computational problem

Moulin (1988, p. 312) clearly states that the Kemeny method is *"the correct method"* for ranking alternatives, and that the *"only drawback of this aggregation method is the difficulty in computing it when the number of candidates grows"*.

One should note that the number of permutations can easily become unmanageable; for example when 10 alternatives are present, it is 10!=3,628,800.

A NP-hard problem

- The <u>complexity class</u> of <u>decision problems</u> that are intrinsically harder than those that can be solved by a <u>nondeterministic Turing machine</u> in <u>polynomial time</u>. When a decision version of a combinatorial <u>optimization problem</u> is proved to belong to the class of <u>NP-complete</u> problems, then the optimization version is NP-hard.
- (definition given by the National Institute of Standards and Technology, http://www.nist.gov/dads/HTML/nphard.html)

- This NP-hardness has discouraged the development of algorithms searching for exact solutions, thus the majority of the algorithms which have been proposed in the literature; are mainly
- heuristics based on artificial intelligence,
- branch and bound approaches and
- multi-stage techniques

(see e.g., Barthelemy et al., 1989; Charon et al., 1997; Cohen et al., 1999; Davenport and Kalagnam, 2004; Dwork et al., 2001; Truchon, 1998b). • A *new numerical algorithm* aimed at solving the computational problem connected to linear median orders by finding *exact solutions* has been proposed by Munda (2005). Main characteristics of this algorithm are that linear median orders are computed by using their theoretical equivalence with maximum *likelihood rankings* and that *outranking matrixes* are used as a starting computational step.

$$\begin{cases} a_{j}Pa_{k} \iff g_{m}(a_{j}) > g_{m}(a_{k}) + p \\ a_{j}Qa_{k} \iff g_{m}(a_{k}) + p \ge g_{m}(a_{j}) > g_{m}(a_{k}) + q \\ a_{j}Ia_{k} \iff \begin{cases} g_{m}(a_{k}) + q \ge g_{m}(a_{j}) \\ g_{m}(a_{j}) + q \ge g_{m}(a_{k}) \end{cases}$$

Taking into account *intensity of preference*

A Real-World Application for 146

Countries



Figure 1. Comparison of rankings obtained by the linear aggregation (ESI2005) and the non-compensatory (NCMA) rules

	Aggregation	ESI rank with LIN	rank with NCMC	Change in Rank
	Azerbaijan	99	61	38
ent	Spain	76	45	31
rovem	Nigeria South	98	69	29
Imp	Africa	93	68	25
	Burundi	130	107	23
	Indonesia	75	114	39
ution	Armenia	44	79	35
rior	Ecuador	51	78	27
Detei	Turkey	91	115	24
П	Sri Lanka	79	101	22
	Average change o	ver 146 countr	ies	8

 Table 3. ESI rankings obtained by linear aggregation (LIN) and non-compensatory rule (NCMC): countries that largely improve or worsen their rank position

Matrix type Impact Case Study				
Alternatives Criteria	Budapest	Moscow	Amsterdam	New York
Houses owned (%)	50.5	40.2	2.2	10.3
Residential density (pers. /hectare)	123.3	225.2	152.1	72
Use of private car (%)	31.1	10	60	32.5
Mean travel time to work (minutes)	40	62	22	36.5
Solid waste generated per capita (t./year)	0.2	0.29	0.4	0.61
City product per person (US\$/year)	4750	5100	28251	30952
Income disparity (Q5/Q!)	9.19	7.61	5.25	14.81
Households below poverty line (%)	36.6	15	20.5	16.3
Crime rate per 1000 (theft)	39.4	4.3	144.05	56.7



Normalisation technique used for the different measurement units dealt with.

Scale adjustment used, for example population or GDP of each country considered.

Common measurement unit used (money, energy, space and so on).

100	78.674	0	16.770
33.485	100	52.28	0
42.2	0	100	45
45	100	0	36.25
0	21.95	48.78	100
0	1.335	89.691	100
41.213	24.686	0	100
100	0	25.462	6.018
25.116	0	100	37.495

Normalized Impact Matrix

100	78.674	0	16.770
66.515	0	47.72	100
57.8	100	0	55
55	0	100	63.75
100	78.05	51.22	0
0	1.335	89.691	100
58.787	75.314	100	0
0	100	74.538	93.982
74.884	100	0	62.505

Normalised Impact Matrix Accounting for Minimisation Objectives

Budapest = 512.986 Moscow = 533.373 Amsterdam = 463.169 New York = 492.052



- Weights in linear aggregation rules have always the meaning of trade-off ratio. In all constructions of a composite indicator, weights are used as importance coefficients; as a consequence, a theoretical inconsistency exists.
- The assumption of **preference independence** is essential for the existence of a linear aggregation rule. Unfortunately, this assumption has very strong consequences which often are not desirable in sustainability indicators. The use of a linear aggregation procedure implies that among the different ecosystem aspects there are not phenomena of synergy or conflict. This appears to be quite an unrealistic assumption.
- In linear aggregation rules, **compensability** among the different individual indicators is always assumed; this implies complete substitutability among the various components considered. For example, in a sustainability index, economic growth can always substitute any environmental destruction or inside e.g., the environmental dimension, clean air can compensate for a loss of potable water. From a descriptive point of view, such a complete compensability is often not desirable.

	Budapest	Moscow	Amsterdam	New York
Budapest	0	4	4	5
Moscow	5	0	5	б
Amsterdam	5	4	0	3
New York	4	3	6	0

Outranking Matrix of the 4 Cities According to the 9 Indicators

В	Α	D	С	31	С	В	D	А	27
В	D	С	Α	31	D	В	А	С	27
A	В	D	С	30	D	С	В	A	27
В	D	A	С	30	A	С	В	D	26
В	С	A	D	29	А	D	С	В	26
В	А	С	D	28	D	А	В	С	26
В	С	D	А	28	D	С	А	В	26
С	В	А	D	28	D	А	С	В	25
D	В	С	А	28	С	А	D	В	24
А	В	С	D	27	С	D	В	А	24
A	D	В	С	27	А	C	D	В	23
С	А	В	D	27	C	D	А	В	23

В	Α	D	С	31
В	D	С	Α	31

Where A is Budapest, B is Moscow, C is Amsterdam and D is New York.

Economic dimension City product per person Environmental dimension Use of private car Solid waste generated per capita Social dimension Houses owned **Residential density** Mean travel time to work Income disparity Households below poverty line Crime rate

A reasonable decision might be to consider the three dimensions equally important. This would imply to give the same weight to each dimension considered and finally to split this weight among the indicators. That is, each dimension has a weight of 0.333; then the economic indicator has a weight of 0.333, the 2 environmental indicators have a weight of 0.1666 each, and each one of the 6 social indicators receives a weight equal to 0.0555. As one can see, if dimensions are considered, weighting indicators by means of importance coefficients is crucial.

	Budapest	Moscow	Amsterdam	New York
Budapest	0	0.3	0.4	0.4
Moscow	0.7	0	0.5	0.6
Amsterdam	0.6	0.5	0	0.3
New York	0.6	0.4	0.7	0

Weighted Outranking Matrix

BDCA

Where A is Budapest, B is Moscow, C is Amsterdam and D is New York.

CONCLUSION:

Results are heavily dependent on the problem structuring step!!



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