



**NEW TOOLS FOR THE
CONSTRUCTION, ANALYSIS
AND INTERPRETATION OF
SOCIAL INDICATORS
BASED ON ORDINAL VARIABLES**

Filomena Maggino – Marco Fattore



PREMISE



Issues in creating indicators representing phenomena, for evaluation and governance aims.



PREMISE → 1 BETWEEN ACCURACY AND AMBIGUITIES



Socio-economic phenomena
can be measured and represented
by means of

- “hard” approaches (e.g., financial analysis) → *sometimes*
- “soft” approaches → *often*



PREMISE → 1 BETWEEN ACCURACY AND AMBIGUITIES



This because the true nature of
socio-economic phenomena



very often **qualitative** and **ordinal**



Ordinal data are
the true expression of real phenomena

not just a rough approximation of true precise,
yet non-observable, variables;



PREMISE → 1

BETWEEN ACCURACY AND AMBIGUITIES



Ambiguities and nuances of socio-economic phenomena are not an obstacle to be removed; they often are what really matters.



PREMISE → 2 BETWEEN OBJECTIVITY AND SUBJECTIVITY



Defining and using data in socio-economic statistics inevitably involves **subjectivity**.

This is also true for decision making purposes.



PREMISE → 2

BETWEEN OBJECTIVITY AND SUBJECTIVITY



This is not an issue in itself, since the knowledge process always involves

- “objectivity”, in observational methods
- “subjectivity”, in definitions and other choices (conceptual framework, data definitions, analytical approaches, ...)

The epistemological research of the last century clearly showed as objectivism cannot account for the knowledge process (just like idealism)



PREMISE → 2 BETWEEN OBJECTIVITY AND SUBJECTIVITY



So, using *subjectivity* is completely consistent with the aims of the socio-economic analysis.

The real issue is not whether using subjectivity or not; it is how to consistently combine subjectivity and the need to observe and analyse data consistently and objectively.



PREMISE → 2 BETWEEN OBJECTIVITY AND SUBJECTIVITY



Subjective choices are unavoidable and their use is fully justifiable.

Real issue → *how to build a sound methodological process, where the subjective choices are clearly stated, while their consequences are worked out in a formal and unambiguous way.*



Final results will be clearly interpretable and the role of subjective inputs and sound formal computations can be clearly distinguished and understood.



PREMISE → 3 DATA METRICS



When dealing with ordinal data, common statistical practice is not quite clearly.

With the aim of pursuing metric analysis out of non-metric data, a lot of arbitrary choices are often taken in data analysis



Final result depends upon subjective choices.



PREMISE → 3 DATA METRICS



In the end, it is not clear whether the results reflect real facts and sound interpretations or are induced by arbitrary methodological choices (e.g., how non-metric data are turned into metric scales).



INDICATORS CONSTRUCTION

1. → the process

2. → methodological critical issues

3. → the alternative approach

AN APPLICATION

4. → traditional approach

5. → alternative approach

6. → state of the art and future perspectives



The process



Two phases:

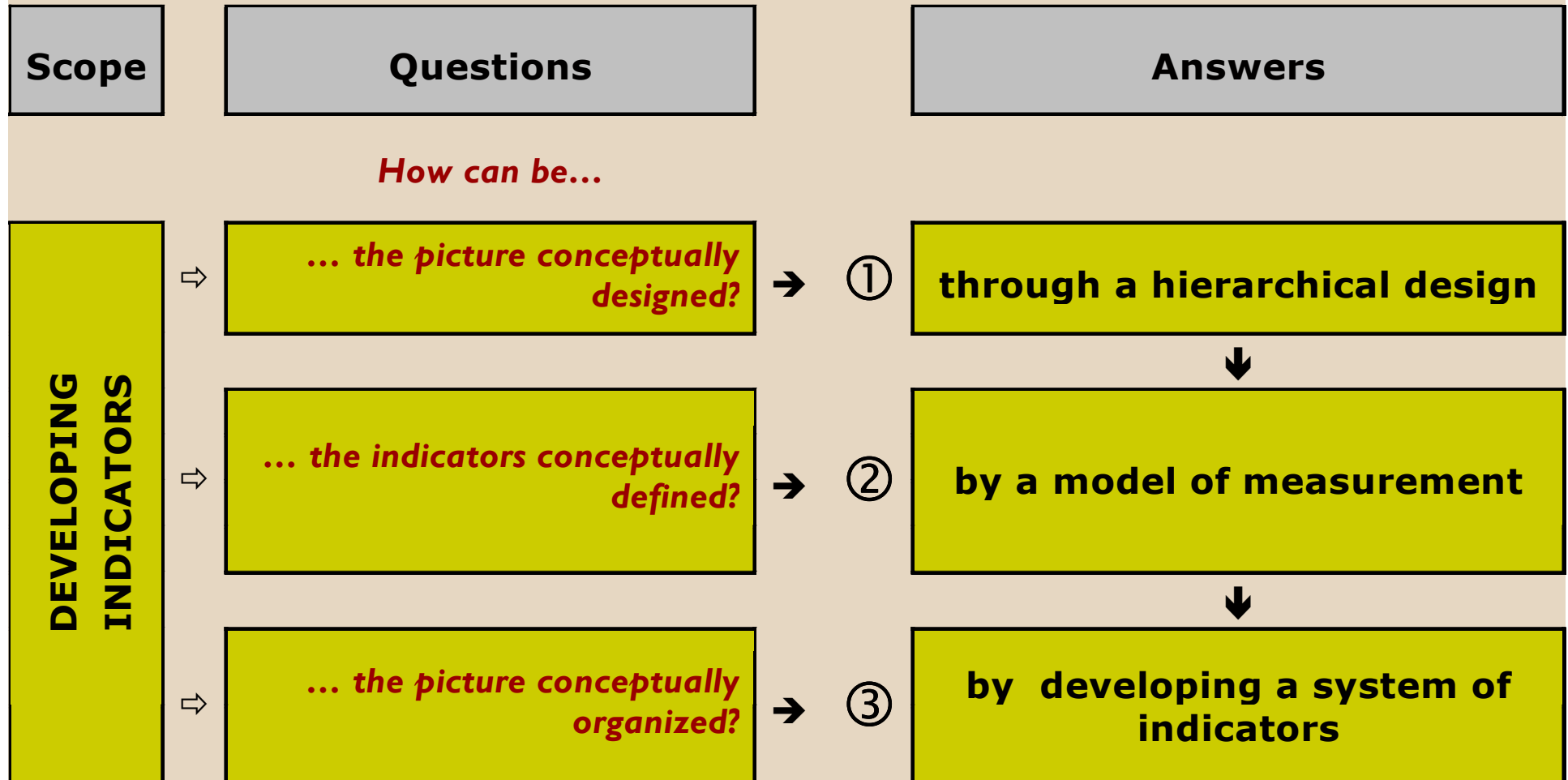
- I. CONCEPTUAL DEFINITION (FRAMEWORK AND STRUCTURE)**
- II. ANALYTICAL TOOLS AND STRATEGIES**



The process



I. CONCEPTUAL DEFINITION (FRAMEWORK AND STRUCTURE)





The process



hierarchical design ← 1

Indicators should be developed through a logical modelling process conducting from concept to measurement. Given its features, this logical design is defined *hierarchical*, since each component is defined and finds its meaning in the ambit of the preceding one.

Conceptually, the hierarchical design is characterized by the following components:

- (i) the conceptual model,
- (ii) the areas to be investigated,
- (iii) the latent variables, and
- (iv) the elementary (basic) indicators.



The process

model of measurement ← ②

A further component of the hierarchical design definition is represented by the relationships between:

- *Latent variables and the corresponding indicators*: these relations define the model of measurement. Consistently with the measurement model, also the relationship between the *elementary indicators* should be defined.
- *Latent variables for a given area*: these relations are defined in the ambit of the conceptual model and identify the structural pattern (modelling indicators).



The process



system of indicators ← ③

A system of indicators represents the fulfilment of the conceptual framework. Moreover, it

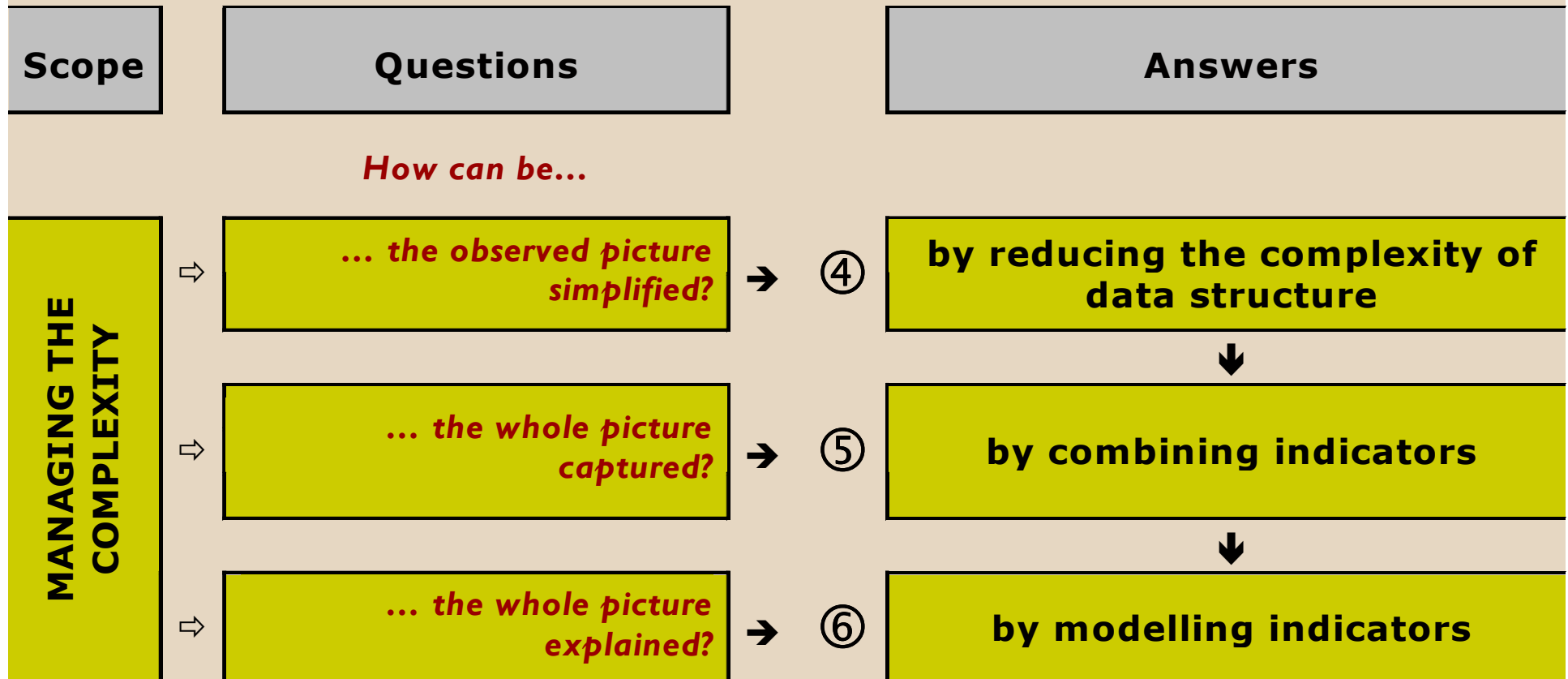
- offers an effective organizational context, relying on methodological supports and allowing data to be managed;
- allows structured and systematic data to be used, observed in long-term longitudinal perspective. This is particularly demanding with reference to subjective data, which require a great use of resources (beyond a solid survey research methodology).



The process



II. ANALYTICAL TOOLS AND STRATEGIES





The process

reducing the complexity of data ← 4

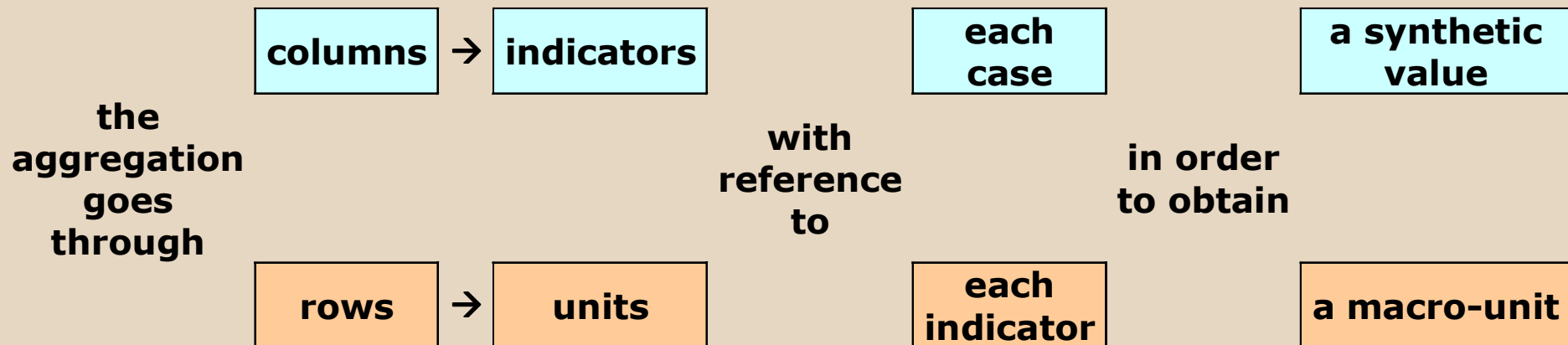
The consistent application of the hierarchical design produces a complex data structure (elementary indicators, cases, variables, areas, etc.). In order to manage the complexity:

- aggregating elementary indicators for each variable → reconstructing the conceptual variables consistently with the approach (reflective or formative) adopted at micro level (construction of synthetic indicators)
- aggregating units/cases: leading information observed at micro-level to the proper macro level (definition of macro-units).



The process

reducing the complexity of data ← 4





The process

combining indicators ← 5

In some occasion, the complexity of the system of indicators may require the indicators allowing for more comprehensive measurement, in order to (Noll, 2009)

- answer the call by 'policy makers' for condensed information
- improve the chance to get into the media (compared to complex indicator systems)
- make multi-dimensional phenomena uni-dimensional
- compare situations across time more easily
- compare cases (e.g. nations) in a transitive way (ranking)
- to observe and record change across time, difference between groups of population or comparison between cities, countries, ...

Dashboards or composite indicators → useful approaches for summarising indicators.



The process



modelling indicators ← ⑥

This stage is aimed at analysing different aspects of the defined model (e.g. objective and subjective indicators) in order to find explanation by identifying the proper analytical approaches.



INDICATORS CONSTRUCTION

1. → the process

2. → methodological critical issues

3. → the alternative approach

AN APPLICATION

4. → traditional approach

5. → alternative approach

6. → state of the art and future perspectives



Methodological critical issues



social indicators construction



consolidated tradition

however

critical issues remained unsolved and unsettled



Methodological critical issues



with reference to difficulty in dealing with data which

- refer to a complex reality
- are ambiguous and softened
- are multidimensional
- are dynamic and evolutionary
- are qualitative also when quantitatively measured
- contain errors and approximations
- are sensitive



Methodological critical issues



new challenges and perspectives
to improve technical tools strategies
with reference to

- reducing data structure in order to aggregate
 - units
 - indicators
- combining indicators
- communicating the “picture” obtained through the indicators (correctly and significantly representing and showing results).



Methodological critical issues



new challenges and perspectives
to improve analytical tools and strategies
which should take into account

- nature of data → generally ordinal
- process and trends of phenomena → monotonic



INDICATORS CONSTRUCTION

1. → the process

2. → methodological critical issues

3. → the alternative approach

AN APPLICATION

4. → traditional approach

5. → alternative approach

6. → state of the art and future perspectives



Introduction



The particular application illustrated here is aimed at
comparing
the traditional and the alternative approach
to reduce the complexity of data structure,
by using subjective and objective data provided by the
European Social Survey project.

In particular, we selected the following information:



Introduction



European Social Survey						
Area	Variable	Items	Item number		Scaling technique	Model of measurement
				R1 (2002)		
Politics	Self-placement	placement on left-right scale	B28	LRSCALE	0 (left) – 10 (right)	
Subjective aspects	Life satisfaction	how satisfied with life as a whole	B29	STFLIFE	0 (extremely dissatisfied) – 10 (extremely satisfied)	
Immigration and asylum issues	Acceptance of immigration: allow	many/few immigrants of same race/ethnic group as majority	D4	IMSMETN	1. allow many 2. allow some 3. allow a few 4. allow none to come and live here	reflective
		many/few immigrants of different race/ethnic group from majority	D5	IMDFETN		
		many/few immigrants from richer countries in Europe	D6	EIMRCNT		
		many/few immigrants from poorer countries in Europe	D7	EIMPCNT		
		many/few immigrants from richer countries outside Europe	D8	IMRCNTR		
		many/few immigrants from poorer countries outside Europe	D9	IMPCNTR		
Socio-demographic profile	Income	feeling about household's income nowadays	F31	HINCFEL	1. living comfortably 2. coping 3. difficult 4. very difficult on present income	



INDICATORS CONSTRUCTION

1. → the process

2. → methodological critical issues

3. → the alternative approach

AN APPLICATION

4. → traditional approach

5. → alternative approach

6. → state of the art and future perspectives



Traditional approach

First stage: synthesizing indicators at individual level

Goal: synthesizing indicators related to each variable consistently with the adopted model of measurement (reflective or formative).



Traditional approach



First stage: synthesizing indicators at individual level

First level

Acceptance of immigration

Variable	Items	Item code	Loading
Acceptance of immigration: allow	many/few immigrants of same race/ethnic group as majority	IMSMETN	.8
	many/few immigrants of different race/ethnic group from majority	IMDFETN	.9
	many/few immigrants from richer countries in Europe	EIMRCNT	.7
	many/few immigrants from poorer countries in Europe	EIMPCNT	.9
	many/few immigrants from richer countries outside Europe	IMRCNTR	.8
	many/few immigrants from poorer countries outside Europe	IMPCNTR	.9
Total variance explained (%)			70
Cronbach's alpha			.94



Traditional approach

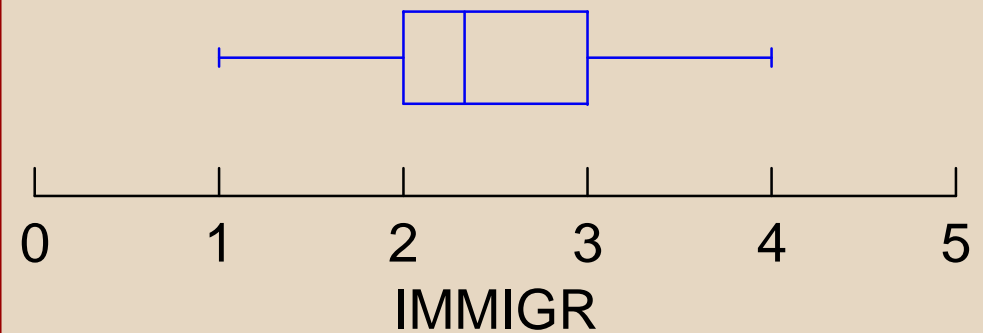
First stage: synthesizing indicators at individual level

First level

Acceptance of immigration

Synthetic score (IMMIGR)

Minimum	1.0
Maximum	4.0
Median	2.3
Mean	2.4
Standard Dev	0.7
Skewness	0.1
Kurtosis	-0.2



1 (allow many) - 4 (allow none)



Traditional approach



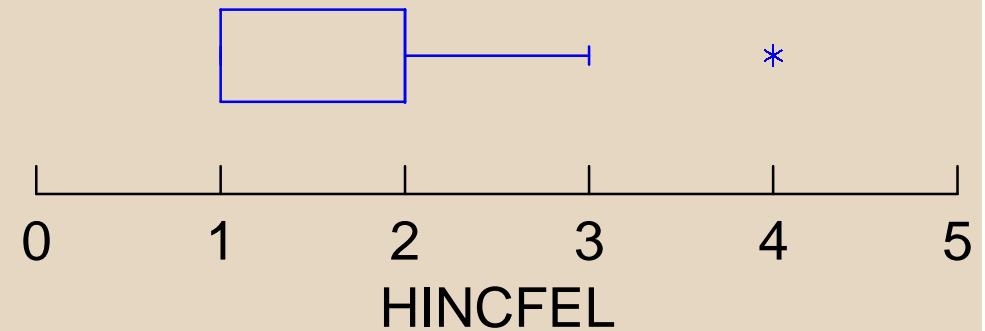
First stage: synthesizing indicators at individual level

First level

feeling about household's income nowadays

Synthetic score (HINCFEL)

Minimum	1.0
Maximum	4.0
Median	2.0
Mean	2.0
Standard Dev	0.8
Skewness	0.6
Kurtosis	-0.2



1 (comfortably) – 4 (difficult)



Traditional approach



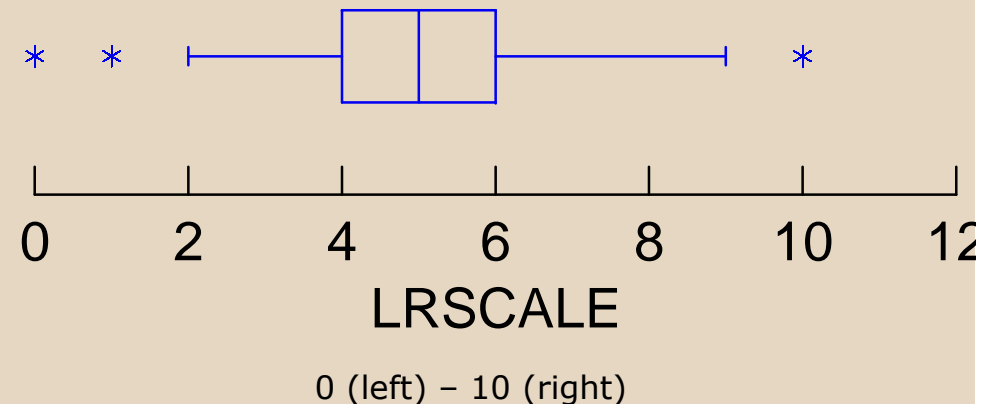
First stage: synthesizing indicators at individual level

First level

Political placement on left-right scale

Synthetic score (LRSCALE)

Minimum	0.0
Maximum	10.0
Median	5.0
Mean	5.1
Standard Dev	2.2
Skewness	-0.0
Kurtosis	0.1





Traditional approach



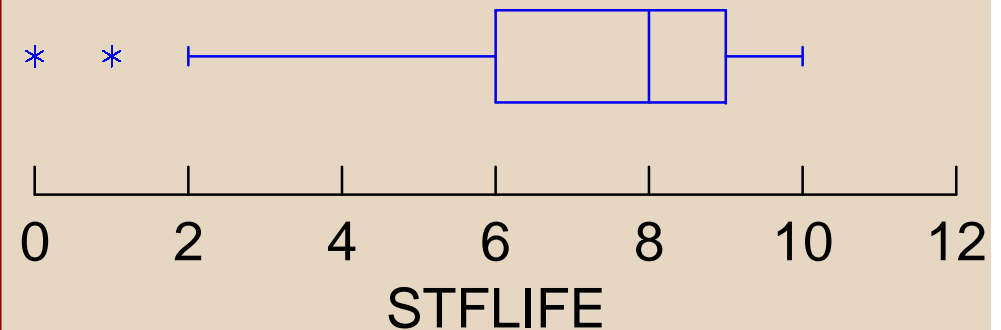
First stage: synthesizing indicators at individual level

First level

how satisfied with life as a whole

Synthetic score (STFLIFE)

Minimum	0.0
Maximum	10.0
Median	8.0
Mean	7.0
Standard Dev	2.3
Skewness	-1.0
Kurtosis	0.7



0 (extremely dissatisfied) – 10 (extremely satisfied)



Traditional approach

First stage: synthesizing indicators at individual level

Second level

In order to synthesise the identified indicators, the traditional approach counts on the application of a synthesis technique (e.g., PCA).

In our case, the PCA results did not allow any meaningful synthesis since it produced two components on four indicators (!!!)



Traditional approach

Second stage: defining macro-units

Goal: synthesizing indicators observed at individual level in order to ascribe a synthetic value to groups.

The aggregation can be done through

- additive approach: a single value synthesizes the values observed at micro level (also through further indicators aggregation processes “second-level indicators aggregation”);
- compositional approach: when micro-units’ macro-units’ values are obtained by aggregating individual values in a certain number of homogeneous sub-groups

In our case, we adopted the latter approach in order to simultaneously aggregate indicators and cases.



Traditional approach



Country level
of
acceptance

Country	Acceptance mean score
AT	2.61 (rank → 8)
BE	2.41 (rank → 5)
CH	2.18 (rank → 1)
CZ	2.46 (rank → 6)
DE	2.32 (rank → 3)
DK	2.31 (rank → 2)
ES	2.38 (rank → 4)
FI	2.53 (rank → 7)
Overall	2.42



INDICATORS CONSTRUCTION

1. → the process

2. → methodological critical issues

3. → the alternative approach

AN APPLICATION

4. → traditional approach

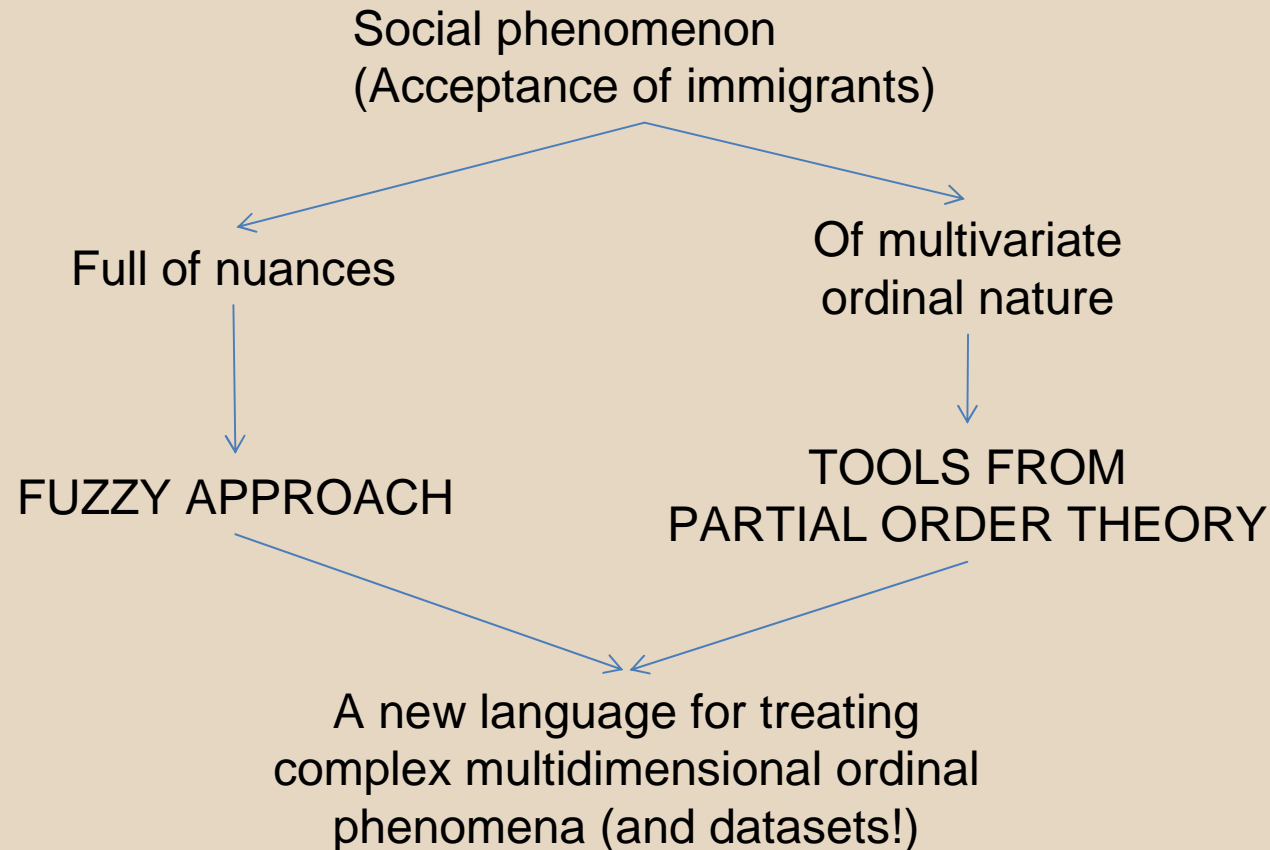
5. → alternative approach

6. → state of the art and future perspectives



Alternative approach

Searching for new formal languages...





Alternative approach

Partial order analysis through a simple example

Many ordinal variables recorded on a population,



individuals cannot be directly ordered,
since each variable is likely to induce different rankings

The most natural way to represent such data is through
a partial order.



Alternative approach

Partial order analysis through a simple example

We introduce basic concepts using a simple example, based on considering two variables from the European Social Survey, namely variables D4 (IMSMETN) and D5 (IMDFETN):

- D4: acceptance of many/few immigrants of same race/ethnic group as majority;
- D5: acceptance of many/few immigrants of different race/ethnic group from majority.

Both variables are recorded on a four grade scale from 1 (allow many) to 4 (allow none).

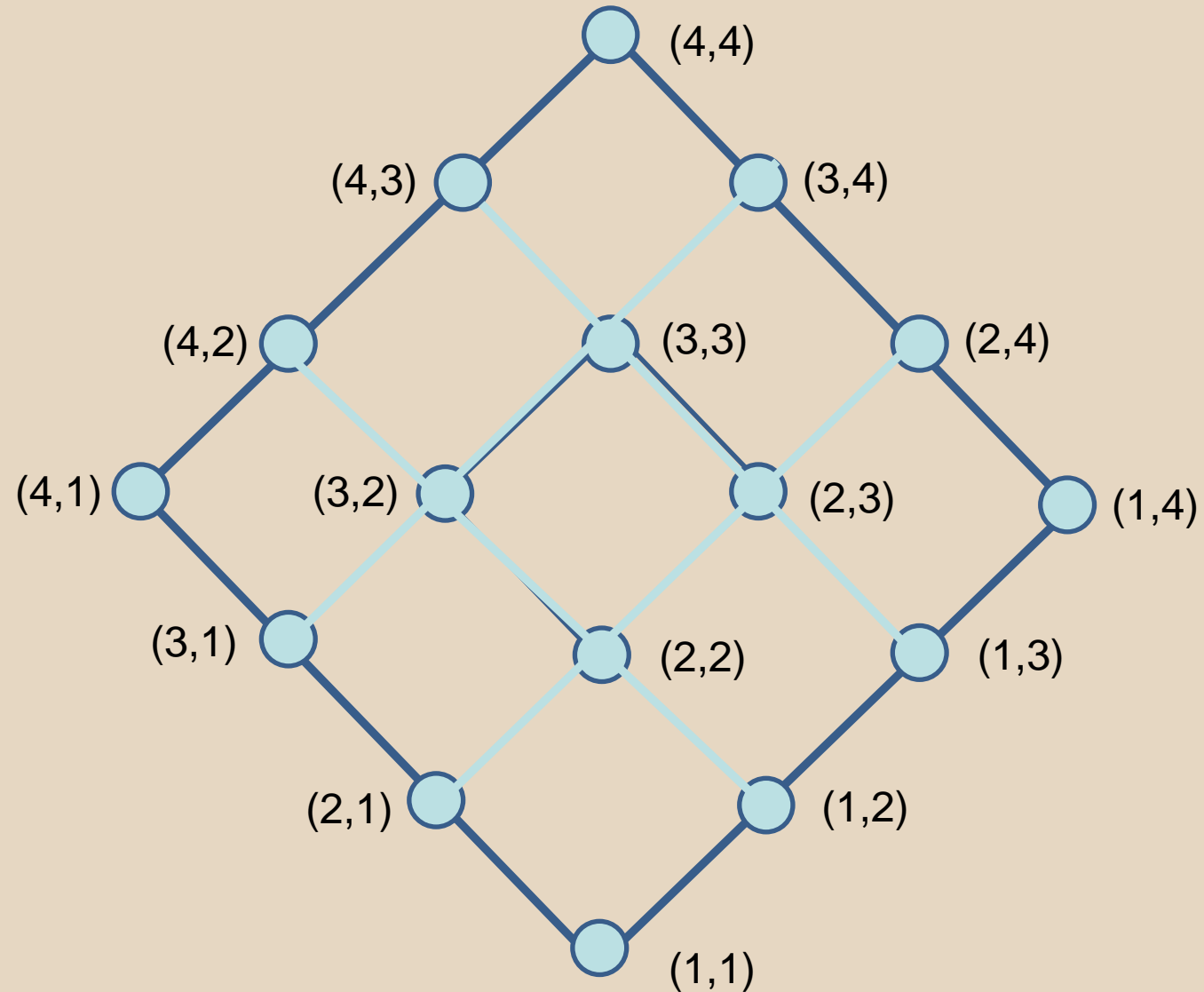
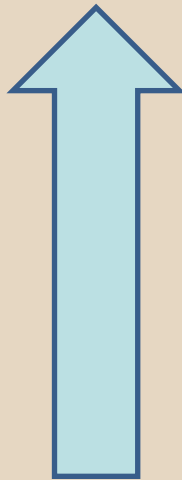


Alternative approach

Acceptance configurations on D4 and D5

example:
(4,2) stands for
(D4=4, D5=2)

Clos to immigrants





Alternative approach



Assessing the degree of acceptance

It is clear that if $(a,b) \leq (c,d)$, then the degree of rejection of immigration of (c,d) is greater than that of (a,b) . But:

- Is it possible to assess to what extent it is greater?
- Is it possible to assign to each configuration the corresponding degree?



Alternative approach

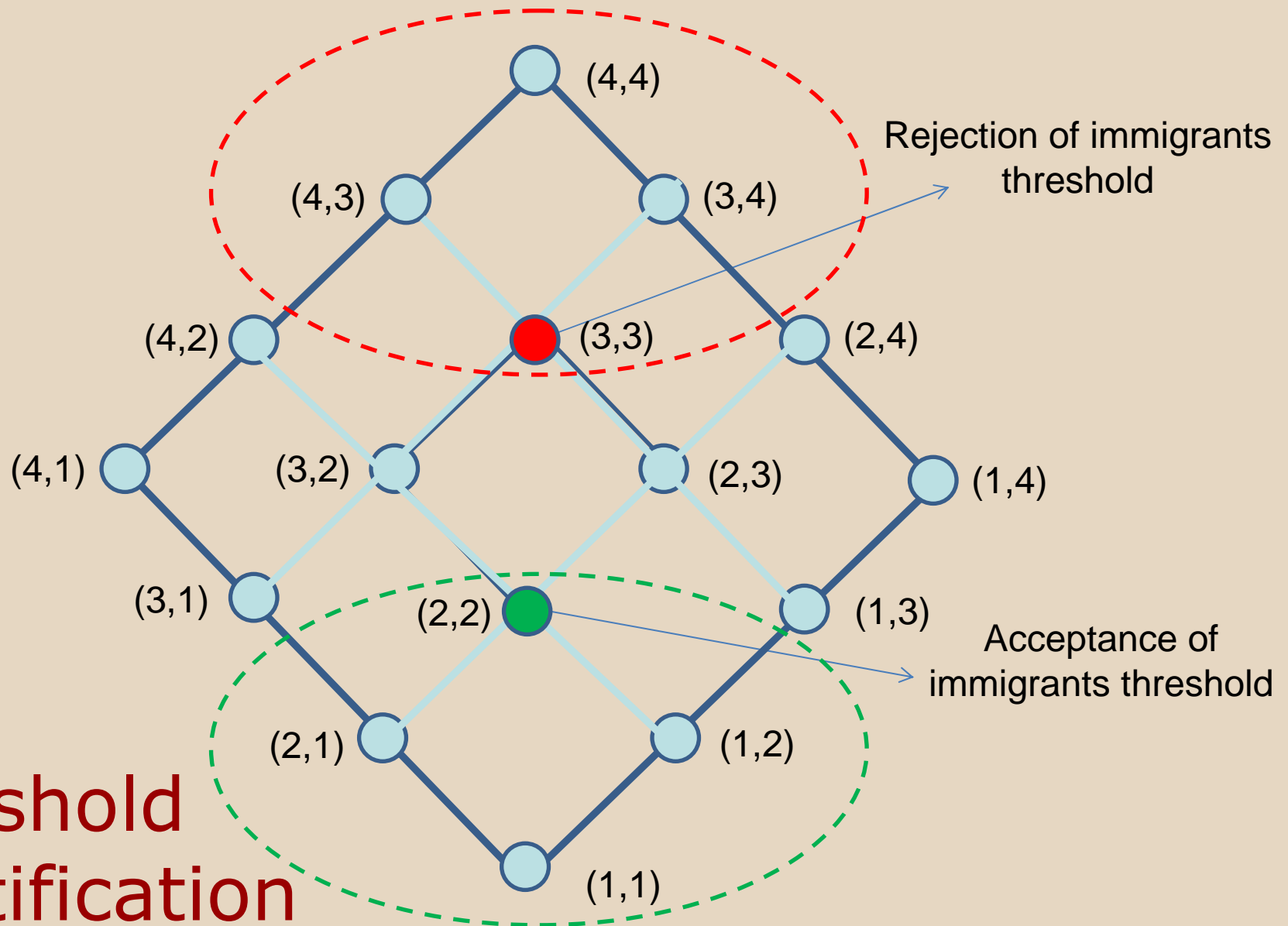
Assessing the degree of acceptance

Yes, if we suppose that some configurations are identified as definitely representing closeness to immigrants or acceptance of immigrants, that is, **if suitable rejection and acceptance thresholds are identified.**

Here **subjectivity** enters but all the implications of the choice of such thresholds are then **derived based only on the data structure.**



Alternative approach



Threshold identification



Alternative approach

Assessing the degree of acceptance

- Let us agree, in a fuzzy evaluation perspective (so as to take explicitly into account nuances), that nodes in the red ellipse has degree of rejection of immigration equal to 1 (the maximum) and that nodes in the green ellipse has degree of rejection of immigration equal to 0 (i.e. the minimum).
- All other nodes should receive a degree of rejection between 0 and 1, reproducing the ambiguities in the phenomenon.



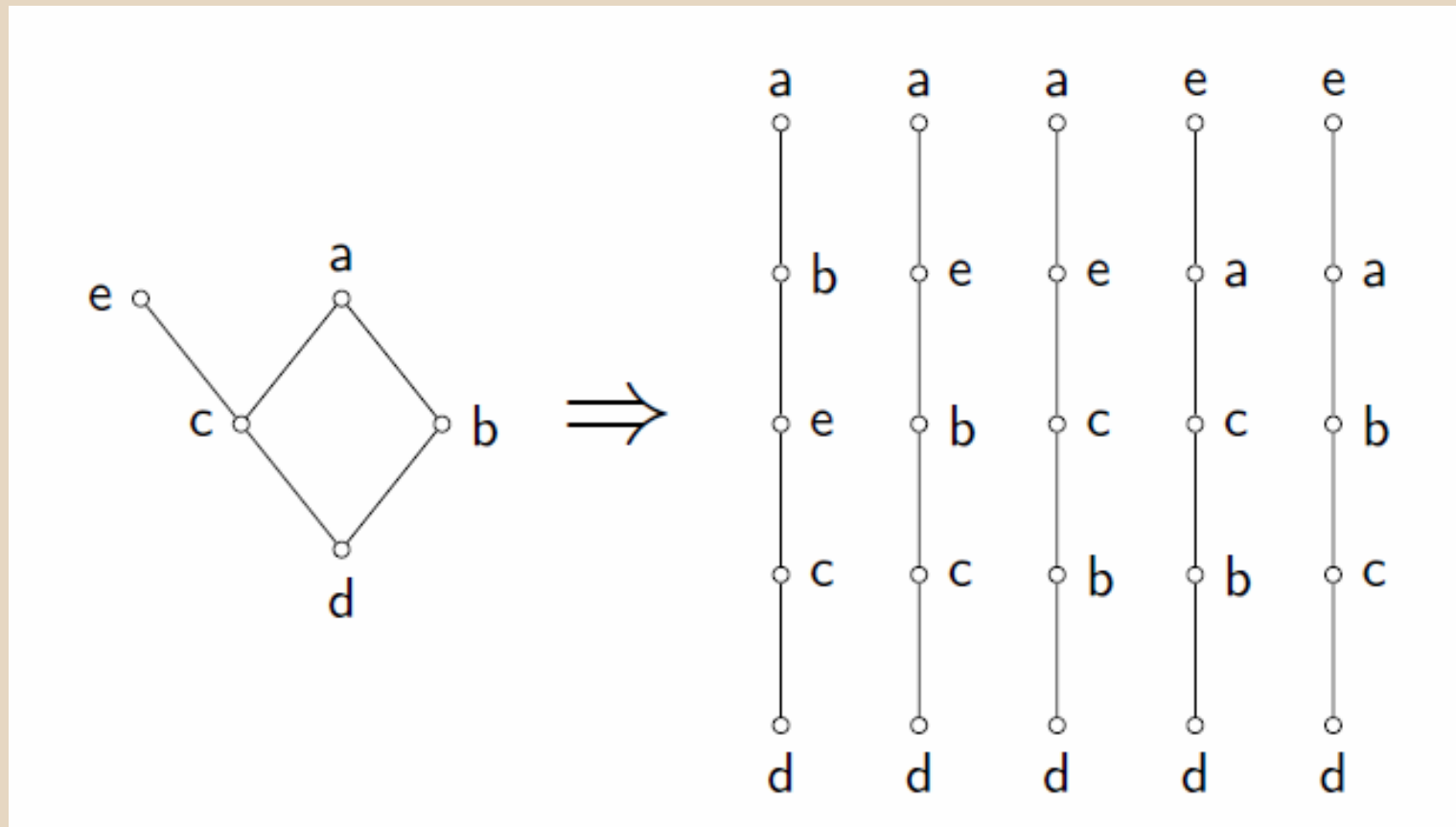
Alternative approach

Assessing the degree of acceptance

- The computation of such degrees is based **only on the analysis of the partial order structure of the poset**, that is, it is based on the analysis of the different **relational position** of each node, with respect to the thresholds selected.
- The required information about the degree of acceptance/rejection of immigration is extracted from the structure of the poset and not on the aggregation of variable scores (which are treated as they are, i.e. as **ordinal** variables).

Alternative approach

Linear extensions of a poset

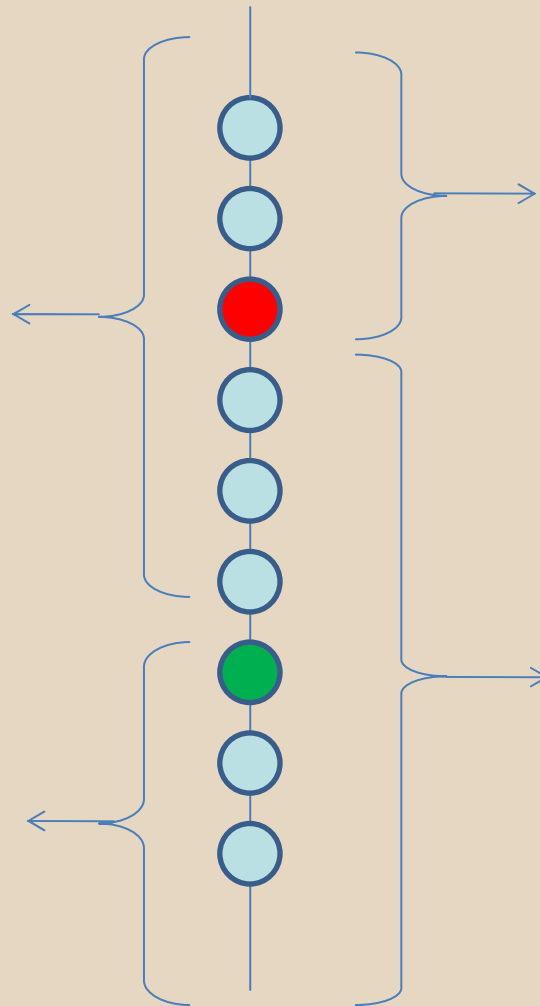




Alternative approach

The basic idea: pick up a linear extension ...

These states receive degree of immigrant acceptance equal to 0.



These states receive degree of closeness to immigrants equal to 1.

These states receive degree of immigrant acceptance equal to 1.

These states receive degree of closeness to immigrants equal to 0.



Alternative approach

The basic idea

- In the end, for each state we get two degrees:
 - the **first** measures to what extent it can be classified as belonging to the group of states representing people who **do not accept** immigrants; → **deg1**
 - the **second** measures to what extent it can be classified as belonging to the group of states representing people who **do accept** immigrants. → **deg2**
- Turning **deg1** into **1-deg1** we get an alternative measure of acceptance of immigrants (in terms of non-rejection of them).



Alternative approach



The basic idea

- So, given the
 - *rejection threshold* (red) and
 - *acceptance threshold* (green)we get two different assessments of the degree of acceptance of immigrants, corresponding to each node.
- To get the final degree of immigration acceptance, for each node we compute the average of **1-deg1** and **deg2** (it can be shown that this is the only way to get a mathematically consistent fuzzy assessment of acceptance degree, out of the two “original” assessments).



Alternative approach



Results

State	Acceptance degree	Rejection degree	Final acceptance degree
	Acceptance threshold (2,2)	Rejection threshold (3,3)	
(1,1)	1,00	0,00	1,00
(1,2)	1,00	0,00	1,00
(1,3)	0,71	0,00	0,86
(1,4)	0,42	0,09	0,66
(2,1)	1,00	0,00	1,00
(2,2)	1,00	0,00	1,00
(2,3)	0,00	0,00	0,50
(2,4)	0,00	0,39	0,31
(3,1)	0,71	0,00	0,86
(3,2)	0,00	0,00	0,50
(3,3)	0,00	1,00	0,00
(3,4)	0,00	1,00	0,00
(4,1)	0,42	0,09	0,66
(4,2)	0,00	0,39	0,31
(4,3)	0,00	1,00	0,00
(4,4)	0,00	1,00	0,00



Alternative approach



Country level
of
acceptance

Country	Acceptance degree (D4 and D5)
AT	0.49 (rank → 7.5)
BE	0.62 (rank → 4)
CH	0.74 (rank → 1)
CZ	0.52 (rank → 6)
DE	0.65 (rank → 2)
DK	0.63 (rank → 3)
ES	0.54 (rank → 5)
FI	0.49 (rank → 7.5)
Overall	0.62



Alternative approach

A more complex example

- A similar analysis has been performed on the four variables D6, D7, D8, D9.
- The resulting poset has $4 \times 4 \times 4 \times 4 = 256$ states and cannot be depicted.
- The red (*rejection*) threshold has been identified as the state $(2, 3, 2, 3)$.
- The green (*acceptance*) threshold has been identified as the state $(2, 2, 2, 2)$.
- In both cases, the thresholds have been identified for explanation purposes. A more meaningful choice requires expert's judgment and/or further analysis.



Alternative approach

Country
level
of
acceptance

Country	Acceptance degree (D6, D7, D8, D9)
AT	0.33 (rank → 8)
BE	0.52 (rank → 3)
CH	0.64 (rank → 1)
CZ	0.50 (rank → 4)
DE	0.53 (rank → 2)
DK	0.49 (rank → 5)
ES	0.48 (rank → 6)
FI	0.37 (rank → 7)
Overall	0.48

(given the thresholds)



Alternative approach



Comments

- We have given a brief example of how poset theory can be used to compute social indicators out of ordinal data, without turning them into numerical scores.
- Due to the exemplificative nature of the slides, the computed numbers should be taken just as rough measures. They depend upon the choice of the thresholds and some sensitivity analysis should be added.
- The poset describing variables D4 and D5 is very small (for presentation purposes), so the variability of the acceptance degrees over its 16 nodes is small compared to that of nodes in the poset concerning D6-D7-D8-D9 (that comprises 256 nodes). Also for this reason, the final numbers obtained in the two cases are not directly comparable.



INDICATORS CONSTRUCTION

1. → the process

2. → methodological critical issues

3. → the alternative approach

AN APPLICATION

4. → traditional approach

5. → alternative approach

6. → state of the art and future perspectives



State-of-the-art and future perspectives



State of the art

- This approach has been (and is being currently) applied to the study of material deprivation, based on EU-SILC data. We are planning to study also other social phenomena.
- Approximated analytical formulas are being developed, so that the computations involved in this approach can be performed without relying on heavy and complex numerical algorithms.
- It is already possible to define thresholds composed of more than a single node. This makes the proposed approach more flexible to real situations.



State-of-the-art and future perspectives



Future perspectives

- Integration of poset analysis and Structural Equation Modeling.
- Definition of algorithms to help identifying thresholds.
- Definition of “weighting” schemes for ordinal variables, i.e. of a way to take into account the different relevance of different variables, without introducing numerical weights.
- Definition of clustering algorithms, for reducing the dimension of posets, when the number of variables and/or the number of possible scores for each variable is high.

In practice: developing a full set of analytical tools for dealing with ordinal variables in a consistent way.



References

- Annoni P., Brueggemann R., “Exploring partial order of European countries”, *Social indicators research*, 93(3), pp. 471-487, 2008.
- Davey B. A.; Priestley H. A. *Introduction to Lattices and Order*, CUP, 2002.
- Fattore M., Brueggemann R., Owsinski J., “Using poset theory to compare fuzzy multidimensional material deprivation across regions”, in S. Ingrassia, R. Rocci, M. Vichi (Eds.), *New Perspectives in Statistical Modeling and Data Analysis*, Springer-Verlag, 2010 (in press).
- Maggino F., *The state of the art in indicators construction in the perspective of a comprehensive approach in measuring well-being of societies*, Firenze University Press, Archivio E-Prints, Firenze, 2009.
- Nardo M., M. Saisana, A. Saltelli and S. Tarantola (EC/JRC), A. Hoffman and E. Giovannini (OECD) *Handbook on Constructing Composite Indicators: Methodology and Userguide*, OECD, Statistics Working Paper, 2005.
- Noll, H.-H. *Measuring and Monitoring the Quality of Life*, Lecture at the Università degli Studi di Firenze, April 23-24, 2009.
- Sharpe A., J. Salzman *Methodological Choices Encountered in the Construction of Composite Indices of Economic and Social Well-Being*, Center for the Study of Living Standards, Ottawa, CAN, 2004.



**FOR ANY FURTHER INFORMATION
PLEASE CONTACT:**

Filomena Maggino

filomena.maggino@unifi.it

Marco Fattore

marco.fattore@unimib.it

Presentation designer: trapani.marco@gmail.com