

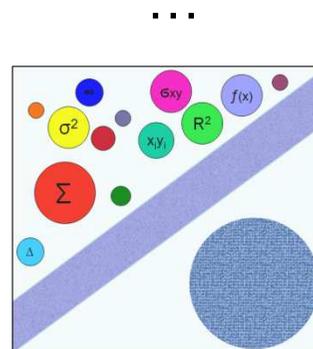
## Making subjectivity explicit

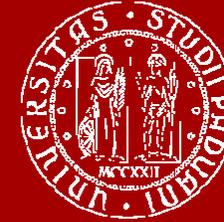
A measure of Social Inclusion for European administrative  
regions

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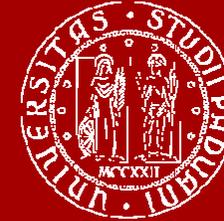
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## The paper in a nutshell

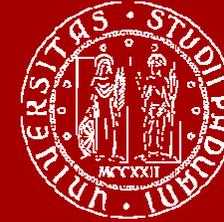
- Main topic: implicit trade-offs resulting from
  - Normalization function (re-scaling)
  - Aggregation function (preference structure)
- Case-study: measure of social inclusion for European regions
- Method: compare strategies for normalization and aggregation
  - Top-down approach: parameters are set by the researcher
  - Expert-based approach: parameters are elicited by experts/stakeholders
- Results
  - Data-driven vs experts-driven strategies lead to very different results and economic interpretations (positive vs normative)



## Case study: 4 variables of social inclusion in Europe

- Theoretical framework on social inclusion starts with Atkinson et al. (2002).
- EUROSTAT DATABASE 2014 for 58 administrative regions, 2004-2012

|                          | <i>Health</i>           | <i>Education</i>         | <i>Labour market</i>      | <i>Economic Resources</i> |
|--------------------------|-------------------------|--------------------------|---------------------------|---------------------------|
| <b>Average 2004-2012</b> | Longevity               | Early school-leaving     | Long-term unemployment    | At-risk-of-poverty rate   |
| <b>Belgium</b>           | 79.8                    | 12.10%                   | 3.80%                     | 14.80%                    |
| <b>Germany</b>           | 80.2                    | 12.30%                   | 4.10%                     | 14.50%                    |
| <b>Italy</b>             | 82                      | 18.40%                   | 3.60%                     | 16.90%                    |
| <b>Spain</b>             | 81.4                    | 29.30%                   | 4.80%                     | 20.80%                    |
| maximum                  | 84.2<br>(Madrid 2012)   | 54.2%<br>(Ceuta 2005)    | 18.2%<br>(Ceuta 2012)     | 44.3%<br>(Sicilia 2011)   |
| minimum                  | 77.5<br>(Wallonie 2004) | 5.4%<br>(Thüringen 2009) | 0.5%<br>(Trent.A.A. 2004) | 5.2%<br>(V.d'Aosta 2006)  |



## The baseline model

- Generic aggregation model of Social Inclusion  $F(\mathbf{x}) = \left[ w_1 v_1(x_1)^\beta + \dots + w_4 v_4(x_4)^\beta \right]^{1/\beta}$

- Min-max normalization function

For «positive» variables

$$v_+^{i,t}(x_+^{i,t}) = 100 \cdot \frac{x_+^{i,t} - b \min(x_+)}{b \max(x_+) - b \min(x_+)}$$

$$v_+^{i,t}(x_+^{i,t}) = 0 \text{ if } x_+^{i,t} \leq b \min(x_+)$$

$$v_+^{i,t}(x_+^{i,t}) = 100 \text{ if } x_+^{i,t} \geq b \max(x_+)$$

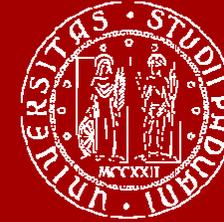
For «negative» variables

$$v_-^{i,t}(x_-^{i,t}) = 100 \cdot \frac{b \max(x_-) - x_-^{i,t}}{b \max(x_-) - b \min(x_-)}$$

$$v_-^{i,t}(x_-^{i,t}) = 0 \text{ if } x_-^{i,t} \geq b \max(x_-)$$

$$v_-^{i,t}(x_-^{i,t}) = 100 \text{ if } x_-^{i,t} \leq b \min(x_-)$$

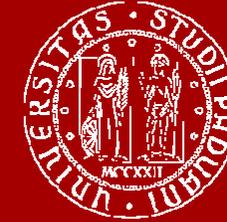
- Normalization benchmarks bmin and bmax can be:
  - Data-driven (i.e., correspond to observed performance in data)
  - Based on preferences of experts/stakeholders



## Elicitation of experts benchmarks

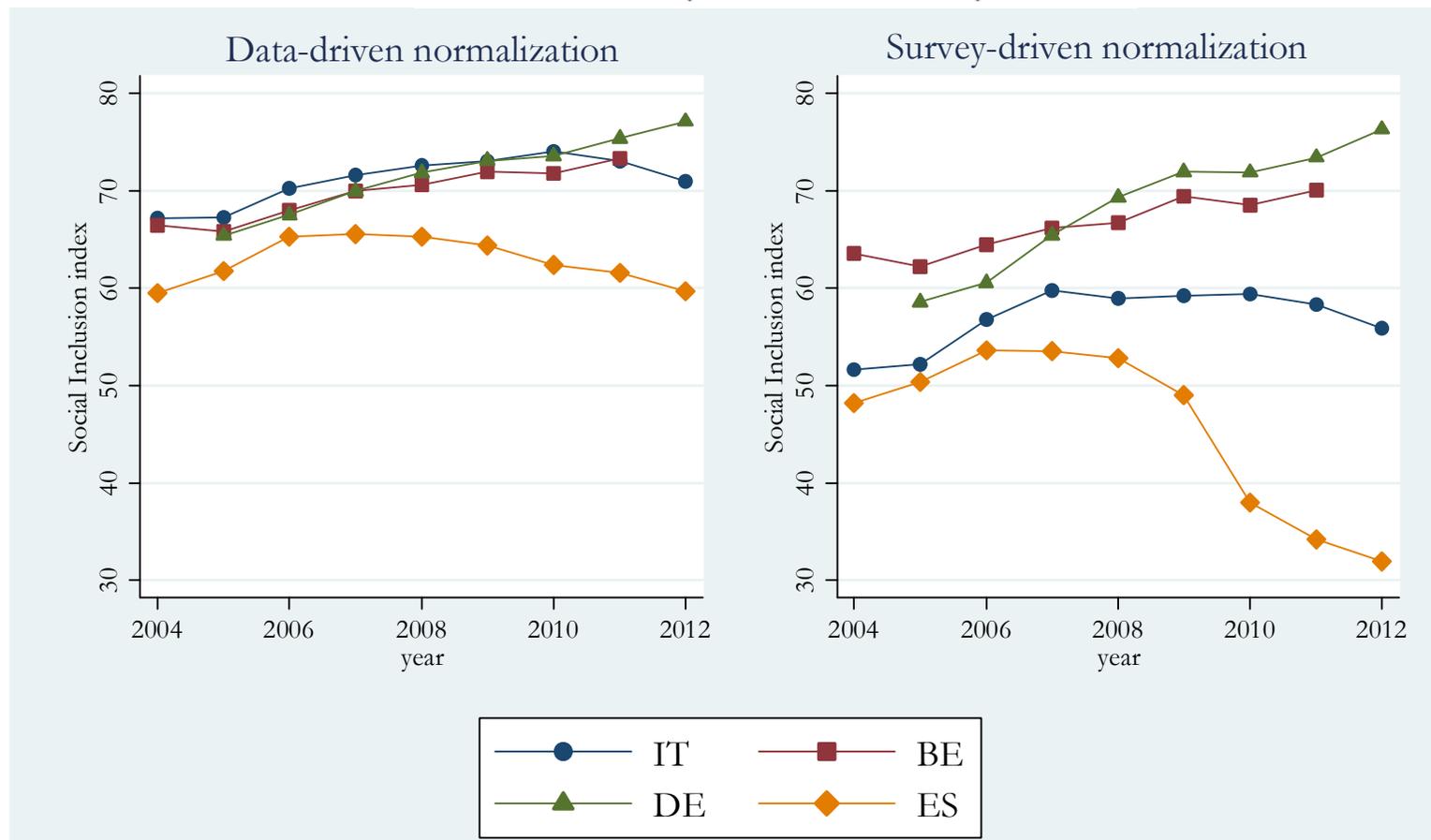
- Population: 143 professors in Economics and Management at the Ca' Foscari University of Venice (88 respondents)
- For each variable, we ask for two thresholds that would represent:
  - a certainly **undesirable** social condition
  - a certainly **desirable** social condition

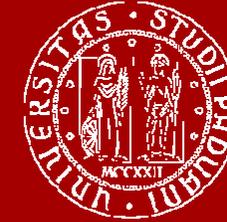
| Variable             | SURVEY benchmarks<br>median responses ( <i>interquartile range</i> ) |                       | DATA-DRIVEN benchmarks |         |
|----------------------|--|-----------------------|------------------------|---------|
|                      | Minimum  | Maximum               | Minimum                | Maximum |
| Early school leaving | 10%<br>[5% - 10%]  | 20%<br>[15% - 25%]    | 5.4%                   | 54.2%   |
| Life expectancy      | 73 years<br>[70 - 75]  | 83 years<br>[80 - 85] | 77.5                   | 84.2    |
| L.T. Unemployment    | 3%<br>[2% - 4%]  | 9%<br>[6% - 10%]      | 0.5%                   | 18.2%   |
| Poverty rate         | 5%<br>[3% - 7%]  | 20%<br>[17% - 21.5%]  | 5.2%                   | 44.3%   |



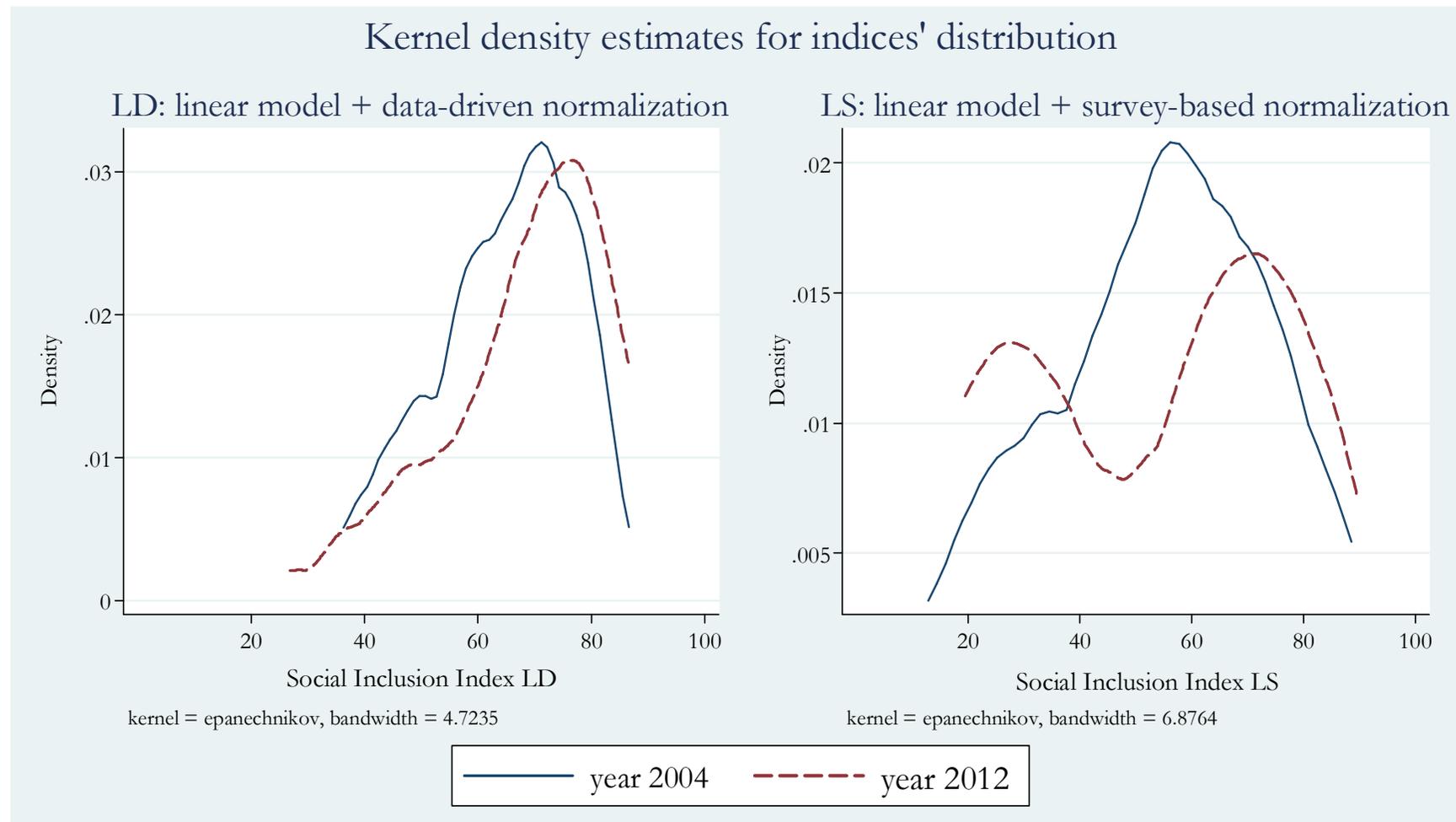
# A simple average model with alternative normalizations

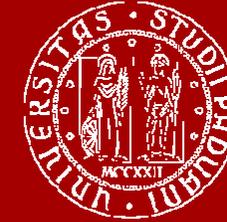
$$F^i(\nu(\mathbf{x}^i)) = \frac{1}{4}\nu_1(x_1^i) + \dots + \frac{1}{4}\nu_4(x_4^i)$$





- Normalization affects the distribution of the index





- Partial derivatives identify variables' relative relevance

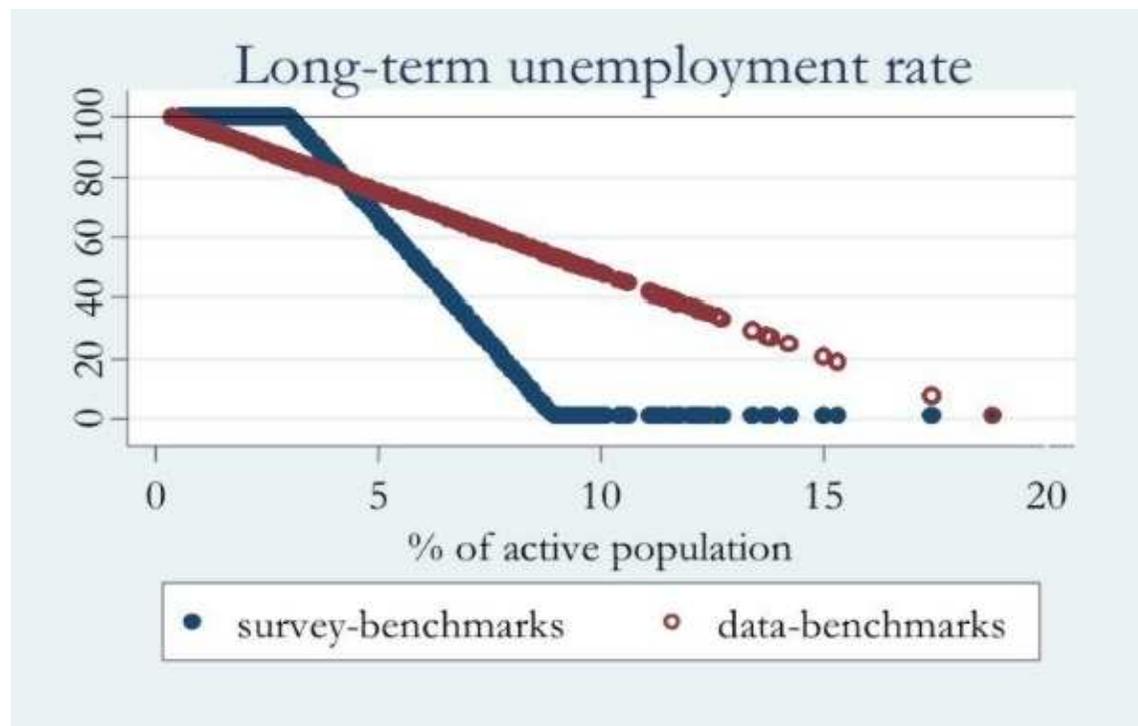
| DATA-DRIVEN<br>normalization | longevity                   | early school<br>leavers | long-term<br>unemployment | poverty-rate |
|------------------------------|-----------------------------|-------------------------|---------------------------|--------------|
|                              | $\partial F / \partial x_j$ | 3.72                    | -0,512                    | -1,412       |
| <b>Relative importance</b>   | 59.8%                       | 8.2%                    | 22.7%                     | 9.1%         |

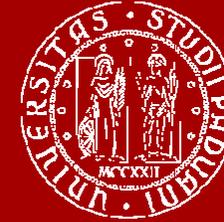
  

| SURVEY<br>normalization    | longevity                   | early school<br>leavers | long-term<br>unemployment | poverty-rate |
|----------------------------|-----------------------------|-------------------------|---------------------------|--------------|
|                            | $\partial F / \partial x_j$ | 2.5                     | -2.5                      | -4           |
| <b>Relative importance</b> | 23.5%                       | 23.5%                   | 37.6%                     | 15.4%        |

- Rank reversal between Italy and Germany because of different relative weights, especially on longevity and unemployment
- These characterizations have different economic justifications. Positive vs normative.

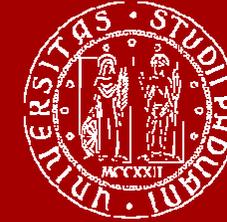
- Normative vs positive





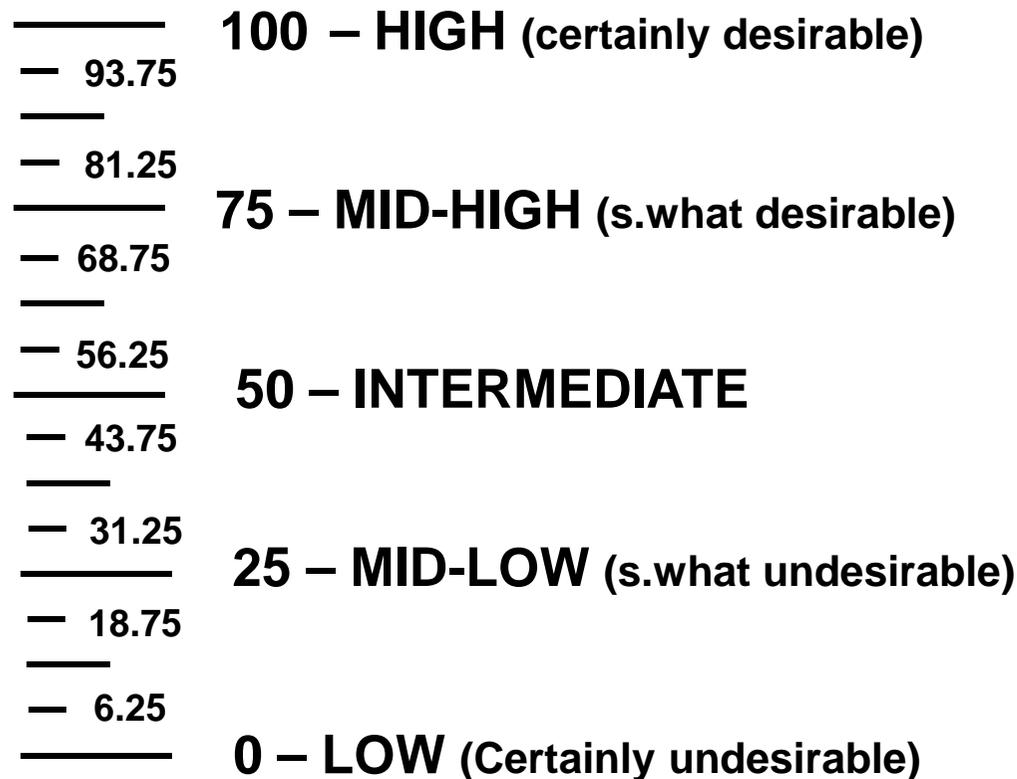
## Estimate the aggregation function through experts' panel

- The general CES model is  $F(\mathbf{x}) = \left[ w_1 v_1 (x_1)^\beta + \dots + w_4 v_4 (x_4)^\beta \right]^{1/\beta}$ 
  - Set of parameters to be estimated:  $P = (\beta, \mathbf{w})$
- Importance of eliciting expert/institutional preferences
  - Kim et al. (SIR, 2015), Decancq, Lugo (Econometric Reviews, 2013)
- Population: 20 Regional Directors General of Social Policy
- Elicitation strategy: Scenarios evaluation (individual interviews)
- Scenario: random combination of normalized values for the 4 variables
  - Assumption: each dimension can take three normalized levels: Certainly desirable (100), Intermediate condition (50), Certainly undesirable (0)
  - *Min-max* normalization function with expert-based benchmarks
  - Set of 27 scenarios, same for every Decision-Maker



## Scenarios' evaluation

- Evaluation on a 0-100 scale, using 5 trivial scenarios as guidelines.

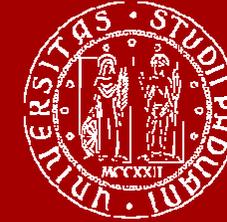


| Istruzione -- Lavoro -- Reddito = Salute ++  |  |  |   | Scenario I0 |
|--|--|--|---|-------------|
| <b>ISTRUZIONE</b><br>condizione <b>BASSA</b> | <b>LAVORO</b><br>condizione <b>BASSA</b> | <b>REDDITO</b><br>condizione <b>INTERMEDIA</b> | <b>SALUTE</b><br>condizione <b>ALTA</b> |             |
| <i>abbandono scolastico</i><br>20 %          | <i>disocc. lunga durata</i><br>9 %       | <i>povertà relativa</i><br>12,5 %              | <i>speranza di vita</i><br>82 anni      |             |

### Catalunia (ES)

| Istruzione + Lavoro + Reddito + Salute +          |   |  |   | Scenario MA |
|---|---|--|---|-------------|
| <b>ISTRUZIONE</b><br>condizione <b>MEDIO-ALTA</b> | <b>LAVORO</b><br>condizione <b>MEDIO-ALTA</b> | <b>REDDITO</b><br>condizione <b>MEDIO-ALTA</b> | <b>SALUTE</b><br>condizione <b>MEDIO-ALTA</b> |             |
| <i>abbandono scolastico</i><br>12,5 %             | <i>disocc. lunga durata</i><br>4,5 %          | <i>povertà relativa</i><br>9 %                 | <i>speranza di vita</i><br>79,5 anni          |             |

### Trivial mid-high (75) scenario

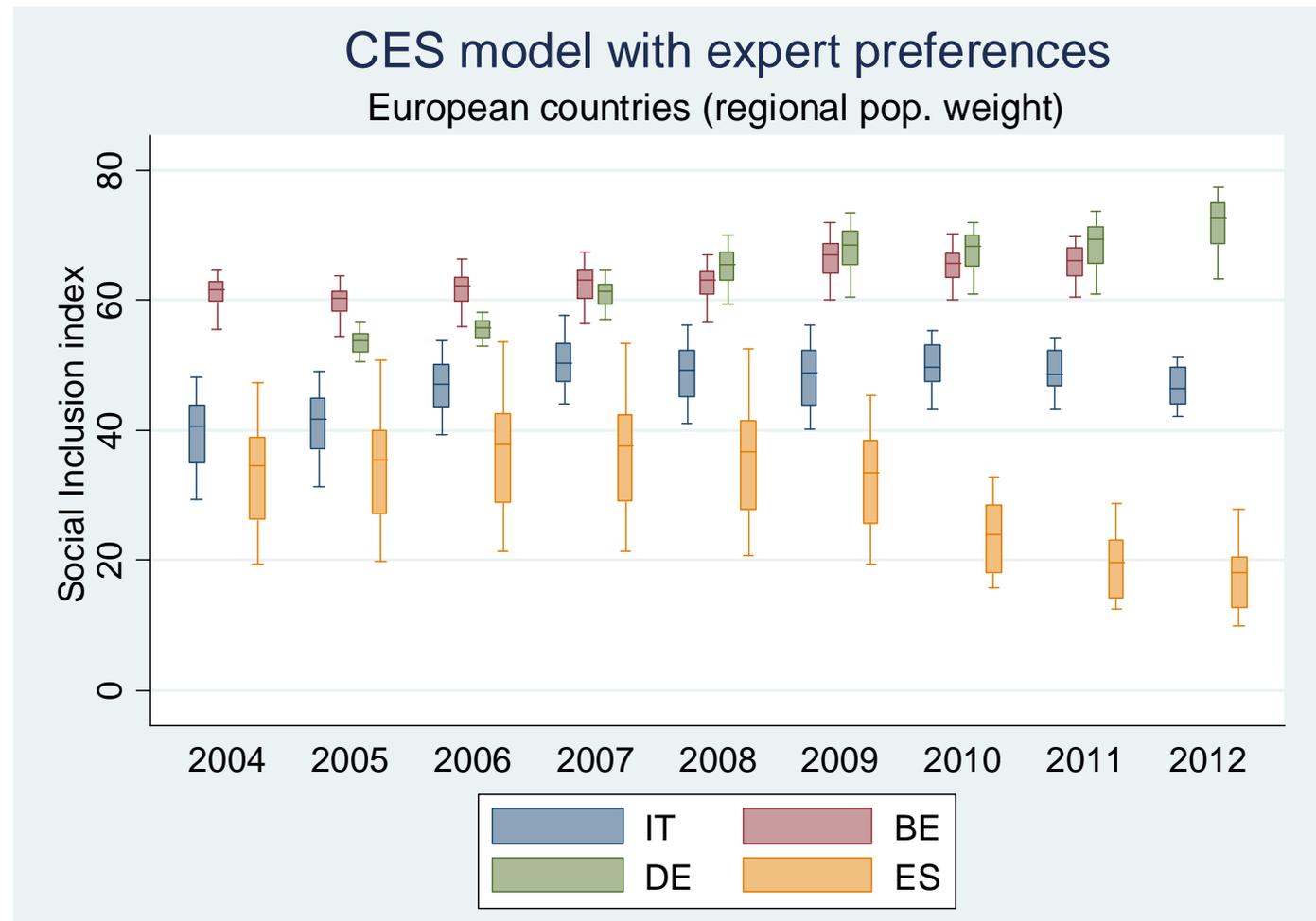


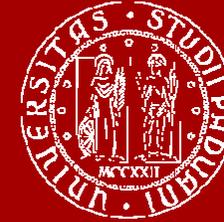
**NLS estimation results  
 per decision-maker**

|                         | $\beta$ | $w_1$<br>Education | $w_2$<br>Lab. mkt | $w_3$<br>Econ. Res. | $w_4$<br>Health | $R^2$ |
|-------------------------|---------|--------------------|-------------------|---------------------|-----------------|-------|
| 1 DM-Piemonte           | 0.95    | 0.38               | 0.20              | 0.30                | 0.13            | 0.95  |
| 2 DM-Lombardia          | 0.66    | 0.11               | 0.2               | 0.37                | 0.33            | 0.93  |
| 3 DM-Liguria            | 0.71    | 0.27               | 0.23              | 0.25                | 0.26            | 0.84  |
| 4 DM-Veneto             | 0.53    | 0.32               | 0.27              | 0.19                | 0.22            | 0.9   |
| 5 DM-TAA                | 0.34    | 0.27               | 0.27              | 0.28                | 0.18            | 0.89  |
| 6 DM-Friuli – V. Giulia | 0.77    | 0.31               | 0.27              | 0.22                | 0.20            | 0.9   |
| 7 DM-Emilia-Romagna     | 0.27    | 0.41               | 0.21              | 0.13                | 0.24            | 0.92  |
| 8 DM-Toscana            | 0.29    | 0.30               | 0.36              | 0.20                | 0.14            | 0.94  |
| 9 DM-Marche             | 0.71    | 0.52               | 0.21              | 0.15                | 0.10            | 0.88  |
| 10 DM-Abruzzo           | 1.00    | 0.22               | 0.36              | 0.28                | 0.14            | 0.82  |
| 11 DM-Campania          | 0.67    | 0.28               | 0.21              | 0.29                | 0.21            | 0.91  |
| 12 DM-Puglia            | 1.00    | 0.27               | 0.26              | 0.39                | 0.08            | 0.84  |

## Results: average national Social Inclusion

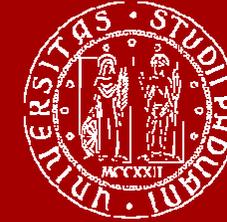
- We do not aggregate experts' preferences
- Each set of preference generate an index
- We graph the indices distribution for each region / nation



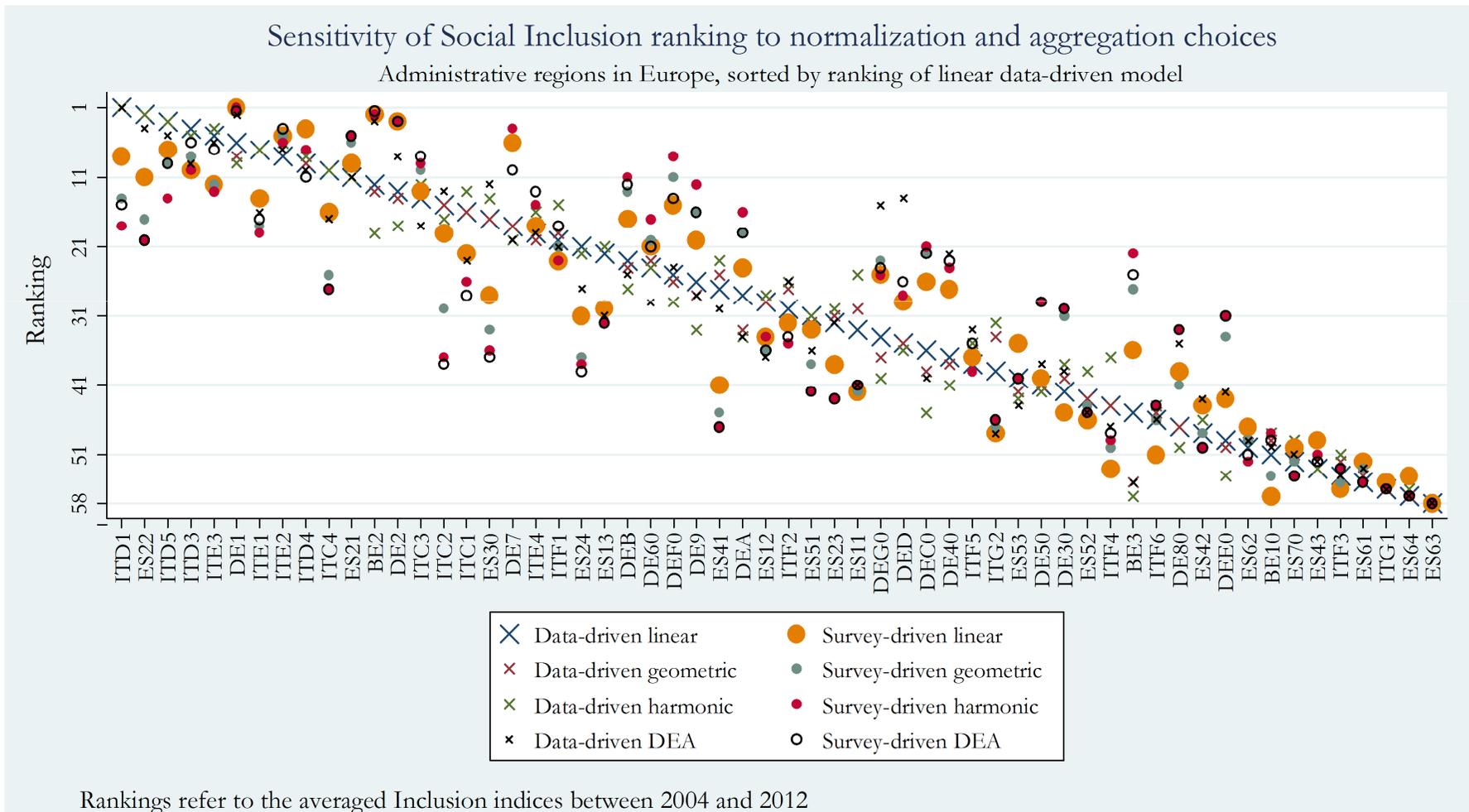


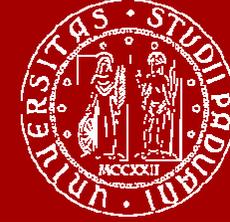
## Conclusions

- There is hardly a neutral method to build synthetic measures of abstract phenomena
- Normalization is a partial weighting stage which can strongly affect results
- Implicit trade-offs can be made transparent
- Economic premises and interpretation of results may largely differ depending on the nature of parameters' selection



# Changing aggregation functions (geometric, harmonic)





# FOLLOW-UP SLIDES

**SOGLIA DI RISCHIO:** qual è un livello di speranza di vita alla nascita che Lei valuterebbe come sicuramente negativo, non desiderabile e problematico in un generico contesto territoriale?

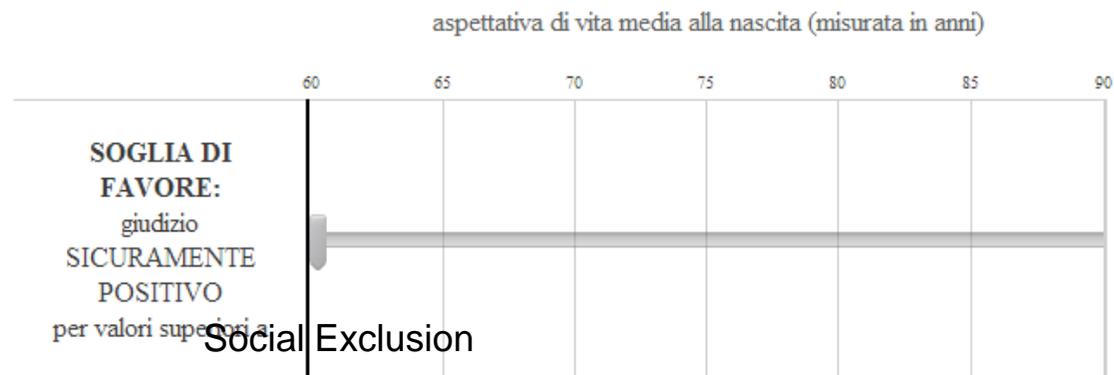
Per individuare la soglia nella tabella seguente occorrerà trascinare il cursore verticale (che assume colore blu non appena lo si sfiora) fino a raggiungere il numero di anni desiderato.



**SOGLIA DI FAVORE:** un livello che Lei valuterebbe come sicuramente positivo, una condizione virtuosa in un generico

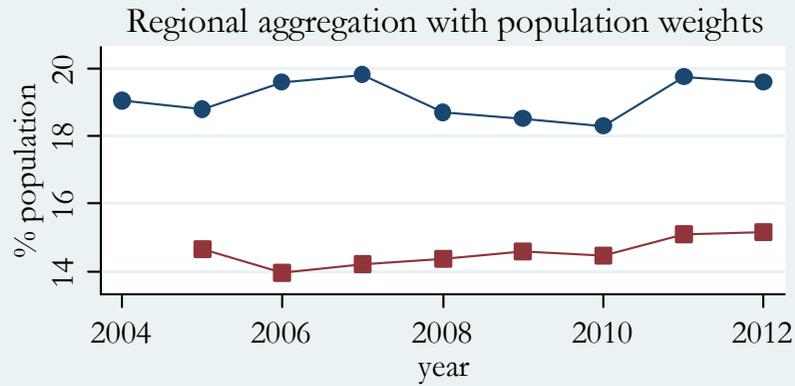
La *soglia di favore* deve essere **maggiore** (o uguale) rispetto alla *soglia di rischio* selezionata nella domanda precedente: una longevità più elevata è preferibile ad una meno elevata.

Per individuare la soglia nella tabella seguente occorrerà trascinare il cursore verticale (che assume colore blu non appena lo si sfiora) fino a raggiungere il numero di anni desiderato.

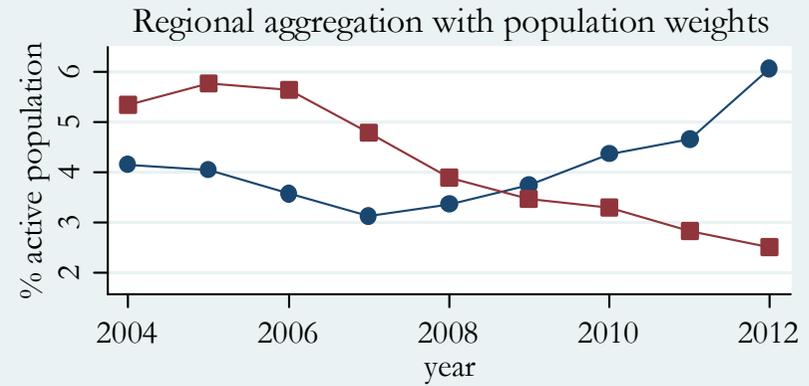


# Time trend of indicators

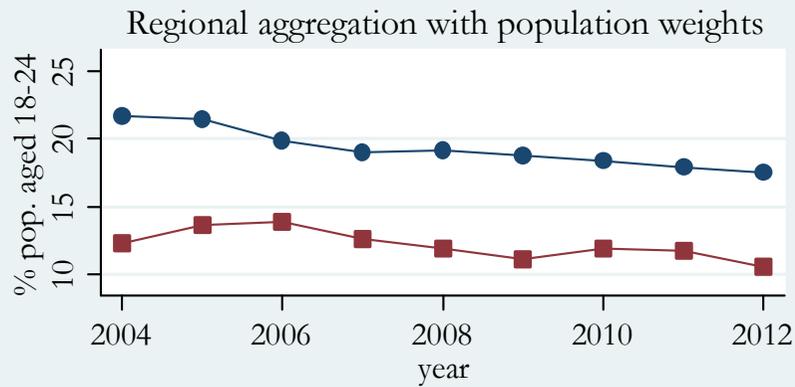
## Poverty rate



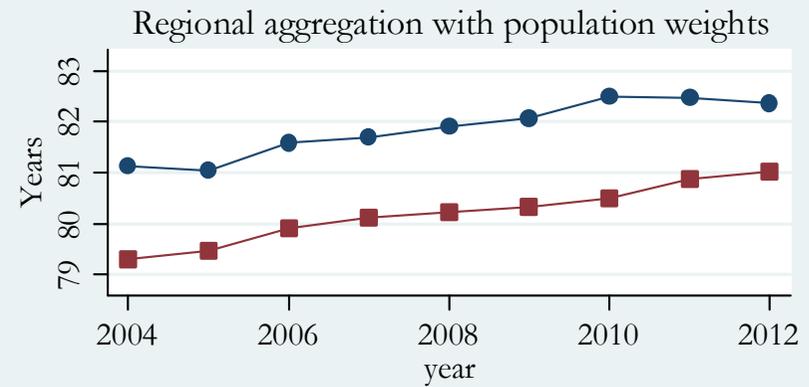
## Long-term unemployment



## Early school-leavers



## Life expectancy at birth



- Recall the CES function and fix a beta < 1

$$F^i(v(\mathbf{x})) = \left[ w_1 v_1(x_1^i)^b + L + w_m v_m(x_m^i)^b \right]^{1/b}$$

- Suppose  $\beta=0$ , we get a geometric mean

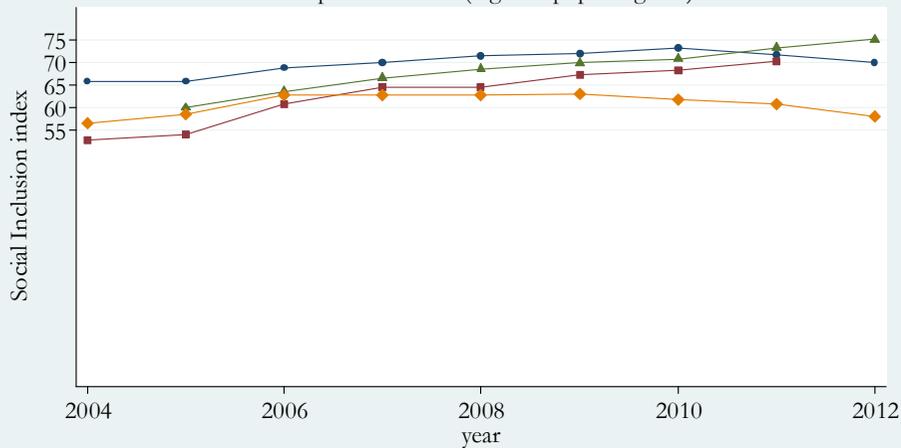
$$G^i(n(\mathbf{x}^i)) = n_1(x_1^i)^w * L * n_m(x_m^i)^w$$

- Higher penalization for bad performances, Extreme case: if one normalized-attribute is zero, the Index collapses to zero. (Ravallion 2012 for a discussion)
- Now an attribute's relevance depends also on its relative performance with respect to the others. Worst performances have higher relevance.

$$\frac{F(v(\mathbf{x}))}{v_j(x_j)} = w_j v_j(x_j) \frac{w_1 v_1(x_1)^b + L + w_m v_m(x_m)^b}{v_j(x_j)}^{1/b} = w_j v_j(x_j) \frac{F(v(\mathbf{x}))}{v_j(x_j)}^{1-b}$$

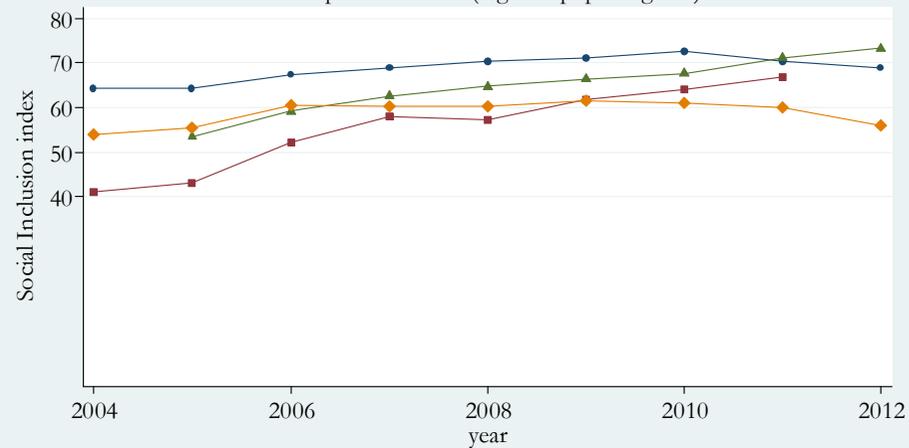
Geometric model with data-driven normalization

European countries (regional pop. weighted)



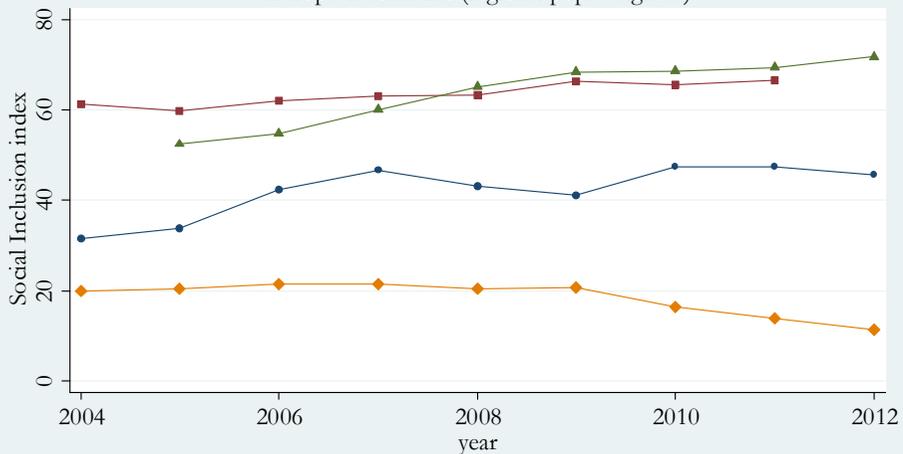
Armonic model with data-driven normalization

European countries (regional pop. weighted)



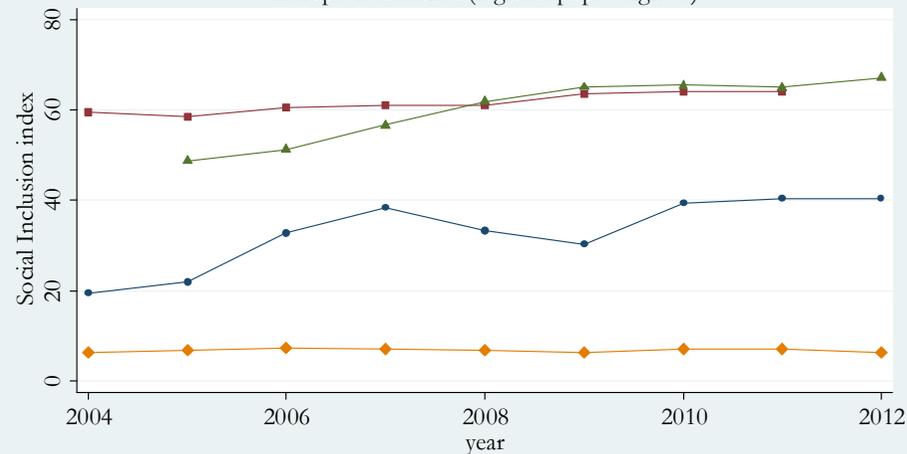
Geometric model with survey-driven normalization

European countries (regional pop. weighted)



Armonic model with survey-driven normalization

European countries (regional pop. weighted)



# Number of scenarios

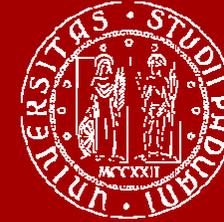
- $3^4 = 81$  possible scenarios
- Reduce them to 27 with fractional factorial (orthogonal arrays)
  - Chen J, Sun DX, Wu CFJ, 1993, International Statistical Review
  - Ferrini, Scarpa, 2007, Journal of Environmental Economics and Management
  - Wu, Amada, 2009, Wiley
  - Street et al, 2005, International Journal of Research in Marketing
- Orthogonal array of strength  $t$  (*integer number*):
  - a "table" whose entries come from a fixed finite set of symbols (ex: 0,1)
  - **for every selection of  $t$  columns, all ordered  $t$ -tuples** of the symbols, formed by taking the entries in each row restricted to these columns, **appear the same number of times.**
- We create an orthogonal array of strength 2, with  $3^3 = 27$  lines

# Interview' stages

- **Preliminary:** introduction to the phenomena of social exclusion, description of the 4 dimensions and indicators, details on the normalization, familiarisation with scenario-cards, clearing the desk:
- **First:** manually allocate scenarios on a 5-levels scale (corresponding to 5 areas on the desk), with option to use  $\pm 6.25$  steps
- **Second:** test for validity of the weak axiom of revealed preferences
  - take a random subset of 10 scenarios and repeat the experiment to check for coherence and independence of irrelevant alternatives
- **Third:** explicit question on relative weights  $w_1, w_2, w_3, w_4$ .
  - Budget allocation: «how would you allocate 100 points...?»
- **Last:** self-assessment of answers' reliability and other questions
  - 5 points scale, from “no confidence” to “very high confidence”;
  - Respondent's education level, years worked in public sector, past experiences,...

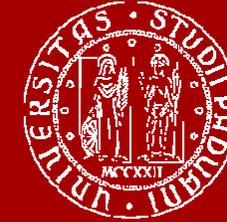
# Example of responses from an expert

| # scenario | RESPONSE | Education<br>$v(x_A)$ | Labour Market<br>$v(x_B)$ | Econ.<br>resources<br>$v(x_C)$ | Health<br>$v(x_D)$ |
|------------|----------|-----------------------|---------------------------|--------------------------------|--------------------|
| 1          | 100      | 100                   | 100                       | 100                            | 100                |
| 3          | 81.25    | 50                    | 100                       | 50                             | 100                |
| 11         | 81.25    | 100                   | 50                        | 50                             | 100                |
| 15         | 68.75    | 50                    | 1                         | 100                            | 100                |
| 21         | 68.75    | 50                    | 50                        | 100                            | 50                 |
| 23         | 68.75    | 100                   | 100                       | 50                             | 50                 |
| 9          | 56.25    | 100                   | 50                        | 1                              | 50                 |
| 17         | 56.25    | 100                   | 50                        | 100                            | 1                  |
| 13         | 56.25    | 1                     | 50                        | 100                            | 100                |
| 14         | 56.25    | 100                   | 1                         | 100                            | 50                 |
| 20         | 43.75    | 1                     | 100                       | 100                            | 50                 |
| 12         | 31.25    | 1                     | 50                        | 50                             | 50                 |
| 22         | 25       | 100                   | 1                         | 50                             | 1                  |
| 10         | 25       | 1                     | 100                       | 50                             | 1                  |
| 16         | 25       | 100                   | 1                         | 1                              | 100                |
| 24         | 25       | 1                     | 100                       | 1                              | 100                |
| 27         | 1        | 1                     | 50                        | 1                              | 1                  |
| 25         | 1        | 50                    | 1                         | 1                              | 1                  |
| 19         | 1        | 1                     | 1                         | 1                              | 50                 |



## Estimation of $\beta$ and $w_i$

- Henningsen & Henningsen, 2012, Economic Letters
- 1. Grid search on  $\beta$ 
  - a. Fix a range of values for  $\beta$ 
    - example: from 1 to -3 with steps of 0,01
  - b. For each  $\beta$  perform a constrained OLS
    - Weights should be non-negative
    - Sum of weights should be 1
    - Get optimal coefficients  $w$  conditioned on  $\beta$
  - c. Choose the  $\beta$  (and related  $w|\beta$ ) that minimize sum of squared residuals
- 2. Perform a NLS estimation with starting values coming from grid search.
- Obtain a set of parameters ( $\beta$ ,  $w$ ) for each expert



## Results: Social Inclusion for specific regions

