

# High Dimensional-Non Additive Municipalities' Performance Index in a Two Levels Approach

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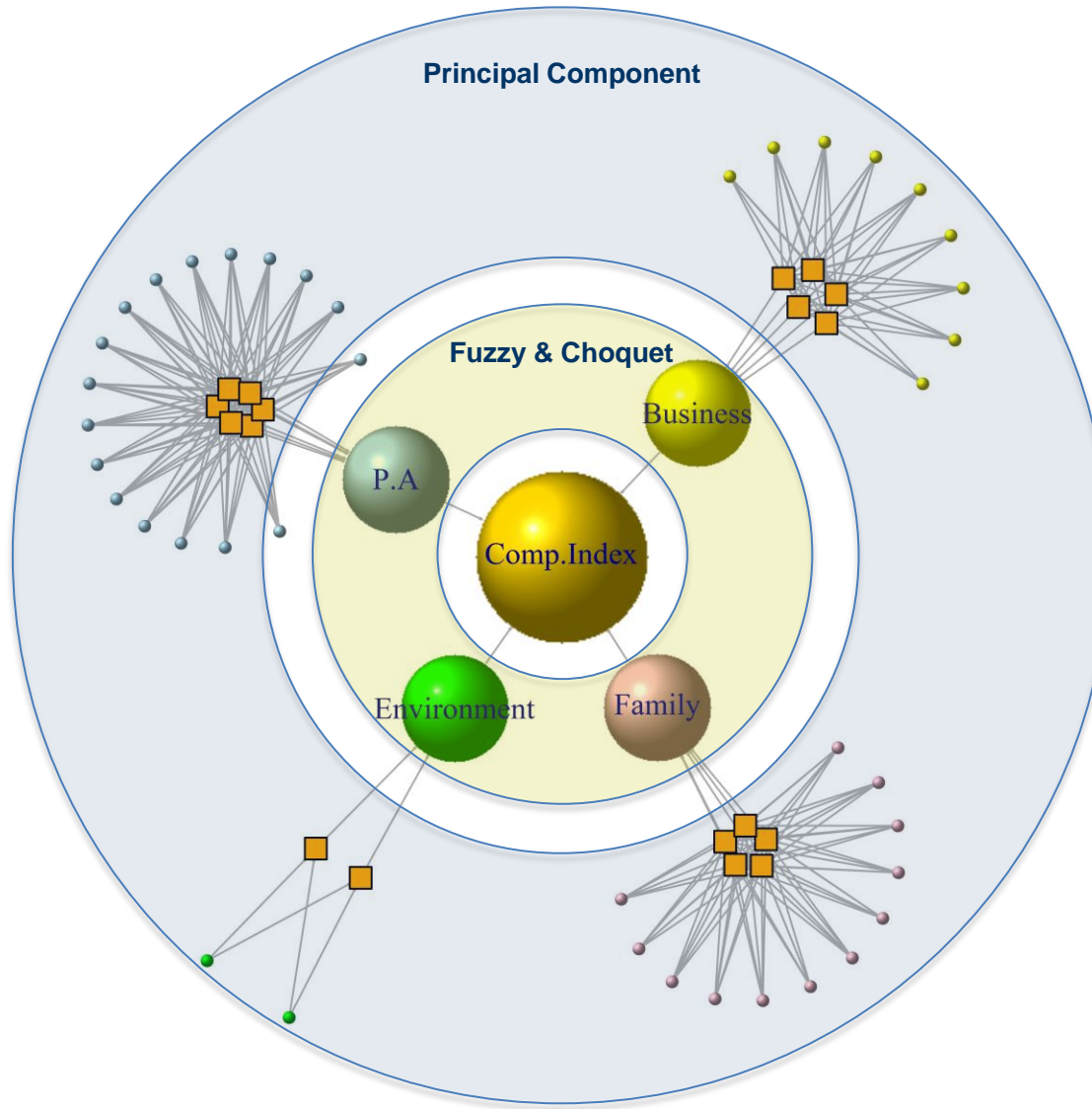
The Role of National Statistical Offices in the production of  
Leading, Composite and Sentiment Indicators



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- High Dimension
  - ❖ Weighting technique
    - Principal Component
  - ❖ Non Additive Aggregation Function
    - Fuzzy Measures and Choquet Integral
    - Preferences' Elicitation

# Composite Index Structure



## Composite Index structure

- 4 Dimensions
- 17 Indicators for *P.A.*
- 2 Indicators for *Environment*
- 12 Indicators for *Family*
- 9 Indicators for *Business*
- 131 Municipalities evaluated

## Computation techniques

- Two levels approach; first within, then between dimensions.
- Additive aggregation within Dimensions (PC technique)
- Non additive aggregation between dimensions (Fuzzy measures and Choquet Integral)

## Dimensions:

- Family: indicators related to demographic dimensions, households' income, internal and external migration flows;
- Business: indicators related to total number of business, business types, survival rate;
- P. A.: indicators related to different municipalities balance sheet items;
- Environment: two indicators related to recycled urban waste and total urban waste

## Data have been:

- Transformed in relative terms to guarantee comparability among municipalities (by number of citizens, firms, taxpayers, etc.)
- Transformed to guarantee the same polarity of indicators (the higher the measure of an indicator the better it is)
- Standardized to ensure zero mean and unit variance (PCA)

# Weighting and Aggregation Principal component Technique

# Indicators' Weight inside a Component

- The weight of each indicator in the  $k$ -th component is proportional to its absolute influence on that component (**not as a function of the explained variance**):

$$z_k = a_{k1}x_1 + \cdots + a_{kn}x_n$$

$$w_{x_j; z_k} = \frac{|a_{kj}|}{\sum_j^n |a_{kj}|} \quad \text{not } w_{x_j; z_k} = a_{kj}^2$$

# Components' Weight Inside the Dimension

- The weight of each component in the dimension is the same; it is not hence a function of the variance explained.

Two reasons:

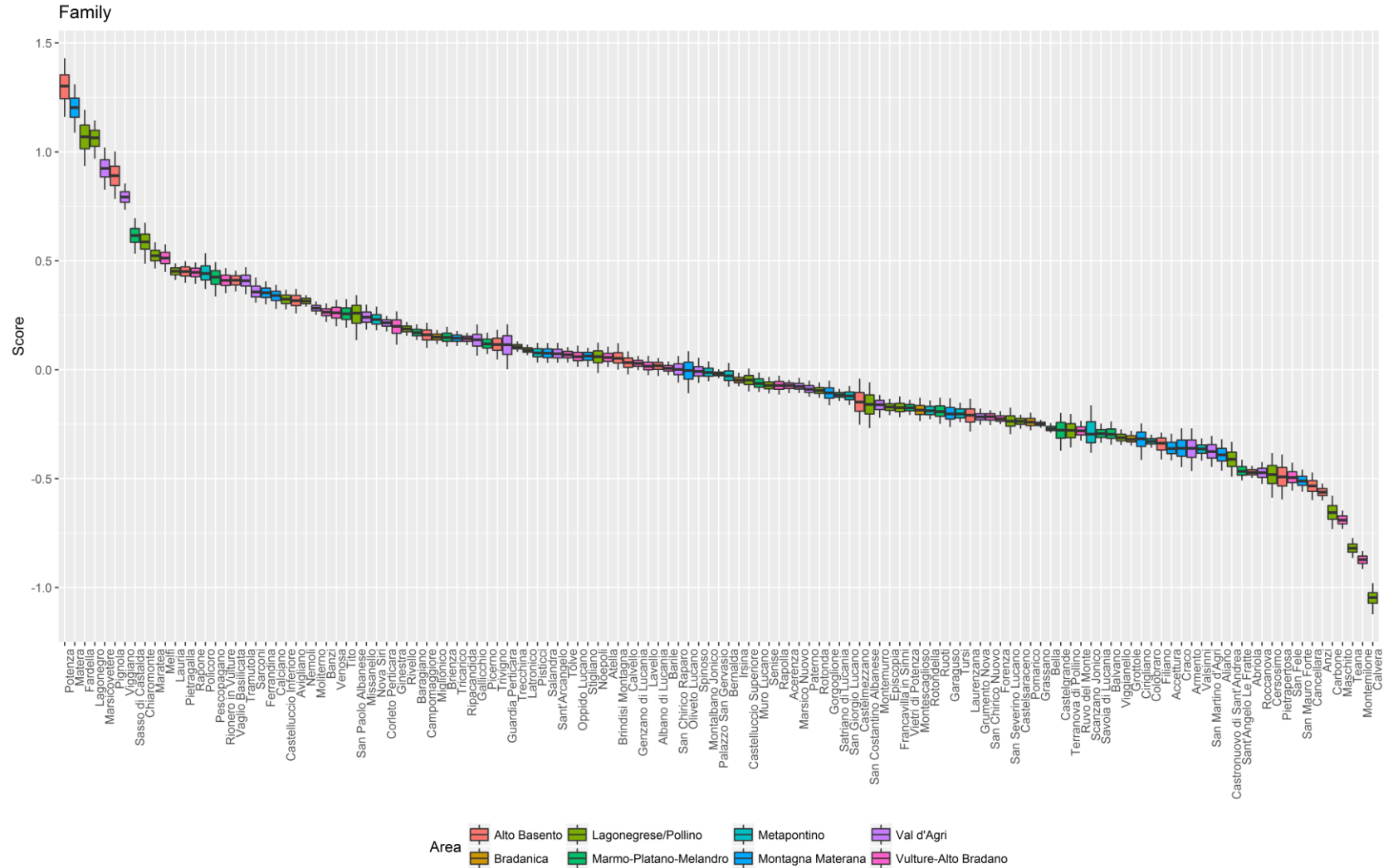
- In the context of C.I. a component cannot be more important than others;
- Mitigate the risk of overweighting correlated variables.

Example:

$$\mathbf{R}_x = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 1 \end{bmatrix} \rightarrow \mathbf{A} = \begin{bmatrix} 0 & 1 \\ -0.5 & 0 \\ -0.5 & 0 \\ -0.5 & 0 \\ -0.5 & 0 \end{bmatrix}; \lambda' = [4 \quad 1]$$

**According to variance explained, the first component weights 80% and the second 20%**

# Family Dimension





## Weighting and Aggregation Fuzzy Measures and Choquet Integral

Many composite indices existing in the literature assume implicitly the **preferential independence** among criteria assumption [D. Scott, P. Suppes, 1958; Fishburn, 1970; Marichal, Roubens 1998]

↔ *Arithmetic Mean, Weighted Mean.*

Depending on the objective of the composite indicator, this assumption could be limited or unrealistic because indicators may have some form of **synergies** and **redundancies** that should be taken into account in the aggregation phase

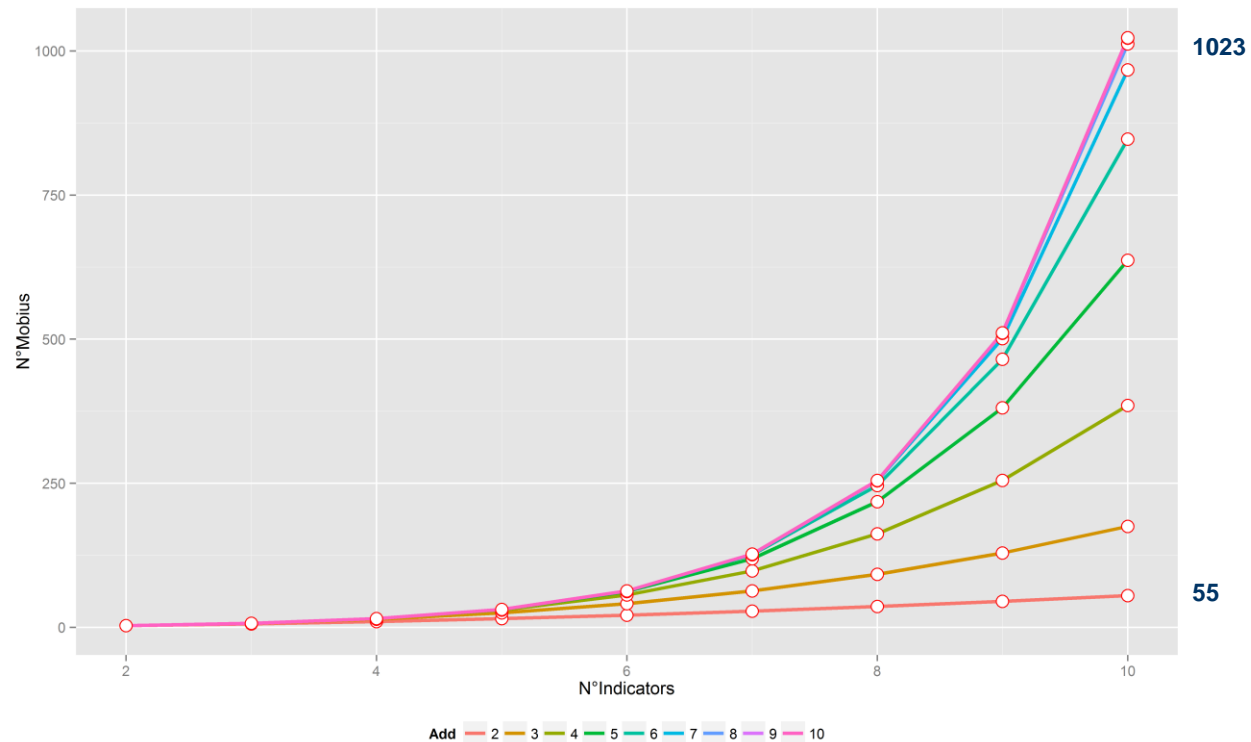
↔ *Geometric Mean, Constant Elasticity of Substitution*

However these aggregator functions are **not flexible enough**.

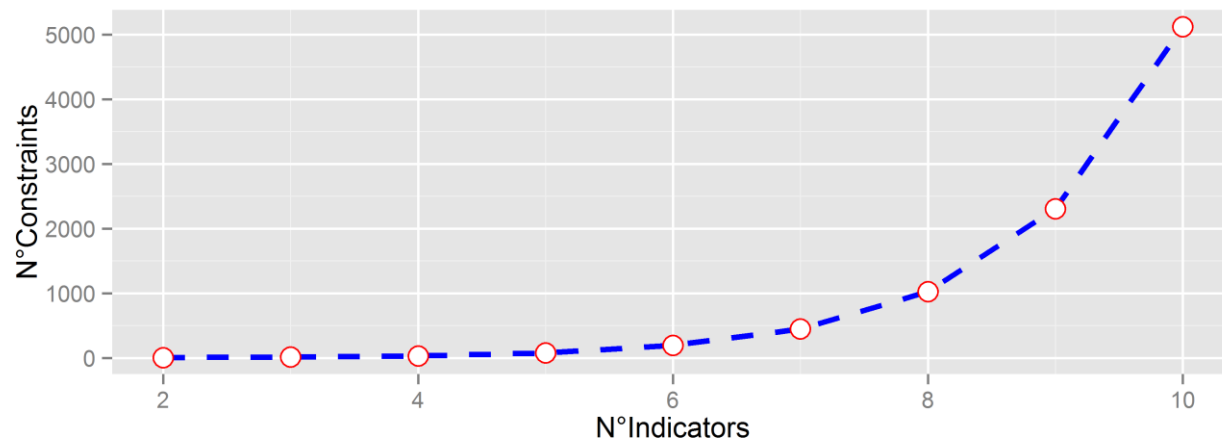
- Allow to model flexible interactions (from redundancy/substitutability to synergy/complementarity) and to consider hence complex dependencies among criteria, hence relaxing the *preferential independence assumption*;
- They can **exactly** replicate:
  - Weighted mean;
  - Ordered Weighted Averaging Operators (OWA);
  - Minimum and Maximum Operator.
- They are a powerful tool to elicit **Experts' preferences**.

# Exponential Complexity

$$N^{\circ} \text{Möbius} = \sum_{j=1}^k \binom{n}{j}$$



$$N^{\circ} \text{Cons.} = n \cdot 2^{(n-1)}$$



# Preferences' Elicitation

# LS Experts' Preference Elicitation

<i>Alternative</i>	<b>Criteria</b>				<b>Expert</b>
	<i>Family</i>	<i>Business</i>	<i>P.A.</i>	<i>Environment</i>	<i>Overall Evaluation</i>
1	Excellent	Very bad	Very bad	Very bad	-
2	Very bad	Excellent	Very bad	Very bad	-
3	Very bad	Very bad	Excellent	Very bad	-
4	Very bad	Very bad	Very bad	Excellent	-
5	Excellent	Excellent	Very bad	Very bad	-
6	Excellent	Very bad	Excellent	Very bad	-
7	Excellent	Very bad	Very bad	Excellent	-
8	Very bad	Excellent	Excellent	Very bad	-
9	Very bad	Excellent	Very bad	Excellent	-
10	Very bad	Very bad	Excellent	Excellent	-

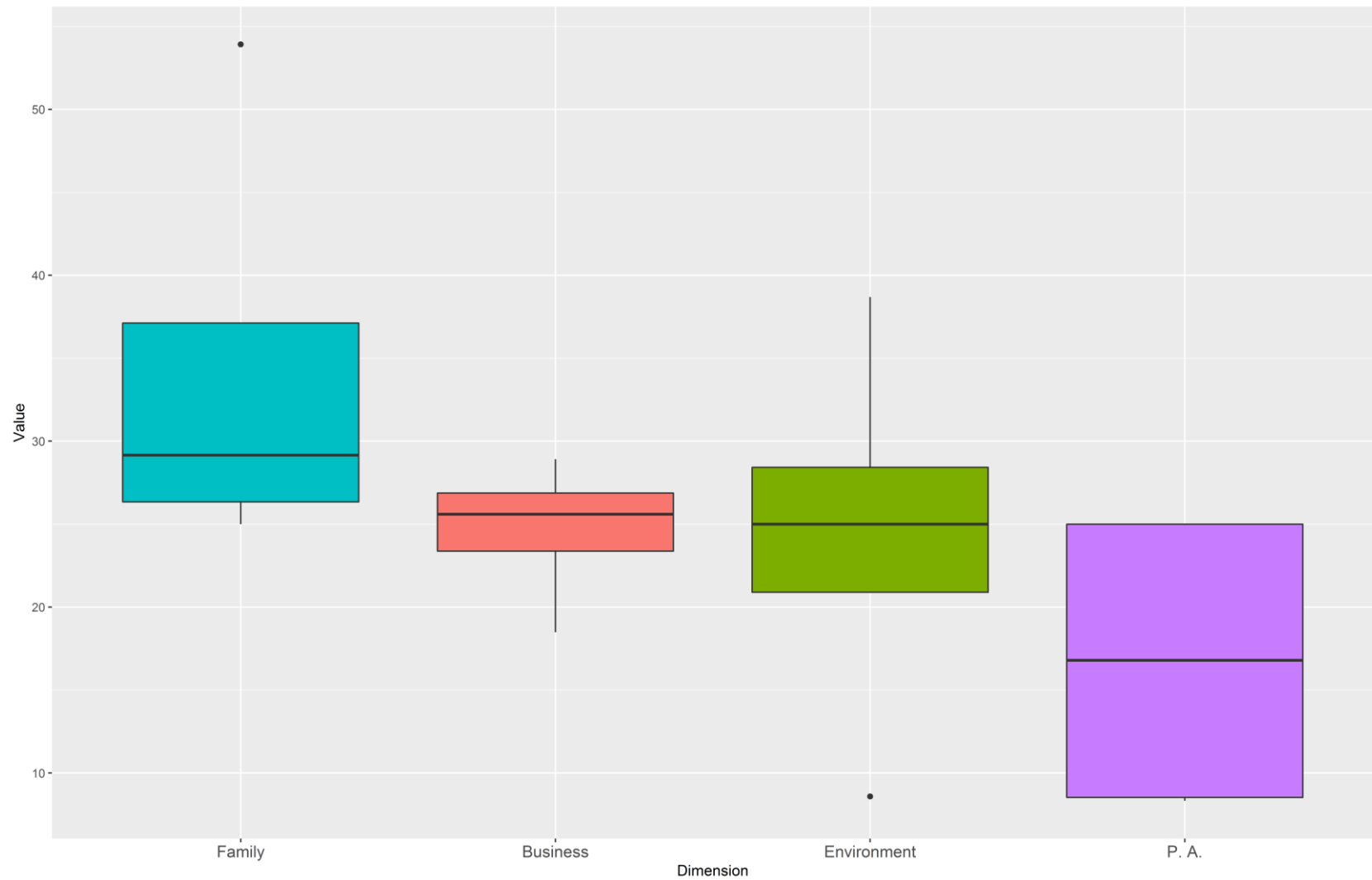
*Expert V. = {Very Sat., Sat., Nor Sat or Disat., Disat., Very Dis.}*

# Test for answer's consistency

<i>Alternative</i>	<b>Criteria</b>				<b>Expert Overall Evaluation</b>
	<i>Family</i>	<i>Business</i>	<i>P.A.</i>	<i>Environment</i>	
1	<b>Excellent</b>	Very bad	Very bad	Very bad	<b>Very Satisf.</b>
2	Very bad	Excellent	Very bad	Very bad	-
3	Very bad	Very bad	Excellent	Very bad	-
4	Very bad	Very bad	Very bad	Excellent	-
5	<b>Excellent</b>	<b>Excellent</b>	Very bad	Very bad	<b>Satisf.</b>
6	Excellent	Very bad	Excellent	Very bad	-
7	Excellent	Very bad	Very bad	Excellent	-
8	Very bad	Excellent	Excellent	Very bad	-
9	Very bad	Excellent	Very bad	Excellent	-
10	Very bad	Very bad	Excellent	Excellent	-

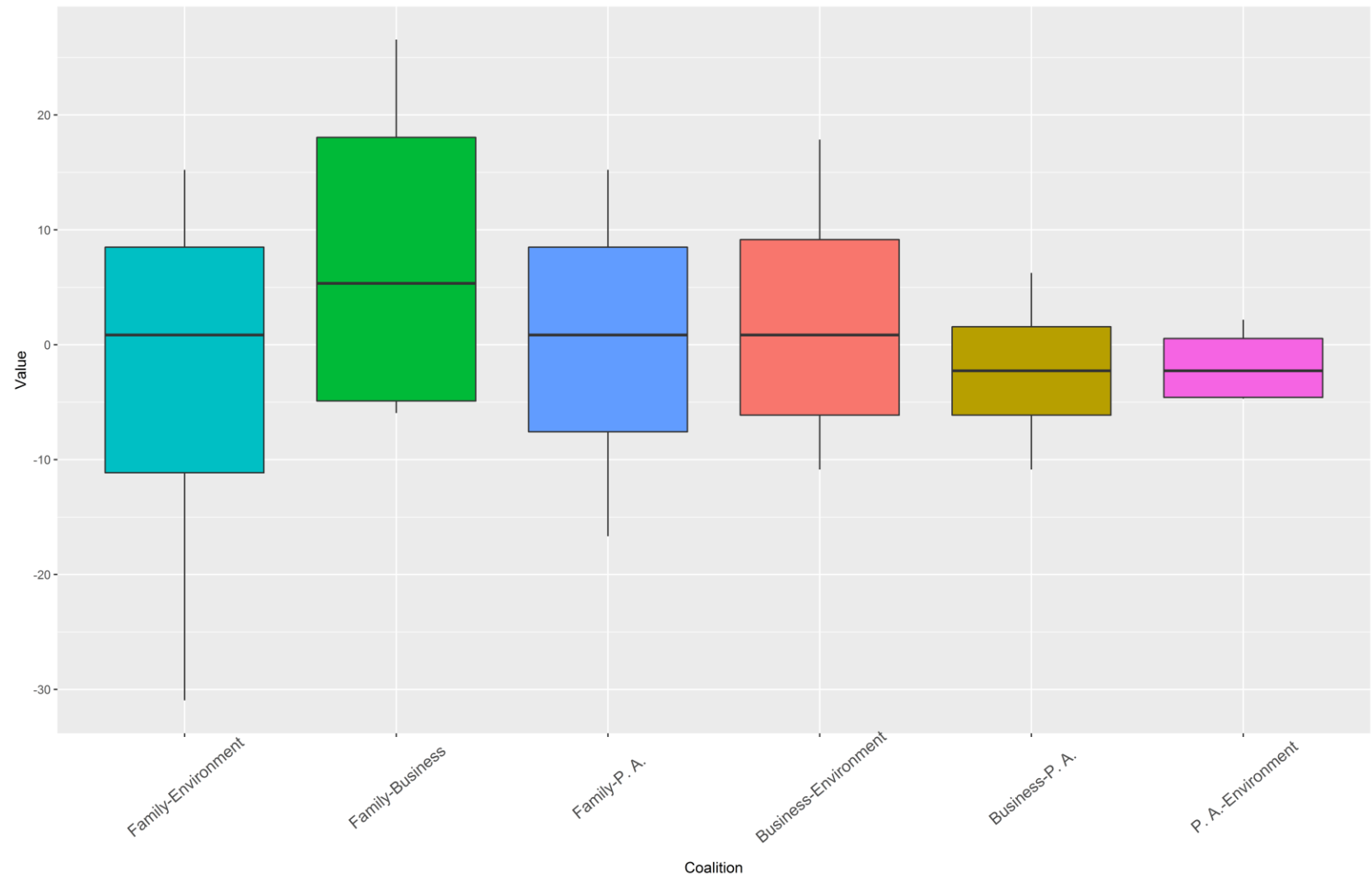


# Shapley Value





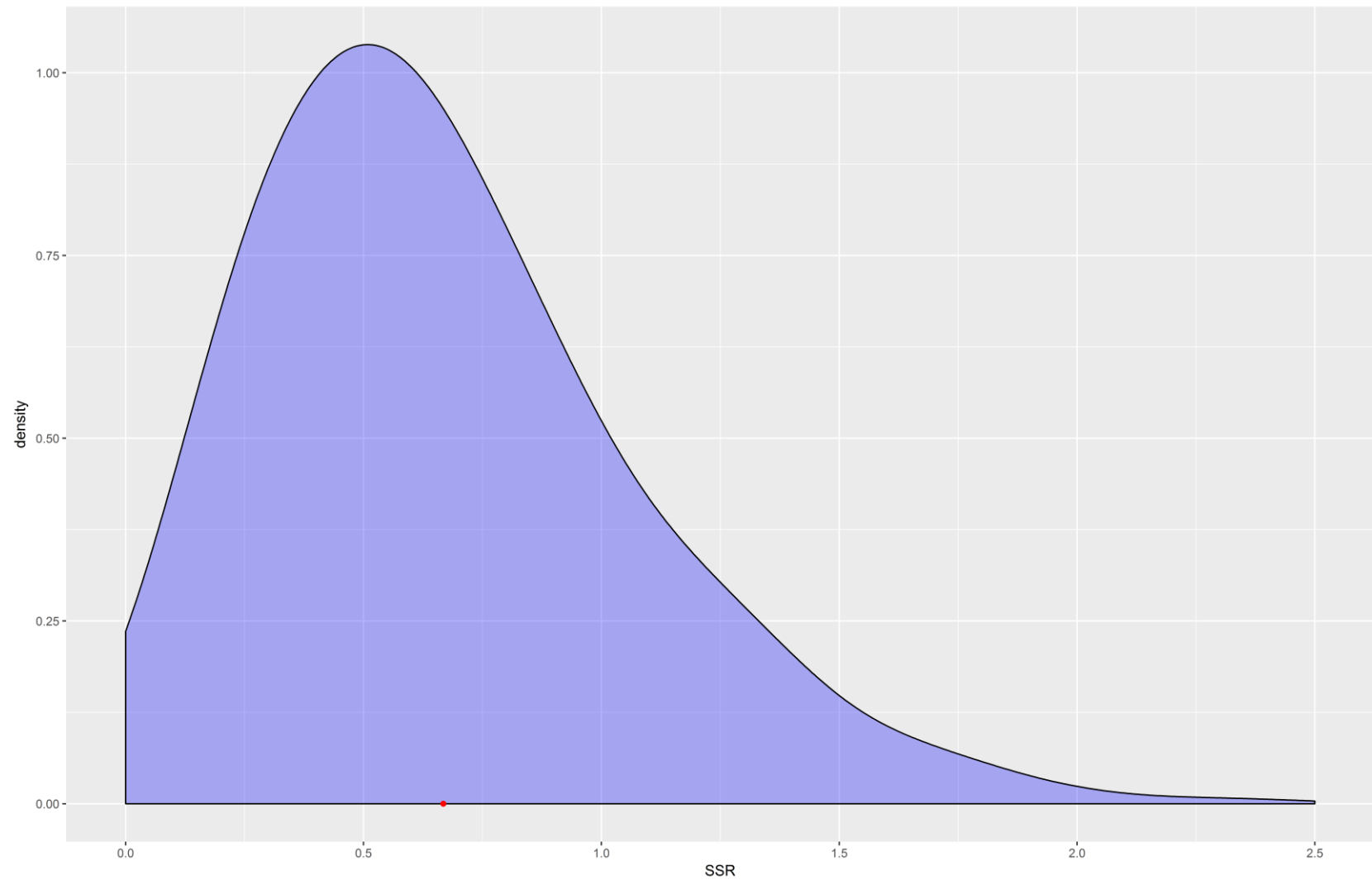
# Interaction Index



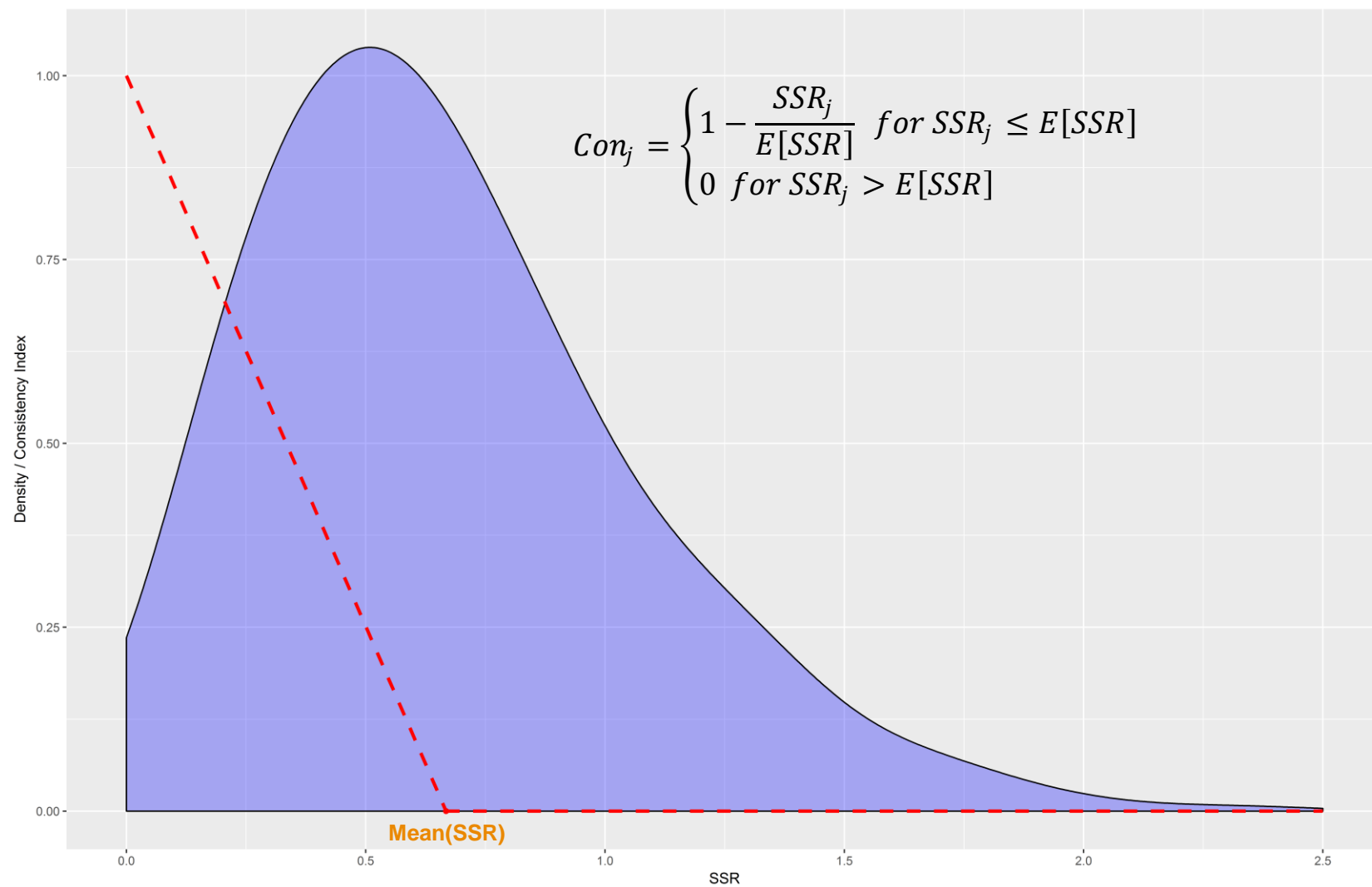
# Aggregation of Individual Opinions

## Consistency Index

# SSR Density Distribution



# Consistency Index



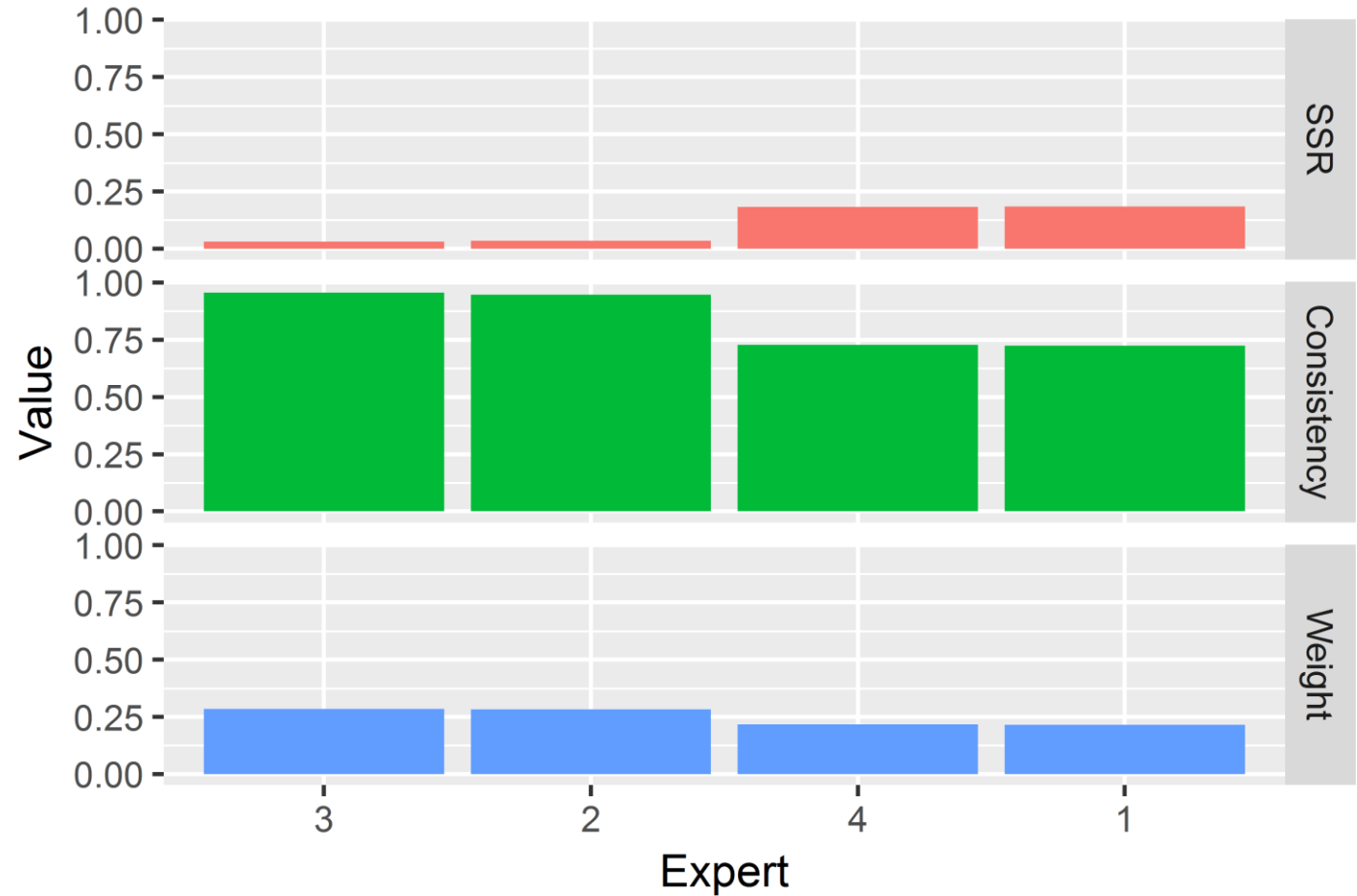
# Aggregation of Individual Opinions

$$\mathbf{y}_j = \mathbf{C}\mathbf{h}_j + \boldsymbol{\varepsilon}_j$$

$$SSR_j = \boldsymbol{\varepsilon}_j' \boldsymbol{\varepsilon}_j$$

$$Con_j = f(SSR_j | E[SSR])$$

$$w_j = \frac{Con_j}{\sum_{j=1}^d Con_j}$$

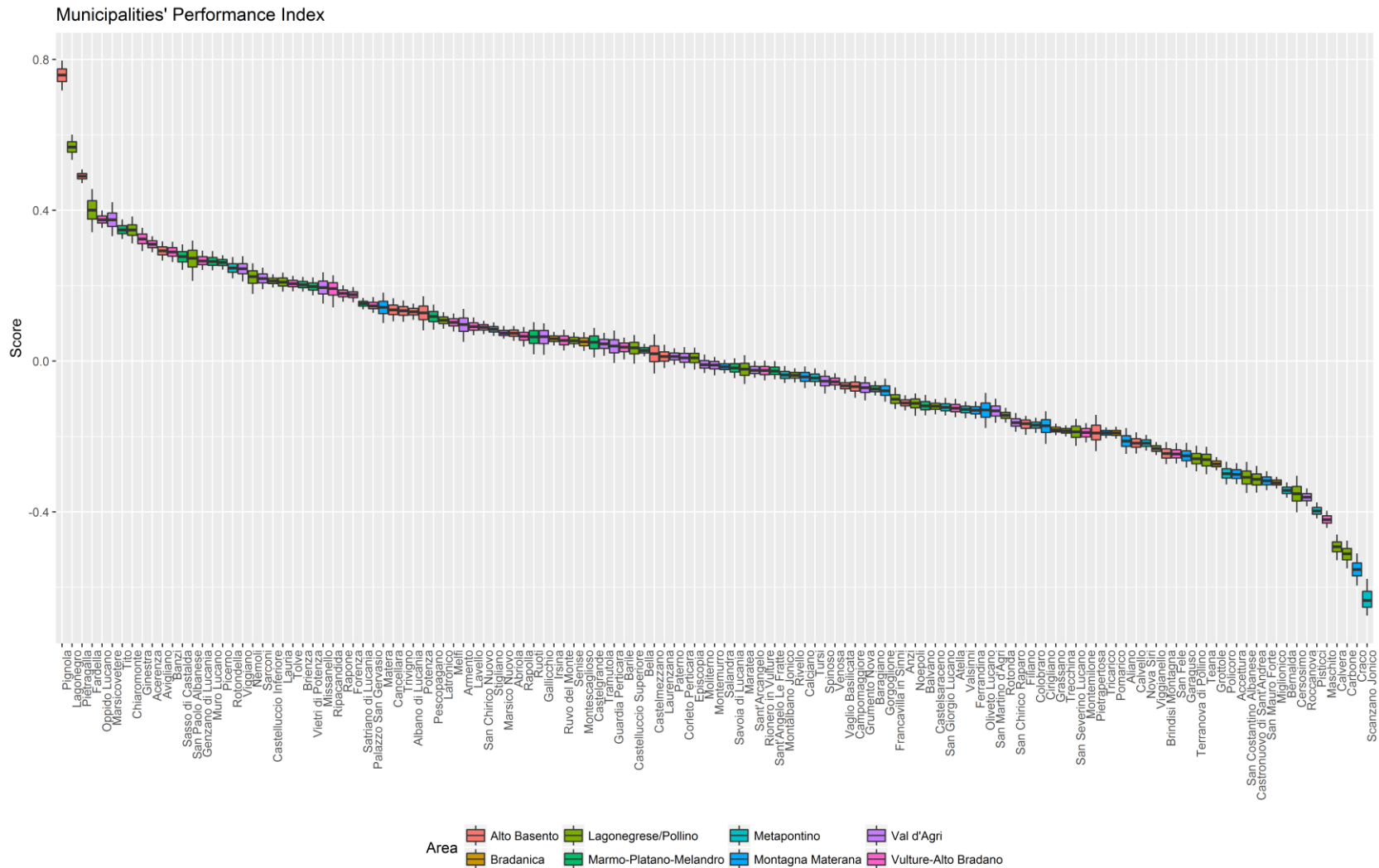


$$\Rightarrow m^*\{T\} = \sum_{j=1}^d w_j m_j\{T\}$$

<b>Dimension</b>	<b>Shapley %</b>	<b>Coalition</b>	<b>Interaction %</b>
<i>Family</i>	35.40	<i>Family - Business</i>	9.56
<i>Business</i>	24.50	<i>Family – P.A.</i>	1.51
<i>Environment</i>	23.32	<i>Family – Environm.</i>	-1.56
<i>P.A.</i>	16.76	<i>Business – P.A.</i>	-2.31
		<i>Business – Environm.</i>	1.53
		<i>P.A. - Environment</i>	-1.69

# Non Additive Aggregation of Dimensions

# Municipalities' Performance Index





# Conclusions

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- In high dimension problems, the use of P. C. could be a good solution to take into account the correlation among variables and hence to reduce the risk of overweighting their influence on the C. I.
- More consensus among researchers is needed to derive the weights of individual indicators by means of P.C.
- Fuzzy Measures and Choquet integral are powerful tool to model interactions among indicators
- It is important to take into account the consistency of the respondents in evaluating a questionnaire

**Thank you**

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## Extra Slides

A Fuzzy measures or **Non Additive Measure**, defined over the set of criteria  $N = \{1, 2, \dots, n\}$ , is a set function  $\mu: 2^N \rightarrow [0, 1]$  satisfying the following boundary and **monotonicity conditions**:

$$\begin{cases} \mu(\emptyset) = 0 \\ \mu(N) = 1 \\ \mu(S) \leq \mu(T) \leq 1, \quad \forall S \subseteq T \subseteq N \end{cases}$$

**Choquet Integral:**

$$Ch(x_1, \dots, x_n) = \sum_{i=1}^n (x_{(i)} - x_{(i-1)}) \cdot \mu(A_{(i)})$$

$$x_{(0)} \equiv 0; \quad x_{(1)} \leq \dots \leq x_{(n)}; \quad A_{(i)} = \{x_{(i)}, \dots, x_{(n)}\}$$

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**Example:** Two criteria  $N = \{1, 2\}$

$$\mu\{\emptyset\} = 0; \mu\{1, 2\} \geq \mu\{1\}; \mu\{1, 2\} \geq \mu\{2\}; \mu\{1, 2\} \equiv 1$$

- If  $\mu\{1\} + \mu\{2\} > \mu\{1, 2\}$  **sub-additive**  $\rightarrow \{1, 2\}$  **substitutes**
- If  $\mu\{1\} + \mu\{2\} < \mu\{1, 2\}$  **super-additive**  $\rightarrow \{1, 2\}$  **complements**
- If  $\mu\{1\} + \mu\{2\} = \mu\{1, 2\}$  **additive**  $\rightarrow \{1, 2\}$  **independent**

# Möbius Representation and Choquet Integral

The Möbius representation of Fuzzy Measures allow to reduce the problem complexity ( **$k$ -additivity**); the monotonicity and boundary conditions are:

$$\left\{ \begin{array}{l} m(\emptyset) = 0 \\ \sum_{T \subseteq N} m(T) = 1 \\ \sum_{\substack{T \subseteq S \\ T \ni i}} m(T) \geq 0, \quad \forall S \subseteq N, \quad \forall i \in S \\ m(T) = 0 \quad \forall t > k \end{array} \right.$$

The Choquet Integral with respect to Möbius representation:

$$C_m(x_1, \dots, x_n) = \sum_{T \subseteq N} m(T) \bigwedge_{i \in T} x_i$$

## Example – Veto Effect

- i) *Mathematics is a **veto** criterion*
- ii) *Statistics  $\succ$  Literature*

<i>Student</i>	<i>Math.</i>	<i>Stat.</i>	<i>Liter.</i>	<i>Choquet</i>
<i>a</i>	10	10	10	10
<i>b</i>	10	10	9	9.7
<i>c</i>	10	9	10	9.5
<i>d</i>	7	8	9	7
<i>e</i>	7	10	5	6.4
<i>f</i>	7	5	10	6
<i>g</i>	6	7	7	6
<i>h</i>	6	10	5	5.7
<i>i</i>	6	5	10	5.5
<i>l</i>	5	10	10	5