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Descriptive profiles for sets of alternatives in MCDA

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3. Finding the profiles
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Introduction

Multiple Criteria Decision Aid

- aims at modelling the **preferences** of decision-makers;
- **aids** them in reaching certain **decisions**;

Alternatives	Criteria			
	Price ↓	Acceleration ↓	Safety ↑	...
Car 1	18,342	30.7s	good	...
Car 2	15,335	30.2s	medium	...
Car 3	16,973	29s	v.good	...
⋮	⋮	⋮	⋮	⋮

Modelling preferences

Value functions (U)

- each alternative receives a **score**;
- $U(x)$ = aggregated criteria evaluations of x ;
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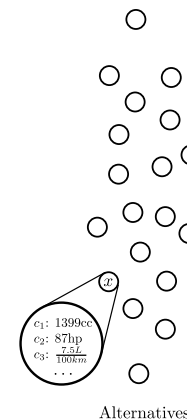
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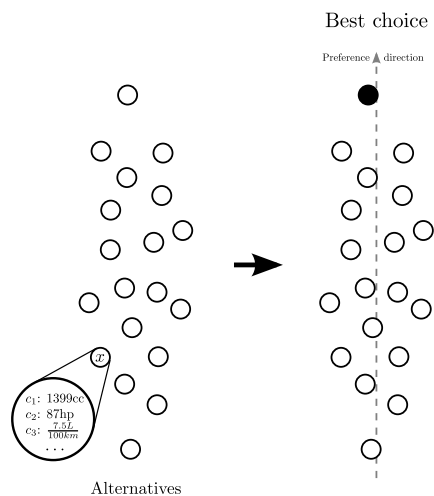
Preferential situations

$U(x) = U(y)$	Indifference (I)	$xSy \wedge ySx$
$U(x) > U(y)$	Strict preference (P)	$xSy \wedge y \not S x$
$U(x) \geq U(y)$	Weak preference (Q)	xSy
	Incomparability (R)	$x \not S y \wedge y \not S x$

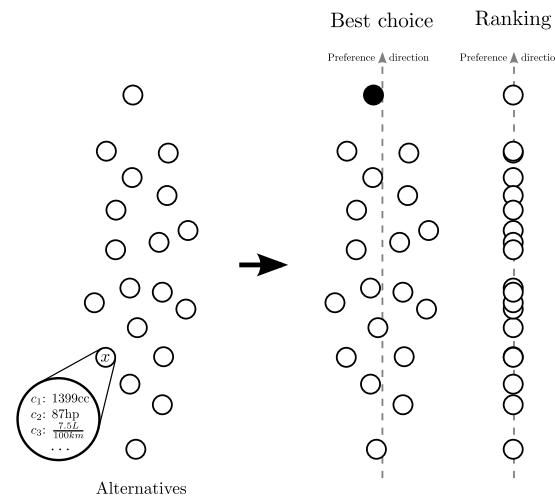
Main typologies of problems



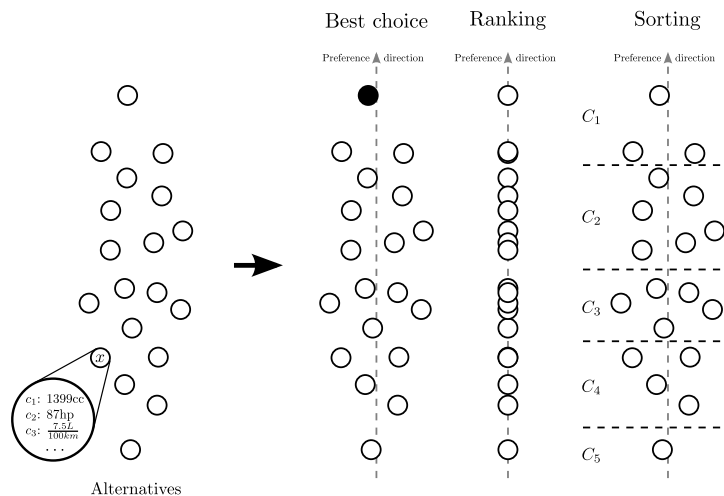
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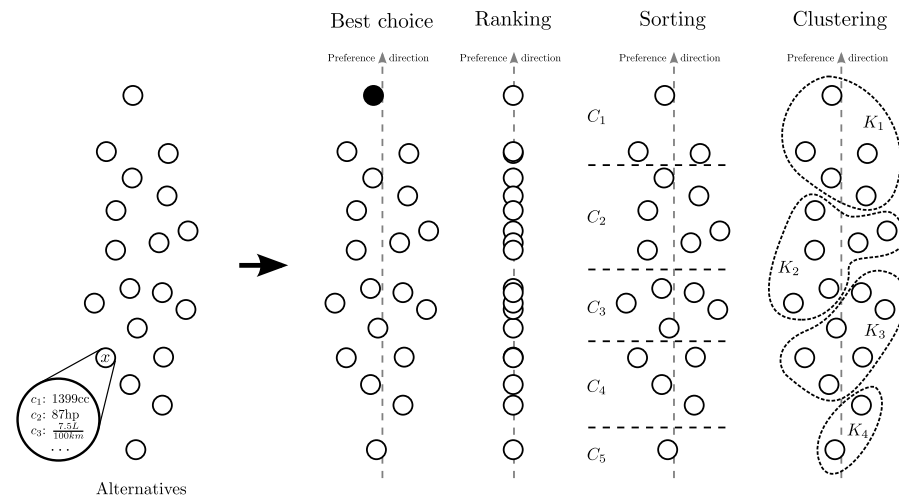
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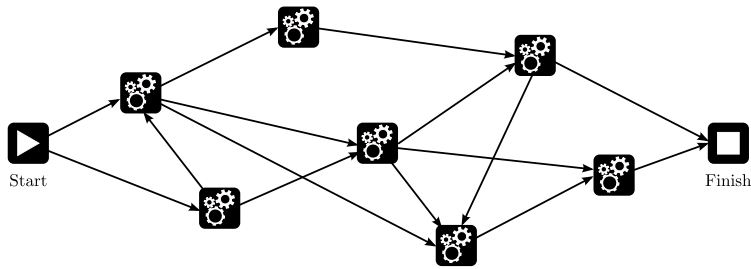
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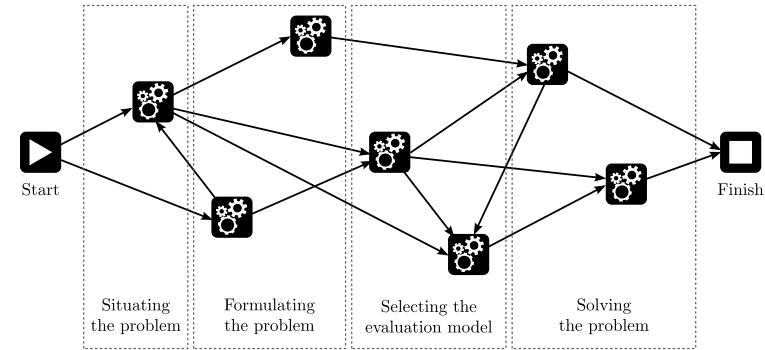
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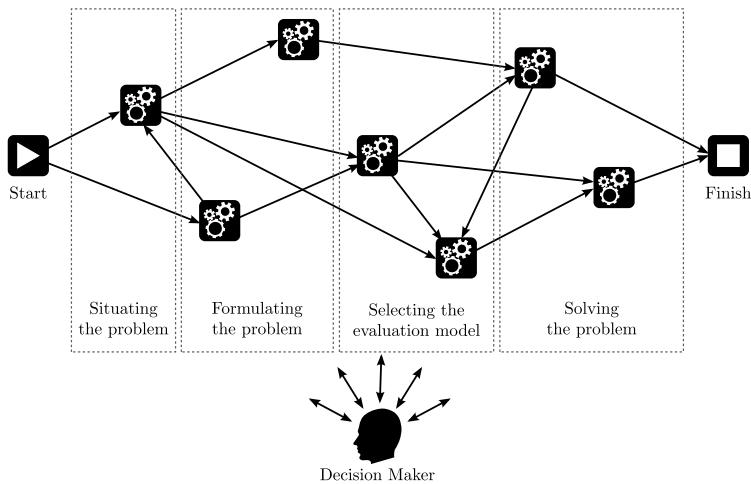
Decision Aiding process



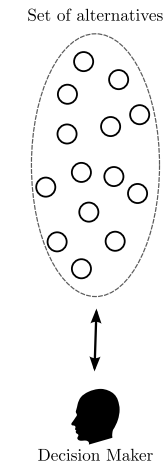
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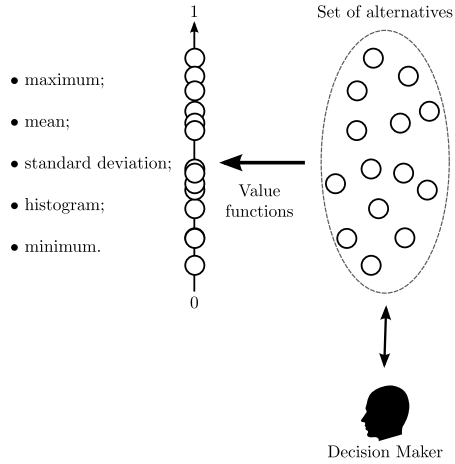
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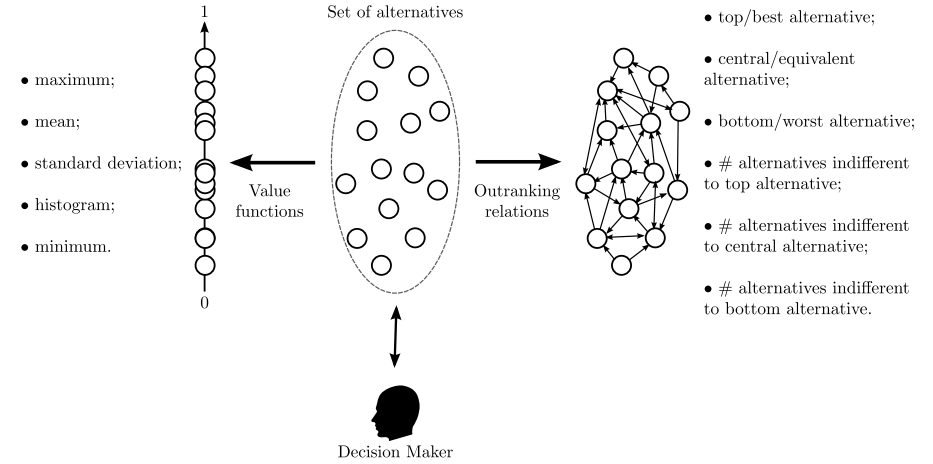
Motivation



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Defining the profiles

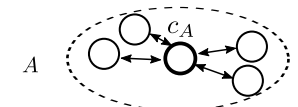
Defining the profiles

Setting

- a set of alternatives A ;
- a set of criteria F ;
- x_i evaluation of $x \in A$ on $i \in F$;
- outranking relation $S \rightarrow$ indifference relation I , strict preference relation P and incomparability R ;

Central profile (c_A)

- **indifferent** to the alternatives in A ;
- ◇ $f(c_A) = |\{x \in A: x I c_A\}|$;
- may be used to **replace** A ;
- useful for representing sets of indifferent alternatives.



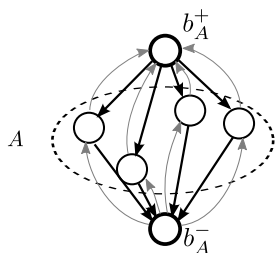
Bounding profiles (b_A^+, b_A^-)

- b_A^+ is above A
(**not strictly preferred by** $\forall x \in A$);
 b_A^- is below A
(**not strictly preferred to** $\forall x \in A$);
- b_A^+, b_A^- are close to A (**indifferent to** as many $x \in A$);

$$\diamond f(b_A^+) = |A| \cdot \left| \left\{ \begin{array}{l} x \in A: b_A^+ S x \\ x \in A: x S b_A^+ \end{array} \right\} \right|;$$

$$\diamond f(b_A^-) = |A| \cdot \left| \left\{ \begin{array}{l} x \in A: x S b_A^- \\ x \in A: b_A^- S x \end{array} \right\} \right|;$$

- **extend** a central profile;
- useful for representing sets of less indifferent alternatives.

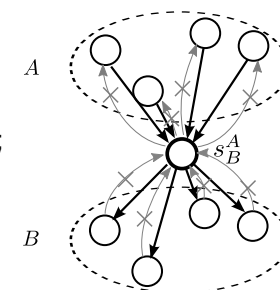


Separating profile (s_B^A)

- s_B^A is between A and B
(**not strictly preferred to** $\forall x \in A$ and **not strictly preferred by** $\forall x \in B$);
- s_B^A is separated from A and B
(**not indifferent to** as many $x \in A \cup B$);

$$\diamond f(s_B^A) = (n + m) \cdot \left| \left\{ \begin{array}{l} x \in A: x S s_B^A \\ x \in B: s_B^A S x \end{array} \right\} \right| + (n + m) \cdot \left| \left\{ \begin{array}{l} x \in A: s_B^A \not S x \\ x \in B: x \not S s_B^A \end{array} \right\} \right|;$$

- useful for delimiting two sets of alternatives that are ordered;



Finding the profiles

Finding the profiles

Exact approaches

Selecting

- an existing alternative from A (or B) that maximizes f ;

Building

- a fictitious alternative from the evaluations of $x \in A$ (or B);

$$c_{Ai} = \frac{1}{n} \sum_{x \in A} x_i \quad b_{Ai}^+ = \max_{x \in A} x_i \quad b_{Ai}^- = \min_{x \in A} x_i \quad s_{Bi}^A = \frac{1}{2} (c_{Ai} + c_{Bi})$$

- using a linear program that models the outranking relation S ([Bisdorff, Meyer, Roubens 07]*, [Bisdorff 12]) between the profiles and the alternatives in A and B .

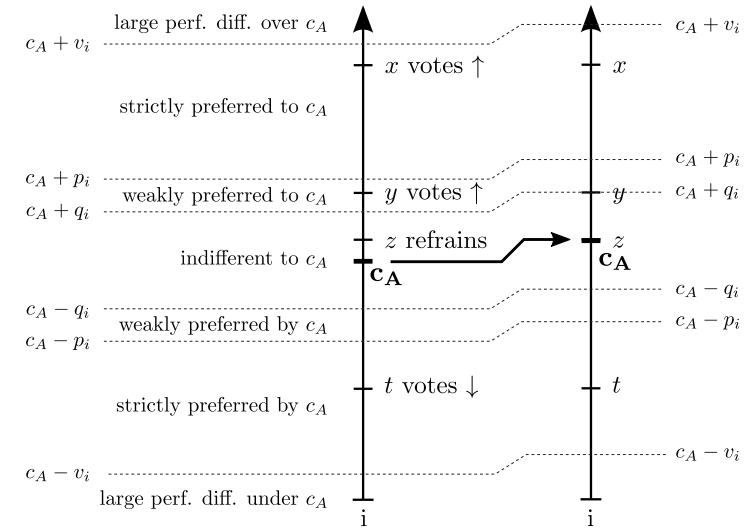
* with only one veto threshold

Approximative approach

Meta-heuristic

- single **solution meta-heuristic**:
 - start from an initial solution;
 - iteratively change it until a stop criterion is met;
- tested **simulated annealing**:
 - ability to **escape** local optima;
 - relatively easy to tune (cooling schedule);
 - may use restarts;
- used the outranking relation S from [Bisdorff, Meyer, Roubens 07] with only one veto threshold;
 - proposed a **heuristic** for the algorithm.

Heuristic



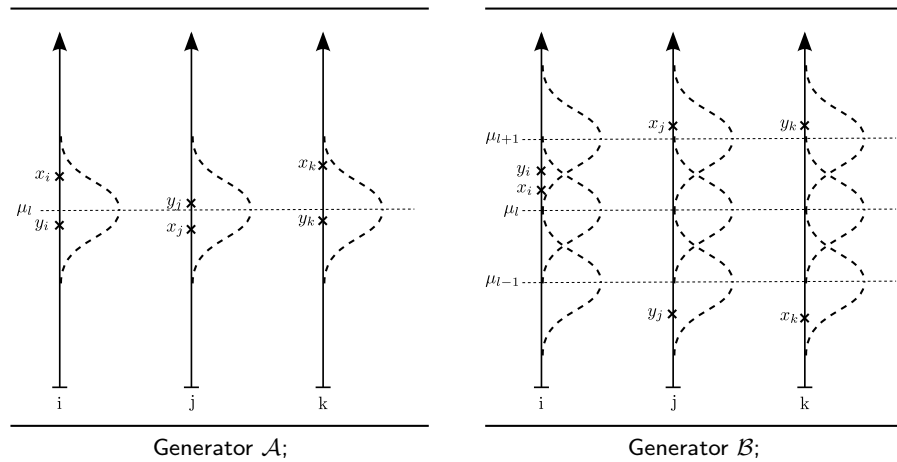
Results

Results

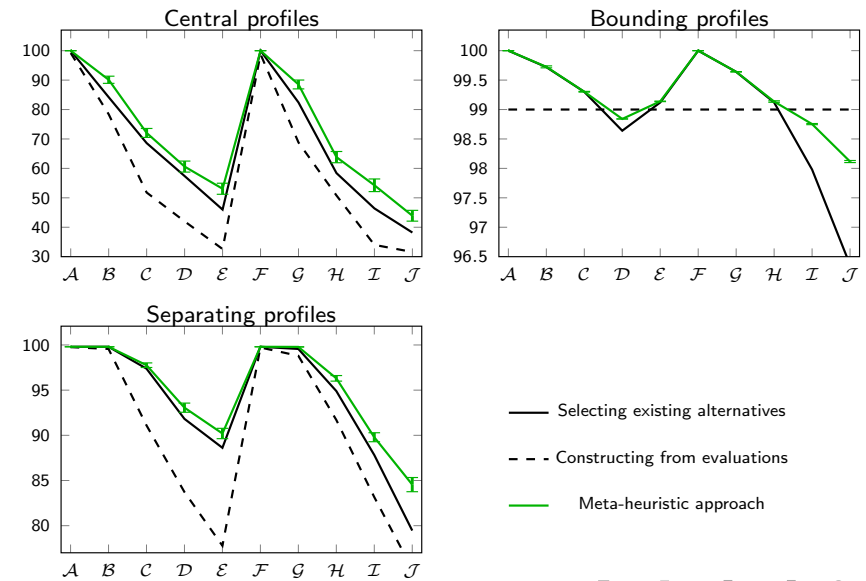
Experiments description:

- constructed a series of 50 benchmarks:
 - 50 alternatives;
 - 11 criteria;
 - $[0, 1]$ ratio scales;
 - 10 classes of difficulty $(\mathcal{A} - \mathcal{J})$;
- considered a fictive DM:
 - outranking relation S from [Bisdorff, Meyer, Roubens 07] with only one veto threshold;
 - equally significant criteria;
 - indifference, preference and veto thresholds;
 - median cut $(\lambda = 0.5)$;
- executed all the approaches (except linear programs > 60 min) (50 executions over each benchmark, 10 seconds each);
- compared results w.r.t. the fitness measures.

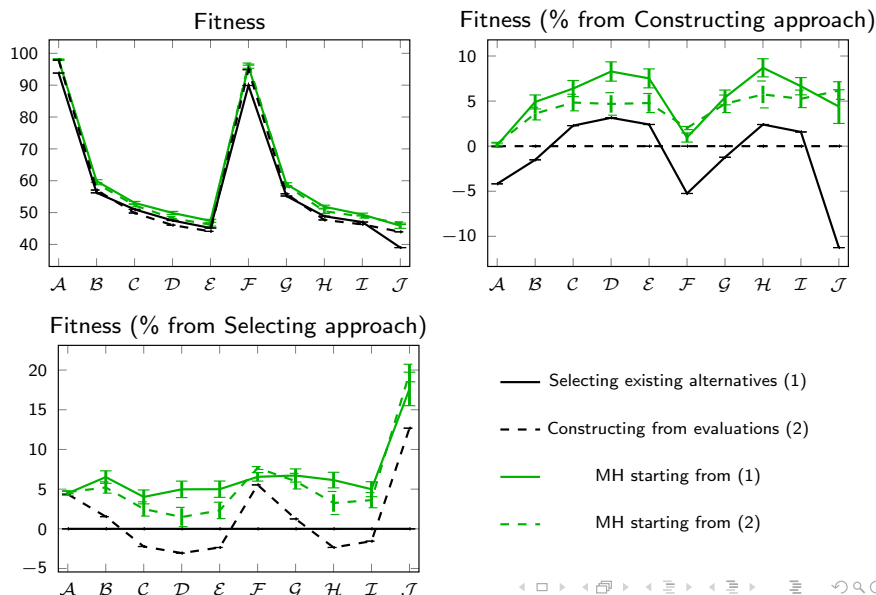
Benchmark construction



Results for all profiles



Results for central profiles using a valued indifference relation



Conclusions and perspectives

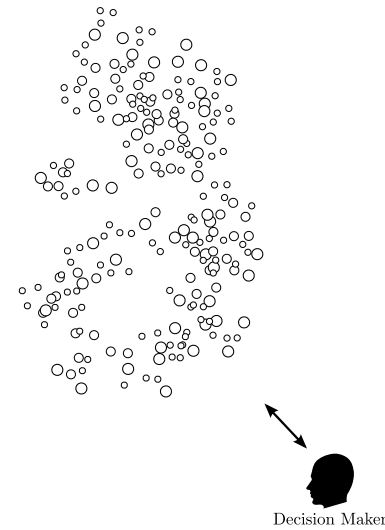
Conclusions

- **selecting** an alternative is generally better than **constructing** one from mean, max or min evaluations;
- meta-heuristic provides **significant improvements** over exact approaches for central profiles (over 5% even when using the credibility of the indifference relation);
- improvements for bounding and separating profiles are not so visible (modelling the two objectives);
- using **min** and **max** evaluations for bounding profiles maximizes the first set of objectives → should only model the second (which brings the profiles closer to the alternatives);

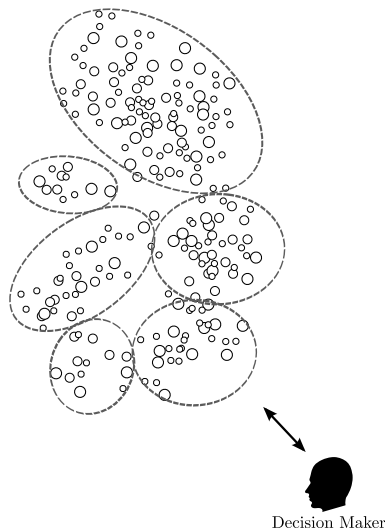
Perspectives

- further investigation into bounding and separating profiles and the representation of their fitness;
- finding the optimum result for each benchmark;
- inclusion of the veto in the heuristic;
- easy extension of using the weights in the heuristic;
- application for describing clustering results.

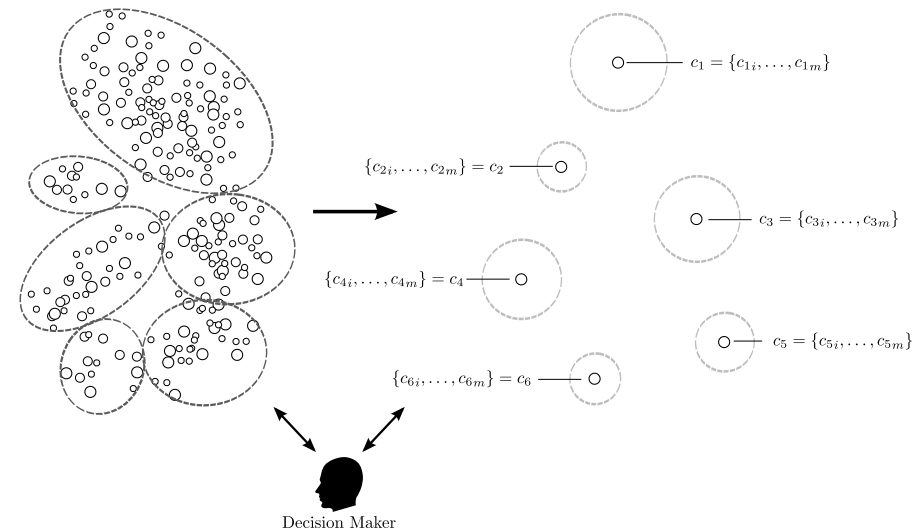
Perspectives



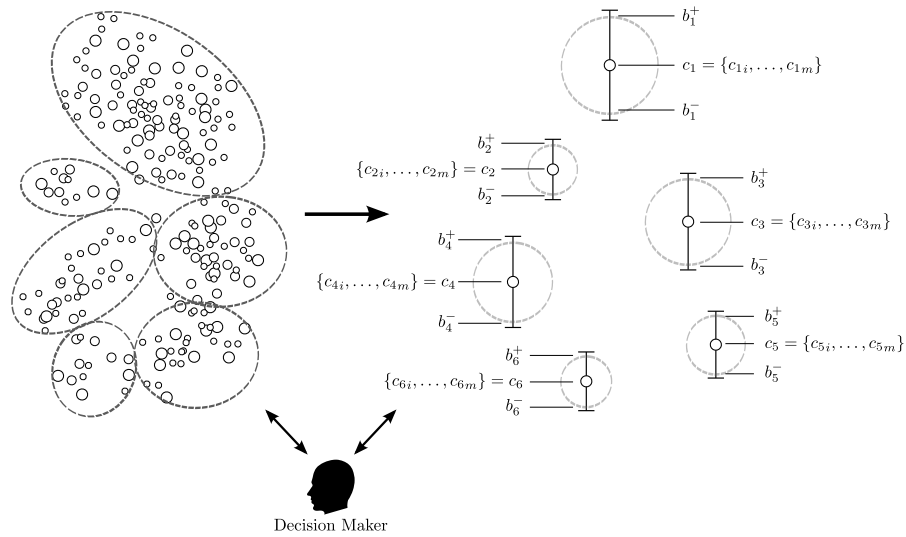
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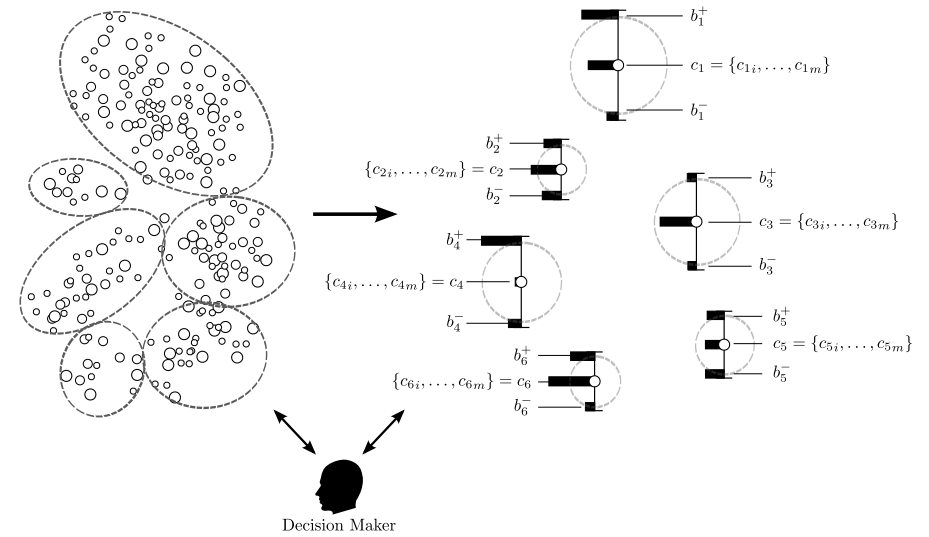
Perspectives



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