

# What is a Decision Problem?

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# Outline

- 1 Motivations
- 2 Methods
- 3 Problem Statements
- 4 Questions

# Problems

- Patients triage in emergency room;
- Identification of classes of similar DNA sequences;
- Star ratings of hotels;
- Waste collection vehicle routing;
- Vendor rating and bids assessment;
- Optimal mix of sausages;
- Chip-set lay out;
- Airplanes priority landing;
- Tennis tournament scheduling ...

# What is a decision problem?

Consider a set  $A$  established as any among the following:

- an enumeration of objects;
- a set of combinations of binary variables (possibly the whole space of combinations);
- a set of profiles within a multi-attribute space (possibly the whole space);
- a vector space in  $\mathbb{R}^n$ .

## Technically:

A Decision Problem is a partitioning of  $A$  under some desired properties.

# What is important?

## What does really matter?

In designing, choosing, applying, implementing, understanding, explaining, justifying, a method?

## What are the primitives?

And what is the derived information and the expected outcomes?

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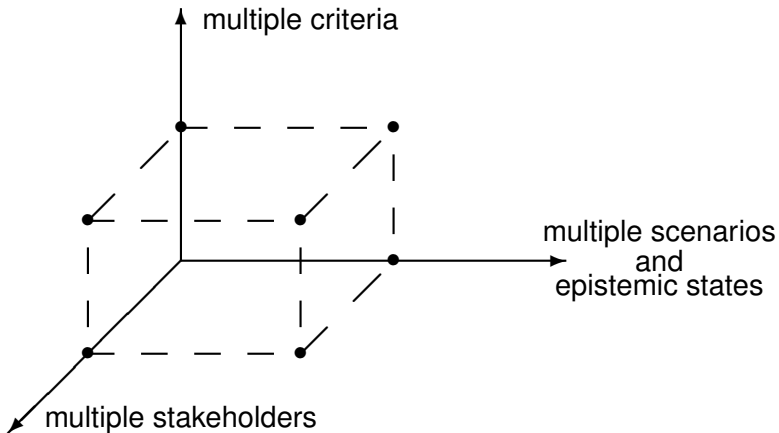
## What are the primitives?

And what is the derived information and the expected outcomes?

# Why is not straightforward?

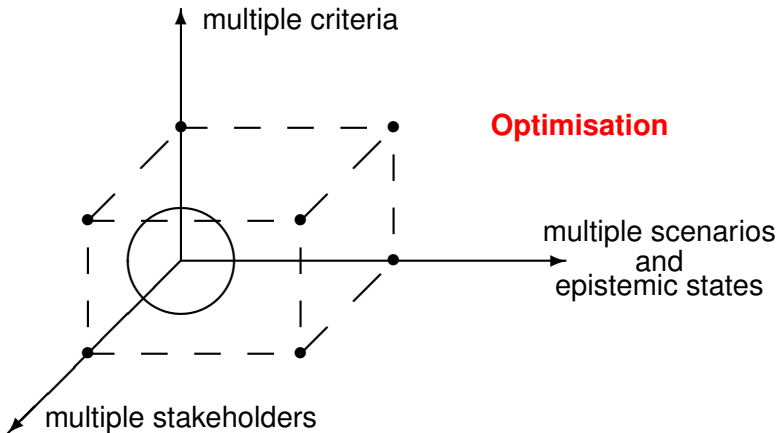
- multiple opinions
- multiple values
- multiple likelihoods
- + algorithmic aspects

# Partitioning? How?

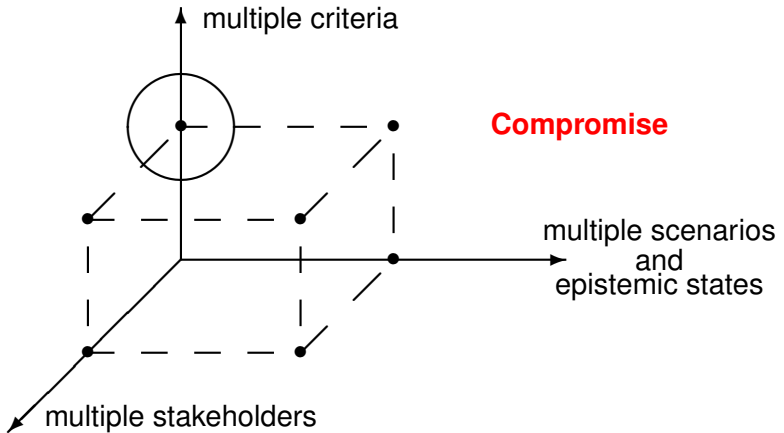




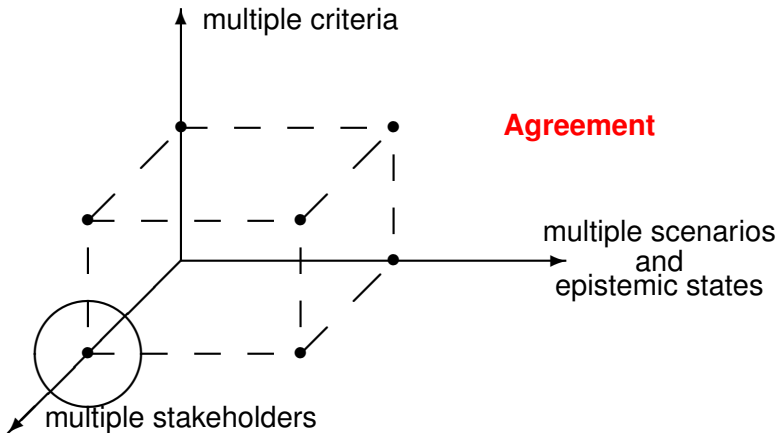
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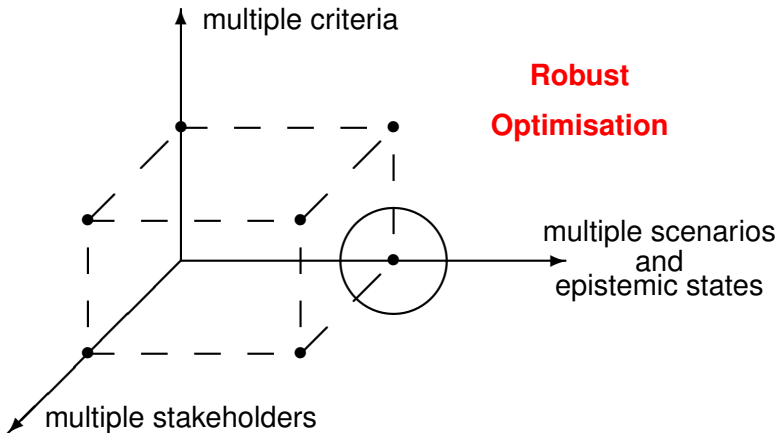
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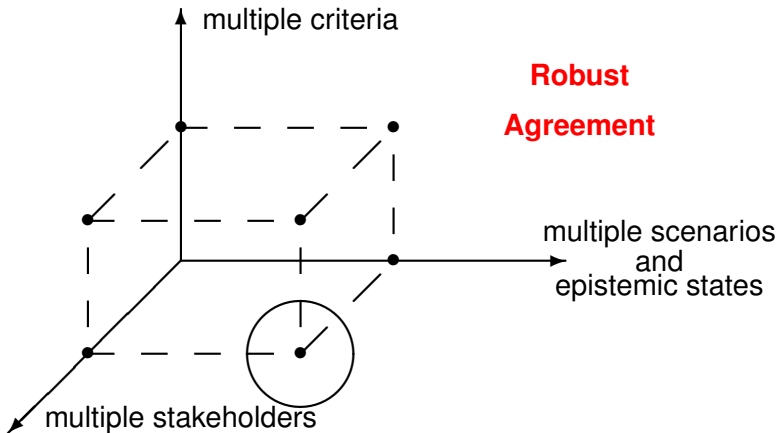
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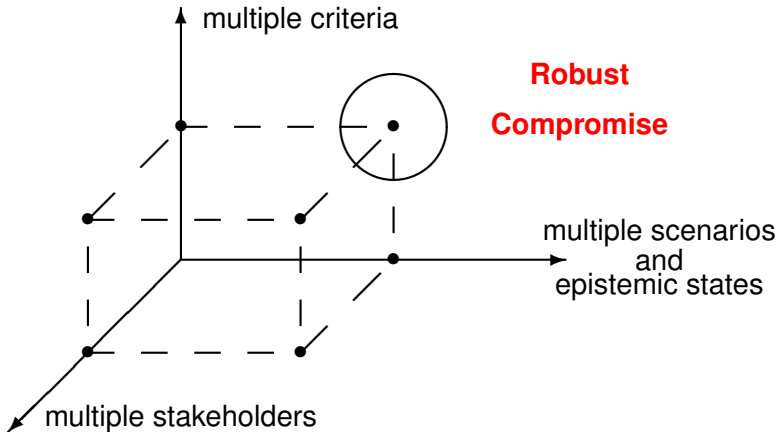
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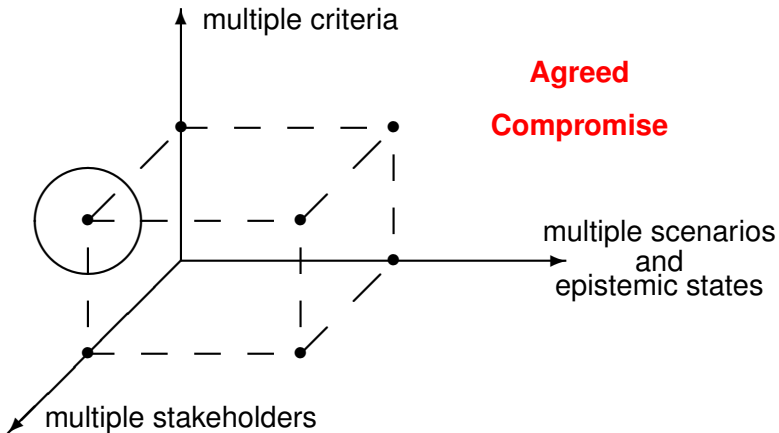
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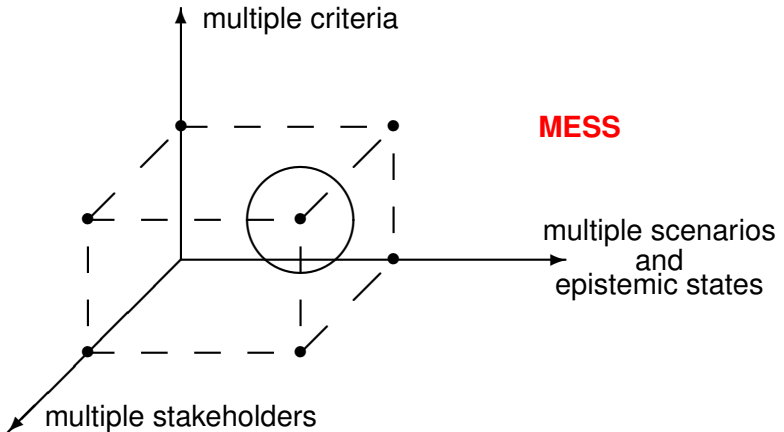
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# Partitioning? How?





# Is that all?

- Behind a criterion other criteria may be considered in a hierarchy of criteria (objectives);
- Behind a stakeholder other actors may have to be considered, that precise stakeholder being a speaker for a community;
- Behind a state of the nature other uncertainties may have to be considered;
- Any combination of the above may in reality occur as complex as possible.

# Claims

## Claim 1

All the previously mentioned problems boil down in aggregating some ordering relations applied on the set  $A$ .

## Claim 2

Establishing the set  $A$  is on its turn a decision problem. We explore one step of the recursion without any loss of generality.

## Claim 3

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# Primitives

- The set  $A$ .
- The description of the elements of  $A$ .
- Preference (ordering) statements about  $A$  and its subsets.
- Preference statements are of two types: relative and absolute ones.

# Critical Issues

- The set of alternatives
- Problem statement
- Differences of preferences
- Hierarchy/Separability/Indipendence
- Positive and Negative Reasons

# Partitioning? For what?

Practically we partition  $A$  in  $n$  classes. These can be:

	Pre-defined wrt some external norm	Defined only through pairwise comparison
Ordered	Rating	Ranking
Not Ordered	Assigning	Clustering

Two special cases:

- there are only two classes (thus complementary);
- the size (cardinality) of the classes is also predefined.

# What is a ranking problem?

## Primitive

The primitive is a binary relation on  $A$ :  $\succeq \subseteq A \times A$  to be read “at least as good as”.

## Result

The result is a partitioning of  $A$  in  $[A_1], \dots, [A_n]$  such that:

$[A_j] \geq [A_i] \Leftrightarrow j \geq i$  and

$\forall x \in [A_j], y \in [A_i] : x \succeq' y$



# Discussion 1

What is a choice problem?

We partition  $A$  in two classes  $[A_1] \geq [A_2]$ . Thus  $[A_1] = \sup_A(\succ')$ .

What is an optimisation problem?

A choice problem for which:

- $\succ = \succ'$
- $x \succ y \Leftrightarrow f(x) \geq f(y)$ .
- Thus  $[A_1] = \max_A f(x)$

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- $\succ = \succ'$
- $x \succ y \Leftrightarrow f(x) \geq f(y)$ .
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## Discussion 2

### Why is $\succ'$ different from $\succ$ ?

Generally speaking  $\succ$  is not an ordering relation since preferences can be partial and or inconsistent. If we have to proceed with some operational procedure we need to transform  $\succ$  to an ordering relation  $\succ'$ .

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How do we learn  $\succ$ ?

What properties should  $\succ'$  fulfill?

# What is a clustering problem?

## Primitive

The primitive is a set of binary relations on  $A$ :  $\approx_I \subseteq A \times A$   
 to be read “similar to”.

## Result

The result is a partitioning of  $A$  in  $[A_1], \dots [A_n]$  such that:

$\exists \approx_I : \forall x, y \in [A_j] \quad x \approx y$  and

$\forall x \in [A_j], y \in [A_i] : \neg(x \approx y)$

# Discussion 1

## Indiscernibility.

In case  $\approx_j$  are equivalence relations then the partitioning of  $A$  results in constructing the indiscernibility relation on  $A$ .  
However, this is not generally the case and  $[A_j] = \sup_A(\approx_j)$ .

In other terms we try to maximise similarity within classes (clusters) and minimise similarity among classes (clusters).

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## Discussion 2

### Distances.

If  $\approx_l$  are nested similarity relations with nice properties then we can establish a metric:

- $s(x, y)$ : how similar is  $x$  to  $y$ ?
- $d(x, y)$ : how distant is  $x$  from  $y$ ?

Then  $[A_y] = \{x \mid \max_A F(s(x, y))\}$ ,

$F$  being a measure of the overall similarity of the elements of  $[A_y]$  with respect to  $y$ .

What properties should  $F$  and the metrics fulfill?

# What is a rating problem?

## Primitive

The primitive is a binary relation on  $A$ :  $\succeq \subseteq A \times P \cup P \times A$   
 to be read “at least as good as”.

$P$  being the set of external “norms” characterising the ordered classes  $C_1 \triangleright \cdots \triangleright C_n$

## Result

The result is to assign each element of  $A$  in a  $C_j$  such that:  
 $x \in C_j \Leftrightarrow x \succeq' p_j, p_{j+1}, \dots, p_n$  and  $p_1 \cdots p_{j-1} \succeq' x$

# Discussion 1

## Constraint Satisfaction

If  $\forall x, y \in A \cup P \ x \succcurlyeq y \Leftrightarrow f(x) \geq f(y)$ .

Then  $x \in C_j \Leftrightarrow f(p_{j-1}) \geq f(x) \geq f(p_j)$ .

This is a Constraint Satisfaction Problem.

Why is  $\succcurlyeq'$  different from  $\succcurlyeq$ ?

Generally speaking  $\succcurlyeq$  is not an ordering relation since preferences can be partial and or inconsistent. If we have to proceed with some operational procedure we need to transform  $\succcurlyeq$  to an ordering relation  $\succcurlyeq'$ .

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# What is an assigning problem?

## Primitive

The primitive is a set of binary relations on  $A$ :  $\approx_I \subseteq A \times P \cup P \times A$   
 to be read “similar to”.

$P$  being the set of external “norms” characterising the classes  
 $C_1 \cdots C_n$

## Result

The result is to assign each element of  $A$  in a  $C_j$  such that:

$$x \in C_j \Leftrightarrow \exists \approx_I: x \approx_I p_j$$

# Discussion 1

## Constraint Satisfaction

If  $\forall x, y \in A \cup P \ x \approx_I y \Leftrightarrow f(x) = f(y)$ .

This is once again a Constraint Satisfaction Problem.

## Basic Claim

- **Any unsupervised decision problem is an optimisation problem.**
- **Any supervised decision problem is a constraint satisfaction problem.**

Since any constraint satisfaction problem can be seen as an optimisation problem,  
we can definitely focus only to the later ones

# Why $x \succeq y$ ?

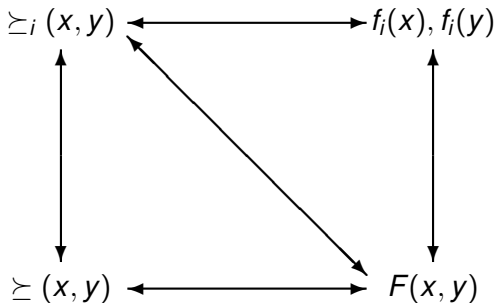
When in reality we just know that:

$$\begin{array}{rcc}
 w \succcurlyeq_1 z \succcurlyeq_1 & x \succcurlyeq_1 y & \succcurlyeq_1 t \\
 w \succcurlyeq_2 & y \succcurlyeq_2 x & \succcurlyeq_2 t \succcurlyeq_2 z \\
 w \succcurlyeq_3 t \succcurlyeq_3 & x \succcurlyeq_3 y & \succcurlyeq_3 z \\
 z \succcurlyeq_4 & y \succcurlyeq_4 x & \succcurlyeq_4 t \succcurlyeq_4 w \\
 & \vdots &
 \end{array}$$



# The Problem

Suppose we have  $n$  ordering relations  $\succeq_1 \cdots \succeq_n$  on the set  $A$ .  
 We are looking for an overall ordering relation  $\succeq$  on  $A$   
 “representing” the different orders.



## Two fundamental questions

- 1 **How do we consider preferences and differences of preferences along a single criterion/dimension?**
- 2 How do we consider preferences and differences of preferences among several different criteria/dimensions?

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