

# Evolving Decision Principles

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# Outline of talk

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1. Motivation
2. Introduction
3. Stability with respect to alternative values
4. Stability with respect to weights
5. Results
6. Conclusions
7. Future work

# Motivation

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- Multicriteria decision problems – many subjective decisions
- Subjective decisions – not very exact
- Sensitivity analysis for small changes in the subjective values
- The need for stable solutions

# Introduction

- Buying a TV set – criteria: cost, quality of image, physical aspect, trade-name, etc.
- Ranking the alternatives – by using some **decision principle** (a function of the values of an alternative with respect to the different criteria)
- Sensitivity analysis – the analysis of the results, when we make small perturbations to the subjective values
- Stability index – takes values between 0 and 1 and measures the stability of a solution

# Multicriteria decision problem

- Criteria  $C_i$  with weights  $w_i$
- Alternatives  $A_j$  with aggregate values  $x_j$

$$\begin{array}{cc} & \begin{array}{cccc} x_1 & \cdot & \cdot & x_n \end{array} \\ & \begin{array}{cccc} A_1 & \cdot & \cdot & A_n \end{array} \\ \begin{array}{cc} w_1 & C_1 \\ \cdot & \cdot \\ \cdot & \cdot \\ w_m & C_m \end{array} & \left[ \begin{array}{cccc} a_{11} & \cdot & \cdot & a_{1n} \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot \\ a_{m1} & \cdot & \cdot & a_{mn} \end{array} \right] \end{array}$$

# Decision principle

$$f : \mathbf{R}_+^m \times \mathbf{R}_+^m \rightarrow \mathbf{R}_+$$

A1.  $f(\lambda \mathbf{v}, \mathbf{z}) = f(\mathbf{v}, \mathbf{z})$  and  $f(\mathbf{v}, \lambda \mathbf{z}) = \lambda f(\mathbf{v}, \mathbf{z})$   
(Homogeneity)

A2.  $f(v_{\sigma(1)}, \dots, v_{\sigma(m)}; z_{\sigma(1)}, \dots, z_{\sigma(m)}) =$   
 $f(v_1, \dots, v_m; z_1, \dots, z_m)$

A3.  $f$  is strictly increasing in variable  $z_i, i = 1, \dots, n$

A4.  $\min_{i=1, \dots, m} z_i \leq f(v_1, \dots, v_m; z_1, \dots, z_m) \leq \max_{i=1, \dots, m} z_i$

# The aggregated value

$$x_j = f(w_1, \dots, w_m; a_{1j}, \dots, a_{mj})$$

$\alpha$  power mean as decision principle:

$$m_\alpha(w_1, \dots, w_m; y_1, \dots, y_m) = \left( \frac{\sum_{i=1}^m w_i y_i^\alpha}{\sum_{i=1}^m w_i} \right)^{1/\alpha}$$

# Sensitivity analysis

- The analysis of the results, when we make small perturbations to the weights  $w_i$  and/or values  $a_{ij}$  of a decision table
- A solution  $\{x_1, \dots, x_n\}$  is **stable** with respect to these perturbations if the decreasing order of the new solution  $\{x'_1, \dots, x'_n\}$  coincides with the decreasing order of  $\{x_1, \dots, x_n\}$



# Sensitivity with respect to alternatives

## Minimum and maximum aggregate value

$$x_j^-(\varepsilon) = f(w_1, \dots, w_m; a_{1j}^-(\varepsilon), \dots, a_{mj}^-(\varepsilon))$$

$$x_j^+(\varepsilon) = f(w_1, \dots, w_m; a_{1j}^+(\varepsilon), \dots, a_{mj}^+(\varepsilon))$$

$x_j^-(\varepsilon)$  is attained for  $a_{ij}^-(\varepsilon) = a_{ij} - \varepsilon a_{ij}$

$x_j^+(\varepsilon)$  is attained for  $a_{ij}^+(\varepsilon) = a_{ij} + \varepsilon a_{ij}$

# Stability with respect to alternatives

Let  $\sigma, x_{\sigma(1)} \geq \cdots \geq x_{\sigma(n)}$

- Stability of the order  $x_{\sigma(j)} \geq x_{\sigma(j+1)}$   
the greatest  $\delta_j \in [0, 1]$  for which  
 $x_{\sigma(j)}^-(\delta_j \varepsilon) \geq x_{\sigma(j+1)}^+(\delta_j \varepsilon)$
- Stability index of decision principle  $f$

$$S(\varepsilon) = \left( \prod_{j=1}^{n-1} \delta_j \right)^{\frac{1}{n-1}}$$

# Genetic programming

- Evolutionary search for the **program** that produces the output corresponding to the input given in the fitness cases
- Algorithm
  1. Generate an initial population of individuals
  2. Execute each program of the population on the fitness cases and assign it a fitness value
  3. Create a new population of individuals by reproduction, crossover and mutation
  4. Iterate through steps 2-3

# Genetic programming setup

- Representation (only feasible solutions)

Terminals -  $\alpha$  power means applied to an alternative's values

Functions -  $\alpha$  power means with equal weights

- Evaluation method

1. For each  $\mathbf{A}_j$  compute  $x_j, x_j^-(\varepsilon), x_j^+(\varepsilon)$

2. Order the alternatives:  $x_{\sigma(1)} \geq \dots \geq x_{\sigma(n)}$

3. For each pair  $\mathbf{A}_{\sigma(j)}, \mathbf{A}_{\sigma(j+1)}$  compute  $\delta_j$

4. Compute  $Fitness = S(\varepsilon)$

# Sensitivity with respect to weights

## Minimum and maximum aggregate value

$$x_j^-(\varepsilon) = f(w_{1j}^-(\varepsilon), \dots, w_{mj}^-(\varepsilon); a_{1j}, \dots, a_{mj})$$

$$x_j^+(\varepsilon) = f(w_{1j}^+(\varepsilon), \dots, w_{mj}^+(\varepsilon); a_{1j}, \dots, a_{mj})$$

- Maximum (minimum) of a monotone function on a subset of the simplex  $S = \sum_{i=1}^m v_i = 1$  ( $W_\varepsilon = [\mathbf{w} - \varepsilon\mathbf{w}, \mathbf{w} + \varepsilon\mathbf{w}]$ )
- Branch and bound technique
- Starting from the initial simplex, we maintain a list of simplices
- We always divide the simplex with the largest upper bound and keep those of the resulting two simplices, which have intersection with the set of interest

# Stability with respect to weights

Let  $\sigma, x_{\sigma(1)} \geq \dots \geq x_{\sigma(n)}$

- Stability of the order  $x_{\sigma(j)} \geq x_{\sigma(j+1)}$   
the greatest  $\delta_j \in [0, 1]$  for which  
 $x_{\sigma(j)}^-(\delta_j \varepsilon) \geq x_{\sigma(j+1)}^+(\delta_j \varepsilon)$
- Stability index of decision principle  $f$

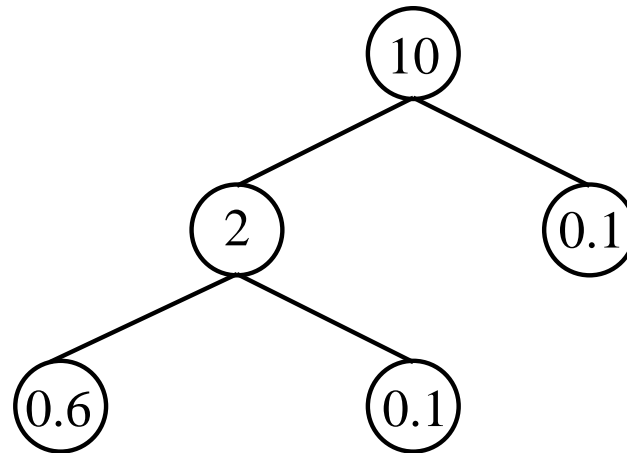
$$S(\varepsilon) = \left( \prod_{j=1}^{n-1} \delta_j \right)^{\frac{1}{n-1}}$$

# Example 1

|             |                    |       |       |       |       |       |       |       |       |       |          |
|-------------|--------------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|----------|
|             |                    | $A_1$ | $A_2$ | $A_3$ | $A_4$ | $A_5$ | $A_6$ | $A_7$ | $A_8$ | $A_9$ | $A_{10}$ |
| $w_1 = 0.1$ | $C_1 =$ aesthetics | 1     | 2     | 9     | 3     | 1     | 4     | 8     | 9     | 5     | 6        |
| $w_2 = 0.2$ | $C_2 =$ warranty   | 3     | 9     | 1     | 6     | 9     | 7     | 8     | 2     | 3     | 1        |
| $w_3 = 0.3$ | $C_3 =$ trade-name | 9     | 1     | 7     | 1     | 8     | 8     | 3     | 6     | 2     | 9        |
| $w_4 = 0.4$ | $C_4 =$ image      | 5     | 6     | 2     | 9     | 1     | 2     | 4     | 3     | 7     | 4        |

|        |            |       |       |          |          |       |       |       |       |       |       |
|--------|------------|-------|-------|----------|----------|-------|-------|-------|-------|-------|-------|
| best   | order      | $A_1$ | $A_7$ | $A_6$    | $A_{10}$ | $A_4$ | $A_9$ | $A_8$ | $A_2$ | $A_5$ | $A_3$ |
| GP     | $x_j$      | 4.80  | 4.60  | 4.38     | 4.31     | 4.28  | 4.06  | 3.91  | 3.75  | 3.39  | 3.20  |
|        | $\delta_j$ | 0.21  | 0.25  | 0.09     | 0.06     | 0.29  | 0.20  | 0.24  | 0.68  | 0.36  |       |
| arith- | order      | $A_1$ | $A_4$ | $A_{10}$ | $A_6$    | $A_7$ | $A_2$ | $A_5$ | $A_9$ | $A_8$ | $A_3$ |
| metic  | $x_j$      | 5.4   | 5.4   | 5.1      | 5        | 4.9   | 4.7   | 4.7   | 4.5   | 4.3   | 4     |
| mean   | $\delta_j$ | 0     | 0.39  | 0.1      | 0.1      | 0.26  | 0     | 0.23  | 0.25  | 0.40  |       |

# Best solution



$$f = \left( \frac{\left( \left( \frac{m_{0.6}^2 + m_{0.1}^2}{2} \right)^{\frac{1}{2}} \right)^{10} + m_{0.1}^{10}}{2} \right)^{\frac{1}{10}}$$

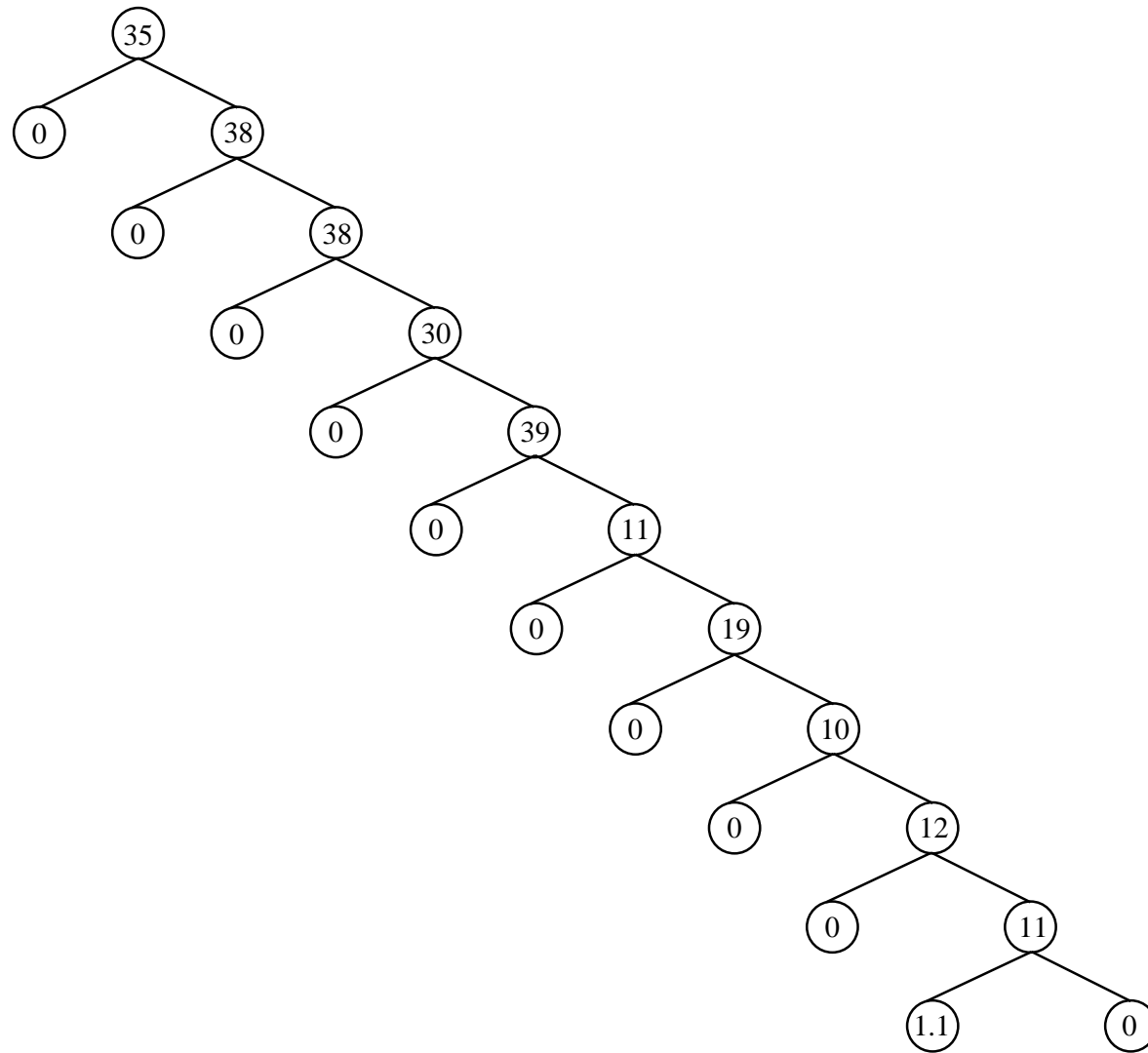


# Example 2

$$\begin{array}{l}
 w_1 = 0.3 \quad \mathbf{C}_1 \\
 w_2 = 0.2 \quad \mathbf{C}_2 \\
 w_3 = 0.5 \quad \mathbf{C}_3
 \end{array}
 \begin{array}{c}
 \mathbf{A}_1 \quad \mathbf{A}_2 \quad \mathbf{A}_3 \quad \mathbf{A}_4 \quad \mathbf{A}_5 \quad \mathbf{A}_6 \\
 \left[ \begin{array}{cccccc}
 1 & 8 & 9 & 9 & 5 & 9 \\
 7 & 9 & 1 & 6 & 9 & 7 \\
 9 & 1 & 7 & 1 & 8 & 8
 \end{array} \right]
 \end{array}$$

|                    |            |                |                |                |                |                |                |
|--------------------|------------|----------------|----------------|----------------|----------------|----------------|----------------|
| best<br>GP         | order      | $\mathbf{A}_6$ | $\mathbf{A}_5$ | $\mathbf{A}_3$ | $\mathbf{A}_1$ | $\mathbf{A}_2$ | $\mathbf{A}_4$ |
|                    | $x_j$      | 8.07           | 7.11           | 5.13           | 4.52           | 3.33           | 3.12           |
|                    | $\delta_j$ | 0.7            | 1              | 1              | 1              | 0.5            |                |
| arithmetic<br>mean | order      | $\mathbf{A}_6$ | $\mathbf{A}_5$ | $\mathbf{A}_3$ | $\mathbf{A}_1$ | $\mathbf{A}_2$ | $\mathbf{A}_4$ |
|                    | $x_j$      | 8.1            | 7.3            | 6.4            | 6.2            | 4.7            | 4.4            |
|                    | $\delta_j$ | 0.59           | 0.82           | 0.25           | 1              | 0.51           |                |

# Best solution



# Conclusions

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- A novel method for **sensitivity** and **stability analysis** of multicriteria decision problems with respect to alternative values and weights
- Alternatives – genetic programming for generating more stable decision principles
- Weights – special algorithm for computing stability

# Future work

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- Evolutionary stability analysis for weights
- Joint stability analysis for alternative values and weights
- More general stability analysis where only a number of alternatives are of interest