



# Empirical studies for comparing interactive multiobjective optimization methods with human participants

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# Outline

- Motivation
- Background
- Experiment design
- Results
- Conclusions

# About me

Giovanni Misitano  
PhD

- Post-doctoral researcher from the University of Jyväskylä.
- Working in the Multiobjective Optimization Group led by prof. Kaisa Miettinen.
- My main research interest lies in the interface between decision makers and (interactive) multiobjective optimization methods.
- One of the main contributors and maintainers of the DESDEO software framework (more on this later).



# Motivation

- Multiobjective optimization methods are tools to support decision-making.
- These tools can help decision makers find solutions to problems with multiple conflicting objective functions and no clear single optimum.
- Because of the large amount of optimal solutions, preferences can be utilized in some methods to guide the optimization process.
- Especially in so-called interactive multiobjective optimization methods, the decision maker can iteratively explore the set of optimal solutions and change their preferences, which allows them to learn about the available solutions and the feasibility of the preferences.





**But how to compare interactive multiobjective optimization methods? How to choose the best interactive method for a decision maker?**



# Background

- Multiobjective optimization problem
- Multiobjective optimization methods
- Interactive methods
- Reference Point method (RPM)
- NIMBUS
- E-NAUTILUS
- Comparing interactive methods

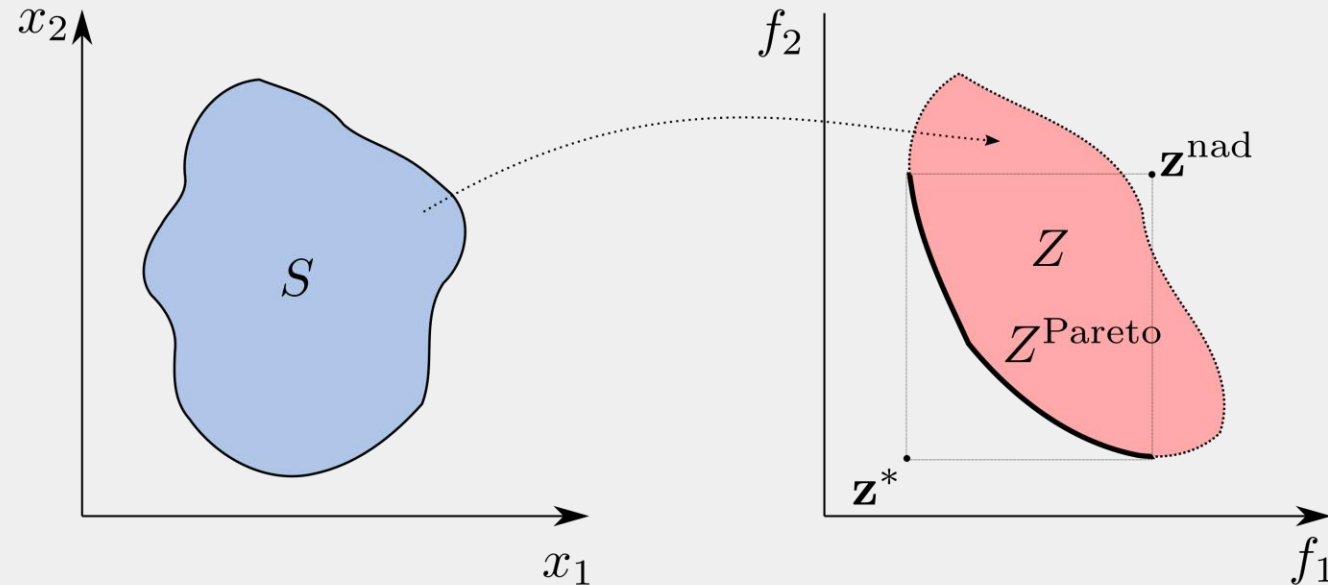


# Multiobjective optimization problem

## Problem definition

$$\begin{aligned} &\text{minimize} && F(\mathbf{x}) = (f_1(\mathbf{x}), f_2(\mathbf{x}), \dots, f_k(\mathbf{x})) \\ &\text{s.t.} && \mathbf{x} \in S \end{aligned}$$

## Central concepts



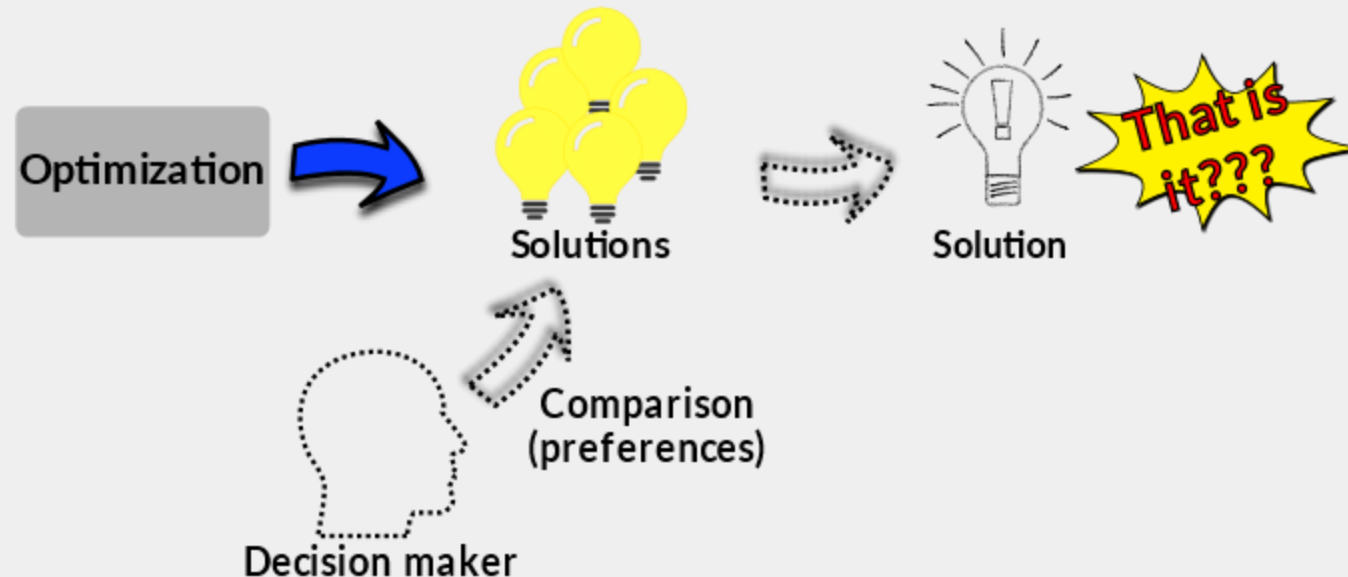


# Multiobjective optimization methods

## A priori methods



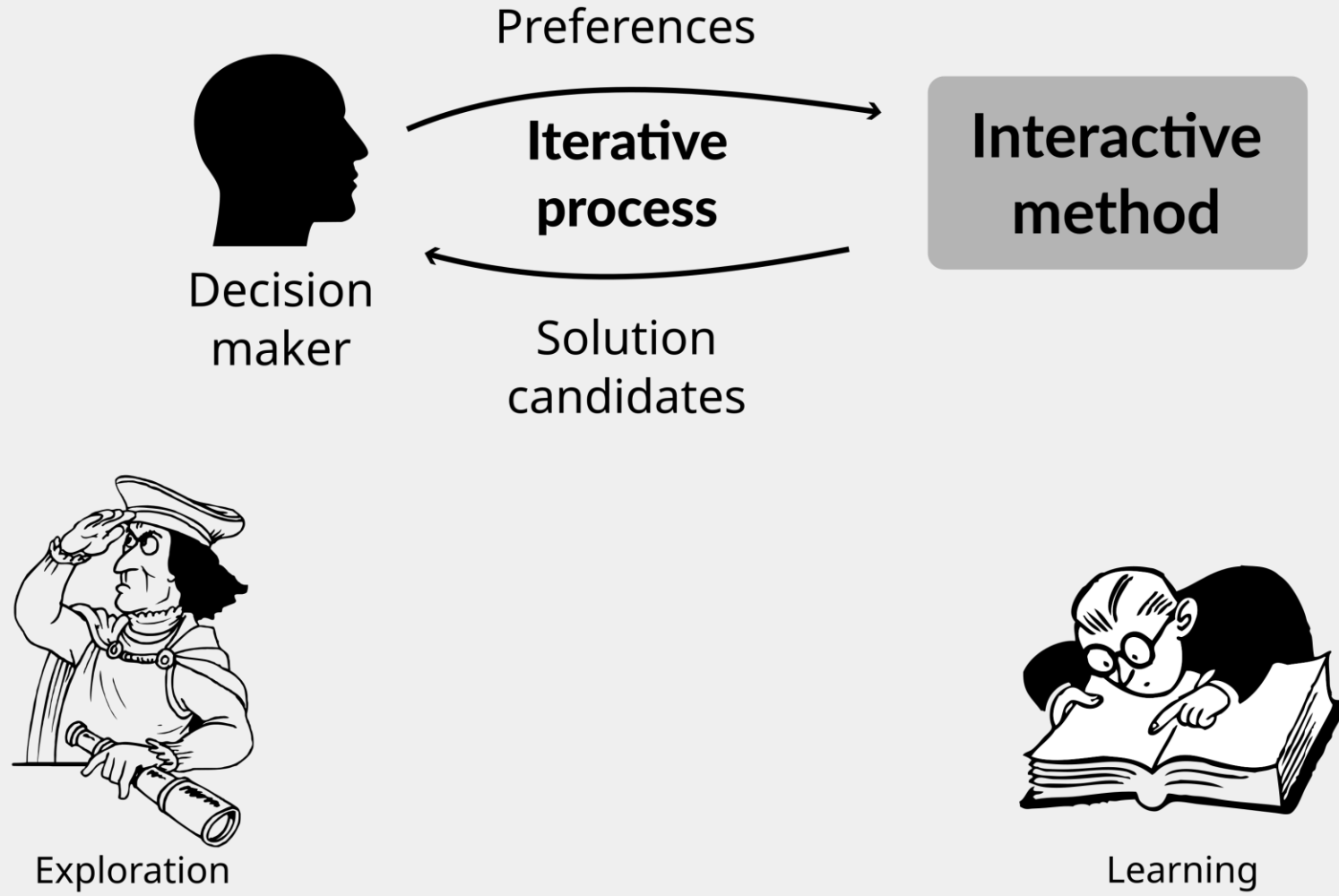
## A posteriori methods





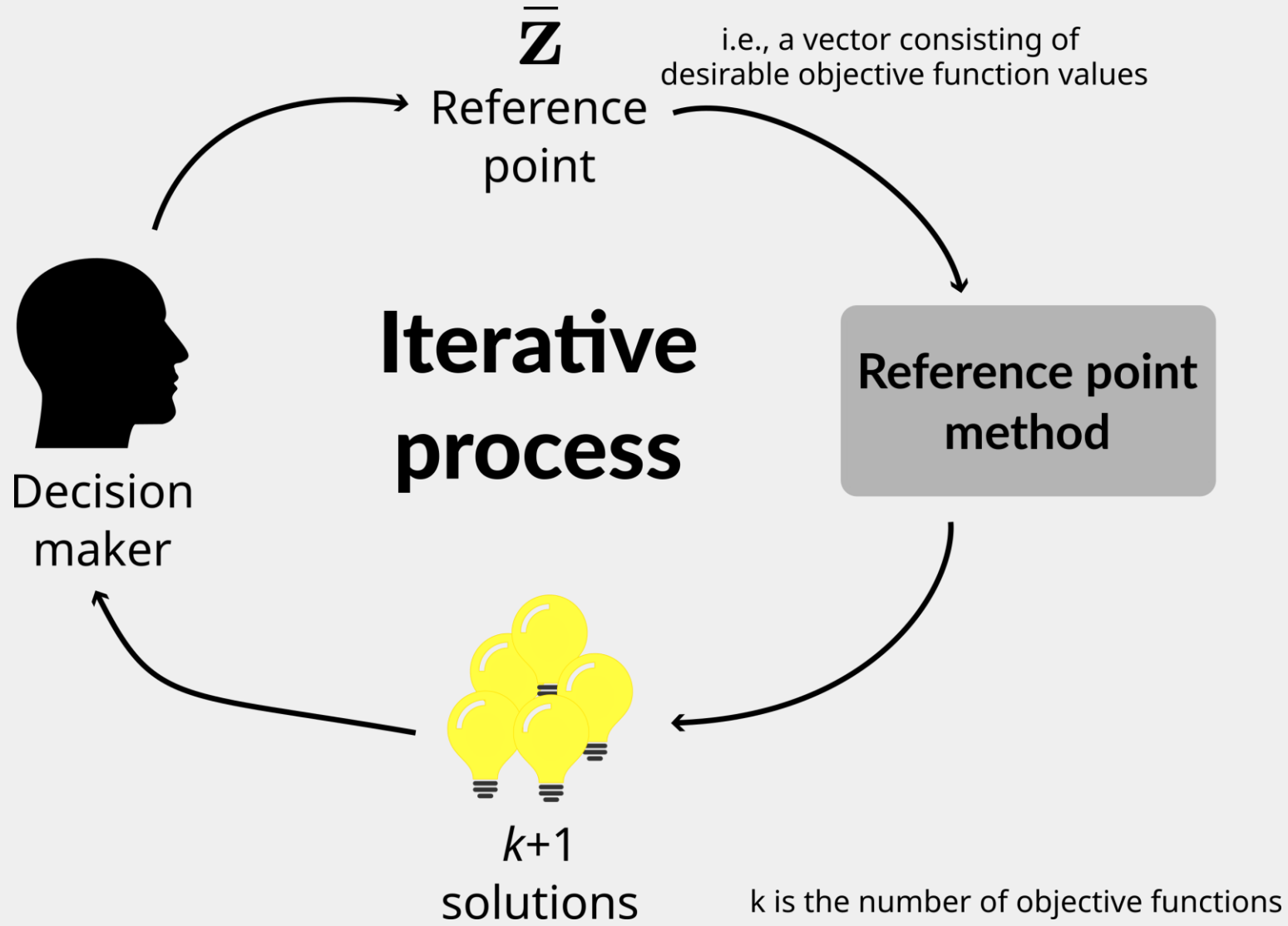


# Interactive methods



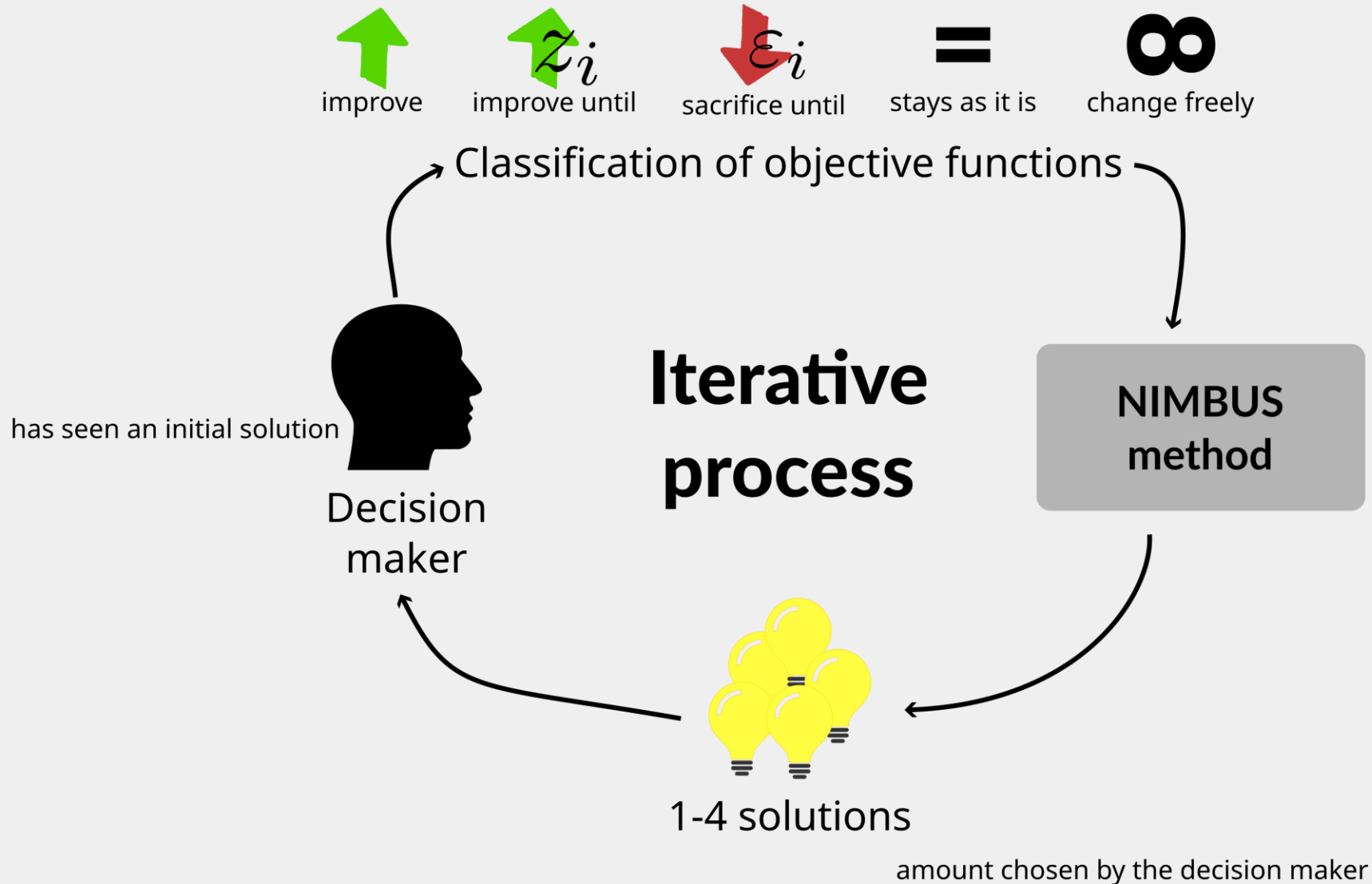


# The Reference Point method





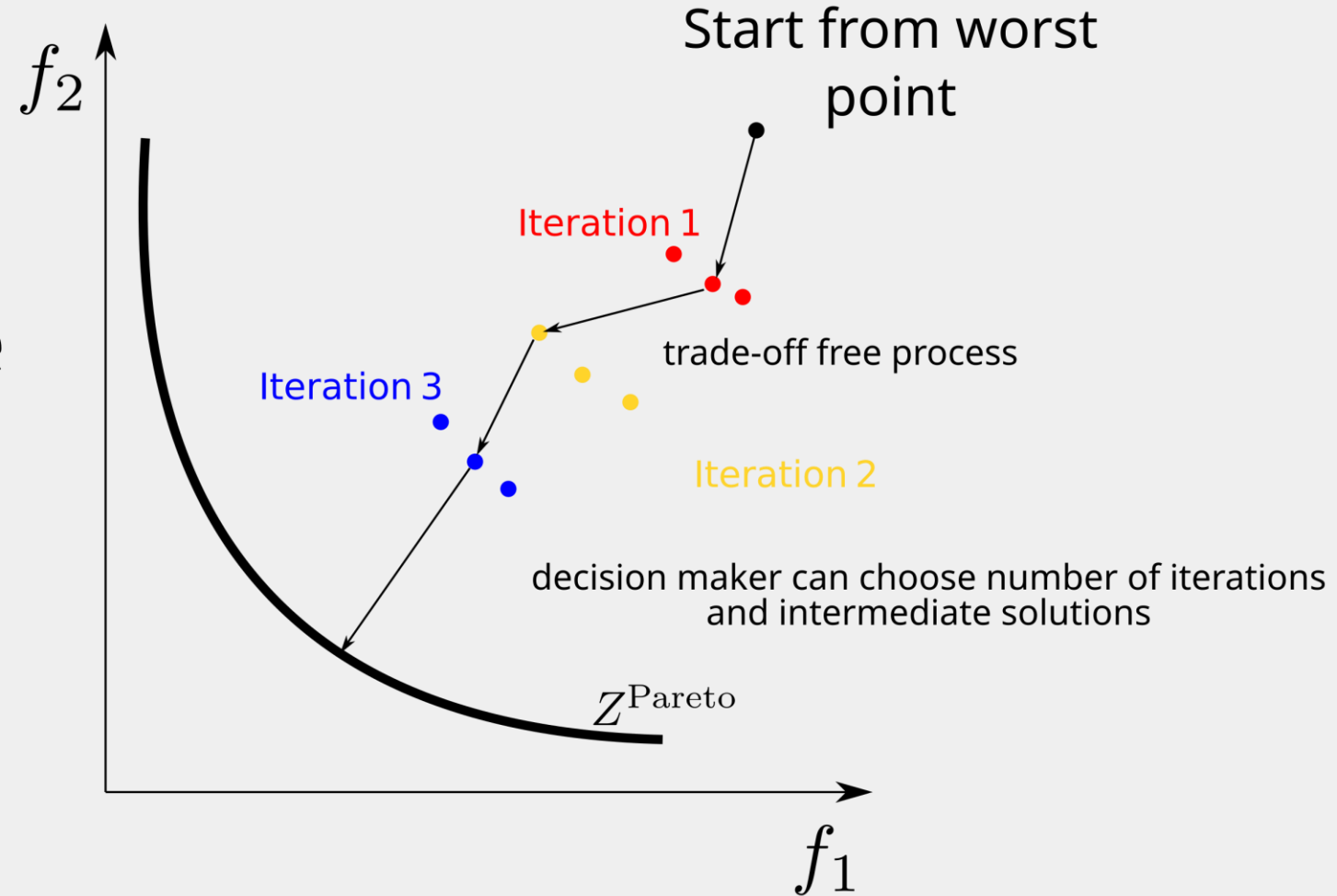
# The (simplified) NIMBUS method





# The E-NAUTILUS method

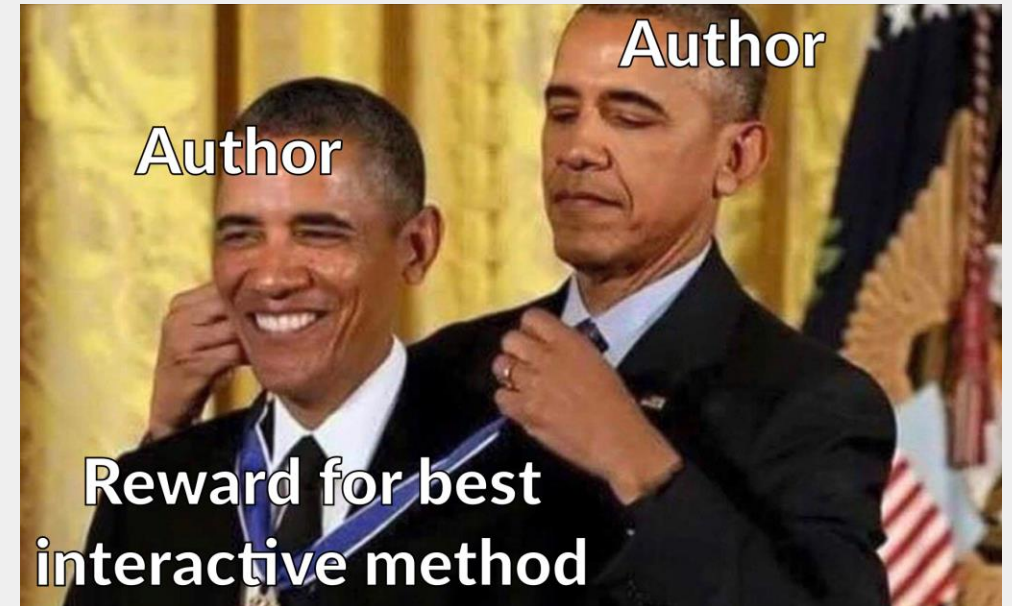
## Iterative process





# Comparing interactive methods

- Empirical research with decision makers is essential for accurately capturing human characteristics when comparing interactive methods.
- Recent studies lack such comparisons, and older studies have shortcomings, such as non-reproducibility because of missing experimental details (i.e., not reporting questions utilized) and reliance on one author simulating decision maker responses.
- This lack of detail prevents the replication of these experiments, as it is unclear how the studied phenomena were operationalized for measurement.



# Experiment design

- Questionnaire
- User interface
- Execution and practicalities



# Questionnaire

- Based on desirable properties identified for interactive multiobjective optimization methods by Afsar et al. 2021.
- In the first study (Afsar et al. 2023), the questionnaire was designed and piloted in a small-scale experiment with two interactive methods considered (within-subjects).
- In the second study (Afsar et al. 2024), it was refined and applied in a much larger experiment with three interactive methods considered (between-subjects).

## Research questions

RQ1: Cognitive load: How extensive is the cognitive load of the whole solution process?

RQ2: Capturing preferences and responsiveness: How well does the method capture and respond to the DM's preferences?

RQ3: Satisfaction and confidence: Is the DM satisfied with the overall solution process and confident with the final solution?



# Questionnaire

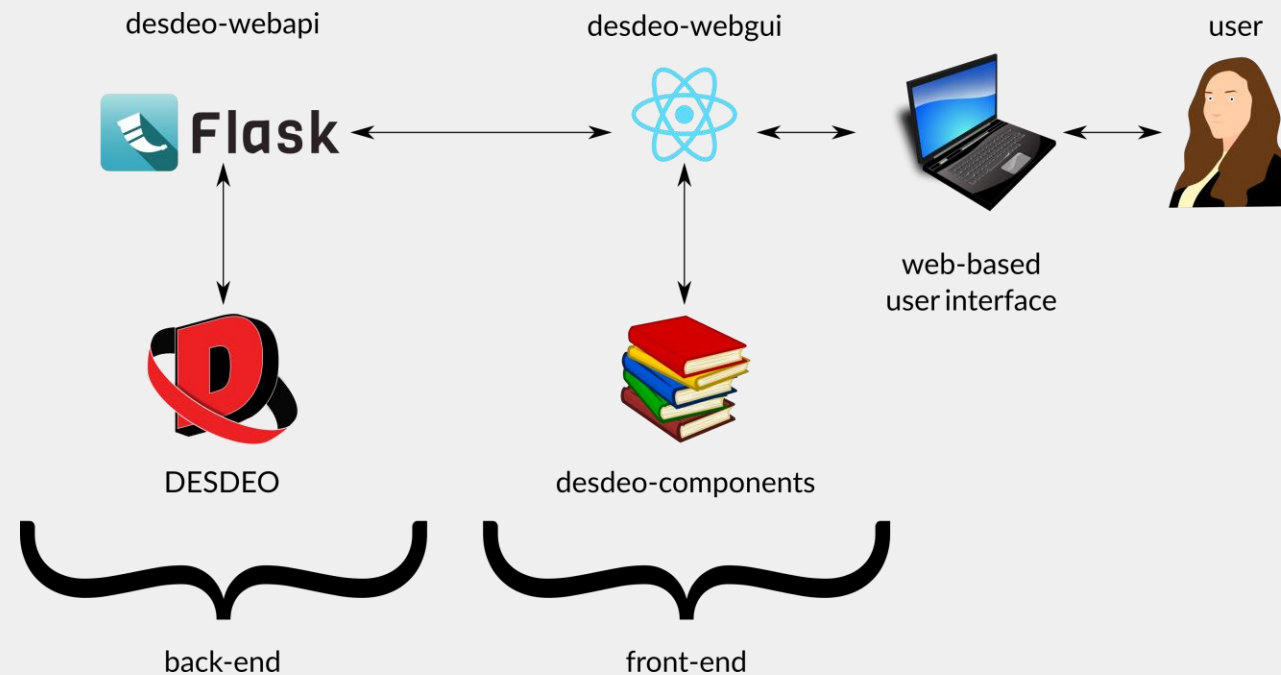
- Each desirable property is assessed by one or more questions on a
  - Likert scale: "**It was easy to explore solutions with different conflicting values of the objective functions.**" (RQ2: "The method easily changes the area explored as a response to a change in the preference information given by the decision maker.")
  - 5-point semantic differential scale: "**What degree of conflict do you think exists between each pair of objectives? Among f1 and f2? Among f1 and f3? Among f2 and f3?**" (RQ3: "The method sets as low a cognitive burden on the decision maker as possible.")
  - Open ended: "**Did some solution(s) surprise you? Why?**" (RQ3: The method does not miss any Pareto optimal solution that is more preferred (with a given tolerance) for the decision maker than the one chosen.")
- Some questions were adapted from the NASA task load index (NASA-TLX), a validated measurement.
- Total of 29 questions. Some to be answered during an interactive method, some after the decision maker has found their final solution.





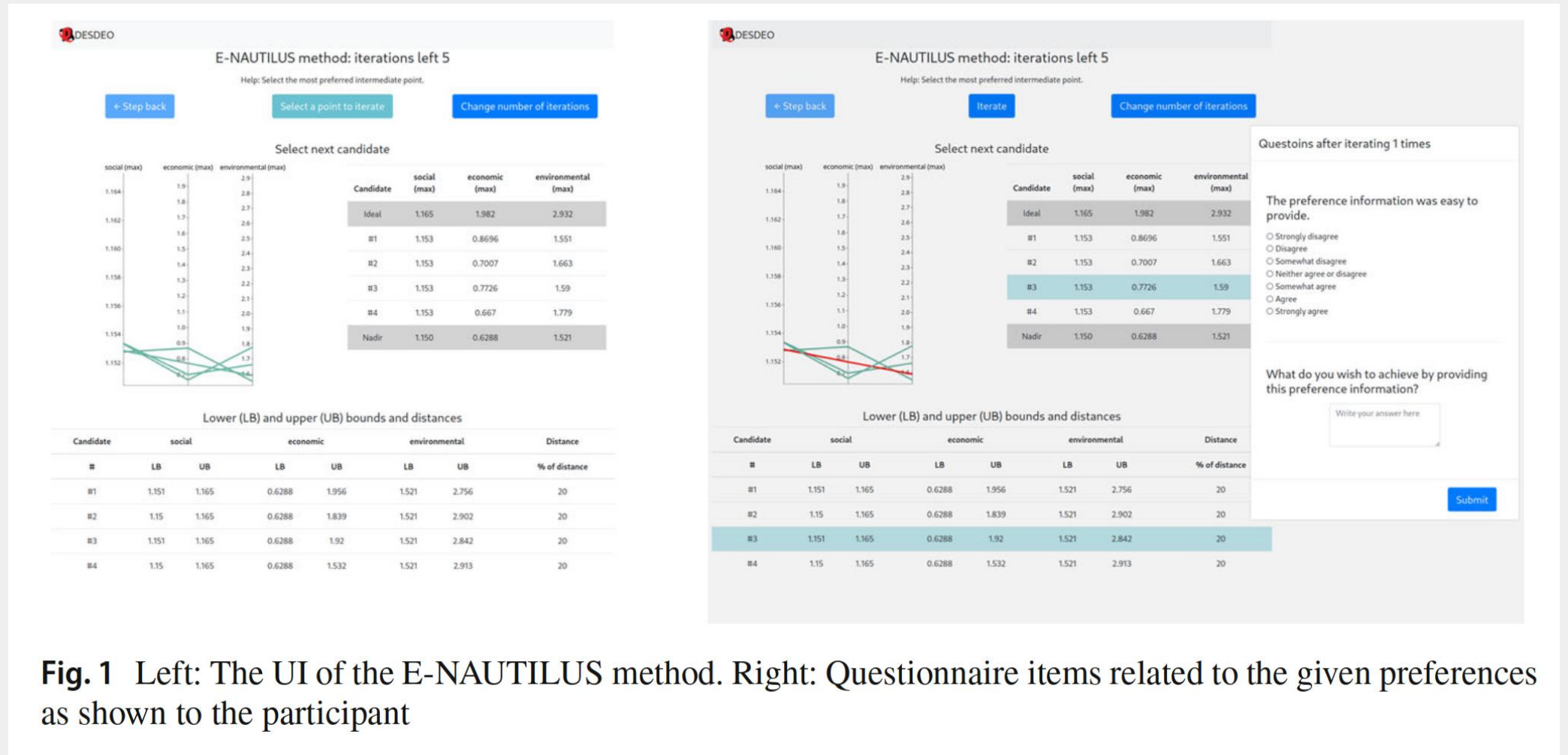
# User interface

- Web-based interfaces for each method (The Reference Point method, NIMBUS, E-NAUTILUS).
- Accessible with any modern web-browser.
- Questionnaire integrated in the interface, no need to switch to other applications. Possibility to log user information.
- We utilized the DESDEO framework and its implementations of the interactive methods under study.



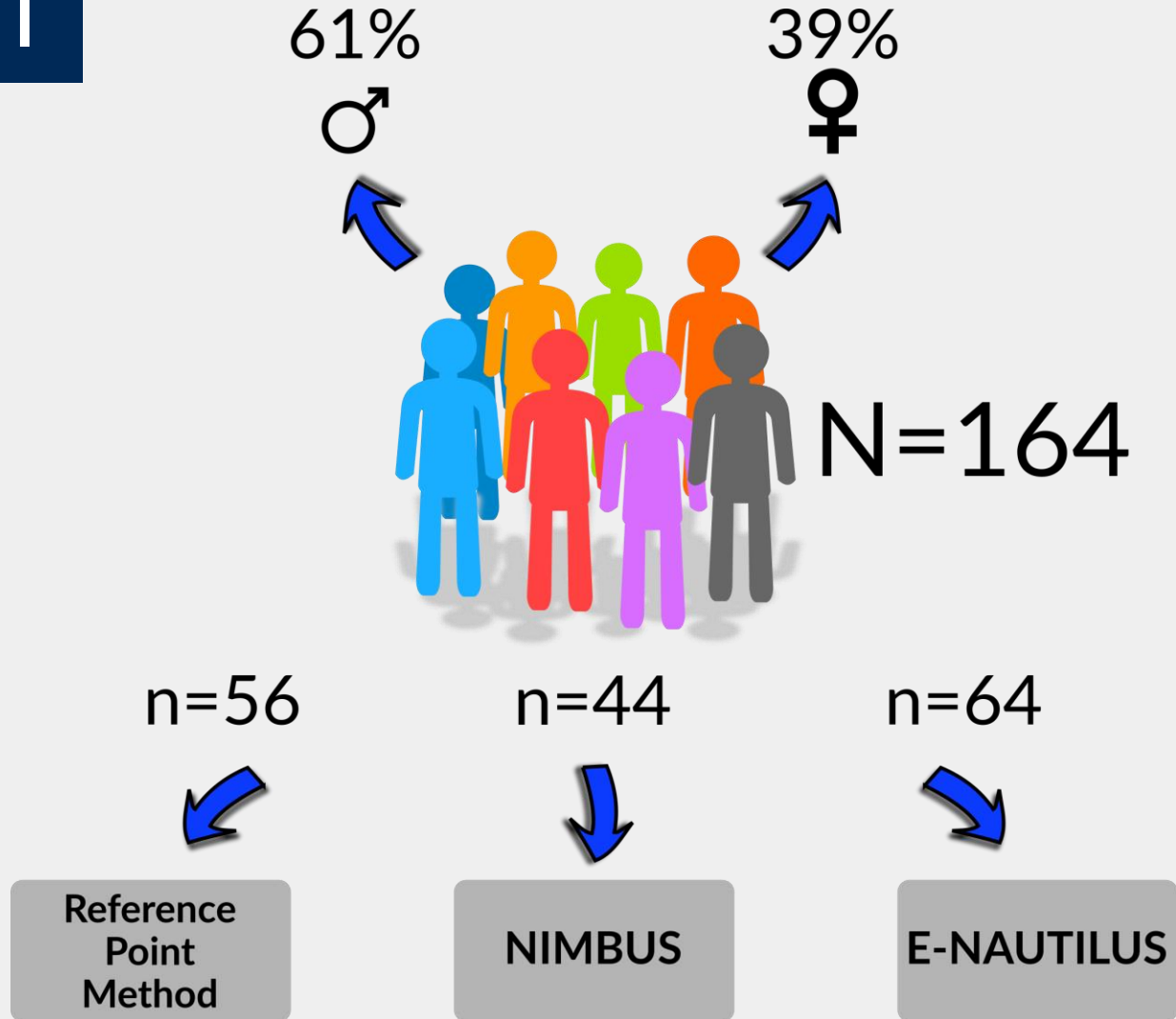


# User interface





# Execution and practicalities (in the second study)



- Between-subjects study
- Student participants from University of Malaga, Faculty of Economics and Business Studies.
- Age range 18-28, mean=19, and SD=2.2.
- Participants were briefed on the interactive methods and the problem being solved in advance to mitigate effects from being unfamiliar with the method or the problem.
- Participants were also provided with supplementary material prior to the experiment to allow them to think carefully about their preferences before the study.



# Execution and practicalities

## Problema de Programación Multiobjetivo para Analizar la Sostenibilidad de España

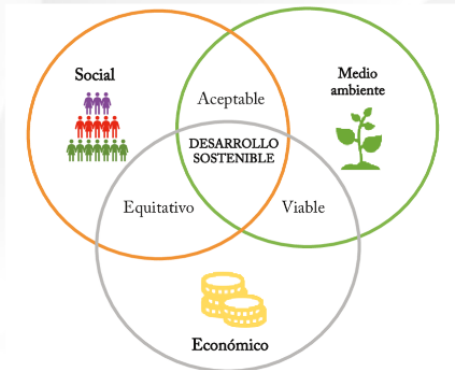
Indicadores individuales de la Agenda 2030 para el Desarrollo Sostenible de las Naciones Unidas



$\mathbf{x} = (x_1, \dots, x_{11})^T$  es el vector de decisión con los 11 indicadores individuales más significativos

Tres dimensiones de la sostenibilidad

$f_j(\mathbf{x})$  = indicador compuesto que representa la dimensión  $j$ , siendo  $j = 1$  (social), 2 (económica), y 3 (mediambiental)



Problema de Optimización Multiobjetivo

Maximizar  $\{f_1(\mathbf{x}), f_2(\mathbf{x}), f_3(\mathbf{x})\}$   
subject to  $\mathbf{x} \in S$

Significado de los valores de los indicadores compuestos para las tres dimensiones ( $f_1$ ,  $f_2$  y  $f_3$ )

Valores	Rendimiento global de la dimensión
[0, 1]	Entre el peor valor y el $P_{25}$ de los países de la UE
(1, 2]	Entre el $P_{25}$ y el $P_{50}$ de los países de la UE
(2, 3]	Entre el $P_{50}$ y el $P_{75}$ de los países de la UE
(3, 4]	Entre el $P_{75}$ y el mejor valor de los países de la UE

Situación de España en 2017

	Social ( $f_1$ )	Económico ( $f_2$ )	Medioambiental ( $f_3$ )
Ideal	1.1653	1.9822	2.9324
Actual	1.1550	1.3053	1.9281
Nadir	1.1504	0.6287	1.5214

- Decisor = Político interesado en encontrar el mejor compromiso posible entre las tres dimensiones para mejorar la sostenibilidad en España.
- ¿Qué sacrificios son necesarios entre las tres dimensiones para alcanzar una situación mejor?
- ¿Cuáles serían los mejores valores que se podrían lograr para las tres dimensiones, en base a tus preferencias?

- Problem with three objective functions to be maximized.
- The three objective functions consists of a social one, an economical one, and an environmental one.
- The goal was to choose a problem that would matter to the students, so that they must make actual choices.
- Based on the sustainable development goals setup by the United Nations.

# Results

- Analysis
- Answering RQ1
- Answering RQ2
- Answering RQ3
- Which method is best?



# Analysis

- To analyse the results, both a quantitative and qualitative approach was used.
  - For quantitative results (Likert scale and 5-point semantic differential scale) we adopted the Kruskal-Wallis test. We report the mean and standard deviation alongside  $p$ -values.
  - For qualitative results, we employed a qualitative content analysis approach.



# RQ1 - Cognitive load: How extensive is the cognitive load of the whole solution process?

## Key Takeaways

- **E-NAUTILUS**: offers a balanced approach with lower effort, frustration, and iterations, although it requires more time.
- **NIMBUS**: preferred for satisfaction but involves higher frustration and effort.
- **The Reference Point method (RPM)**: quick but less satisfactory and mentally demanding.



# RQ2 - Capturing preferences and responsiveness: How well does the method capture and respond to the DM's preferences?

## Key Takeaways

- **E-NAUTILUS:** easiest for providing preferences, learning to use, and exploring conflicting solutions; better at returning to previous solutions.
- **NIMBUS:** best at generating solutions that reflect preferences and reacting to changes; good at capturing and responding to preferences.
- **RPM:** adequate performance but not exceptional in any specific area; good at representing user preferences but with some difficulties due to trade-offs.





# RQ3 - Satisfaction and confidence: Is the DM satisfied with the overall solution process and confident with the final solution?

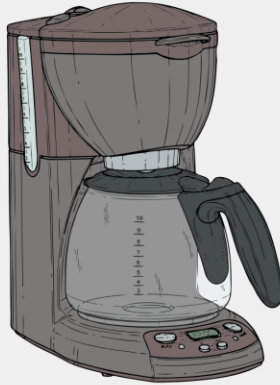
## Key Takeaways

- **E-NAUTILUS:** highest overall satisfaction, easy to learn and return to solutions; participants found solutions similar to imagined.
- **NIMBUS:** best for final solution satisfaction and confidence; moderate knowledge gain and surprise levels.
- **RPM:** adequate performance with lower satisfaction and more surprises in final solutions; less knowledge gain.



# Which method is best?

## E-NAUTILUS



**Best For:** Low mental effort, exploring trade-offs, clear understanding of overall solutions.

**Recommended When:** Ease of use and minimal frustration are priorities.

**Best For:** Fine-tuning final solutions, accurate preference reflection.

**Recommended When:** High satisfaction with final solution and detailed control are needed.

## NIMBUS



## RPM



(instant coffee)

**Best For:** Quick solutions with less complexity in preference information..

**Recommended When:** Time efficiency is crucial, but willing to accept lower satisfaction.

# Conclusions



# Conclusions

- Comparing interactive multiobjective optimization methods is challenging.
- We have designed, implemented, piloted, and applied our method.
- We have shared openly our method to others to use in their own experiments.
- More experiments are needed to truly say anything definitive on which interactive method is suitable for different situations. **More experiments are needed!**
- We are also exploring on how to compare interactive methods when the method is switched between iterations, and when multiple decision makers are involved in a group decision-making setting.



# To support your endeavors

- DESDEO has played an important supporting role in enabling the works discussed in this presentation.
- DESDEO is currently going through a complete overhaul, which will make it more usable and welcoming to wild new ideas, including explainability.

## The Multiobjective Optimization Group



We regularly post about  
our activities on LinkedIn!





## An experimental design for comparing interactive methods based on their desirable properties

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### Abstract

In multiobjective optimization problems, Pareto optimal solutions representing different tradeoffs cannot be ordered without incorporating preference information of a decision maker (DM). In interactive methods, the DM takes an active part in the solution process and provides preference information iteratively. Between iterations, the DM can learn how achievable the preferences are, learn about the tradeoffs, and adjust the preferences. Different interactive methods have been proposed in the literature, but the question of how to select the best-suited method for a problem to be solved remains partly open. We propose an experimental design for evaluating interactive methods according to several desirable properties related to the cognitive load experienced by the DM, the method's ability to capture preferences and its responsiveness to changes in the preferences, the DM's satisfaction in the overall solution process, and their confidence in the final solution. In the questionnaire designed, we connect each questionnaire item to be asked with a relevant research question characterizing these desirable properties of interactive methods. We also conduct a between-subjects experiment to compare three interactive methods and report interesting findings. In particular, we find out that trade-off-free methods may be more suitable for exploring the whole set of Pareto

Link to paper (open access!)



The Multiobjective Optimization Group



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