# Computationally Complex Multiobjective Problems: Experiences on Industrial Optimization

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## **Outline**

- Background
- Experiences on Industrial Optimization
- IND-NIMBUS Optimization Framework
- Future Directions

# Background

- I've been involved with industrial optimization since 1999
- MSc, Information Technology, University of Jyväskylä, 2008
- Currently as Researcher
  - Strategic Development of Multiobjective Optimization:
     Theory and Software, Academy of Finland
  - 2009–2012

## PhD Thesis

- Optimization framework for computationally complex multiobjective problems
  - Collection of papers
- Supervisor: prof. Kaisa Miettinen
- Area: Multiobjective optimization
- One published article
  - Laukkanen, T.; Tveit, T.-M.; Ojalehto, V.; Miettinen, K. & Fogelholm, C.-J. An interactive multi-objective approach to heat exchanger network synthesis Computers & Chemical Engineering, 2010, 34, 943-952
- Four articles under progress
  - GAMS-NIMBUS Tool for Multiobjective Optimization in the GAMS Modeling Environment Ojalehto, V. et al
  - Solving a Computationally Expensive Multiobjective Wastewater Treatment Plant Desing and Operation Problem with a Novel PAINT Method and the Interactive Method NIMBUS Hartikainen, M.; Ojalehto, V.
- Two articles under consideration
  - Testing framework for multi/single objective optimization

# Multiobjective Optimization

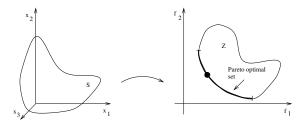
In multiobjective Optimization we consider a problem of the form

$$\begin{array}{ll} \text{minimize} & \{f_1(x),...,f_k(x)\} \\ & Ax \leq b \\ & g_1(\mathbf{x}) \leq 0 \\ & \text{subject to} & \vdots \\ & g_m(\mathbf{x}) \leq 0 \\ & x^l \leq \mathbf{x} \leq x^u \end{array} \right\} \quad \text{linear constraints} \quad \begin{cases} \mathbf{x} \in S. \\ & \text{subject to} \\ & \text{subject to} \end{cases}$$

 (Objective functions f can be either minimized or maximized)

# Pareto Optimality

- We consider several conflicting objective functions
- Concept of optimality Pareto optimality



 Other concepts: ideal vector (z\*), nadir vector (z<sup>nad</sup>), objective vector (z\*\*).

## NIMBUS method

- Classification is the central idea of the NIMBUS method
- The decision maker is asked to divide the functions into up to five different classes:
  - < objective value should be improved,
  - objective value should be improved till some aspiration level,
  - objective value is satisfactory at the moment,
  - objective value is allowed to impair up till some bound,
  - objective value is allowed to change freely.
- Upto four single objective subproblems

## **Experiences on Industrial Optimization**

- 1997 2001: DECISION, EU
  - Heterogeneous optimization oriented integration platform
  - Participants
    - End User: Dassault Aviation, Messet, Nokka Tume
    - Software Developer: NAG
    - Researchers: Inria, University of Jyväskylä, VTT
  - Main results
    - DEEP Platform IRIS Explorer with PBEXPE
    - Optimal Design of the Structure of Grapple Loader
- 2001 2002: Methodological and Implementational Challenges in Nonlinear Multiobjective Optimization and Decision Support, Academy of Finland
  - WWW-NIMBUS Development

## Experiences on Industrial Optimization with TEKES, I

- 2002 2005: Multiobjective Optimization in Product Development
  - The NIMBUS method for industry, that is, for non-academic use
  - Participants
    - End User: Metso Paper, Jyväskylän Teknologiakeskus Oy,(Liqum Oy,)
    - Software Developer: Numerola Oy
    - Researchers: University of Jyväskylä, VTT Prosessit
  - Results
    - IND-NIMBUS Software for multiobjective optimization
    - NIMBalas multiobjective optimization for BALAS Process Simulation Software
    - MOP Multiobjective process line optimization for paper making

## Experiences on Industrial Optimization with TEKES, II

- 2005–2008: Multiobjective Optimization and Multidisciplinary Decision Support
  - Developing a new intelligent decision support system and apply it to new application areas
  - Participants
    - End User: Andritz Oy, Foster Wheeler Energia Oy, Kuopio University Hospital, Kvaerner Power Oy, M-real Oyj, Patria Aerostructures Oy, Varian Medical Systems Finland Oy, Wärtsilä Finland Ov
    - Researchers: University of Jyväskylä, Helsinki School of Economics, Helsinki University of Technology, Tampere University of Technology, University of Kuopio
  - Results:
    - Pareto Navigator Interactive approximation method for nonlinear multiobjective optimization
    - SynHEAT Multiobjective optimization of heat exchanger network synthesis

Research Projects
Obstacles encountered
IND-NIMBUS Software framework
Challenges

#### And Others

- MCDM Cases
  - Continuous Casting of Steel, Timo Männikkö
  - Heuristic Solution Methods for Housing Location Models, Michael P. Johnson
  - Multiobjective Optimization of an Ultrasonic Transducer using NIMBUS, Paavo Nieminen et al.
  - Optimization of Internal Combustion Engine, Timo Aittokoski
  - Simulated moving bed processes, Jussi Hakanen
  - Engine control system optimization, Markus Inkeroinen et al.
  - Wastewater Treatment, Kristian Sahlstedt
  - Multi-criteria model for intensity modeled radiotherapy planning, Henri Ruotsalainen
    - And more
- Coworkers
   Tommi Ronkainen, Tommi Myöhänen, Jari Huikari, Ville Tirronen, Timo Tarvainen, Antoine Le Hyaric,
   Juha-Pekka Koskinen, Aki Järvinen, Kirsi Holopainen, Heikki Maaranen, Tero Oravasaari, and many more

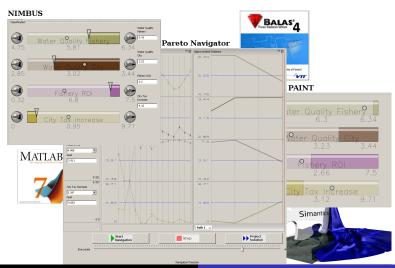
# Major obstacles encountered

- Research project buy-in
  - Management engineer
- Black-box optimization
  - Are current methods sufficiently robust?
  - Wider set of methods
  - Testing
- Computationally demanding problems
  - Parallel computing
  - Methods based on Pareto Frontier approximation
- Problem formulation
  - Different formulation for different methods
  - Not difficult, but time consuming
- Knowledge domain differences

# **Proposed Solution**

- Software framework for multiobjective optimization: IND-NIMBUS
- Different multiobjective methods
  - Interactive methods: NIMBUS, Nautilus
  - Approximation Methods: Pareto Navigator, PAINT
  - EMO Methods: Hybrid NSGAII, UPS-EMO
- General interface for problem formulation
  - Programming languages
  - Simulator: Matlab, Balas
  - Problem Modelling: GAMS, CPLEX
  - Simulator Platforms: Simantics, CapeOpen

#### **IND-NIMBUS Software framework**



# **Future Challenges**

- Closed source NIMBUS development
  - Set restrictions on usable tool set
  - Framework future?
- Black-box problem pre-analysis
  - DM Agent with an utility function?
- Platform implementation
  - Resources?
  - Motivation?

Research Projects
Obstacles encountered
IND-NIMBUS Software framework
Challenges

# Thank You!

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http://www.mit.jyu.fi/optgroup/
http://ind-nimbus.it.jyu.fi/
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## References

Miettinen, K., Mäkelä, M.M., Synchronous Approach in Interactive Multiobjective Optimization, European Journal of Operational Research. 170(3), 909-922. 2006.

Eskelinen, P., Miettinen, K., Klamroth, K., Hakanen, J., *Pareto Navigator for Interactive Nonlinear Multiobjective Optimization*, OR Spectrum, 23, 211-227, 2010.

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Hartikainen, M., Miettinen, K., Wiecek, M.M., Constructing a Pareto Front Approximation for Decision Making, Mathematical Methods of Operations Reserach, 73(2), 209-234, 2011.