

A Direct Multisearch Inexact Restoration Filter Method for Biobjective Optimization

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Outline

- ① Introduction
- ② DMS-FILTER-IR
- ③ Convergence Analysis
- ④ Numerical Results
- ⑤ Conclusions and Future Work

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Multiobjective Constrained Derivative-free Optimization

$$\min_{x \in \Omega \subset \mathbb{R}^n} F(x) = (f_1(x), f_2(x), \dots, f_m(x))^{\top}$$
$$f_j : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}, \quad j = 1, 2, \dots, m \geq 2$$

with $\Omega = X \cap \{x \in \mathbb{R}^n \mid C(x) \leq 0\}$, where X is a full dimensional polyhedron and $C : \mathbb{R}^n \rightarrow (\mathbb{R} \cup \{+\infty\})^p$

- several conflicting objectives
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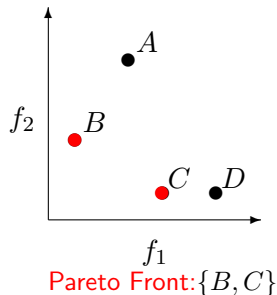
Direct MultiSearch (DMS) Main Lines

- does **not aggregate** any of the objective function components
- makes use of **Pareto dominance**

Pareto Dominance (x dominates y)

$$F(x) \leq F(y), \text{ with } F(x) \neq F(y)$$

- generalizes directional direct-search to MOO
- considers the **search/poll** paradigm with an optional search step
- computes approximations to the complete Pareto front



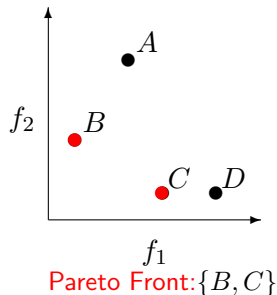
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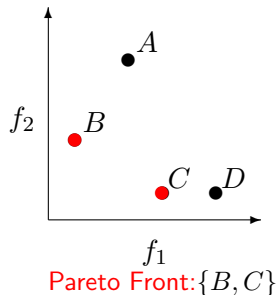
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Direct MultiSearch (DMS) Main Lines

- constraints are addressed by an **extreme barrier approach**

$$F_{\Omega}(x) = \begin{cases} F(x) & \text{if } x \in \Omega, \\ (+\infty, +\infty, \dots, +\infty)^{\top} & \text{otherwise} \end{cases}$$

- keeps a **list of feasible nondominated points**
- **poll centers** are chosen **from the list**
- **successful iterations** correspond to **list changes**

successful iteration \Leftrightarrow new feasible nondominated point

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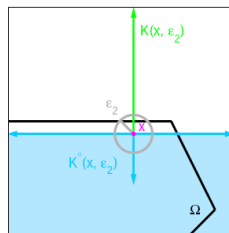
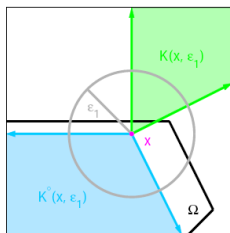
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Poll Directions and Constraints

- Unconstrained or variable bounds: coordinate search

$$D = D_{\oplus} = [I \ -I]$$

- Linear constraints: directions must conform to the geometry of nearby constraints



Abramson, Brezhneva, Dennis, and Pingel [2008]

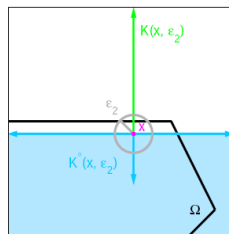
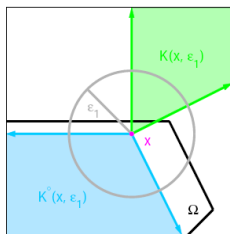
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- DMS → A. L. Custódio, J. F. A. Madeira, A. I. F. Vaz, and L. N. Vicente, *Direct multisearch for multiobjective optimization*, SIAM J. Optim. (2011), 21, 1109–1140
- DFMO → G. Liuzzi, S. Lucidi, and F. Rinaldi, *A derivative-free approach to constrained multiobjective nonsmooth optimization*, SIAM J. Optim. (2016), 26, 2744–2774
- 93 biobjective problems with nonlinear constraints and variable bounds
 - number of variables between 3 and 30
 - number of nonlinear constraints between 1 and 29
 - maximum of 20000 function evaluations

Metrics for Performance Profiles (Dolan and Moré [2002])

- Purity

$$\frac{|F_{p,s} \cap F_p|}{|F_{p,s}|}$$

- Spreads Γ and Δ

$$\Gamma_{p,s} = \max_{j \in \{1, \dots, m\}} \left(\max_{i \in \{0, \dots, N\}} \{d_i\} \right)$$

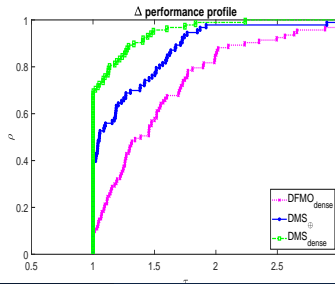
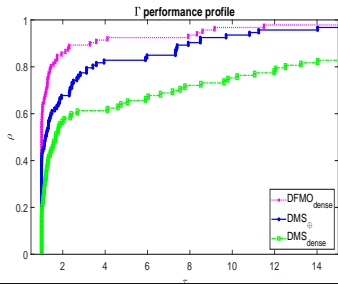
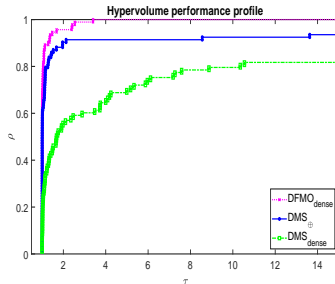
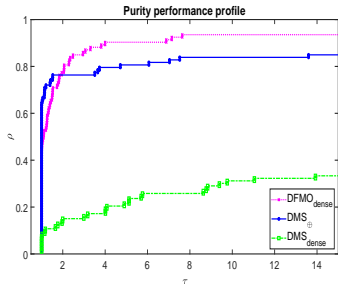
$$\Delta = \max_{j \in \{1, \dots, m\}} \left(\frac{d_0 + d_N + \sum_{i=1}^{N-1} |d_i - \bar{d}|}{d_0 + d_N + (N-1)\bar{d}} \right)$$

- Hypervolume

$$HI_{p,s} = Vol\{b \in \mathbb{R}^m \mid b \leq U_p \wedge \exists a \in F_{p,s} : a \leq b\}$$

Nonlinear + Bound Constraints (Biobjective Problems)

DFMO – Liuzzi, Lucidi, and Rinaldi (2016)



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Problem Reformulation – Filter Approach

$$\min_{x \in X} (F(x); h(x)) = (f_1(x), f_2(x), \dots, f_m(x), h(x))^{\top}$$

where X is a full dimensional polyhedron and

$$h(x) = \|C(x)_+\|_2^2 = \sum_{i=1}^p \max\{0, c_i(x)\}^2$$

Constraints in X continue to be addressed by an extreme barrier approach.

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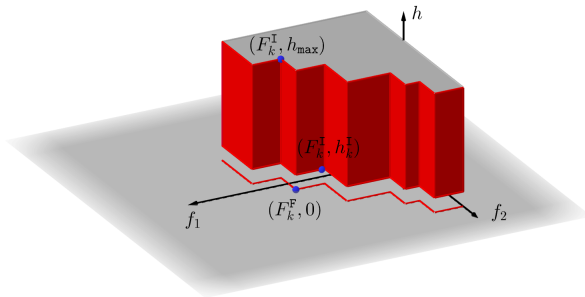
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New Dominance Relation

A point x' is filtered by \mathcal{F} if:

- $h(x') > h_{\max}$ (for some fixed $h_{\max} > 0$)
- there is $x \in \mathcal{F}$ such that $x' \succeq_{(F;h)} x$
- $x' \notin \Omega$ and there is $x \in \mathcal{F}$ such that

$$h(x') \geq h(x) \text{ and } x' \succeq_F x$$



DMS Filter and Inexact Restoration Approach

- **Nonlinear feasibility** is treated as an **additional objective**
- **Priority** given to **feasible poll centers**
- When all poll points associated with a poll center x_k are infeasible, **switches to an infeasible poll center**

Attempts to **restore feasibility** by solving:

$$\begin{aligned} \min_{y \in X} \quad & \frac{1}{2} \|y - x_k\|^2 \\ \text{s.t.} \quad & h(y) \leq \xi(\alpha_k) h(x_k), \end{aligned}$$

where $\xi : (0, +\infty) \rightarrow (0, 1)$, is continuous, nondecreasing, and satisfies

$$\xi(t) \rightarrow 0 \text{ when } t \downarrow 0$$

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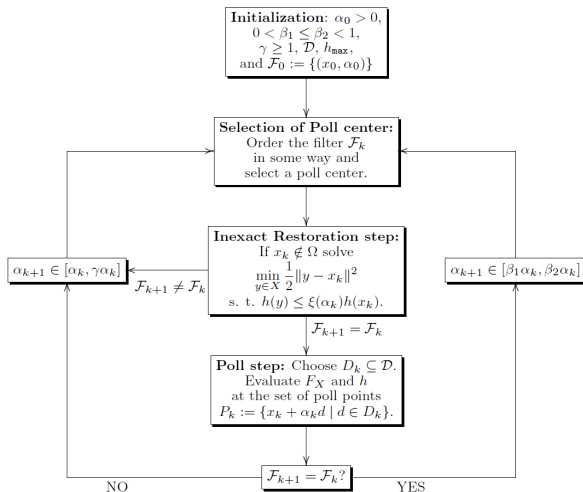
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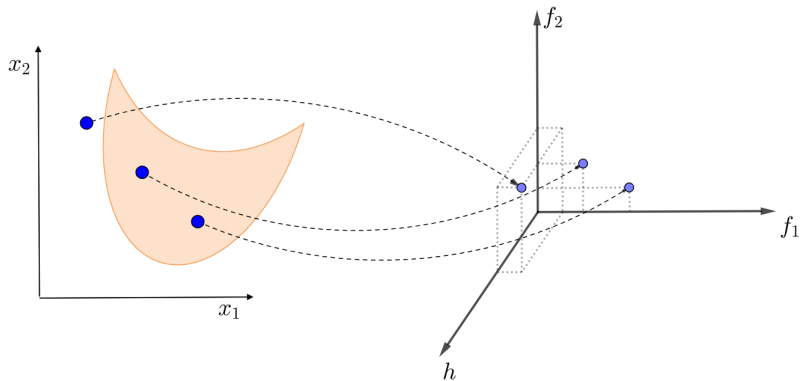
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DMS-FILTER-IR – Algorithmic Structure

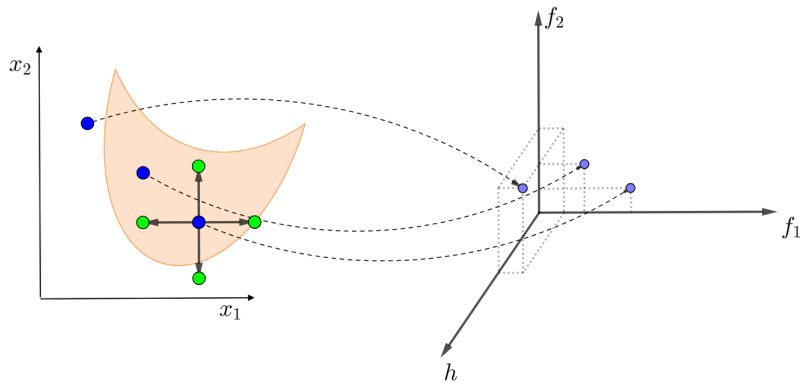


Solutions: $L := \{(x, \alpha) \in \mathcal{F} \mid (F(x); h(x)) = (F(x); 0)\}$

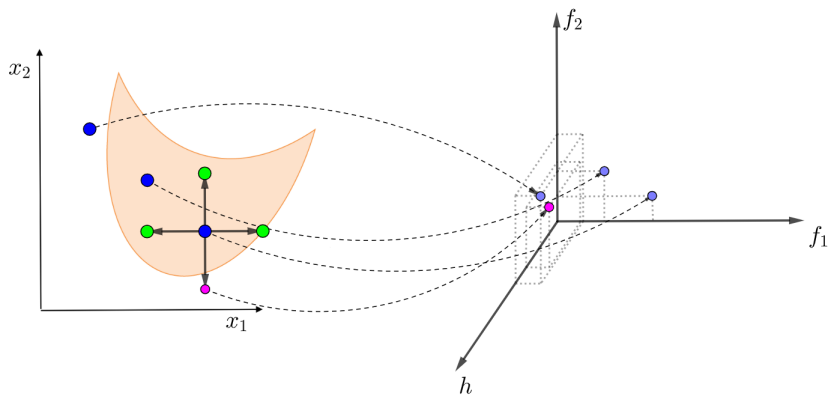
DMS-FILTER-IR – Poll Step



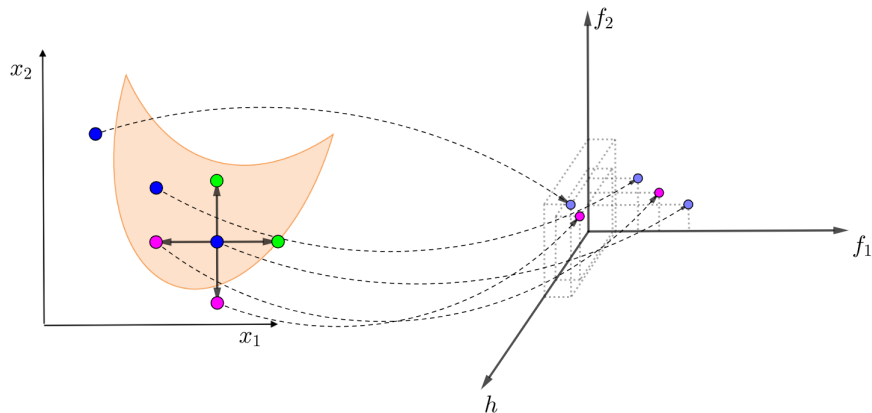
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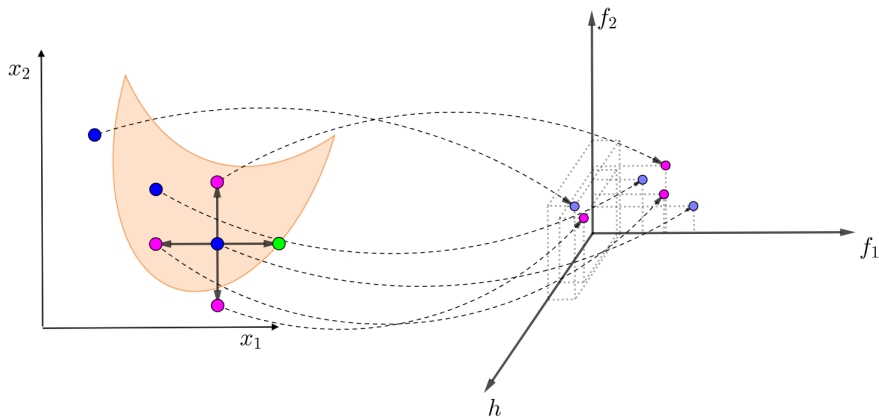
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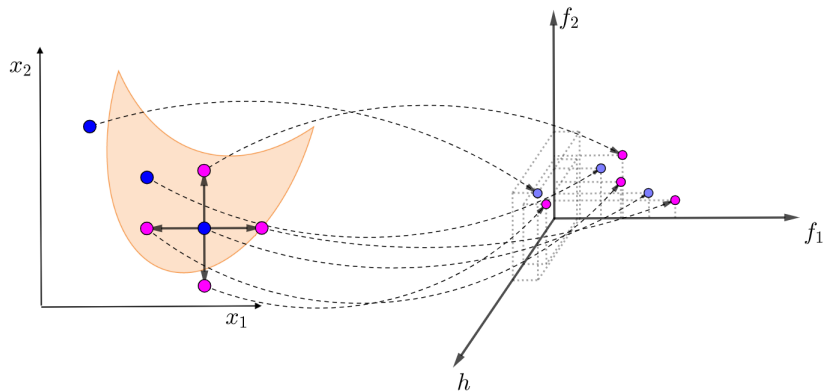
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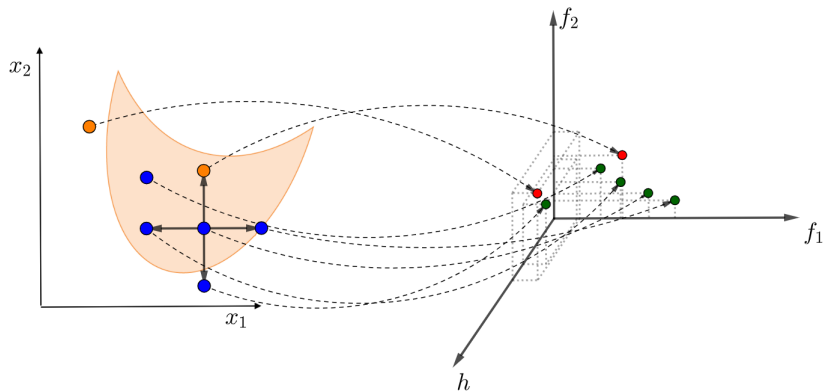
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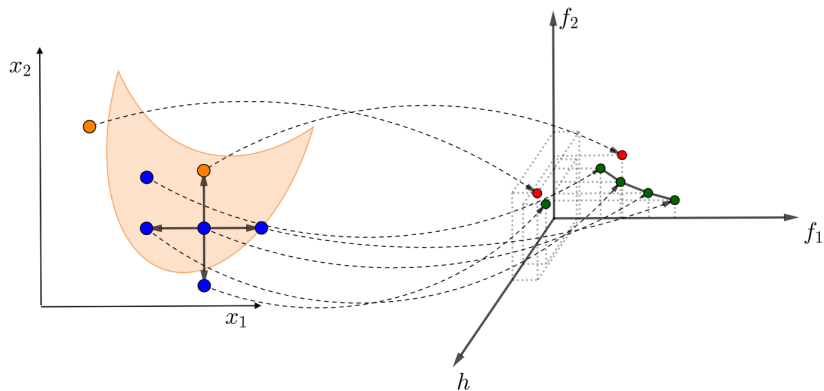
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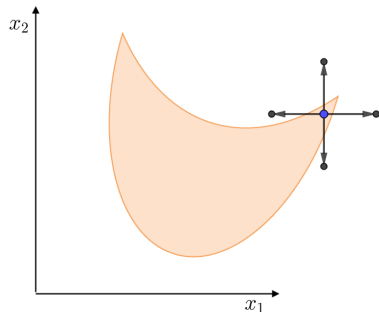
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Poll Center Selection

Feasible to Infeasible

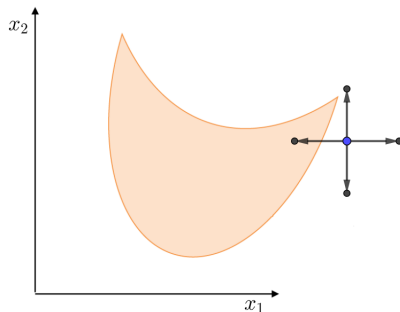
Polling only generates infeasible points



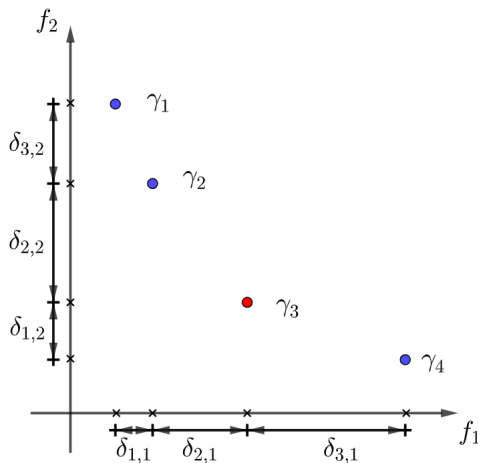
Infeasible to Feasible

Infeasible x_k generates feasible point by inexact restoration or polling

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Feasible Poll Center – Most Isolated Point



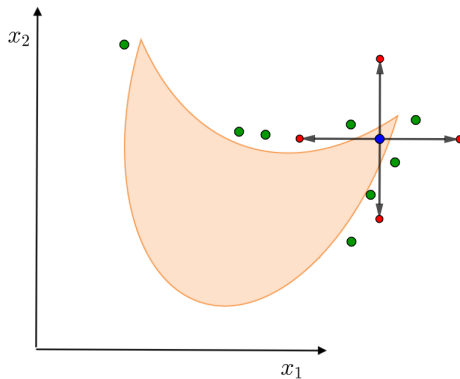
$$\gamma_1 = \frac{\delta_{3,2} + \delta_{1,1}}{2}$$

$$\gamma_2 = \frac{\frac{\delta_{3,2} + \delta_{2,2}}{2} + \frac{\delta_{1,1} + \delta_{2,1}}{2}}{2}$$

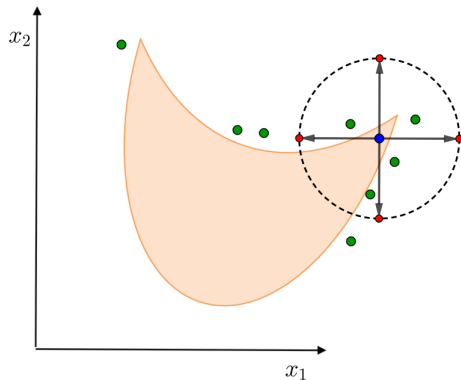
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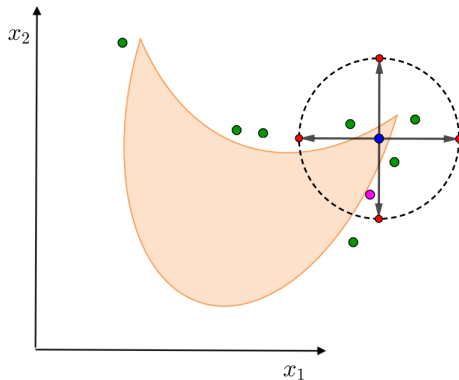
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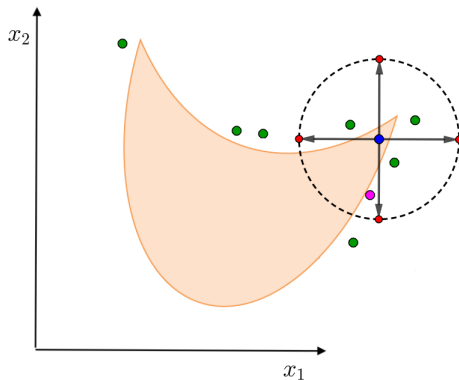
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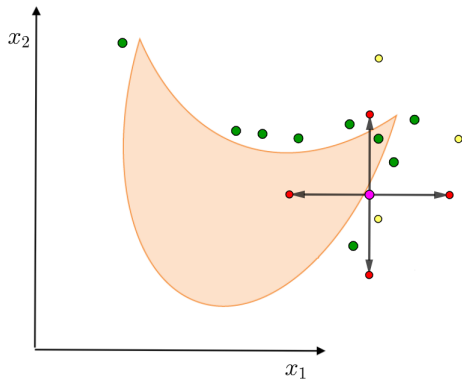


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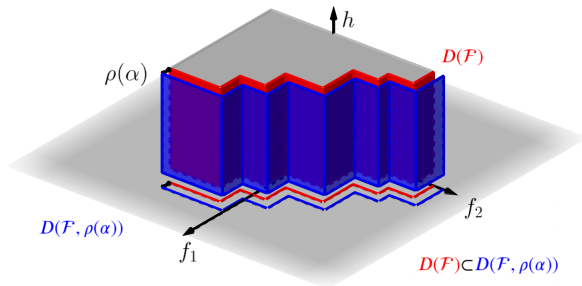
Globalization Strategies

Using Integer Lattices (Torczon [1997], Audet and Dennis [2002])

- requires only simple decrease
- poll directions and step size must satisfy integer/rational requirements

Imposing Sufficient Decrease (Kolda, Lewis, and Torczon [2003])

- use of a forcing function
 $\rho : (0, +\infty) \rightarrow (0, +\infty)$, continuous, nondecreasing, and satisfying $\rho(t)/t \rightarrow 0$ when $t \downarrow 0$
- x is nondominated $\Leftrightarrow (F_X(x), h(x)) \notin D(\mathcal{F}, \rho(\alpha))$



Convergence Results – Sequences

Linked sequence

A sequence $\{(x_k, \alpha_k)\}_{k \in K}$ such that (x_k, α_k) is generated from (x_{k-1}, α_{k-1}) at a poll or at an inexact restoration step.

Refining sequence

A sequence $\{(x_k, \alpha_k)\}_{k \in K}$, such that $k \in K$ is an unsuccessful iteration and $\lim_{k \in K} \alpha_k = 0$.

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Theorem

From any **linked sequence** it is possible to extract a **convergent refining subsequence** (under any of the two globalization strategies).

Let \bar{x} be the limit point of a convergent refining subsequence $\{x_k\}_{k \in K}$.

Definition (Refining Directions)

Refining directions for \bar{x} are **limit points of $\{d_k / \|d_k\|\}_{k \in K}$** , where $d_k \in D_k$ and $x_k + \alpha_k d_k \in \mathcal{S} := \{x \in X \mid h(x) \leq h_{\max}\}$.

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Consider a linked sequence $\{(x_k, \alpha_k)\}_{k \in K}$ generated by DMS-FILTER-IR. Let $\{(x_k^F, \alpha_k^F)\}_{k \in K' \subseteq K}$ be a **feasible refining subsequence converging to $\bar{x} \in \Omega$** . Assume that F and h are Lipschitz continuous near \bar{x} .

Theorem

- If $d \in \text{int}(T_\Omega^{Cl}(\bar{x}))$ is a refining direction for \bar{x} then:

$$\exists j = j(d) \in \{1, \dots, m\} \text{ such that } f_j^\circ(\bar{x}; d) \geq 0$$

- If the set of refining directions for \bar{x} is dense in $\text{int}(T_\Omega^{Cl}(\bar{x})) \neq \emptyset$ then \bar{x} is a Pareto-Clarke critical point of F (in Ω):

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- If the **set of refining directions for \bar{x} is dense in $\text{int}(T_{\mathcal{S}}^{Cl}(\bar{x})) \neq \emptyset$** then \bar{x} is a **Pareto-Clarke critical point of F in $\mathcal{S} := \{x \in X \mid h(x) \leq h_{\max}\}$** :

$$\forall d \in T_{\mathcal{S}}^{Cl}(\bar{x}), \exists j = j(d) \in \{1, \dots, m\} \text{ such that } f_j^\circ(\bar{x}; d) \geq 0$$

Outline

- ① Introduction
- ② DMS-FILTER-IR
- ③ Convergence Analysis
- ④ Numerical Results**
- ⑤ Conclusions and Future Work

Numerical Settings

- Comparison among **DFMO**, **DMS-EB**, **DmultiMads-PB*** and **DMS-FILTER-IR**
- 93 biobjective problems with nonlinear constraints and bounds
 - number of variables between 3 and 30
 - number of nonlinear constraints between 1 and 29
- Initialization:
 - **DMS-EB**: Feasible point provided by Kar Mitsa [2007]
 - **Others**: n -points equally spaced in the line segment, joining the variable upper and lower bounds
- Stopping criterion
 - **DMS-EB** and **DMS-FILTER-IR**:
 - $\alpha_k < 10^{-3}$ for all points in the filter
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* **DmultiMads-PB** → Jean Bignon, Sébastien Le Digabel, Ludovic Salomon. *Handling of constraints in multiobjective blackbox optimization*. ArXiv:2204.00904

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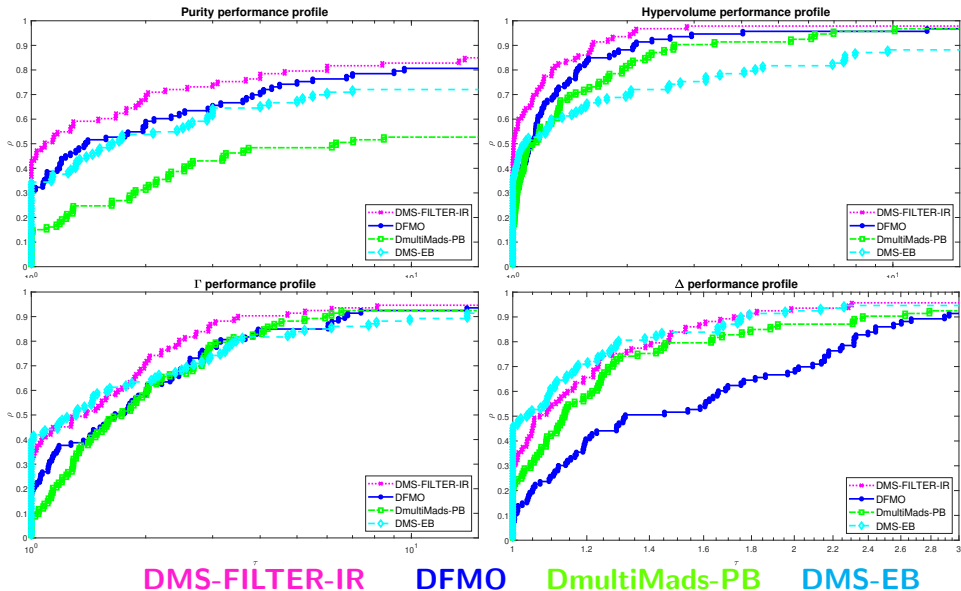
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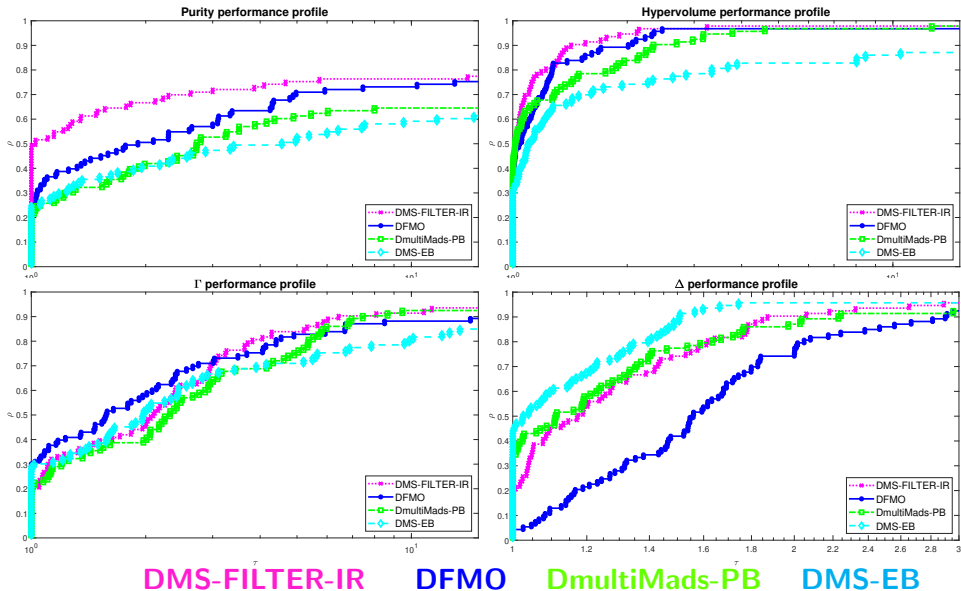
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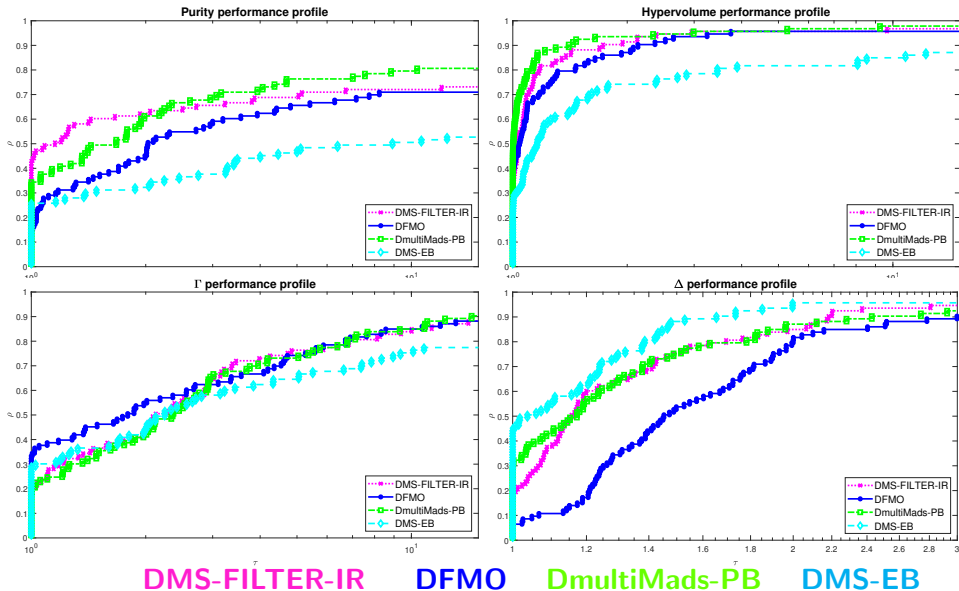
Results - 500 function evaluations



Results - 5000 function evaluations



Results - 20000 function evaluations



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- DMS-FILTER-IR presents a well-supported convergence analysis
- DMS-FILTER-IR presents competitive numerical results for constrained biobjective derivative-free optimization problems

Future Work

- developing a competitive numerical implementation for problems with more than two objectives

Thank you for your attention!

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