

A Direct Multisearch Filter Method for Biobjective Optimization

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Outline

- ① Introduction
- ② Direct Multisearch Filter (DMS-Filter)
- ③ Convergence Results
- ④ Computational Results
- ⑤ Conclusions and Future Work

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Multiobjective Optimization

$$\min_{x \in \Omega \subseteq \mathbb{R}^n} F(x) = (f_1(x), f_2(x), \dots, f_p(x))^T$$
$$f_j : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}, \quad j = 1, 2, \dots, p \geq 2$$

- $\Omega = X \cap \{x \in \mathbb{R}^n \mid C(x) \leq 0\}$ where X is a full dimensional polyhedron and $C : \mathbb{R}^n \rightarrow (\mathbb{R} \cup \{+\infty\})^m$
- objectives often conflicting
- expensive function evaluation
- impossible to use or approximate derivatives

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Motivation

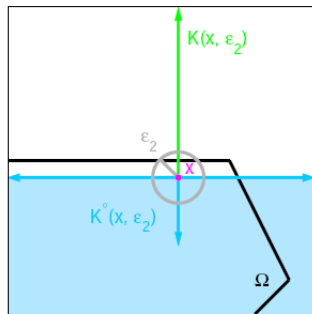
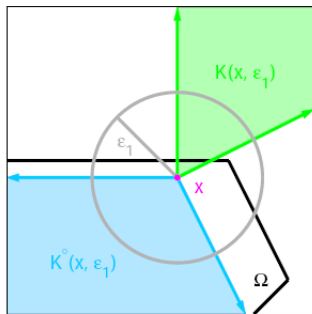
- DMS \rightarrow A. L. Custódio, J. F. A. Madeira, A. I. F. Vaz, and L. N. Vicente. *Direct multisearch for multiobjective optimization*, SIAM J. Optim. (2011), 21, 1109-1140
 - $\text{DMS}_{\text{dense}}$ \rightarrow Asymptotically dense in the unit sphere
 - DMS_{\oplus} \rightarrow Coordinate directions

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DMS - General Linear Constraints

- Set of poll directions **conforms to the geometry** of nearby constraints
- Proposed for single objective optimization in Abramson, Brezhneva, Dennis, and Pingel [2008].



(in Kolda, Lewis, and Torczon [2003])

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

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

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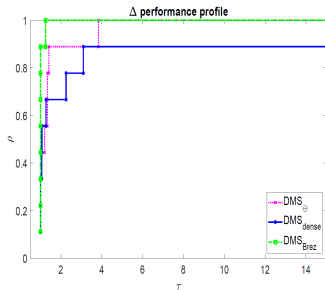
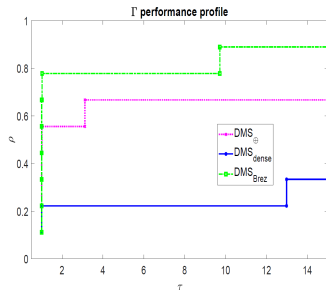
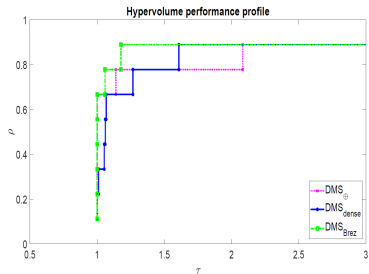
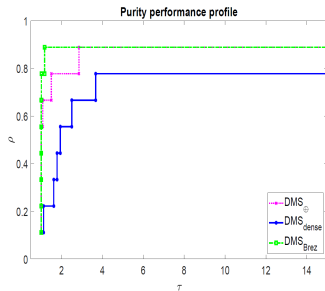
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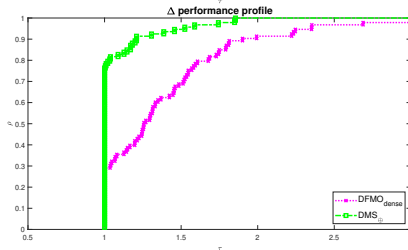
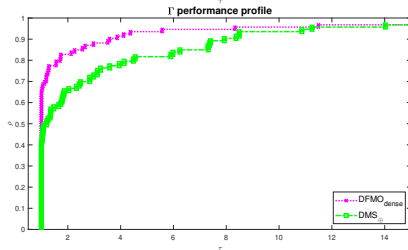
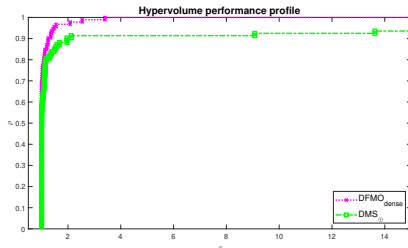
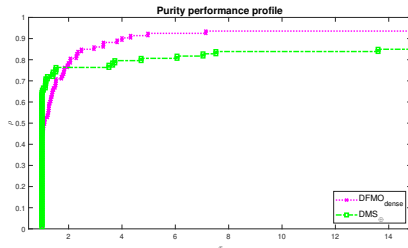
DMS - General Linear Constraints



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DMS - Nonlinear + Bound Constraints



-DFMO

-DMS

- Extreme Barrier Function:

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

- Constraint Violation function:

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

$$\min_{x \in X} (f_1(x), f_2(x), \dots, f_p(x), h(x))^\top$$

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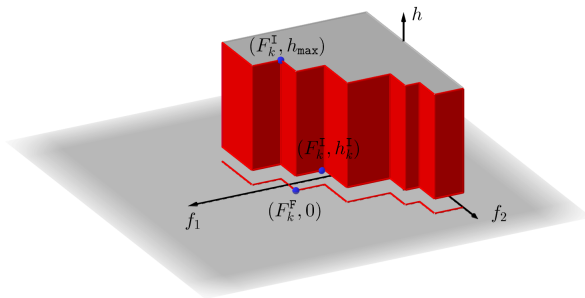
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Filter Approach

The filter \mathcal{F} is a set of nondominated points

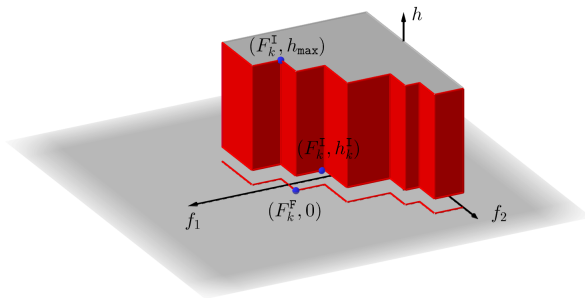


A point x' is said to be filtered by a filter \mathcal{F} if any of the following properties hold:

- There exists a point $x \in \mathcal{F}$ such that $x' \geq x$;
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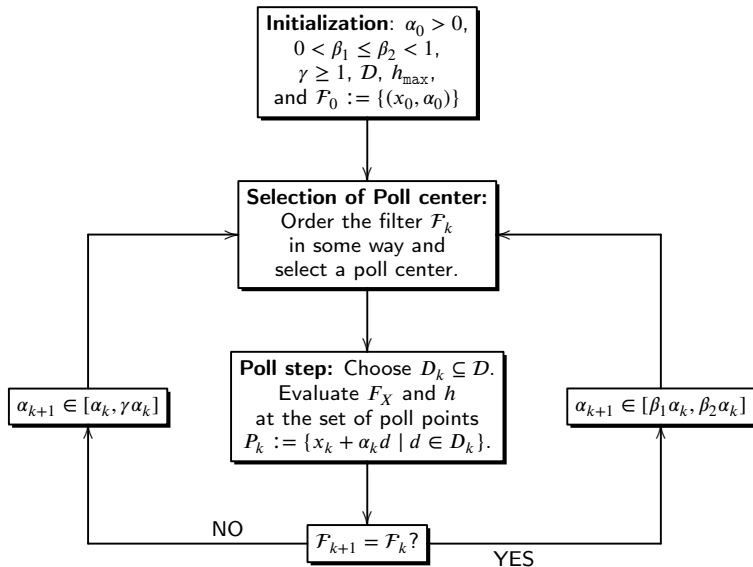
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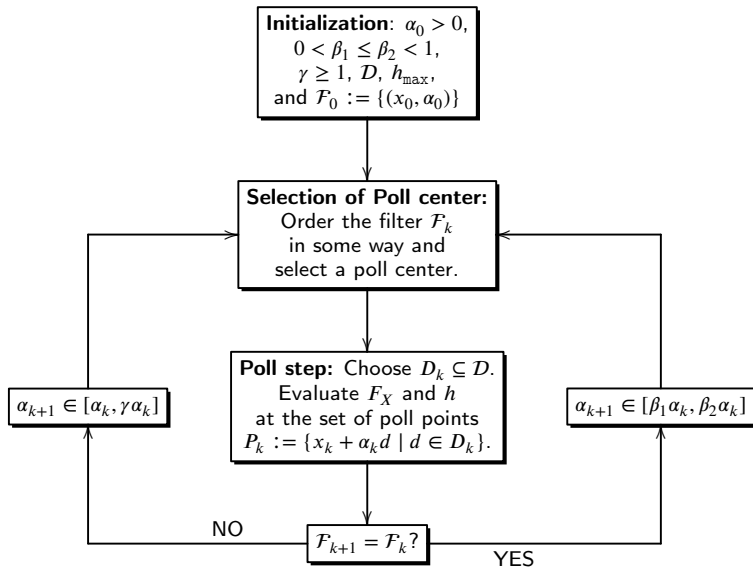
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DMS-Filter - Algorithmic Structure



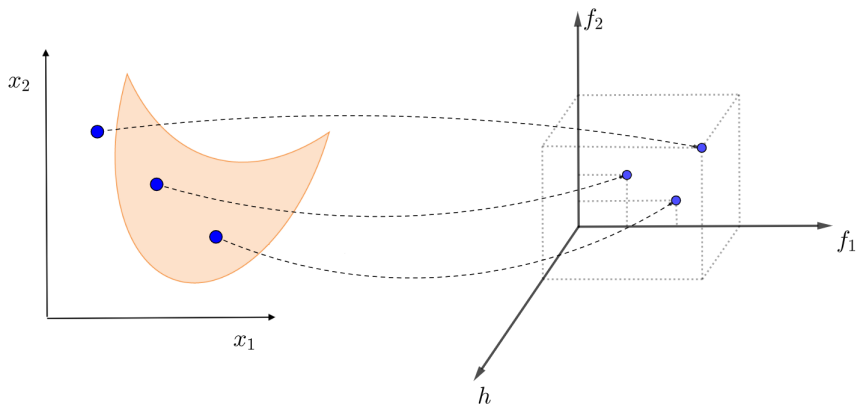
Solutions: $L := \{(x, \alpha) \in \mathcal{F} \mid (F_X(x), h(x)) = (F(x), 0)\}$.

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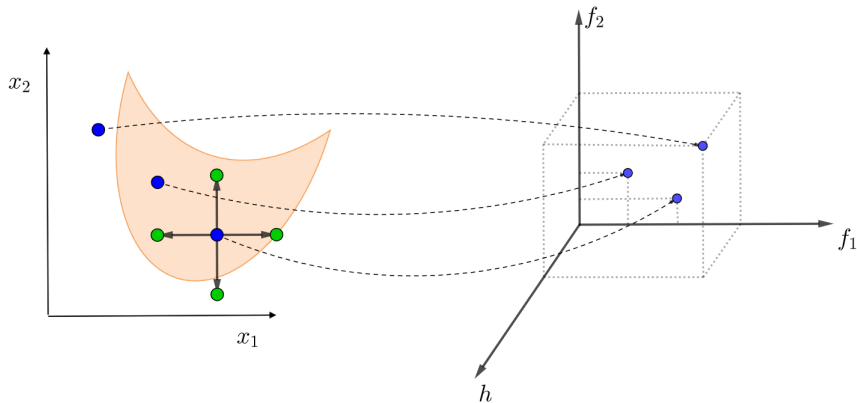


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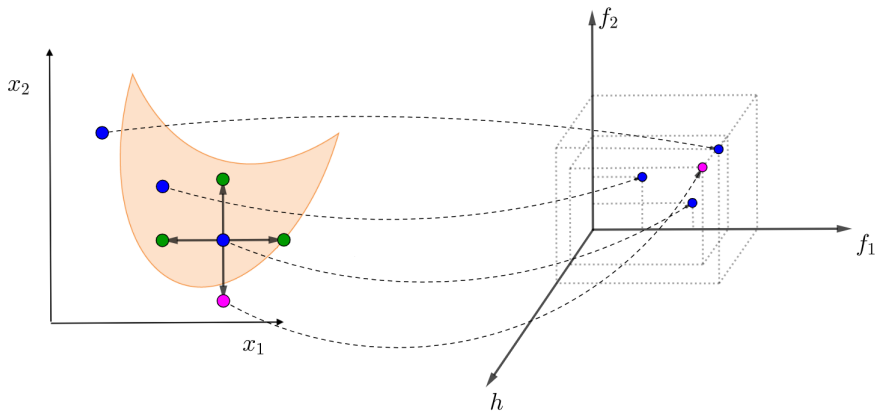
DMS-Filter – Poll Step



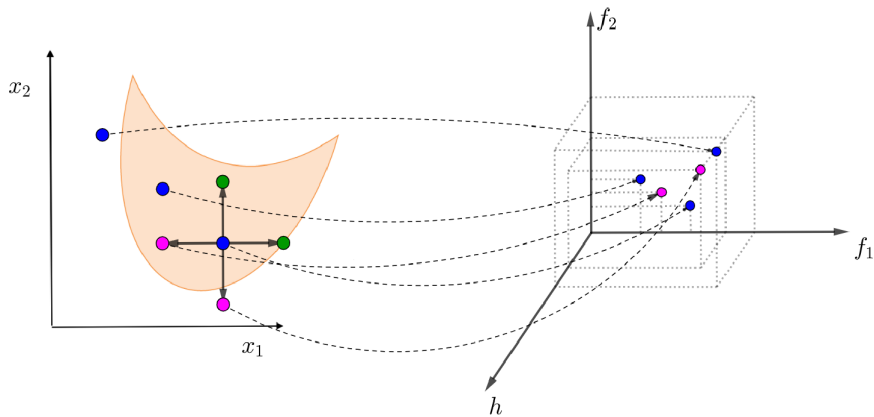
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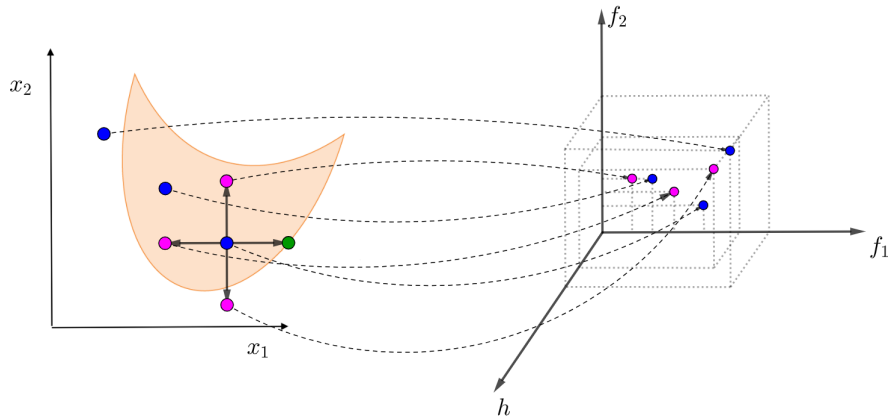
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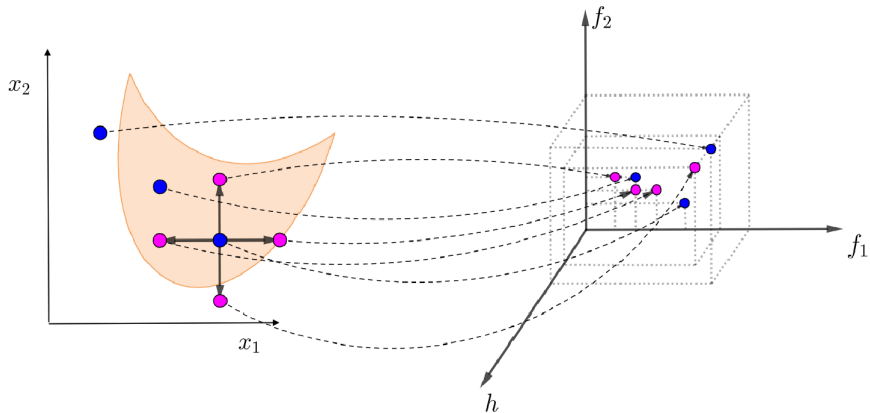
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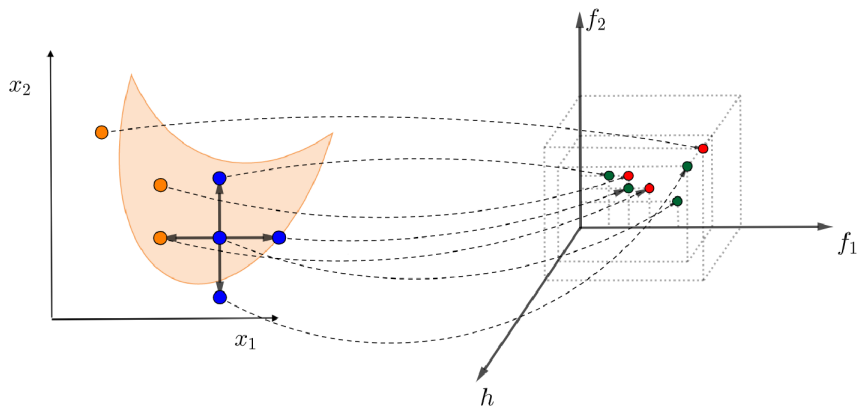
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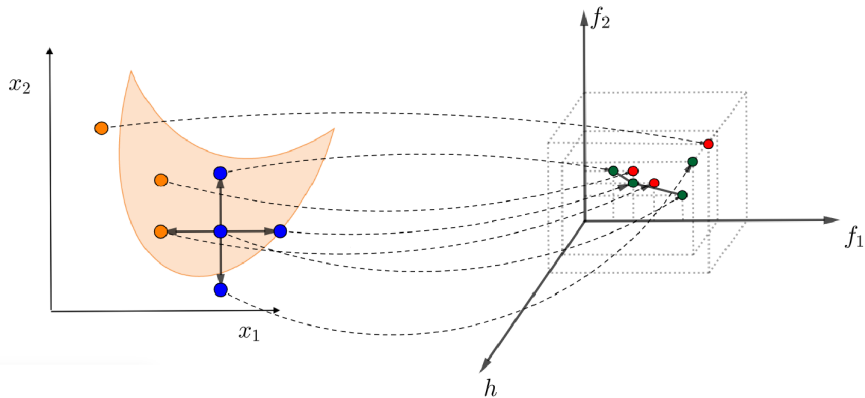
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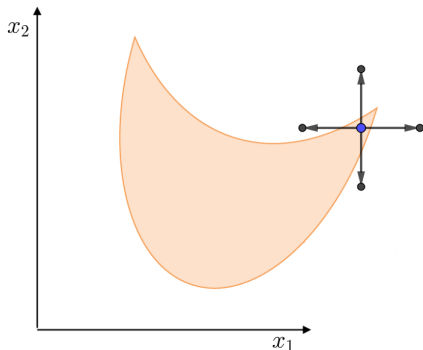


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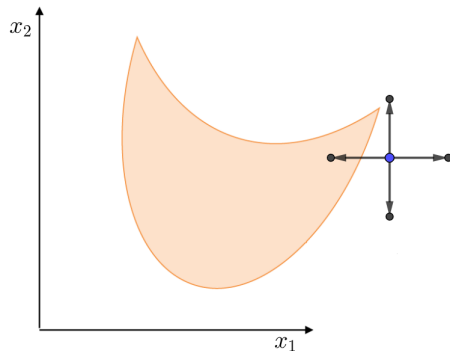


Poll Center Selection

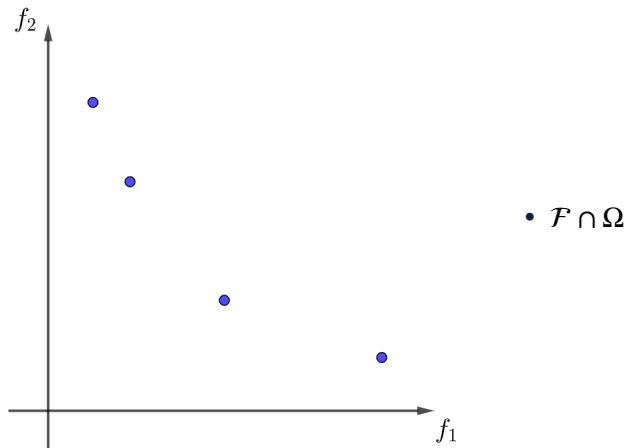
- Feasible to Infeasible



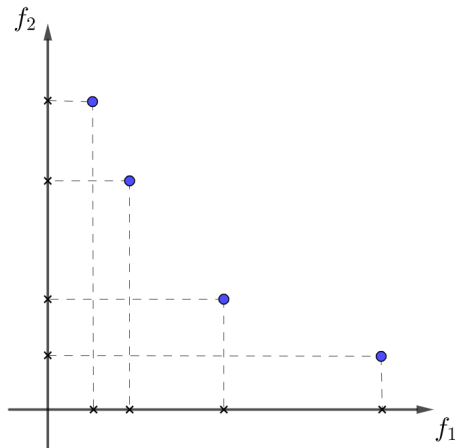
- Infeasible to Feasible



Feasible poll center - Most Isolated Point

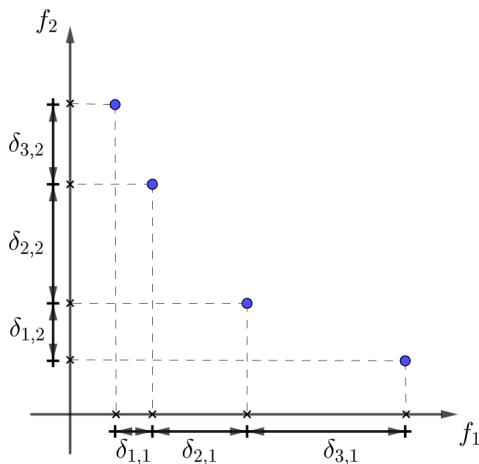


Feasible poll center - Most Isolated Point



• $\mathcal{F} \cap \Omega$

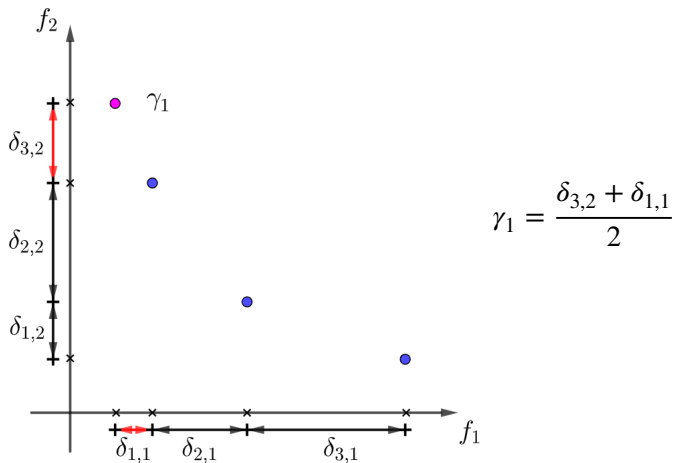
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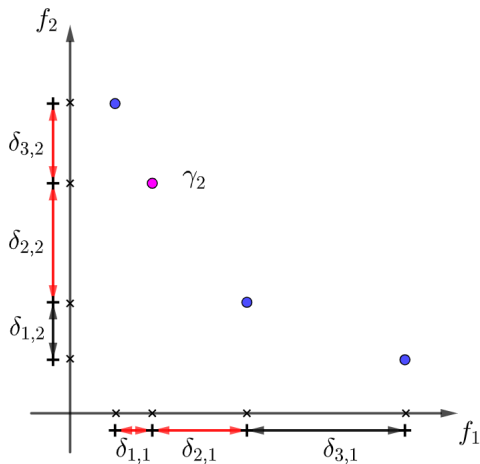
$$\delta_{i,j} = f_{i+1,j} - f_{i,j}$$

for $i = 1, 2, 3$ and $j = 1, 2$.

Feasible poll center - Most Isolated Point

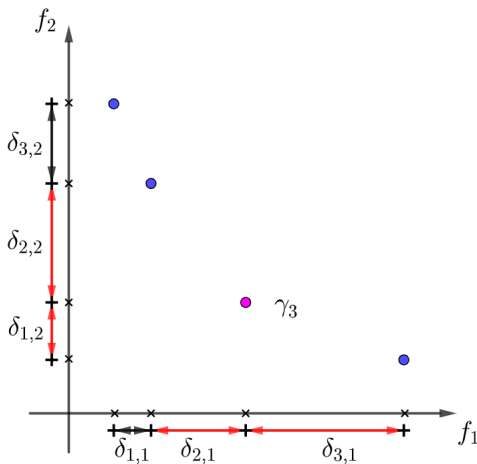


Feasible poll center - Most Isolated Point



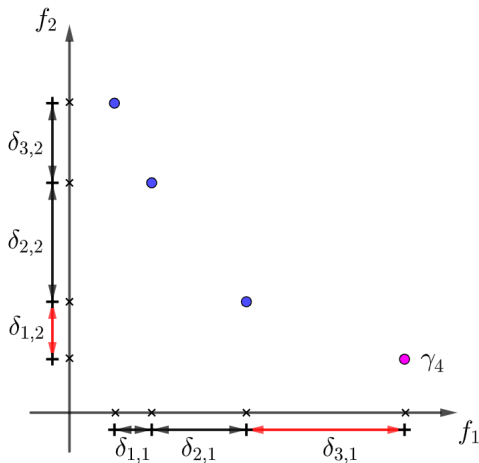
$$\gamma_2 = \frac{\frac{\delta_{3,2} + \delta_{2,2}}{2} + \frac{\delta_{1,1} + \delta_{2,1}}{2}}{2}$$

Feasible poll center - Most Isolated Point



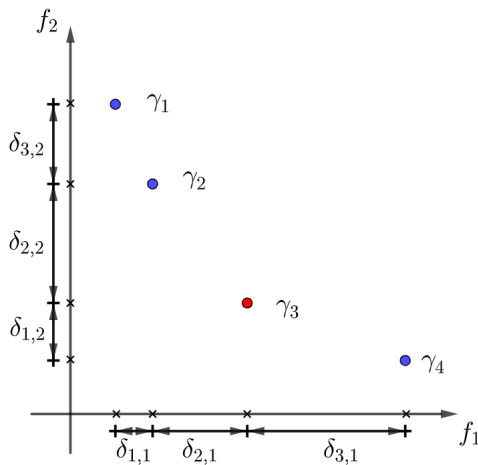
$$\gamma_3 = \frac{\frac{\delta_{2,2} + \delta_{1,2}}{2} + \frac{\delta_{2,1} + \delta_{3,1}}{2}}{2}$$

Feasible poll center - Most Isolated Point



$$\gamma_4 = \frac{\delta_{1,2} + \delta_{3,1}}{2}$$

Feasible poll center - Most Isolated Point



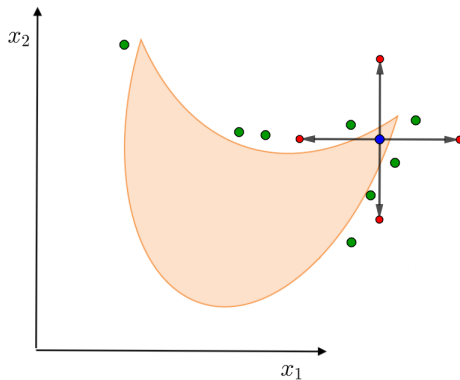
$$\gamma_1 = \frac{\delta_{3,2} + \delta_{1,1}}{2}$$

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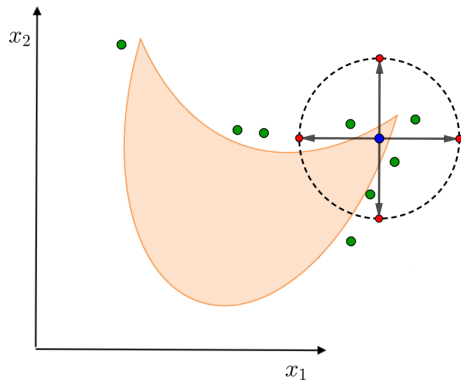
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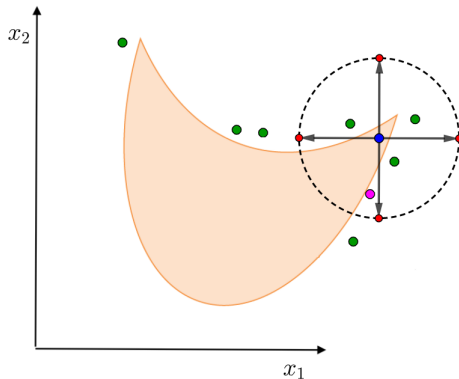
Infeasible poll center



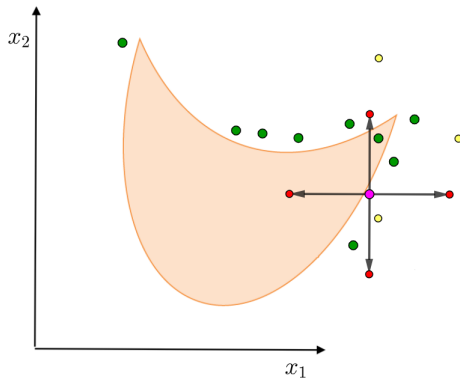
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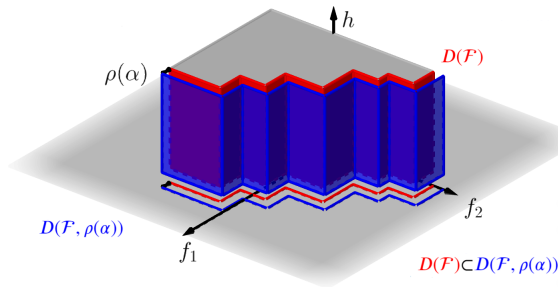
Globalization Strategies

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 and nondecreasing, satisfying $\rho(t)/t \rightarrow 0$ when $t \downarrow 0$
- x is nondominated $\Leftrightarrow (F_X(x), h(x)) \notin D(\mathcal{F}, \rho(\alpha))$



Theorem (Refining Subsequences)

There is at least a **convergent refining subsequence of iterates** $\{x_k\}_{k \in K}$, corresponding to unsuccessful poll steps, with $\lim_{k \in K} \alpha_k = 0$.

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 $x_k + \alpha_k d_k \in S := \{x \in X \mid h(x) \leq h_{\max}\}$

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Convergence Results

Assume that F and h are Lipschitz continuous near \bar{x}

Theorem

- $\{x_k^I\}_{k \in K}$ an infeasible refining subsequence converging to $\bar{x} \in \mathcal{S}$. If $d \in \text{int}(T_{\mathcal{S}}^{CI}(\bar{x}))$ is a refining direction for \bar{x} then:

$$h^\circ(\bar{x}; d) \geq 0$$

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Numerical Settings

- Comparison among DFMO, DMS and DMS-Filter
- 93 biobjective problems with nonlinear constraints and bounds
 - number of variables between 3 and 30
 - number of constraints between 1 and 29
- Initialization with a feasible point
 - Feasible point provided by Kar Mitsa [2007]
- Initialization in line
 - n -points equally spaced in the line segment, joining the variable upper and lower bounds
- Stopping criterion
 - **DMS and DMS-Filter:**
 - $\alpha_k < 10^{-3}$ for all points in the filter
 - **DFMO:**
 - default values
 - maximum of 20000 function evaluations

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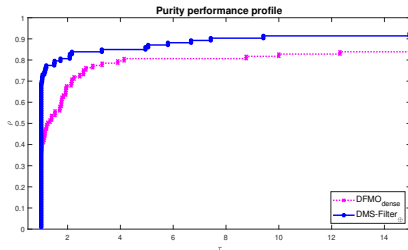
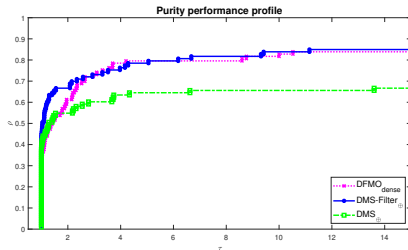
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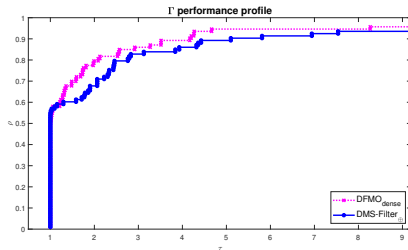
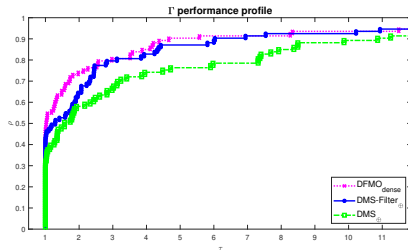
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Results - Purity



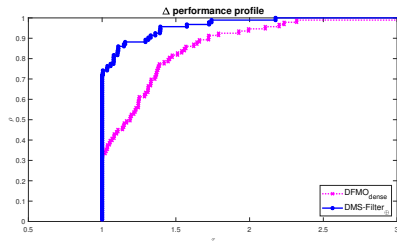
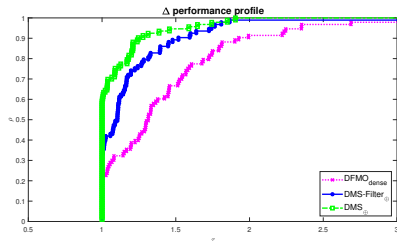
- DFMO
- DMS-Filter
- DMS

Results - Spread Gamma (Γ)



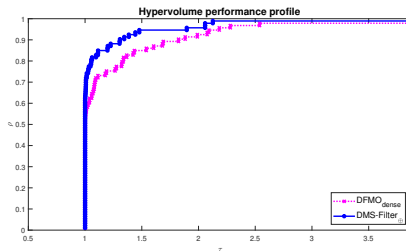
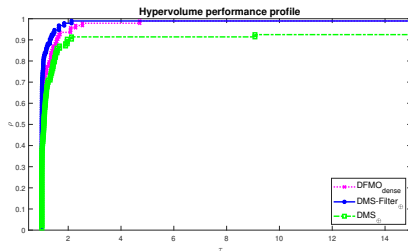
- DFMO
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- DMS

Results - Spread Delta (Δ)



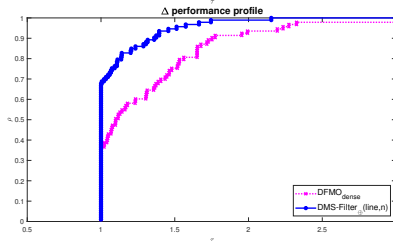
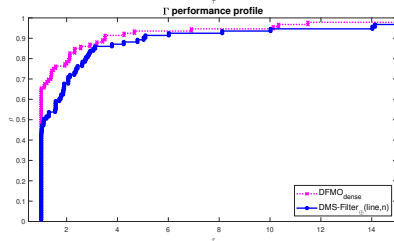
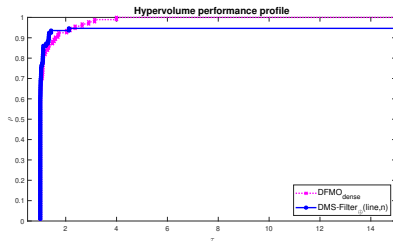
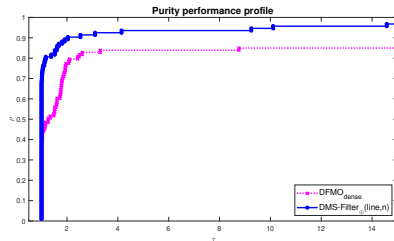
- DFMO
- DMS-Filter
- DMS

Results - Hypervolume



- DFMO
- DMS-Filter
- DMS

Results - DMS-Filter(line,n) VS DFMO



- DFMO
- DMS-Filter

Outline

- ① Introduction
- ② Direct Multisearch Filter (DMS-Filter)
- ③ Convergence Results
- ④ Computational Results
- ⑤ Conclusions and Future Work

Conclusions and Future Work

- DMS-Filter extends filter methods to constrained Multiobjective Derivative-free Optimization
- DMS-Filter presents a well-supported convergence analysis for both globalization strategies
- DMS-Filter presents competitive numerical results for constrained Biobjective Derivative-free Optimization Problems

- Future work comprises extending the approach to problems with more than two objectives

THANKS FOR YOUR ATTENTION!

Any comments or questions?

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