

# 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

$$\min_{x \in \Omega \subseteq \mathbb{R}^n} F(x) = (f_1(x), f_2(x), \dots, f_m(x))^T$$

$$f_j : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}, \quad j = 1, 2, \dots, m \geq 2$$

- $\Omega = X \cap \{x \in \mathbb{R}^n \mid C(x) \leq 0\}$  where  $X$  is a full dimensional polyhedron and  $C : \mathbb{R}^n \rightarrow (\mathbb{R} \cup \{+\infty\})^p$
- 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
  - $\text{DMS}_{dense}$   $\rightarrow$  Asymptotically dense in the unit sphere
  - $\text{DMS}_{\oplus}$   $\rightarrow$  Coordinate directions

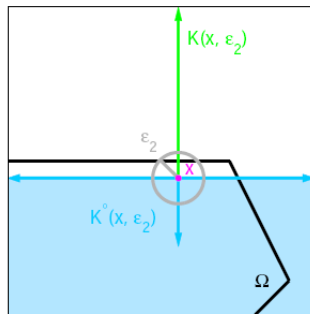
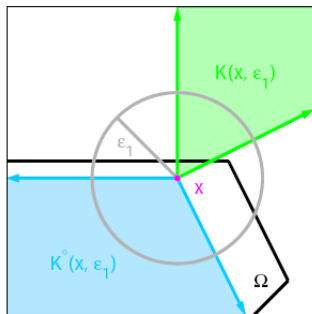
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# DMS - General Linear Constraints

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



(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  $\Gamma$  and  $\Delta$

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

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

- 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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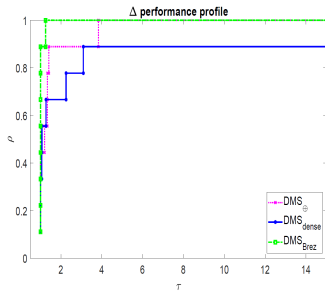
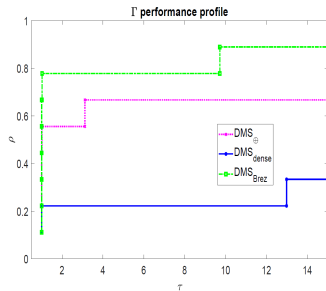
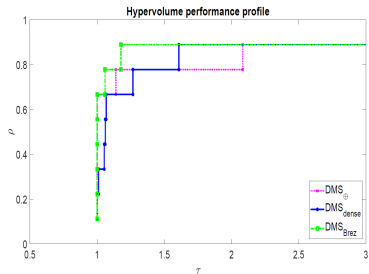
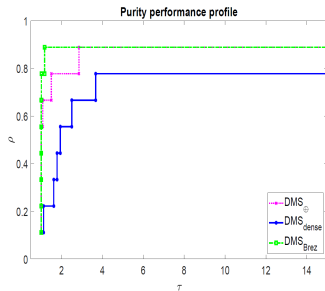
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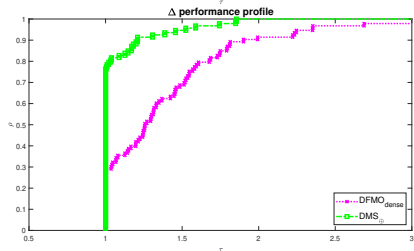
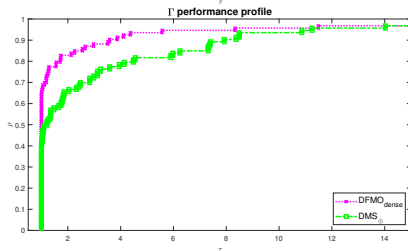
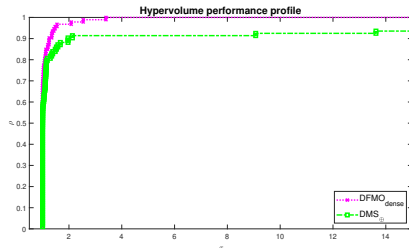
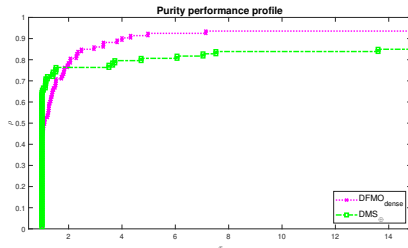
# DMS - General Linear Constraints



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- 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 bounds
  - number of variables between 3 and 30
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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:

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

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

# New Problem

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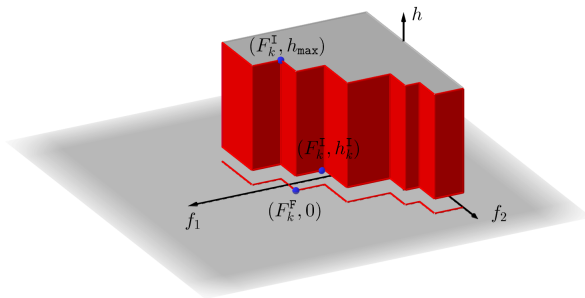
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# Filter Approach

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

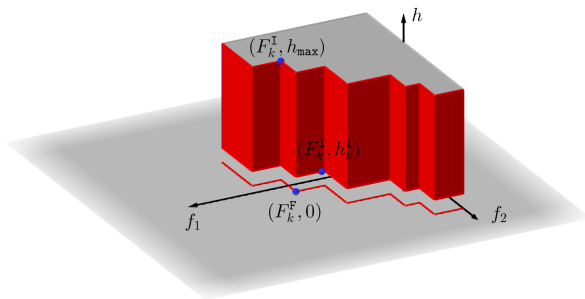


A point  $x'$  is said to be filtered by a filter  $\mathcal{F}$  if any of the following properties hold:

- There exists a point  $x \in \mathcal{F}$  such that  $x' \geq x$
- $h(x') > h_{\max}$  for some positive finite upper bound  $h_{\max}$

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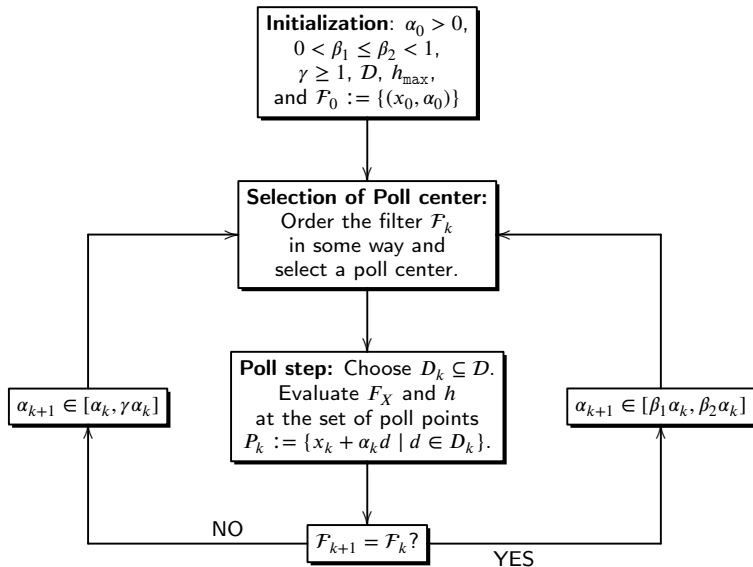
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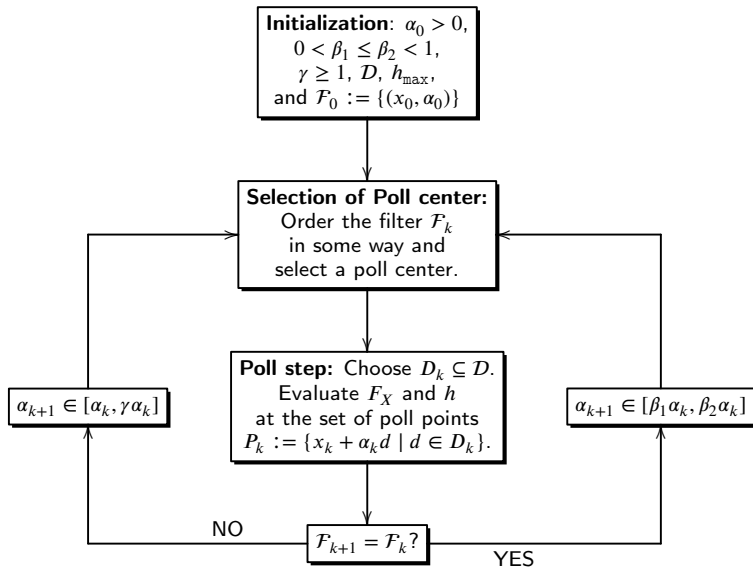
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# DMS-Filter - Algorithmic Structure



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

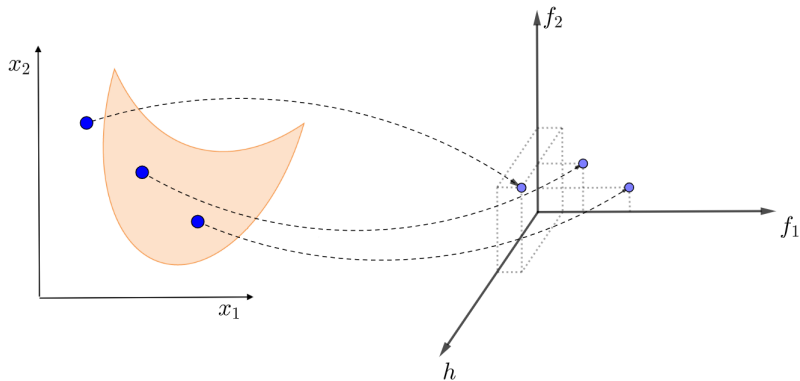
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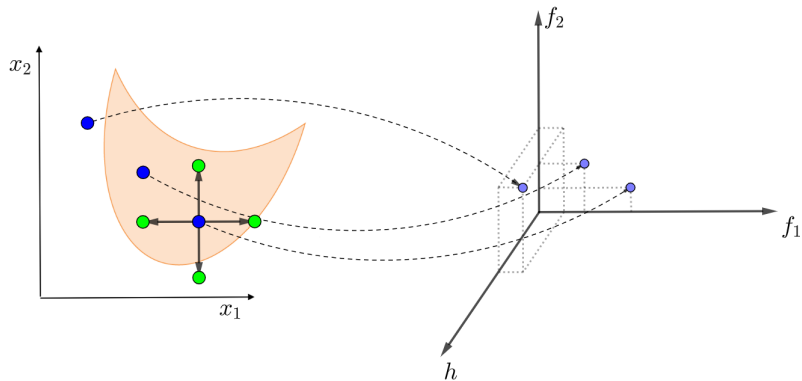
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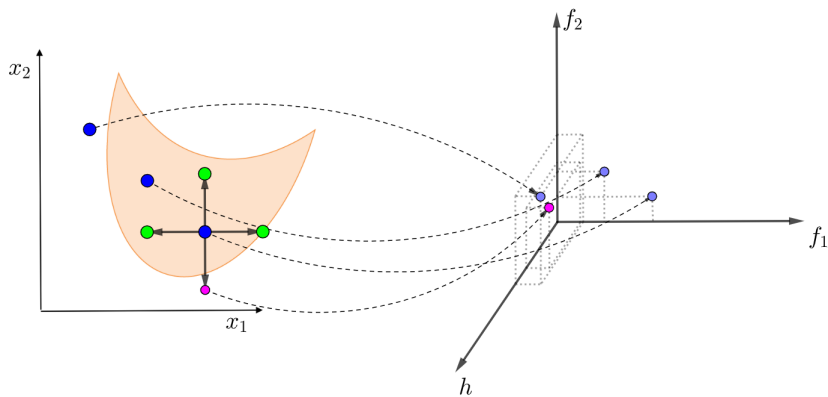
# DMS-Filter – Poll Step



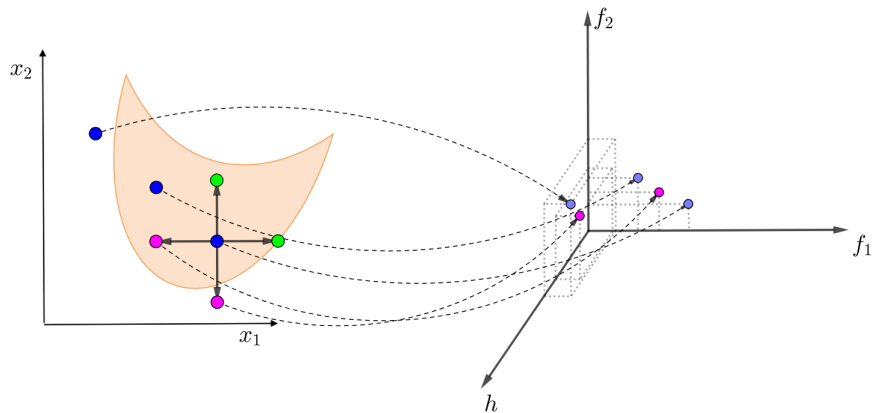
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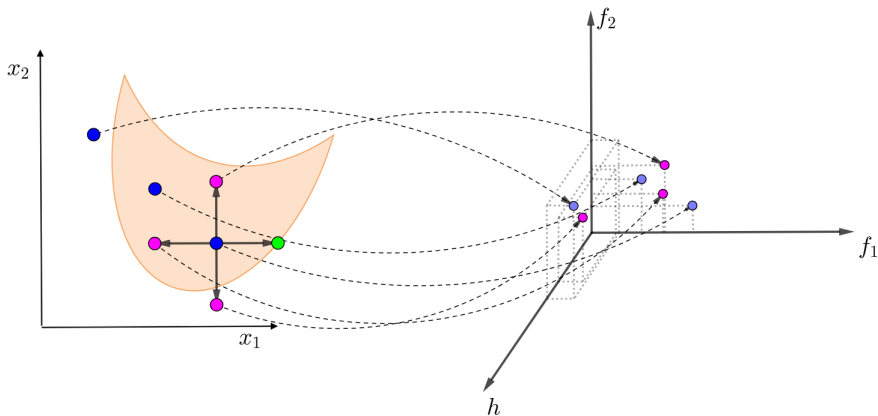
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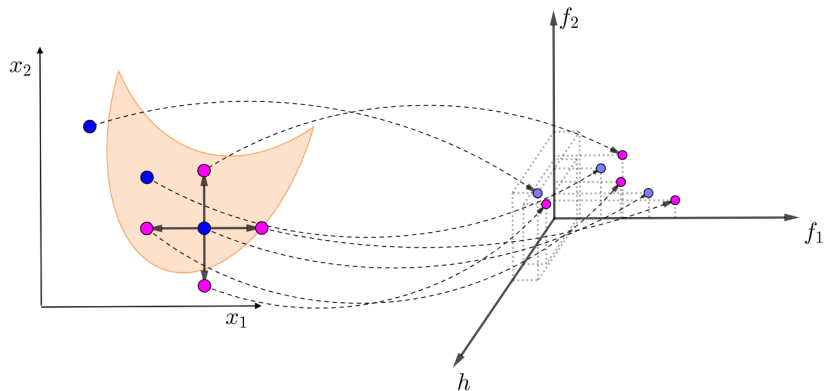
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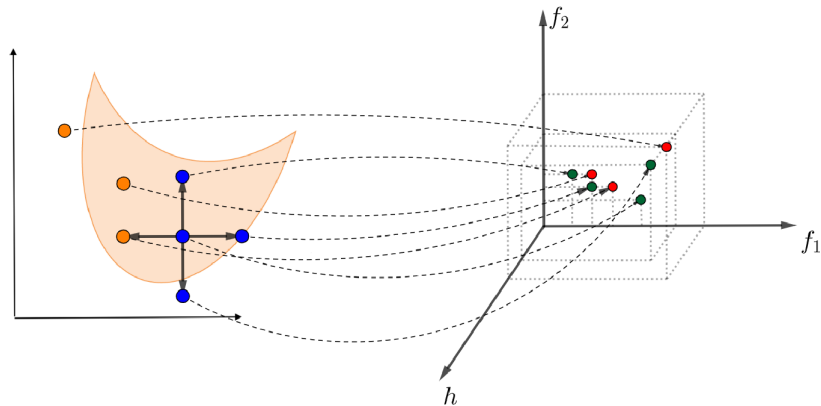
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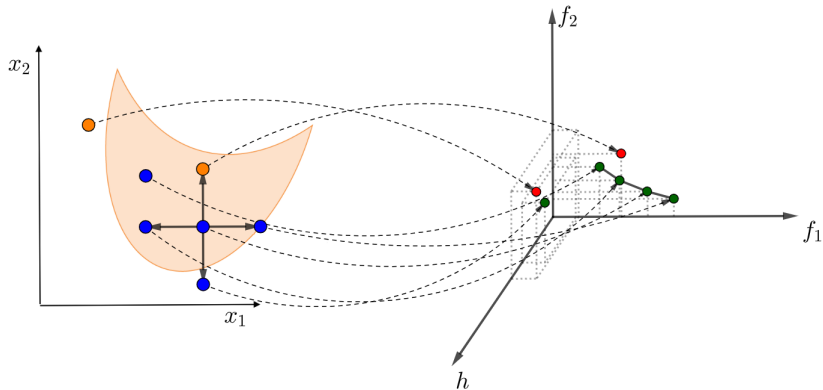
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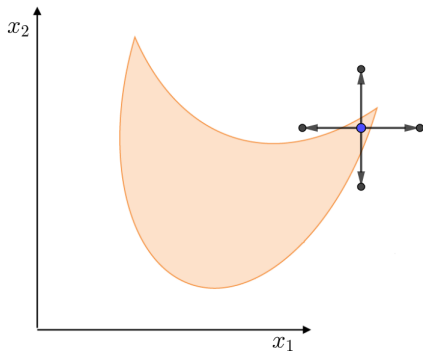
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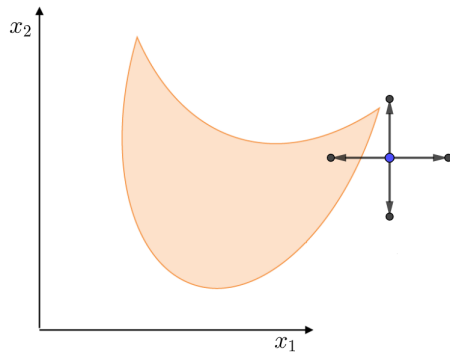


# Poll Center Selection

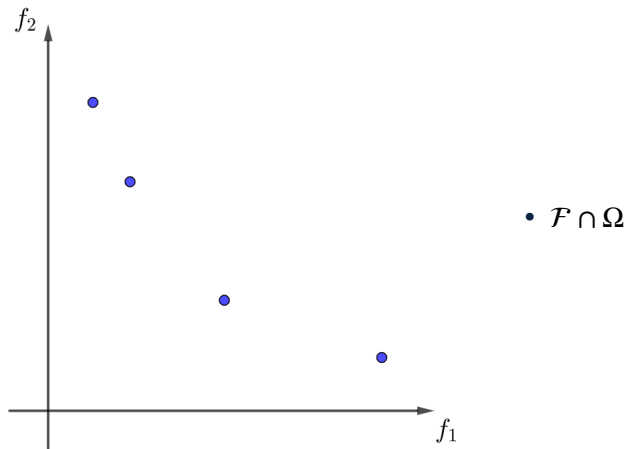
- Feasible to Infeasible



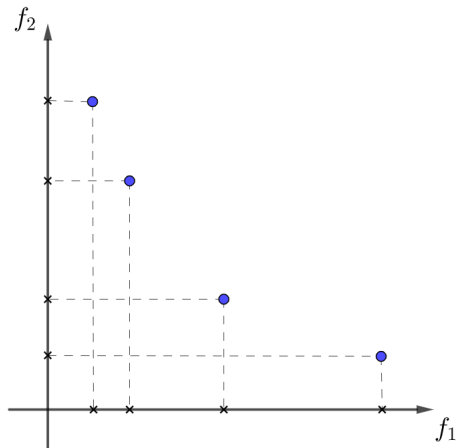
- Infeasible to Feasible



# Feasible poll center - Most Isolated Point

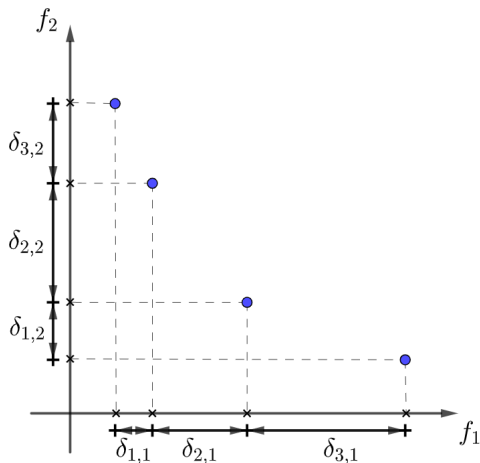


# Feasible poll center - Most Isolated Point



•  $\mathcal{F} \cap \Omega$

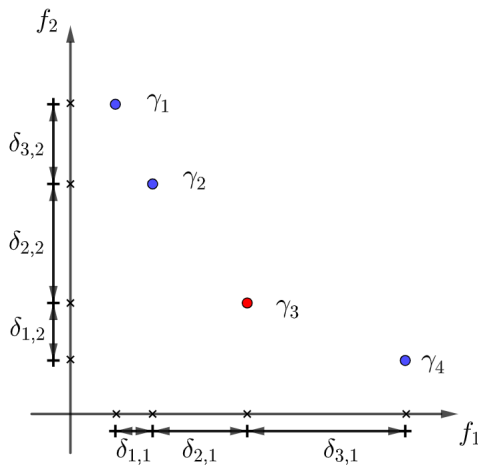
# Feasible poll center - Most Isolated Point



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

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

# Feasible poll center - Most Isolated Point



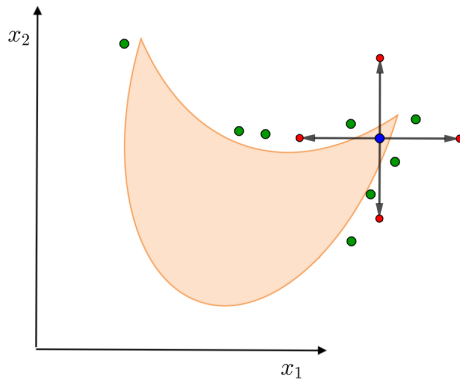
$$\gamma_1 = \frac{\delta_{3,2} + \delta_{1,1}}{2}$$

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

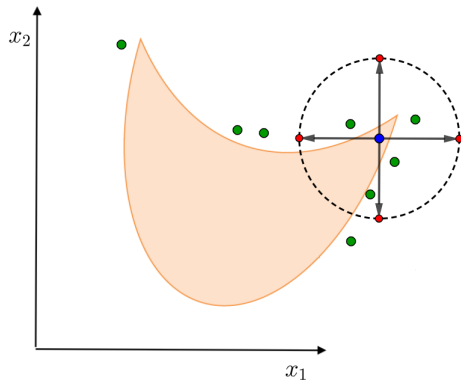
$$\gamma_3 = \frac{\frac{\delta_{2,2} + \delta_{1,2}}{2} + \frac{\delta_{2,1} + \delta_{3,1}}{2}}{2}$$

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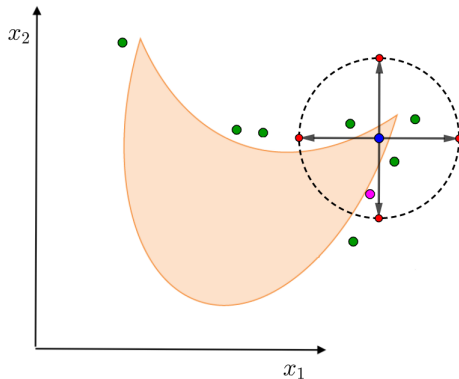
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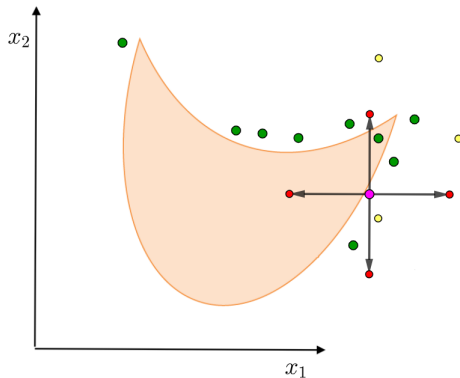


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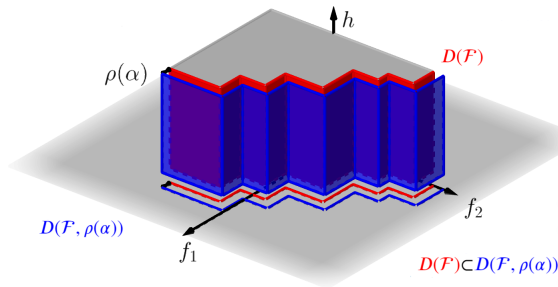
# Globalization Strategies

## 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 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

- Let  $\{x_k^I\}_{k \in K}$  be 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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  - number of variables between 3 and 30
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- 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

# Numerical Settings

- Comparison among DFMO, DMS and DMS-Filter
- 93 biobjective problems with nonlinear constraints and bounds
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  - number of constraints between 1 and 29
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  - Feasible point provided by Kar Mitsa [2007]
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  - **DFMO:**
    - default values
  - maximum of 20000 function evaluations

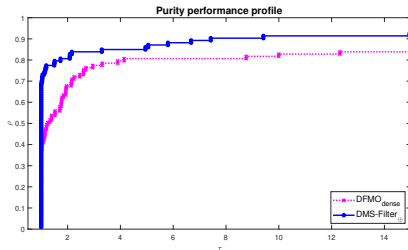
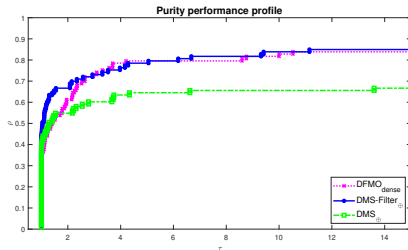
# Numerical Settings

- Comparison among DFMO, DMS and DMS-Filter
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  - **DFMO**:
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  - maximum of 20000 function evaluations

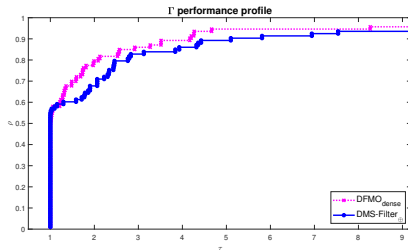
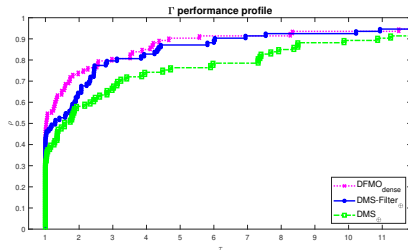
# Results - Purity



- DFMO
- DMS-Filter
- DMS

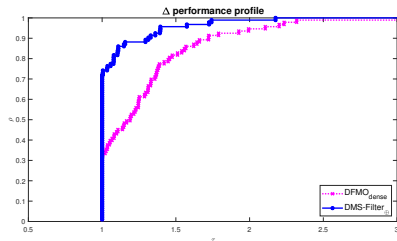
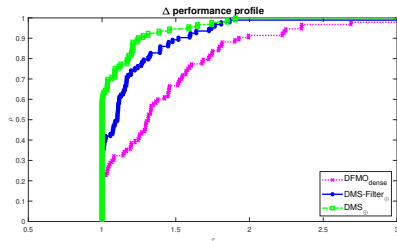


# Results - Spread Gamma ( $\Gamma$ )



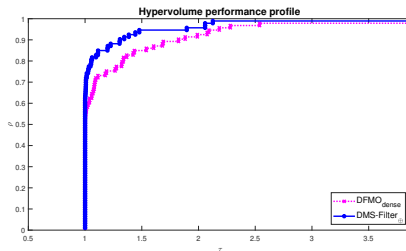
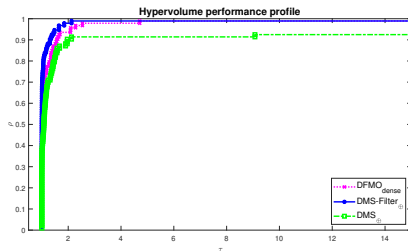
- DFMO
- DMS-Filter
- DMS

# Results - Spread Delta ( $\Delta$ )



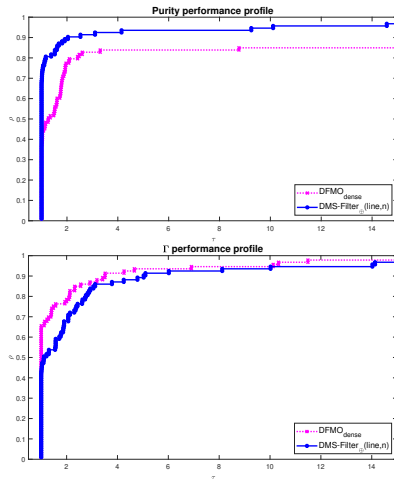
- DFMO
- DMS-Filter
- DMS

# Results - Hypervolume



- DFMO
- DMS-Filter
- DMS

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

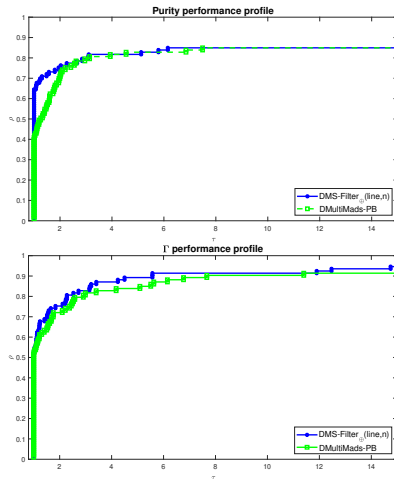


DFMO

DMS-Filter

- Comparison between DMS-Filter and DMultiMads-PB
  - **DMultiMads-PB** → Jean Bignon, Sébastien Le Digabel, Ludovic Salomon. *Handling of constraints in multiobjective blackbox optimization*. ArXiv:2204.00904
- 93 biobjective problems with nonlinear constraints and bounds
  - number of variables between 3 and 30
  - number of constraints between 1 and 29
- Initialization in line
  - $n$ -points equally spaced in the line segment, joining the variable upper and lower bounds
- Stopping criterion
  - **DMS-Filter** and **DMultiMads-PB**
    - $\alpha_k < 10^{-9}$  for one point
    - maximum of 20000 function evaluations

# DMS-Filter VS DMultiMads-PB



DMS-Filter

DMultiMads-PB

# 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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