

Julia tutorial (3)

MultiObjective Metaheuristic Implementation

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Multiojective Optimization Problem (MOP)

$$\begin{array}{ll}\min & F(x) \\ \text{s.t.} & x \in X\end{array}$$

where:

$$\begin{array}{ll}X \subseteq \mathbb{R}^n & \longrightarrow \text{the set of feasible solutions} \\ Y = F(X) \subseteq \mathbb{R}^p & \longrightarrow \text{the set of images}\end{array}$$

Computing sets Y_N and X_E for MOP:

$$y = F(x)$$

$y^* \in Y$ is **nondominated**, if $\nexists y \in Y$ such that $y_i \leq y_i^*, \forall i$ and $y \neq y^*$.

Y_N is the set of nondominated points.

$x^* \in X$ is **efficient** if y^* is nondominated.

X_E is a complete set of efficient solutions.

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Example 1

Compute X_E and Y_N for the following 2-IP:

$$\begin{array}{llllll} \max z_1 & = & x_1 & + & x_2 & \\ \min z_2 & = & x_1 & + & 3x_2 & \\ s.t. & & 2x_1 & + & 3x_2 & \leq 30 \\ & & 3x_1 & + & 2x_2 & \leq 30 \\ & & x_1 & - & x_2 & \leq 5.5 \\ & & x_1 & , & x_2 & \in \mathbb{N} \end{array}$$

1. Exact resolution



Exact resolution

Easy in Julia with these packages:

- ▶ Model:
JuMP
- ▶ Algorithm:
MultiObjectiveAlgorithms
- ▶ Solver:
HiGHS, Gurobi, etc.



From vOptGeneric.jl to MultiObjectiveAlgorithms.jl

vOptSolver (vOptGeneric and vOptSpecific)

<https://github.com/vOptSolver>

2015: Kick-off of vOpt ANR/DFG project

2017: Introduction of vOptSolver (vOptGeneric, vOptSpecific)

Xavier Gandibleux, Gauthier Soleilhac, Anthony Przybylski, Stefan Ruzika. vOptSolver: an open source software environment for multiobjective mathematical optimization. *IFORS2017: 21st Conference of the International Federation of Operational Research Societies*. July 17-21, 2017. Quebec City (Canada).

2018: Julia 1.0

2019: End of vOpt project

2019: First discussions on discourse/new design of MOO in JuMP

2022: JuMP 1.0

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JuMP

<https://jump.dev/>

Miles Lubin. The state of JuMP. *JuliaCon 2023: The 10th Annual JuliaCon - Co-Located with JuMP-dev*, July 25th - 29th, 2023. MIT, Cambridge, (USA).

MultiObjectiveAlgorithms (MOA)

<https://github.com/jump-dev/MultiObjectiveAlgorithms.jl>

Oscar Dowson, Xavier Gandibleux, Gokhan Kof. From vOptGeneric.jl to MultiObjectiveAlgorithms.jl *JuliaDays 2023*, 4-6 October 2023. Paris (France).



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Getting Started

Install:

```
julia> using Pkg
```

```
julia> Pkg.add("JuMP")
```

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julia> Pkg.add("MultiObjectiveAlgorithms")
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julia> Pkg.add("HiGHS")
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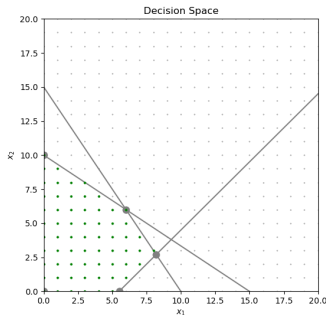
```
julia> Pkg.add("HiGHS")
```

Example 1 (Con't)

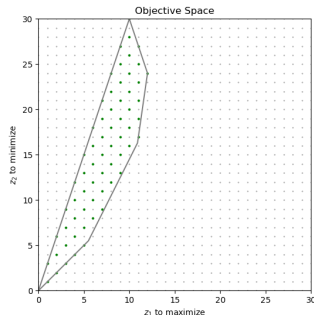
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```
julia> c1 = [1,1]
julia> c2 = [1,3]
julia> A = [2 3; 3 2; 1 -1]
julia> b = [30, 30, 5.5]
```

Decision space:



Objective space:



Example 1 (Con't)

Write the corresponding 2-IP model with JuMP

```
julia> using JuMP, HiGHS
julia> import MultiObjectiveAlgorithms as MOA
```

```
julia> model = Model( )
julia> @variable(model, x1 ≥ 0, Int)
julia> @variable(model, x2 ≥ 0, Int)
julia> @expression(model, fct1, x1 + x2)           # to maximize
julia> @expression(model, fct2, x1 + 3 * x2)       # to minimize
julia> @objective(model, Max, [fct1, (-1) * fct2]))
julia> @constraint(model, 2*x1 + 3*x2 ≤ 30))
julia> @constraint(model, 3*x1 + 2*x2 ≤ 30))
julia> @constraint(model, x1 - x2 ≤ 5.5))
```

```
julia> print(model)
```

Example 1 (Con't)

Setup the MIP solver, e.g. HiGHS

```
julia> set_optimizer(model, ()->MOA.Optimizer(HiGHS.Optimizer))
```

Setup the algorithm: ϵ -constraint; step=1 (default value)

```
julia> set_attribute(model, MOA.Algorithm(), MOA.EpsilonConstraint())
```

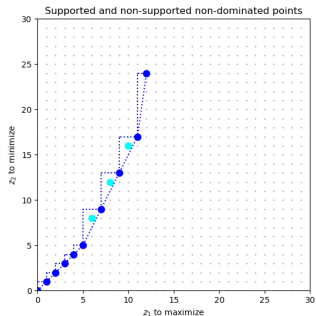
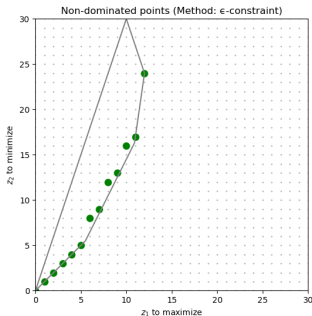
Optimize the MOO problem

```
julia> optimize!(model)
```

Example 1 (Con't)

Get X_E and Y_N

```
julia> for i in 1:result_count(model)
julia>     z1_opt = objective_value(model; result = i)[1]
julia>     z2_opt = -1 * objective_value(model; result = i)[2]
julia>     x1_opt = value(x1; result = i)
julia>     x2_opt = value(x2; result = i)
julia> end
```



1. Heuristic resolution



You enter on the scene Antonio...

During your talk, could you illustrate the resolution of the example 1 with MetaJul?



References

- ▶ Jeff Bezanson, Alan Edelman, Stefan Karpinski and Viral B. Shah. Julia: A Fresh Approach to Numerical Computing. *SIAM Review*, 59: 65–98. 2017.
- ▶ Miles Lubin, Oscar Dowson, Joaquim Dias Garcia, Joey Huchette, Benoît Legat, Juan Pablo Vielma. JuMP 1.0: Recent improvements to a modeling language for mathematical optimization. *Mathematical Programming Computation*. 2023.
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