



JuMP-dev 2024 at Montreal, Canada, July 19th, 2024

What?

$$\theta_0, \theta_1, ..., \theta_m)$$

$$\mathcal{P} : \min_{x} F_0(x, \theta_0)$$

s.t. $G_i(x, \theta_i) \in \mathcal{S}_i$

Solve:
$$x^* \in \operatorname{argmin}_x \mathcal{P}$$

Compute: $\frac{\partial x^*}{\partial \theta} \equiv \left(\frac{\partial x_j^*}{\partial \theta_i}\right)_{ij}$

Or... Differentiate the solution of parametrized optimization problems



Why?

Classic OR

Sensitivities

Perturbation analysis

Bilevel Optimization

New trends in ML

Optimization problem as layer

Hyper parameter tuning

Learning



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Home > INFORMS Journal on Computing > Vol. 36, No. 2 >

BilevelJuMP.jl: Modeling and Solving Bilevel Optimization Problems in Julia

Joaquim Dias Garcia 🕩, Guilherme Bodin 🕩, Alexandre Street 🕩

Published Online: 13 Dec 2023 https://doi.org/10.1287/ijoc.2022.0135

Mathematics > Optimization and Control

[Submitted on 26 Feb 2021 (v1), last revised 8 Apr 2024 (this version, v5)]

Application-Driven Learning: A Closed-Loop Prediction and Optimization Approach Applied to Dynamic Reserves and Demand Forecasting

Joaquim Dias Garcia, Alexandre Street, Tito Homem-de-Mello, Francisco D. Muñoz

(At arXiv but just accepted in Operations Research)



In JuMP

But remember the sentence:

Differentiate the solution of **parametrized** optimization problems



In JuMP

But remember the sentence:

Differentiate the solution of **parametrized** optimization problems

Which is a lot of work, but we can get some help



In JuMP

But remember the sentence:

Differentiate the solution of **parametrized** optimization problems

Which is a lot of work, but we can get some help



Google Summer of Code



Differentiable optimization: Differentiate

2020 Program | NumFOCUS

Contributor AkshaySharma

View Code

Optimization problem differentiation

Mentors

Organization

Joaquim, Benoît Legat, Mario Souto

NumFOCUS



Differentiable optimization: Differentiate

2020 Program | NumFOCUS

Contributor AkshaySharma

View Code

Optimization problem differentiation

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Flexible Differentiable Optimization via Model Transformations

Mathieu Besançon 🔍, Joaquim Dias Garcia 🔍, Benoît Legat 🔍, Akshay Sharma 🔍

Published Online: 14 Nov 2023 https://doi.org/10.1287/ijoc.2022.0283



Differentiable optimization: Differentiate

```
using JuMP, DiffOpt, HiGHS
model = Model(() -> DiffOpt.Optimizer(HiGHS.Optimizer()))
set_silent(model)
@variable(model, x)
@constraint(model, cons, x >= 3 * 3)
@objective(model, Min, 2x)
optimize!(model)
```

```
sensitivity = convert(MOI.ScalarAffineFunction{Float64}, 1.0)
MOI.set(model, DiffOpt.ForwardConstraintFunction(), cons, sensitivity)
DiffOpt.forward_differentiate!(model)
MOI.get(model, DiffOpt.ForwardVariablePrimal(), x) # ~ - 1.0
```

Necessary to differentiate with respect to arbitrary coefficients.

But might be cumbersome to use.

```
The function is "normalized"
So, it is a sensitivity wrt changing a
constant on the LEFT-hand-side
```



Differentiable optimization: Parametric

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View Code

Adding parameters to JuMP and MathOptInterface

Mentors

Organization

Joaquim, Oscar Dowson, Benoît Legat

NumFOCUS



Differentiable optimization: Parametric

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Contributor Tomás Gutierrez



Adding parameters to JuMP and MathOptInterface









Differentiable optimization: *Parametric*

```
using JuMP, HiGHS
import ParametricOptInterface as POI
model = Model(() -> POI.Optimizer(HiGHS.Optimizer()))
set_silent(model)
@variable(model, x)
@variable(model, p in Parameter(3.0))
(aconstraint(model, cons, x >= 3 * p))
@objective(model, Min, 2x)
optimize!(model)
value(x) # ≈ 9
set_parameter_value(p, 2.0)
optimize!(model)
value(x) \# \approx 6
```



```
using JuMP, DiffOpt, HiGHS
import ParametricOptInterface as POI
model = Model(() -> POI.Optimizer(DiffOpt.Optimizer(HiGHS.Optimizer())))
set_silent(model)
@variable(model, x)
@variable(model, p in MOI.Parameter(3.0))
@constraint(model, cons, x >= 3 * p)
@objective(model, Min, 2x)
optimize!(model)
value(x) \# \approx 9
# the function is: x(p) = 3p, hence x'(p) = 3
# differentiate w.r.t. p
MOI.set(model, POI.ForwardParameter(), p, 1)
DiffOpt.forward_differentiate!(model)
MOI.get(model, DiffOpt.ForwardVariablePrimal(), x) \# \approx 3
```



```
using JuMP, DiffOpt, HiGHS
import ParametricOptInterface as POI
model = Model(() -> POI.Optimizer(DiffOpt.Optimizer(HiGHS.Optimizer())))
set_silent(model)
@variable(model, x)
@variable(model, p in MOI.Parameter(3.0))
(aconstraint(model, cons, x >= 3 * p))
@objective(model, Min, 2x)
optimize!(model)
value(x) \# \approx 9
                                # update parameter
# the function is: x(p) = 3p
                                 set_parameter_value(p, 2.0)
# differentiate w.r.t. p
                                optimize!(model)
MOI.set(model, POI.ForwardPar
                                value(x) # ≈ 6
                                # differentiate w.r.t. p
DiffOpt.forward differentiate
                                MOI.set(model, POI.ForwardParameter(), p, 1)
MOI.get(model, DiffOpt.Forwar
                                DiffOpt.forward_differentiate!(model)
                                MOI.get(model, DiffOpt.ForwardVariablePrimal(), x) \# \approx 3
```

[WIP] POI + DiffOpt = S2 #143

;) Open joaquimg wants to merge 12 commits into master from jg/diff .

Currently, DiffOpt needs to be a dependency of POI

But POI only has MOI as deps DiffOpt is heavy

Next step, Move to weakdep + extension approach

Also, Needs more adjustments in DiffOpt (add bridges and tests)



[WIP] POI + DiffOpt = S2 #143

;) Open joaquimg wants to merge 12 commits into master from jg/diff .

Simpler to set sensitivities More direct approach

Simpler to query derivatives One parameter might appear in multiple places Derivatives are accumulated

Use cases typically require multiple solves Parameters facilitate and speedup the updates

Easier to generalize

To nonlinear(!) (thanks to nonlinear refactor) See <u>https://github.com/andrewrosemberg/DiffOpt.jl/pull/1</u>

(WIP) Test script sIPOPT #1

