

Package ‘ROI.plugin.deoptim’

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Version 1.0-2

Title 'DEoptim' and 'DEoptimR' Plugin for the 'R' Optimization Interface

Description Enhances the R Optimization Infrastructure ('ROI') package with the 'DEoptim' and 'DEoptimR' package. 'DEoptim' is used for unconstrained optimization and 'DEoptimR' for constrained optimization.

Imports methods, stats, utils, ROI (>= 1.0-0), DEoptim, DEoptimR (>= 1.0-10)

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URL <https://roigrp.gitlab.io>,
<https://gitlab.com/roigrp/solver/ROI.plugin.deoptim>

NeedsCompilation no

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ROI.plugin.deoptim-package
deoptimr

Description

This package is part of the **R Optimization Infrastructure ROI**

References

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See Also

Function `JDEoptim()` in the **DEoptimR** package.

 Example-1

Banana

Description

The following example is also known as Rosenbrock's banana function (https://en.wikipedia.org/wiki/Rosenbrock_function).

$$\text{minimize } f(x) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2$$

Solution: c(1, 1)

Examples

```
Sys.setenv(ROI_LOAD_PLUGINS = FALSE)
library(ROI)
library(ROI.plugin.deoptim)

f <- function(x) {
  return( 100 * (x[2] - x[1]^2)^2 + (1 - x[1])^2 )
}

x <- OP( objective = F_objective(f, n=2L, names=c("x_1", "x_2")),
        bounds = V_bound(li=1:2, ui=1:2, lb=c(-3, -3), ub=c(3, 3)) )

nlp <- ROI_solve(x, solver = "deoptim")
nlp
## Optimal solution found.
## The objective value is: 3.828383e-22
solution(nlp)
## x_1 x_2
## 1 1
```

 Example-2

Hock-Schittkowski-Collection Problem 16

Description

The following example solves problem 16 from the Hock-Schittkowski-Collection.

$$\text{minimize } f(x) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2$$

$$\text{subject to: } x_1 + x_2^2 \geq 0 \quad x_1^2 + x_2 \geq 0$$

$$-2 \geq x_1 \geq 0.5 \quad x_2 \geq 1$$

Solution: c(0.5, 0.25)

Examples

```

Sys.setenv(ROI_LOAD_PLUGINS = FALSE)
library(ROI)
library(ROI.plugin.deoptim)

f <- function(x) {
  return( 100 * (x[2] - x[1]^2)^2 + (1 - x[1])^2 )
}

f.gradient <- function(x) {
  return( c( -400 * x[1] * (x[2] - x[1] * x[1]) - 2 * (1 - x[1]),
            200 * (x[2] - x[1] * x[1])) )
}

x <- OP( objective = F_objective(f, n=2L, G=f.gradient),
        constraints = c(F_constraint(F=function(x) x[1] + x[2]^2, ">=", 0,
                                   J=function(x) c(1, 2*x[2])),
                       F_constraint(F=function(x) x[1]^2 + x[2], ">=", 0,
                                   J=function(x) c(2*x[1], x[2]))),
        bounds = V_bound(li=1:2, ui=1:2, lb=c(-2, -Inf), ub=c(0.5, 1)) )

nlp <- ROI_solve(x, solver="deoptimr", start=c(0.4, 0.3))
nlp
## Optimal solution found.
## The objective value is: 2.499999e-01
solution(nlp)
## [1] 0.5000001 0.2499994

```

Example-3

Hock-Schittkowski-Collection Problem 36

Description

The following example solves example 36 from the Hock-Schittkowski-Collection.

$$\text{minimize } -x_1x_2x_3$$

$$\text{subject to: } x_1 + 2x_2 + x_3 \leq 72$$

$$0 \leq x_1 \leq 20, 0 \leq x_2 \leq 11, 0 \leq x_3 \leq 42$$

Examples

```

Sys.setenv(ROI_LOAD_PLUGINS = FALSE)
library(ROI)
library(ROI.plugin.deoptim)

hs036_obj <- function(x) {
  -x[1] * x[2] * x[3]
}

```

```
hs036_con <- function(x) {  
  x[1] + 2 * x[2] + 2 * x[3]  
}  
  
x <- OP( objective = F_objective(hs036_obj, n = 3L),  
        constraints = F_constraint(hs036_con, "<=", 72),  
        bounds = V_bound(ub = c(20, 11, 42)) )  
  
nlp <- ROI_solve(x, solver = "deoptimr", start = c(10, 10, 10),  
                max_iter = 2000)  
  
nlp  
## Optimal solution found.  
## The objective value is: -3.300000e+03  
solution(nlp, "objval")  
## [1] -3300  
solution(nlp)  
## [1] 20 11 15
```

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