

Package: evoper (via r-universe)

September 30, 2024

Type Package

Title Evolutionary Parameter Estimation for 'Repast Symphony' Models

Version 0.5.0

Date 2018-08-30

Author Antonio Prestes Garcia [aut, cre], Alfonso Rodriguez-Paton
[aut, ths]

Maintainer Antonio Prestes Garcia <antonio.pgarcia@alumnos.upm.es>

URL <https://github.com/antonio-pgarcia/evoper>

BugReports <https://github.com/antonio-pgarcia/evoper/issues>

Description The EvoPER, Evolutionary Parameter Estimation for Individual-based Models is an extensible package providing optimization driven parameter estimation methods using metaheuristics and evolutionary computation techniques (Particle Swarm Optimization, Simulated Annealing, Ant Colony Optimization for continuous domains, Tabu Search, Evolutionary Strategies, ...) which could be more efficient and require, in some cases, fewer model evaluations than alternatives relying on experimental design. Currently there are built in support for models developed with 'Repast Symphony' Agent-Based framework (<<https://repast.github.io/>>) and with NetLogo (<<https://ccl.northwestern.edu/netlogo/>>) which are the most used frameworks for Agent-based modeling.

License MIT + file LICENSE

LazyData TRUE

Depends rrepast

Imports methods, futile.logger, boot, reshape, ggplot2, deSolve,
plot3D, plyr, data.table, utils, RNetLogo

RoxygenNote 6.0.1

Suggests testthat

NeedsCompilation no

Repository CRAN

Date/Publication 2018-08-30 23:20:06 UTC

Contents

abm.acor	5
abm.ees1	5
abm.ees2	6
abm.pso	7
abm.saa	7
abm.tabu	9
acor.archive	9
acor.F	10
acor.lthgaussian	11
acor.N	11
acor.probabilities	12
acor.S	12
acor.sigma	13
acor.updateants	13
acor.W	14
acor.weigth	14
assert	15
bestFitness	15
bestSolution	16
cbuf	16
compare.algorithms1	17
contourplothelper	17
ees1.challenge	18
ees1.explore	18
ees1.mating	19
ees1.mating1	19
ees1.mutation	20
ees1.recombination	20
ees1.selection	21
elog.debug	21
elog.error	21
elog.info	22
elog.level	22
enforceBounds	23
es.evaluate	23
Estimates-class	24
extremize	24
f0.ackley	25
f0.ackley4	25
f0.adtn.rosenbrock2	26
f0.bohachevsky	26
f0.bohachevsky4	27
f0.cigar	27
f0.cigar4	28
f0.griewank	28
f0.griewank4	29

f0.nlnn.rosenbrock2	29
f0.periodtuningpp	30
f0.periodtuningpp12	30
f0.periodtuningpp24	31
f0.periodtuningpp48	32
f0.periodtuningpp72	33
f0.rosenbrock2	34
f0.rosenbrock4	34
f0.rosenbrockn	35
f0.schaffer	35
f0.schaffer4	36
f0.schwefel	36
f0.schwefel4	37
f0.test	37
f1.ackley	38
f1.adtn.rosenbrock2	38
f1.bohachevsky	39
f1.cigar	39
f1.griewank	40
f1.nlnn.rosenbrock2	40
f1.rosenbrock2	41
f1.rosenbrockn	41
f1.schaffer	42
f1.schwefel	42
f1.test	43
fixdfcolumns	43
generateSolution	44
getFitness	44
getSolution	45
gm.mean	45
gm.sd	46
histplotheper	46
initSolution	47
lowerBound	47
Magnitude	48
naiveperiod	48
NetLogoFunction-class	48
NLWrapper.FindJar	49
NLWrapper.GetParameter	49
NLWrapper.Model	50
NLWrapper.Run	51
NLWrapper.RunExperiment	51
NLWrapper.SetParameter	52
NLWrapper.SetRandomSeed	53
NLWrapper.Shutdown	53
ObjectiveFunction-class	54
Options-class	54
OptionsACOR-class	54

OptionsEES1-class	55
OptionsEES2-class	55
OptionsFactory	55
OptionsPSO-class	56
OptionsSAA-class	56
OptionsTS-class	56
paramconverter	56
partSolutionSpace	57
PlainFunction-class	57
pop.first	57
pop.last	58
predatorprey	58
predatorprey.plot0	59
predatorprey.plot1	59
pso.best	60
pso.chi	61
pso.lbest	61
pso.neighborhood.K2	62
pso.neighborhood.K4	62
pso.neighborhood.KN	63
pso.printbest	63
pso.Velocity	64
push	64
random.wheel	65
RepastFunction-class	65
saa.bolt	65
saa.neighborhood	66
saa.neighborhood1	66
saa.neighborhoodH	67
saa.neighborhoodN	67
saa.tbyk	68
saa.tcte	68
saa.texp	69
scatterplotlohelper	69
searchrow	70
show.comp1	70
slope	71
slopes	71
sortSolution	72
summarize.comp1	72
tabu.getNeighbors	73
tabu.istabu	73
upperBound	74
xmeanci1	74
xmeanci2	75
xyplohelper	75

abm.acor	<i>Ant colony optimization for continuous domains</i>
----------	---

Description

An implementation of Ant Colony Optimization algorithm for continuous variables.

Usage

```
abm.acor(objective, options = NULL)
```

Arguments

objective	An instance of ObjectiveFunction (or subclass) class ObjectiveFunction
options	An appropriate instance from a subclass of Options class

References

[1] Socha, K., & Dorigo, M. (2008). Ant colony optimization for continuous domains. *European Journal of Operational Research*, 185(3), 1155-1173. <http://doi.org/10.1016/j.ejor.2006.06.046>

Examples

```
## Not run:
f<- PlainFunction$new(f0.rosenbrock2)

f$Parameter(name="x1",min=-100,max=100)
f$Parameter(name="x2",min=-100,max=100)

extremize("acor", f)

## End(Not run)
```

abm.ees1	<i>EvoPER Evolutionary Strategy 1</i>
----------	---------------------------------------

Description

This function tries to provide a rough approximation to best solution when no information is available for the correct range of input parameters for the objective function. It can be useful for studying the behavior of individual-based models with high variability in the output variables showing non-linear behaviors.

Usage

```
abm.ees1(objective, options = NULL)
```

Arguments

objective An instance of ObjectiveFunction (or subclass) class [ObjectiveFunction](#)
options An appropriate instance from a subclass of [Options](#) class

Examples

```
## Not run:
f<- PlainFunction$new(f0.rosenbrock2)

f$Parameter(name="x1",min=-100,max=100)
f$Parameter(name="x2",min=-100,max=100)

extremize("ees1", f)

## End(Not run)
```

abm.ees2

EvoPER Evolutionary Strategy 2

Description

This function tries to provide a rough approximation to best solution when no information is available for the correct range of input parameters for the objective function. It can be useful for studying the behavior of individual-based models with high variability in the output variables showing non-linear behaviors.

Usage

```
abm.ees2(objective, options = NULL)
```

Arguments

objective An instance of ObjectiveFunction (or subclass) class [ObjectiveFunction](#)
options An appropriate instance from a subclass of [Options](#) class

Examples

```
## Not run:
f<- PlainFunction$new(f0.rosenbrock2)

f$Parameter(name="x1",min=-100,max=100)
f$Parameter(name="x2",min=-100,max=100)

extremize("ees2", f)

## End(Not run)
```

*abm.pso**abm.pso*

Description

An implementaion of Particle Swarm Optimization method for parameter estimation of Individual-based models.

Usage

```
abm.pso(objective, options = NULL)
```

Arguments

`objective` An instance of ObjectiveFunction (or subclass) class [ObjectiveFunction](#)
`options` An apropiate instance from a subclass of [Options](#) class

References

- [1] Kennedy, J., & Eberhart, R. (1995). Particle swarm optimization. In Proceedings of ICNN 95 - International Conference on Neural Networks (Vol. 4, pp. 1942-1948). IEEE.
- [2] Poli, R., Kennedy, J., & Blackwell, T. (2007). Particle swarm optimization. Swarm Intelligence, 1(1), 33-57.

Examples

```
## Not run:  
f<- PlainFunction$new(f0.rosenbrock2)  
  
f$Parameter(name="x1",min=-100,max=100)  
f$Parameter(name="x2",min=-100,max=100)  
  
extremize("pso", f)  
  
## End(Not run)
```

*abm.saa**abm.saa*

Description

An implementation of Simulated Annealing Algorithm optimization method for parameter estimation of Individual-based models.

Usage

```
abm.saa(objective, options = NULL)
```

Arguments

objective An instance of ObjectiveFunction (or subclass) class [ObjectiveFunction](#)
options An appropriate instance from a subclass of [Options](#) class

Value

The best solution.

References

[1] Kirkpatrick, S., Gelatt, C. D., & Vecchi, M. P. (1983). Optimization by Simulated Annealing. *Science*, 220(4598).

Examples

```
## Not run:
f<- PlainFunction$new(f0.rosenbrock2)

f$Parameter(name="x1",min=-100,max=100)
f$Parameter(name="x2",min=-100,max=100)

extremize("saa", f)

## End(Not run)

## Not run:
## A Repast defined function
f<- RepastFunction$new("/usr/models/BactoSim(HaldaneEngine-1.0)","ds::Output",300)

## or a plain function

f1<- function(x1,x2,x3,x4) {
  10 * (x1 - 1)^2 + 20 * (x2 - 2)^2 + 30 * (x3 - 3)^2 + 40 * (x4 - 4)^2
}

f<- PlainFunction$new(f1)

f$addFactor(name="cyclePoint",min=0,max=90)
f$addFactor(name="conjugationCost",min=0,max=100)
f$addFactor(name="pilusExpressionCost",min=0,max=100)
f$addFactor(name="gamma0",min=1,max=10)

abm.saa(f, 100, 1, 100, 0.75)

## End(Not run)
```

 abm.tabu

Tabu Search metaheuristic

Description

An implementation of Tabu Search algorithm for parameter estimation

Usage

```
abm.tabu(objective, options = NULL)
```

Arguments

objective	An instance of ObjectiveFunction (or subclass) class ObjectiveFunction
options	An appropriate instance from a subclass of Options class

References

- [1] Fred Glover (1989). "Tabu Search - Part 1". ORSA Journal on Computing, 190-206. doi:10.1287/ijoc.1.3.190.
- [2] Fred Glover (1990). "Tabu Search - Part 2". ORSA Journal on Computing, 4-32. doi:10.1287/ijoc.2.1.4.

Examples

```
## Not run:
f<- PlainFunction$new(f0.rosenbrock2)

f$Parameter(name="x1",min=-100,max=100)
f$Parameter(name="x2",min=-100,max=100)

or

f$Parameter0(name="x1",levels=c(0:4))
f$Parameter0(name="x2",levels=c(-2,-1,0,1,2))

extremize("tabu", f)

## End(Not run)
```

 acor.archive

acor.archive

Description

This function is used for creating and maintaining the ACO archive 'T'. The function keeps the track of 'k' solution in the archive.

Usage

```
acor.archive(s, f, w, k, T = NULL)
```

Arguments

s	The solution 'ants'
f	The evaluation of solution
w	The weight vector
k	The archive size
T	The current archive

Value

The solution archive

References

[1] Socha, K., & Dorigo, M. (2008). Ant colony optimization for continuous domains. European Journal of Operational Research, 185(3), 1155-1173. <http://doi.org/10.1016/j.ejor.2006.06.046>

acor.F

acor.F

Description

Helper function for extracting the 'F' function evaluations from archive ACOOr 'T'

Usage

```
acor.F(T)
```

Arguments

T	The solution archive
---	----------------------

Value

The F matrix

acor.lthgaussian	Select the lth gaussian function
------------------	----------------------------------

Description

Given a weight vector calculate the probabilities of selecting the lth gaussian function and return the index of lth gaussian selected with probability p

Usage

```
acor.lthgaussian(W)
```

Arguments

W The vector of weights

Value

The index of lth gaussian function

References

[1] Socha, K., & Dorigo, M. (2008). Ant colony optimization for continuous domains. European Journal of Operational Research, 185(3), 1155-1173. <http://doi.org/10.1016/j.ejor.2006.06.046>

acor.N	acor.N
--------	--------

Description

Helper function for getting the size of solution

Usage

```
acor.N(T)
```

Arguments

T The solution archive

Value

The size 'n' of a solution 's'

acor.probabilities *Gaussian kernel choosing probability*

Description

Calculate the probability of choosing the lth Gaussian function

Usage

```
acor.probabilities(W, l = NULL)
```

Arguments

W The vector of weights
l The lth element of algorithm solution archive T

Value

The vector of probabilities 'p'

References

[1] Socha, K., & Dorigo, M. (2008). Ant colony optimization for continuous domains. *European Journal of Operational Research*, 185(3), 1155-1173. <http://doi.org/10.1016/j.ejor.2006.06.046>

acor.S *acor.S*

Description

Helper function for extracting solution 'S' from archive 'T'

Usage

```
acor.S(T)
```

Arguments

T The solution archive

Value

The solution matrix

acor.sigma

Sigma calculation for ACOr

Description

Calculate the value of sigma

Usage

acor.sigma(Xi, k, T)

Arguments

Xi	The algorithm parameter
k	The solution archive size
T	The solution archive

Value

The sigma value

References

[1] Socha, K., & Dorigo, M. (2008). Ant colony optimization for continuous domains. *European Journal of Operational Research*, 185(3), 1155-1173. <http://doi.org/10.1016/j.ejor.2006.06.046>

acor.updateants

acor.updateants

Description

Update the solution using the gaussian kernel

Usage

acor.updateants(S, N, W, t.mu, t.sigma)

Arguments

S	The current solution ants
N	The number of required ants in solution
W	The weight vector
t.mu	The 'mean' from solution archive
t.sigma	The value of sigma from solution archive

Value

The new solution ants

References

[1] Socha, K., & Dorigo, M. (2008). Ant colony optimization for continuous domains. *European Journal of Operational Research*, 185(3), 1155-1173. <http://doi.org/10.1016/j.ejor.2006.06.046>

acor.W

acor.W

Description

Helper function for extracting the 'W' function evaluations from archive ACO_r 'T'

Usage

acor.W(T)

Arguments

T The solution archive

Value

The weight vector

acor.weight

Weight calculation for ant colony optimization

Description

Calculates the weight element of ACO_r algorithm for the solution archive.

Usage

acor.weight(q, k, l)

Arguments

q The Algorithm parameter. When small best-ranked solution is preferred
 k The Archive size
 l The lth element of algorithm solution archive T

Value

A scalar or a vector with calculated weighth.

References

[1] Socha, K., & Dorigo, M. (2008). Ant colony optimization for continuous domains. European Journal of Operational Research, 185(3), 1155-1173. <http://doi.org/10.1016/j.ejor.2006.06.046>

 assert

assert

Description

The assert function stop the execution if the logical expression given by the parameter expresion is false.

Usage

```
assert(expresion, string)
```

Arguments

expresion	Some logical expression
string	The text message to show if expresion does not hold

 bestFitness

bestFitness

Description

Given a set S of N solutions created with sortSolution, this function returns the fitness component fot the best solution.

Usage

```
bestFitness(S)
```

Arguments

S	The solution set
---	------------------

Value

The best fitness value

bestSolution	<i>bestSolution</i>
--------------	---------------------

Description

Given a set S of N solutions created with sortSolution, this function returns the best solution found.

Usage

```
bestSolution(S)
```

Arguments

S	The solution set
---	------------------

Value

The best solution

cbuf	<i>cbuf</i>
------	-------------

Description

Simple implementation of a circular buffer.

Usage

```
cbuf(b, v, e)
```

Arguments

b	The variable holding the current buffer content
v	The new valued to be added to b
e	The length of circular buffer

Value

The buffer b plus the element v minus the least recently added element

compare.algorithms1 *compare.algorithms1*

Description

Compare the number of function evaluations and convergence for the following optimization algorithms, ("saa","pso","acor","ees1").

Usage

```
compare.algorithms1(F, seeds = c(27, 2718282, 36190727, 3141593, -91190721,
-140743, 1321))
```

Arguments

F	The function to be tested
seeds	The random seeds which will be used for testing algorithms

Examples

```
## Not run:
rm(list=ls())
d.cigar4<- compare.algorithms1(f0.cigar4)
d.schaffer4<- compare.algorithms1(f0.schaffer4)
d.griewank4<- compare.algorithms1(f0.griewank4)
d.bohachevsky4<- compare.algorithms1(f0.bohachevsky4)
d.rosenbrock4<- compare.algorithms1(f0.rosenbrock4)

## End(Not run)
```

contourplotheper *contourplotheper*

Description

Simple helper function for countour plots

Usage

```
contourplotheper(d, x, y, z, nbins = 32, binwidth = c(10, 10),
points = c(300, 300), title = NULL)
```

Arguments

d	A data frame.
x	A string with the dataframe column name for x axis.
y	A string with the dataframe column name for y axis.
z	A string with the dataframe column name for z axis.
nbins	The number bins. The default is 32.
binwidth	The binwidths for 'kde2d'. Can be an scalar or a vector.
points	The number of grid points. Can be an scalar or a vector.
title	The optional plot title. May be omitted.

ees1.challenge	<i>ees1.challenge</i>
----------------	-----------------------

Description

Repeat the evaluation of best solution to tackle with variability.

Usage

```
ees1.challenge(solution, objective)
```

Arguments

solution	The Problem solution
objective	The objective function

ees1.explore	<i>ees1.explore</i>
--------------	---------------------

Description

Explore the solution space on the neighborhood of solution 's' in order to find a new best.

Usage

```
ees1.explore(s, weight, p = 0.01)
```

Arguments

s	The Problem solution
weight	The exploration intensity
p	The mutation probability

 ees1.mating

ees1.mating

Description

This function 'mix' the elements present in the solution. The parameter 'mu' controls the intensity of mixing. Low values give preference to best solution components and high values make the values being select randomly.

Usage

```
ees1.mating(solution, mu)
```

Arguments

solution	The Problem solution
mu	The mixing intensity ratio, from 0 to 1. The mix intensity controls de the probability of chosing a worst solutions

 ees1.mating1

ees1.mating1

Description

This function 'mix' the elements present in the solution. The parameter 'mu' controls the intensity of mixing. Low values give preference to best solution components and high values make the values being select randomly.

Usage

```
ees1.mating1(solution, mu)
```

Arguments

solution	The Problem solution
mu	The mixing intensity ratio, from 0 to 1. The mix intensity controls de the probability of chosing a worst solutions

ees1.mutation *ees1.mutation*

Description

Performs the mutation on generated solution

Usage

```
ees1.mutation(solution, mates, p = 0.01)
```

Arguments

solution	The Problem solution
mates	The mixed parents
p	The mutation probability

ees1.recombination *ees1.recombination*

Description

Performs the recombination on solution

Usage

```
ees1.recombination(solution, mates)
```

Arguments

solution	The Problem solution
mates	The mixed parents

ees1.selection	<i>ees.selection</i>
----------------	----------------------

Description

Select the elements with best fitness but accept uphill moves with probability 'kkappa'.

Usage

```
ees1.selection(s0, s1, kkappa)
```

Arguments

s0	The current best solution set
s1	The new solution
kkappa	The selection pressure

elog.debug	<i>elog.debug</i>
------------	-------------------

Description

Wrapper for logging debug messages.

Usage

```
elog.debug(...)
```

Arguments

...	Variable number of arguments including a format string.
-----	---

elog.error	<i>elog.error</i>
------------	-------------------

Description

Wrapper for logging error messages.

Usage

```
elog.error(...)
```

Arguments

...	Variable number of arguments including a format string.
-----	---

<code>elog.info</code>	<i>elog.info</i>
------------------------	------------------

Description

Wrapper for logging info messages.

Usage

```
elog.info(...)
```

Arguments

... Variable number of arguments including a format string.

<code>elog.level</code>	<i>elog.level</i>
-------------------------	-------------------

Description

Configure the current log level

Usage

```
elog.level(level = NULL)
```

Arguments

level The log level (ERROR|WARN|INFO|DEBUG)

Value

The log level

enforceBounds	<i>enforceBounds</i>
---------------	----------------------

Description

Checks if parameters fall within upper and lower bounds

Usage

```
enforceBounds(particles, factors)
```

Arguments

particles	The particle set
factors	the defined range for objective function parameters

Value

The particle inside the valid limits

es.evaluate	<i>es.evaluate</i>
-------------	--------------------

Description

For each element in solution 's' evaluate the respective fitness.

Usage

```
es.evaluate(f, s, enforce = TRUE)
```

Arguments

f	A reference to an instance of objective function
s	The set of solutions
enforce	If true the values are enforced to fall within provided range

Value

The solution ordered by its fitness.

Estimates-class	<i>Estimates</i>
-----------------	------------------

Description

A simple class for encapsulating the return of metaheuristic methods

extremize	<i>extremize</i>
-----------	------------------

Description

Entry point for optimization functions

Usage

```
extremize(type, objective, options = NULL)
```

Arguments

type	The optimization method (aco,pso,saa,sda)
objective	An instance of ObjectiveFunction (or subclass) class ObjectiveFunction
options	An appropriate instance from a subclass of Options class

Examples

```
## Not run:
f<- PlainFunction$new(f0.rosenbrock2)

f$Parameter(name="x1",min=-100,max=100)
f$Parameter(name="x2",min=-100,max=100)

extremize("pso", f)

## End(Not run)
```

f0.ackley	<i>f0.ackley</i>
-----------	------------------

Description

The ackley function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$. Domain $x_i \in [-32.768, 32.768]$, for all $i = 1, \dots, d$

Usage

```
f0.ackley(...)
```

Arguments

... The variadic list of function variables.

Value

The function value

References

<https://www.sfu.ca/~ssurjano/ackley.html>

f0.ackley4	<i>f0.ackley4</i>
------------	-------------------

Description

The ackley function of four variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

```
f0.ackley4(x1, x2, x3, x4)
```

Arguments

x1	The first function variable
x2	The second function variable
x3	The third function variable
x4	The fourth function variable

Value

The function value

f0.adtn.rosenbrock2 *f0.adtn.rosenbrock2*

Description

Two variable Rosenbrock function with random additive noise.

Usage

```
f0.adtn.rosenbrock2(x1, x2)
```

Arguments

x1	Parameter 1
x2	Parameter 2

f0.bohachevsky *f0.bohachevsky*

Description

The Bohachevsky function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

```
f0.bohachevsky(...)
```

Arguments

... The variadic list of function variables.

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f0.bohachevsky4	<i>f0.bohachevsky4</i>
-----------------	------------------------

Description

The Bohachevsky function of four variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

f0.bohachevsky4(x1, x2, x3, x4)

Arguments

x1	The first function variable
x2	The second function variable
x3	The third function variable
x4	The fourth function variable

Value

The function value

f0.cigar	<i>f0.cigar</i>
----------	-----------------

Description

The Cigar function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

f0.cigar(...)

Arguments

...	The variadic list of function variables.
-----	--

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

 f0.cigar4

f0.cigar4

Description

The Cigar function of four variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in \{1, \dots, N\}$, $f(x) = 0$.

Usage

f0.cigar4(x1, x2, x3, x4)

Arguments

x1	The first function variable
x2	The second function variable
x3	The third function variable
x4	The fourth function variable

Value

The function value

f0.griewank

f0.griewank

Description

The griewank function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in \{1, \dots, N\}$, $f(x) = 0$.

Usage

f0.griewank(...)

Arguments

...	The variadic list of function variables.
-----	--

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f0.griewank4	<i>f0.griewank4</i>
--------------	---------------------

Description

The griewank function of four variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

```
f0.griewank4(x1, x2, x3, x4)
```

Arguments

x1	The first function variable
x2	The second function variable
x3	The third function variable
x4	The fourth function variable

Value

The function value

f0.nlnn.rosenbrock2	<i>f0.nlnn.rosenbrock2</i>
---------------------	----------------------------

Description

Two variable Rosenbrock function with random additive noise.

Usage

```
f0.nlnn.rosenbrock2(x1, x2)
```

Arguments

x1	Parameter 1
x2	Parameter 2

f0.periodtuningpp *Period tuning for Predator-Prey base*

Description

This function is an example on how EvoPER can be used for estimating the parameter values in order to produce oscillations with the desired period. It is not intended to be used directly, the provided wrappers should be instead.

Usage

```
f0.periodtuningpp(x1, x2, x3, x4, period)
```

Arguments

x1	The growth rate of prey
x2	The decay rate of predator
x3	The predating effect on prey
x4	The predating effecto on predator
period	The desired oscilation period

Value

The solution fitness cost

f0.periodtuningpp12 *Period tuning of 12 time units for Predator-Prey*

Description

This function is an example on how EvoPER can be used for estimating the parameter values in order to produce oscillations with the desired period.

Usage

```
f0.periodtuningpp12(x1, x2, x3, x4)
```

Arguments

x1	The growth rate of prey
x2	The decay rate of predator
x3	The predating effect on prey
x4	The predating effecto on predator

Value

The solution fitness cost

Examples

```
## Not run:
rm(list=ls())
set.seed(-27262565)
f<- PlainFunction$new(f0.periodtuningpp12)
f$Parameter(name="x1",min=0.5,max=2)
f$Parameter(name="x2",min=0.5,max=2)
f$Parameter(name="x3",min=0.5,max=2)
f$Parameter(name="x4",min=0.5,max=2)
extremize("pso", f)

## End(Not run)
```

f0.periodtuningpp24 *Period tuning of 24 time units for Predator-Prey*

Description

This function is an example on how EvoPER can be used for estimating the parameter values in order to produce oscillations with the desired period.

Usage

```
f0.periodtuningpp24(x1, x2, x3, x4)
```

Arguments

x1	The growth rate of prey
x2	The decay rate of predator
x3	The predating effect on prey
x4	The predating effecto on predator

Value

The solution fitness cost

Examples

```
## Not run:
rm(list=ls())
set.seed(-27262565)
f<- PlainFunction$new(f0.periodtuningpp24)
f$Parameter(name="x1",min=0.5,max=2)
f$Parameter(name="x2",min=0.5,max=2)
f$Parameter(name="x3",min=0.5,max=2)
f$Parameter(name="x4",min=0.5,max=2)
extremize("pso", f)

## End(Not run)
```

f0.periodtuningpp48 *Period tuning of 48 time units for Predator-Prey*

Description

This function is an example on how EvoPER can be used for estimating the parameter values in order to produce oscillations with the desired period.

Usage

```
f0.periodtuningpp48(x1, x2, x3, x4)
```

Arguments

x1	The growth rate of prey
x2	The decay rate of predator
x3	The predating effect on prey
x4	The predating effecto on predator

Value

The solution fitness cost

Examples

```
## Not run:
rm(list=ls())
set.seed(-27262565)
f<- PlainFunction$new(f0.periodtuningpp24)
f$Parameter(name="x1",min=0.5,max=2)
f$Parameter(name="x2",min=0.5,max=2)
f$Parameter(name="x3",min=0.5,max=2)
f$Parameter(name="x4",min=0.5,max=2)
extremize("pso", f)
```



```
## End(Not run)
```

f0.periodtuningpp72 *Period tuning of 72 time units for Predator-Prey*

Description

This function is an example on how EvoPER can be used for estimating the parameter values in order to produce oscillations with the desired period.

Usage

```
f0.periodtuningpp72(x1, x2, x3, x4)
```

Arguments

x1	The growth rate of prey
x2	The decay rate of predator
x3	The predating effect on prey
x4	The predating effecto on predator

Value

The solution fitness cost

Examples

```
## Not run:  
rm(list=ls())  
set.seed(-27262565)  
f<- PlainFunction$new(f0.periodtuningpp24)  
f$Parameter(name="x1",min=0.5,max=2)  
f$Parameter(name="x2",min=0.5,max=2)  
f$Parameter(name="x3",min=0.5,max=2)  
f$Parameter(name="x4",min=0.5,max=2)  
extremize("pso", f)  
  
## End(Not run)
```

`f0.rosenbrock2`*f0.rosenbrock2*

Description

Two variable Rosenbrock function, where $f(1,1) = 0$

Usage

```
f0.rosenbrock2(x1, x2)
```

Arguments

x1 Parameter 1

x2 Parameter 2

`f0.rosenbrock4`*f0.rosenbrock4*

Description

The rosenbrock function of 4 variables for testing optimization methods. The global optima for the function is given by $x_i = 1$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

```
f0.rosenbrock4(x1, x2, x3, x4)
```

Arguments

x1 The first function variable

x2 The second function variable

x3 The third function variable

x4 The fourth function variable

Value

The function value

f0.rosenbrockn	<i>f0.rosenbrockn</i>
----------------	-----------------------

Description

The rosenbrock function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 1$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

```
f0.rosenbrockn(...)
```

Arguments

... The variadic list of function variables.

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f0.schaffer	<i>f0.schaffer</i>
-------------	--------------------

Description

The schaffer function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

```
f0.schaffer(...)
```

Arguments

... The variadic list of function variables.

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f0.schaffer4	<i>f0.schaffer4</i>
--------------	---------------------

Description

The Schaffer function of four variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

`f0.schaffer4(x1, x2, x3, x4)`

Arguments

x1	The first function variable
x2	The second function variable
x3	The third function variable
x4	The fourth function variable

Value

The function value

f0.schwefel	<i>f0.schwefel</i>
-------------	--------------------

Description

The schwefel function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 420.96874636$, for all $i \in 1 \dots N$, $f(x) = 0$. The range of x_i is $[-500, 500]$

Usage

`f0.schwefel(...)`

Arguments

...	The variadic list of function variables.
-----	--

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f0.schwefel4	<i>f0.schwefel4</i>
--------------	---------------------

Description

The schwefel function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 420.96874636$, for all $i \in 1 \dots N$, $f(x) = 0$. The range of x_i is $[-500, 500]$

Usage

```
f0.schwefel4(x1, x2, x3, x4)
```

Arguments

x1	The first function variable
x2	The second function variable
x3	The third function variable
x4	The fourth function variable

Value

The function value

f0.test	<i>f0.test</i>
---------	----------------

Description

Simple test function $f(1,2,3,4) = 0$

Usage

```
f0.test(x1, x2, x3, x4)
```

Arguments

x1	Parameter 1
x2	Parameter 2
x3	Parameter 3
x4	Parameter 4

`f1.ackley`*f1.ackley*

Description

The ackley function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$. Domain $x_i \in [-32.768, 32.768]$, for all $i = 1, \dots, d$

Usage`f1.ackley(x)`**Arguments**

`x` The vector of function parameters

Value

The function value

References

<https://www.sfu.ca/~ssurjano/ackley.html>

`f1.adtn.rosenbrock2`*f1.adtn.rosenbrock2*

Description

Two variable Rosenbrock function with random additive noise.

Usage`f1.adtn.rosenbrock2(x)`**Arguments**

`x` Parameter vector

f1.bohachevsky	<i>f1.bohachevsky</i>
----------------	-----------------------

Description

The Bohachevsky function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in \{1 \dots N\}$, $f(x) = 0$.

Usage

f1.bohachevsky(x)

Arguments

x The vector of function parameters

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f1.cigar	<i>f1.cigar</i>
----------	-----------------

Description

The Cigar function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in \{1 \dots N\}$, $f(x) = 0$.

Usage

f1.cigar(x)

Arguments

x The vector of function variables.

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f1.griewank

f1.griewank

Description

The griewank function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

```
f1.griewank(x)
```

Arguments

x The vector of function parameters

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f1.nlnn.rosenbrock2

f1.nlnn.rosenbrock2

Description

Two variable Rosenbrock function with random additive noise.

Usage

```
f1.nlnn.rosenbrock2(x)
```

Arguments

x Parameter vector

f1.rosenbrock2	<i>f1.rosenbrock2</i>
----------------	-----------------------

Description

Two variable Rosenbrock function, where $f(c(1,1)) = 0$

Usage

f1.rosenbrock2(x)

Arguments

x Parameter vector

f1.rosenbrockn	<i>f1.rosenbrockn</i>
----------------	-----------------------

Description

The rosenbrock function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 1$, for all $i \in \{1, \dots, N\}$, $f(x) = 0$.

Usage

f1.rosenbrockn(x)

Arguments

x The vector of function parameters

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f1.schaffer	<i>f1.schaffer</i>
-------------	--------------------

Description

The schaffer function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 0$, for all $i \in 1 \dots N$, $f(x) = 0$.

Usage

f1.schaffer(x)

Arguments

x The vector of function parameters

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f1.schwefel	<i>f1.schwefel</i>
-------------	--------------------

Description

The schwefel function of N variables for testing optimization methods. The global optima for the function is given by $x_i = 420.96874636$, for all $i \in 1 \dots N$, $f(x) = 0$. The range of x_i is $[-500, 500]$

Usage

f1.schwefel(x)

Arguments

x The vector of function variables.

Value

The function value

References

<http://deap.gel.ulaval.ca/doc/dev/api/benchmarks.html>

f1.test	<i>f1.test</i>
---------	----------------

Description

Simple test function $f(c(1,2,3,4)) = 0$

Usage

```
f1.test(x)
```

Arguments

x	Parameter vector
---	------------------

fixdfcolumns	<i>fixdfcolumns</i>
--------------	---------------------

Description

Coerce dataframe columns to a specific type.

Usage

```
fixdfcolumns(df, cols = c(), skip = TRUE, type = as.numeric)
```

Arguments

df	The data frame.
cols	The dataframe columns to be skipped or included.
skip	If TRUE the column names in 'cols' are skipped. When FALSE logic is inverted.
type	The type for which data frame columns must be converted.

Value

The data frame with converted column types.

generateSolution	<i>generateSolution</i>
------------------	-------------------------

Description

Generates a problema solution using discrete leves

Usage

```
generateSolution(parameters, size)
```

Arguments

parameters	The Objective Function parameter list
size	The solution size

Value

The solution set

getFitness	<i>getFitness</i>
------------	-------------------

Description

Given a set S of N solutions created with sortSolution, this function returns the solution component for the best solution.

Usage

```
getFitness(S, i = NULL)
```

Arguments

S	The solution set
i	The fitness index, if null return the whole column.

Value

The selected fitness entry

getSolution *getSolution*

Description

Given a set S of N solutions created with *sortSolution*, this function returns the solution component.
A solutions is a set of solutions and their associated fitness

Usage

getSolution(S)

Arguments

S The solution set

Value

The solution set

gm.mean *gm.mean*

Description

Simple implementation for geometric mean

Usage

gm.mean(x)

Arguments

x data

Value

geometric mean for data

*gm.sd**gm.sd*

Description

Simple implementation for geometric standard deviation

Usage

```
gm.sd(x, mu = NULL)
```

Arguments

<code>x</code>	data
<code>mu</code>	The geometric mean. If not provided it is calculated.

Value

geometric standard deviation for data

*histplothelper**histplothelper*

Description

Simple helper for plotting histograms

Usage

```
histplothelper(d, x, title = NULL)
```

Arguments

<code>d</code>	A data frame.
<code>x</code>	A string with the dataframe column name for histogram
<code>title</code>	The plot title

Value

A ggplot2 plot object

initSolution	<i>initSolution</i>
--------------	---------------------

Description

Creates the initial Solution population taking into account the lower and upper bounds of provided experiment factors.

Usage

```
initSolution(parameters, N = 20, sampling = "mcs")
```

Arguments

parameters	The Objective Function parameter list
N	The size of Solution population
sampling	The population sampling scheme, namely <mcs lhs ffs> standing respectively for montecarlo sampling, latin hypercube sampling and full factorial sampling

Value

A random set of solutions

lowerBound	<i>lowerBound</i>
------------	-------------------

Description

Checks if parameters is greater than the lower bounds

Usage

```
lowerBound(particles, factors)
```

Arguments

particles	The particle set
factors	the defined range for objective function parameters

Value

The particle greater than or equal to lower limit

Magnitude	<i>Magnitude</i>
-----------	------------------

Description

Calculates the magnitude order for a given value

Usage

Magnitude(v)

Arguments

v	The numerical value
---	---------------------

Value

The magnitude order

naiveperiod	<i>naiveperiod</i>
-------------	--------------------

Description

A naive approach for finding the period in a series of data points

Usage

naiveperiod(d)

Arguments

d	The data to search period
---	---------------------------

Value

A list with the average period and amplitude

NetLogoFunction-class	<i>NetLogoFunction</i>
-----------------------	------------------------

Description

NetLogoFunction class

NLWrapper.FindJar *NLWrapper.FindJar*

Description

Search for the netlogo jar file on the provided path

Usage

NLWrapper.FindJar(path)

Arguments

path The base path for searching

Value

The path for NetLogo jar file

NLWrapper.GetParameter
NLWrapper.GetParameter

Description

Gets the value of a model parameter

Usage

NLWrapper.GetParameter(obj, name)

Arguments

obj The object returned by [NLWrapper.Model](#)
name The parameter name string or the collection of parameter names

Value

The parameter values

Examples

```
## Not run:
rm(list=ls())
p<- "C:/Program Files/NetLogo 6.0.4/app"
m<- file.path(nlpath, "models", "Sample Models", "Biology", "Wolf Sheep Predation.nlogo")
o<- NLWrapper.Model(p, m)
v<- NLWrapper.GetParameter(o, c("initial-number-sheep"))

or

v<- NLWrapper.GetParameter(o, c("initial-number-sheep","initial-number-wolves"))

## End(Not run)
```

NLWrapper.Model

NLWrapper.Model

Description

This wrapper prepares the environment and instantiates the model

Usage

```
NLWrapper.Model(netlogodir, modelfile, dataset, maxtime)
```

Arguments

netlogodir	The base path of NetLogo installation
modelfile	The absolute path for NetLogo model file
dataset	The names of model variables
maxtime	The total number of iterations

Examples

```
## Not run:
rm(list=ls())
p<- "C:/Program Files/NetLogo 6.0.4/app"
output<- c("count sheep", "count wolves")
m<- file.path(nlpath, "models", "Sample Models", "Biology", model, "Wolf Sheep Predation.nlogo")
o<- NLWrapper.Model(p, m, output, 150)

## End(Not run)
```

NLWrapper.Run	<i>NLWrapper.Run</i>
---------------	----------------------

Description

Executes a NetLogo Model using rNetLogo

Usage

```
NLWrapper.Run(obj, r = 1, seed = c())
```

Arguments

obj	The object returned by NLWrapper.Model
r	The number of replications
seed	The collection of random seeds

NLWrapper.RunExperiment	<i>NLWrapper.RunExperiment</i>
-------------------------	--------------------------------

Description

Executes a NetLogo Model using rNetLogo

Usage

```
NLWrapper.RunExperiment(obj, r = 1, design, FUN)
```

Arguments

obj	The object returned by NLWrapper.Model
r	The number of replications
design	The desing matrix holding parameter sampling
FUN	THE calibration function.

Value

A list containing the the parameters, the calibration functio output and the whole resultset

Examples

```
## Not run:
rm(list=ls())
objectivefn<- function(params, results) { 0 }

f<- AddFactor(name="initial-number-sheep",min=100,max=250)
f<- AddFactor(factors=f, name="initial-number-wolves",min=50,max=150)
f<- AddFactor(factors=f, name="grass-regrowth-time",min=30,max=100)
f<- AddFactor(factors=f, name="sheep-gain-from-food",min=1,max=50)
f<- AddFactor(factors=f, name="wolf-gain-from-food",min=1,max=100)
f<- AddFactor(factors=f, name="sheep-reproduce",min=1,max=20)
f<- AddFactor(factors=f, name="wolf-reproduce",min=1,max=20)

design<- AoE.LatinHypercube(factors=f)

p<- "C:/Program Files/NetLogo 6.0.4/app"
m<- file.path(p, "models", "Sample Models", "Biology", "Wolf Sheep Predation.nlogo")
output<- c("count sheep", "count wolves")
o<- NLWrapper.Model(p, m, output, 150)
v<- RunExperiment(o, r=1, design, objectivefn)
NLWrapper.Shutdown(o)

## End(Not run)
```

NLWrapper.SetParameter

NLWrapper.SetParameter

Description

Set parameter values

Usage

```
NLWrapper.SetParameter(obj, parameters)
```

Arguments

obj	The object returned by NLWrapper.Model
parameters	The data frame containing the parameters

Examples

```
## Not run:
rm(list=ls())
p<- "C:/Program Files/NetLogo 6.0.4/app"
m<- file.path(nlpath, "models", "Sample Models", "Biology", "Wolf Sheep Predation.nlogo")
o<- NLWrapper.Model(p, m)
```

```
## End(Not run)
```

```
NLWrapper.SetRandomSeed  
NLWrapper.SetRandomSeed
```

Description

Configures the random seed

Usage

```
NLWrapper.SetRandomSeed(obj, seed)
```

Arguments

obj	The object returned by NLWrapper.Model
seed	The new random seed

```
NLWrapper.Shutdown NLWrapper.Shutdown
```

Description

This wrapper terminates RNetLogo execution environment

Usage

```
NLWrapper.Shutdown(obj)
```

Arguments

obj	The object returned by NLWrapper.Model
-----	--

ObjectiveFunction-class
ObjectiveFunction class

Description

The base class for optimization functions.

Fields

object The raw output of objective function
 objective The objective function
 parameters The parameter list for objective function
 value The results from objective function

Options-class *Options*

Description

The base class for the options for the optimization metaheuristics

Fields

type The configuration type
 neighborhood The neighborhood function for population methods
 discrete Flag indicating that and specific algorithm is discrete or continuous
 nlevelz Default value for generating parameter levels when range is provided, default value is 5
 container The object holding the configuration otions

OptionsACOR-class *OptionsACOR*

Description

Options for ACOR method

OptionsEES1-class	<i>OptionsEES1</i>
-------------------	--------------------

Description

Options for EvoPER Evolutionary Strategy 1

OptionsEES2-class	<i>OptionsEES2</i>
-------------------	--------------------

Description

Options for Serial Dilutions method

Fields

dilutions The desired dilutions

OptionsFactory	<i>OptionsFactory</i>
----------------	-----------------------

Description

Instantiate the Options class required for the specific metaheuristic method.

Usage

```
OptionsFactory(type, v = NULL)
```

Arguments

type	The metaheuristic method
v	The options object

Value

Options object

OptionsPSO-class	<i>OptionsPSO</i>
------------------	-------------------

Description

Options for PSO optimization metaheuristic

OptionsSAA-class	<i>OptionsSAA</i>
------------------	-------------------

Description

Options for SAA method

Fields

temperature The temperature decay function

OptionsTS-class	<i>OptionsTS</i>
-----------------	------------------

Description

Options for Tabu search optimization metaheuristic

paramconverter	<i>paramconverter</i>
----------------	-----------------------

Description

Convert parameter from continuous to discrete and vice-versa if needed

Usage

```
paramconverter(parameters, discrete, levelz = 5)
```

Arguments

parameters	The current parameter set
discrete	The desired parameter type
levelz	When discrete is true the number of levels to be generated

Value

The parameter collection casted to desired mode

partSolutionSpace	<i>partSolutionSpace</i>
-------------------	--------------------------

Description

Creates the initial Solution population taking into account the lower and upper bounds of provided experiment factors. This method works by dividing the solution space into partitions of size 'd' and then creating a full factorial combination of partitions.

Usage

```
partSolutionSpace(parameters, d = 4)
```

Arguments

parameters	The Objective Function parameter list
d	The partition size. Default value 4.

Value

A set of solutions

PlainFunction-class	<i>PlainFunction</i>
---------------------	----------------------

Description

PlainFunction Class

pop.first	<i>pop.first</i>
-----------	------------------

Description

pop an element

Usage

```
pop.first(x)
```

Arguments

x	The element collection
---	------------------------

Value

The first element added to list FIFO

pop.last

pop.last

Description

pop an element

Usage

pop.last(x)

Arguments

x The element collection

Value

The last element added to list LIFO

predatorprey

predatorprey

Description

The solver for Lotka-Volterra differential equation.

Usage

predatorprey(x1, x2, x3, x4)

Arguments

x1 The growth rate of prey
x2 The decay rate of predator
x3 The predating effect on prey
x4 The predating effecto on predator

Value

The ODE solution

predatorprey.plot0 *predatorprey.plot0*

Description

Generate a plot for the predator-prey ODE output.

Usage

```
predatorprey.plot0(x1, x2, x3, x4, title = NULL)
```

Arguments

x1	The growth rate of prey
x2	The decay rate of predator
x3	The predating effect on prey
x4	The predating effect on predator
title	The optional plot title. May be omitted.

Value

An ggplot2 object

Examples

```
## Not run:  
predatorprey.plot0(1.351888, 1.439185, 1.337083, 0.9079049)  
  
## End(Not run)
```

predatorprey.plot1 *predatorprey.plot1*

Description

Simple wrapper for 'predatorprey.plot0' accepting the parameters as a list.

Usage

```
predatorprey.plot1(x, title = NULL)
```

Arguments

x	A list containing the values of predator/prey parameters c1, c2, c3 and c4 denoting respectively the growth rate of prey, the decay rate of predator, the predating effect on prey and the predating effect on predator
title	The optional plot title. May be omitted.

Value

An ggplot2 object

Examples

```
## Not run:
rm(list=ls())
predatorprey.plot1(v$getBest()[1:4])

## End(Not run)
```

pso.best

pso.best

Description

Search for the best particle solution which minimize the objective function.

Usage

```
pso.best(objective, particles)
```

Arguments

objective	The results of evaluating the objective function
particles	The particles tested

Value

The best particle

pso.chi

pso.chi

Description

Implementation of constriction coefficient

Usage

pso.chi(phi1, phi2)

Arguments

phi1 Acceleration coefficient toward the previous best

phi2 Acceleration coefficient toward the global best

Value

The calculated constriction coefficient

pso.lbest

pso.lbest

Description

Finds the lbest for the particle 'i' using the topology function given by the topology parameter.

Usage

pso.lbest(i, pbest, topology)

Arguments

i The particle position

pbest The pbest particle collection

topology The desired topology function

Value

The lbes for i th particle

`pso.neighborhood.K2` *pso.neighborhood.K2*

Description

The neighborhood function for a simple linear topology where every particle has $k = 2$ neighbors

Usage

`pso.neighborhood.K2(i, n)`

Arguments

<code>i</code>	The particle position
<code>n</code>	the size of particle population

`pso.neighborhood.K4` *pso.neighborhood.K4*

Description

The von neumann neighborhood function for a lattice-based topology where every particle has $k = 4$ neighbors

Usage

`pso.neighborhood.K4(i, n)`

Arguments

<code>i</code>	The particle position
<code>n</code>	the size of particle population

pso.neighborhood.KN *pso.neighborhood.KN*

Description

Simple helper method for 'gbest' neighborhood

Usage

pso.neighborhood.KN(i, n)

Arguments

i	The particle position
n	the size of particle population

pso.printbest *pso.printbest*

Description

Shows the best particle of each of simulated generations

Usage

pso.printbest(objective, particles, generation, title)

Arguments

objective	An instance of ObjectiveFunction (or subclass) class ObjectiveFunction
particles	The current particle population
generation	The current generation
title	Some informational text to be shown

pso.Velocity *pso.velocity*

Description

Calculates the PSO Velocity

Usage

pso.Velocity(W = 1, Vi, phi1, phi2, Pi, Pg, Xi)

Arguments

W	Weight (Inertia weight or constriction coefficient)
Vi	Current Velocity vector
phi1	Acceleration coefficient toward the previous best
phi2	Acceleration coefficient toward the global best
Pi	Personal best
Pg	Neighborhood best
Xi	Particle vector

Value

Updated velocity

push *push*

Description

push an element

Usage

push(x, v)

Arguments

x	The collection of elements
v	The value to be pushed

Value

The collection of elements

random.wheel	<i>random.wheel</i>
--------------	---------------------

Description

A simple random seed generator

Usage

random.wheel()

Value

A random number for seeding

RepastFunction-class	<i>RepastFunction</i>
----------------------	-----------------------

Description

RepastFunction class

saa.bolt	<i>saa.bolt</i>
----------	-----------------

Description

Temperature function boltzmann

Usage

saa.bolt(t_0 , k)

Arguments

t_0	The current temperature
k	The annealing value

Value

The new temperature

saa.neighborhood *saa.neighborhood*

Description

Generates neighbor solutions for simulated annealing

Usage

saa.neighborhood(f, S, d, n)

Arguments

f	An instance of ObjectiveFunction (or subclass) class ObjectiveFunction
S	The current solution to find a neighbor
d	The distance from current solution S distance = (max - min) * d
n	The number of parameters to be perturbed

Value

The neighbor of solution S

saa.neighborhood1 *saa.neighborhood1*

Description

Generates neighbor solutions perturbing one parameter from current solution S picked randomly.

Usage

saa.neighborhood1(f, S, d)

Arguments

f	An instance of ObjectiveFunction (or subclass) class ObjectiveFunction
S	The current solution to find a neighbor
d	The distance from current solution S distance = (max - min) * d

Value

The neighbor of solution of S

saa.neighborhoodH *saa.neighborhoodH*

Description

Generates neighbor solutions perturbing half parameters from current solution S.

Usage

```
saa.neighborhoodH(f, S, d)
```

Arguments

f	An instance of ObjectiveFunction (or subclass) class ObjectiveFunction
S	The current solution to find a neighbor
d	The distance from current solution S distance = (max - min) * d

Value

The neighbor of solution of S

saa.neighborhoodN *saa.neighborhoodN*

Description

Generates neighbor solutions perturbing all parameters from current solution S.

Usage

```
saa.neighborhoodN(f, S, d)
```

Arguments

f	An instance of ObjectiveFunction (or subclass) class ObjectiveFunction
S	The current solution to find a neighbor
d	The distance from current solution S distance = (max - min) * d

Value

The neighbor of solution of S

saa.tbyk

saa.tbyk

Description

Temperature function t/k

Usage

saa.tbyk(t_0 , k)

Arguments

t_0 The current temperature

k The annealing value

Value

The new temperature

saa.tcte

saa.tcte

Description

Temperature function $cte * t_0$

Usage

saa.tcte(t_0 , k)

Arguments

t_0 The current temperature

k The annealing value

Value

The new temperature

saa.texp	<i>saa.texp</i>
----------	-----------------

Description

Temperature function exponential

Usage

```
saa.texp(t0, k)
```

Arguments

t0	The current temperature
k	The annealing value

Value

The new temperature

scatterplotlohelper	<i>scatterplotlohelper</i>
---------------------	----------------------------

Description

Simple helper for plotting 3d scatterplots

Usage

```
scatterplotlohelper(d, x, y, z, title = NULL)
```

Arguments

d	A data frame.
x	A string with the dataframe column name for x axis
y	A string with the dataframe column name for y axis
z	A string with the dataframe column name for z axis
title	The optional plot title. May be omitted.

Value

A scatter3D plot

searchrow	<i>searchrow</i>
-----------	------------------

Description

Search for a value value on a matrix

Usage

```
searchrow(ddata, value)
```

Arguments

ddata	The matrix containing the dataset
value	The value to search for

Value

Boolean TRUE for those indexes matching value

show.comp1	<i>show.comp1</i>
------------	-------------------

Description

Generates a barplot comparing the number of evaluations for algorithms ("saa", "pso", "acor", "ees1").

Usage

```
show.comp1(mydata, what, title = NULL)
```

Arguments

mydata	The data generated with 'summarize.comp1'
what	The name of variable to plot on 'y' axis
title	the plot title

Examples

```
## Not run:
p.a<- show.comp1(d.cigar4,"evals","(a) Cigar function")
p.b<- show.comp1(d.schaffer4,"evals","(b) Schafer function")
p.c<- show.comp1(d.griewank4,"evals","(c) Griewank function")
p.d<- show.comp1(d.bohachevsky4,"evals","(d) Bohachevsky function")

## End(Not run)
```

slope

slope

Description

Simple function for calculate the slope on the ith element position

Usage

slope(x, y, i)

Arguments

x	The x vector
y	The y vector
i	The position

Value

The slope

slopes

slopes

Description

Calculate all slopes for the discrete x,y series

Usage

slopes(x, y)

Arguments

x	The x vector
y	The y vector

Value

A vector with all slopes

sortSolution *sortSolution*

Description

Sort solution by its respective fitness

Usage

```
sortSolution(s, f)
```

Arguments

s	Problem solution
f	The function evaluation for s

summarize.comp1 *summarize.comp1*

Description

Provides as summary with averaged values of experimental setup

Usage

```
summarize.comp1(mydata)
```

Arguments

mydata	The data frame generated with 'compare.algorithms1'
--------	---

Value

The summarized data

tabu.getNeighbors	<i>tabu.getNeighbors</i>
-------------------	--------------------------

Description

create neighbor solutions

Usage

```
tabu.getNeighbors(tabu, parameters, solution, size)
```

Arguments

tabu	The tabu list
parameters	The parameter set
solution	The current solution
size	The neighborhood size

Value

The neighbor for solution

tabu.istabu	<i>tabu.istabu</i>
-------------	--------------------

Description

Check whether a solution is present on tabulist

Usage

```
tabu.istabu(tabulist, solution)
```

Arguments

tabulist	The matrix of tabu solutions
solution	The solution value to be checked

Value

Boolean TRUE tabulist contains the solution

upperBound	<i>upperBound</i>
------------	-------------------

Description

Checks if parameters is below the upper bounds

Usage

```
upperBound(particles, factors)
```

Arguments

particles	The particle set
factors	the defined range for objective function parameters

Value

The particle inside the valid upper bound

xmeanci1	<i>xmeanci1</i>
----------	-----------------

Description

Calculates confidence interval of mean for provided data with desired confidence level. This functions uses bootstrap resampling scheme for estimating the CI.

Usage

```
xmeanci1(x, alpha = 0.95)
```

Arguments

x	The data set for which CI will be calculated
alpha	The confidence level. The default value is 0.95 (95%)

Value

The confidence interval for the mean calculated using 'boot.ci'

`xmeanci2`*xmeanci2*

Description

Calculates confidence interval of mean for provided data with desired confidence level.

Usage

```
xmeanci2(x, alpha = 0.95)
```

Arguments

<code>x</code>	The data set for which CI will be calculated
<code>alpha</code>	The confidence level. The default value is 0.95 (95%)

Value

The confidence interval for the mean

`xyplothelper`*xyplothelper*

Description

Simple helper for plotting xy dispersion points.

Usage

```
xyplothelper(d, x, y, title = NULL)
```

Arguments

<code>d</code>	A data frame.
<code>x</code>	A string with the dataframe column name for x axis
<code>y</code>	A string with the dataframe column name for y axis
<code>title</code>	The optional plot title. May be omitted.

Value

A ggplot2 plot object

Index

abm.acor, 5
abm.ees1, 5
abm.ees2, 6
abm.pso, 7
abm.saa, 7
abm.tabu, 9
acor.archive, 9
acor.F, 10
acor.lthgaussian, 11
acor.N, 11
acor.probabilities, 12
acor.S, 12
acor.sigma, 13
acor.updateants, 13
acor.W, 14
acor.weigth, 14
assert, 15

bestFitness, 15
bestSolution, 16

cbuf, 16
compare.algorithms1, 17
contourplotheper, 17

ees1.challenge, 18
ees1.explore, 18
ees1.mating, 19
ees1.mating1, 19
ees1.mutation, 20
ees1.recombination, 20
ees1.selection, 21
elog.debug, 21
elog.error, 21
elog.info, 22
elog.level, 22
enforceBounds, 23
es.evaluate, 23
Estimates (Estimates-class), 24
Estimates-class, 24

extremize, 24

f0.ackley, 25
f0.ackley4, 25
f0.adtn.rosenbrock2, 26
f0.bohachevsky, 26
f0.bohachevsky4, 27
f0.cigar, 27
f0.cigar4, 28
f0.griewank, 28
f0.griewank4, 29
f0.nlnn.rosenbrock2, 29
f0.periodtuningpp, 30
f0.periodtuningpp12, 30
f0.periodtuningpp24, 31
f0.periodtuningpp48, 32
f0.periodtuningpp72, 33
f0.rosenbrock2, 34
f0.rosenbrock4, 34
f0.rosenbrockn, 35
f0.schaffer, 35
f0.schaffer4, 36
f0.schwefel, 36
f0.schwefel4, 37
f0.test, 37
f1.ackley, 38
f1.adtn.rosenbrock2, 38
f1.bohachevsky, 39
f1.cigar, 39
f1.griewank, 40
f1.nlnn.rosenbrock2, 40
f1.rosenbrock2, 41
f1.rosenbrockn, 41
f1.schaffer, 42
f1.schwefel, 42
f1.test, 43
fixdfcolumns, 43

generateSolution, 44
getFitness, 44

- getSolution, 45
- gm.mean, 45
- gm.sd, 46
- histplotheper, 46
- initSolution, 47
- lowerBound, 47
- Magnitude, 48
- naiveperiod, 48
- NetLogoFunction
 - (NetLogoFunction-class), 48
- NetLogoFunction-class, 48
- NLWrapper.FindJar, 49
- NLWrapper.GetParameter, 49
- NLWrapper.Model, 49, 50, 51–53
- NLWrapper.Run, 51
- NLWrapper.RunExperiment, 51
- NLWrapper.SetParameter, 52
- NLWrapper.SetRandomSeed, 53
- NLWrapper.Shutdown, 53
- ObjectiveFunction, 5–9, 24, 63, 66, 67
- ObjectiveFunction
 - (ObjectiveFunction-class), 54
- ObjectiveFunction-class, 54
- Options, 5–9, 24
- Options (Options-class), 54
- Options-class, 54
- OptionsACOR (OptionsACOR-class), 54
- OptionsACOR-class, 54
- OptionsEES1 (OptionsEES1-class), 55
- OptionsEES1-class, 55
- OptionsEES2 (OptionsEES2-class), 55
- OptionsEES2-class, 55
- OptionsFactory, 55
- OptionsPSO (OptionsPSO-class), 56
- OptionsPSO-class, 56
- OptionsSAA (OptionsSAA-class), 56
- OptionsSAA-class, 56
- OptionsTS (OptionsTS-class), 56
- OptionsTS-class, 56
- paramconverter, 56
- partSolutionSpace, 57
- PlainFunction (PlainFunction-class), 57
- PlainFunction-class, 57
- pop.first, 57
- pop.last, 58
- predatorprey, 58
- predatorprey.plot0, 59
- predatorprey.plot1, 59
- pso.best, 60
- pso.chi, 61
- pso.lbest, 61
- pso.neighborhood.K2, 62
- pso.neighborhood.K4, 62
- pso.neighborhood.KN, 63
- pso.printbest, 63
- pso.Velocity, 64
- push, 64
- random.wheel, 65
- RepastFunction (RepastFunction-class), 65
- RepastFunction-class, 65
- saa.bolt, 65
- saa.neighborhood, 66
- saa.neighborhood1, 66
- saa.neighborhoodH, 67
- saa.neighborhoodN, 67
- saa.tbyk, 68
- saa.tcte, 68
- saa.texp, 69
- scatterplotlotheper, 69
- searchrow, 70
- show.comp1, 70
- slope, 71
- slopes, 71
- sortSolution, 72
- summarize.comp1, 72
- tabu.getNeighbors, 73
- tabu.istabu, 73
- upperBound, 74
- xmeanci1, 74
- xmeanci2, 75
- xyplotheper, 75