Package: optimg (via r-universe)

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Description This package is a general purpose tool for helping users to implement gradient descent methods for function optimization. Currently, the Steepest 2-Groups Gradient Descent and the Adaptive Moment Estimation (Adam) are the methods implemented. Other methods will be implemented in the future.	
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Description

General-purpose optimization based on gradient descent algorithms. It is greatly inspired on the stats::optim function, aiming at increased usability and adaptability for optim's users.

Usage

```
optimg(par, fn, gr=NULL, ..., method=c("STGD","ADAM"),
Interval=1e-6, maxit=100, tol=1e-8, full=F, verbose=F)
```

Arguments

par	Initial values for the parameters to be optimized over.
fn	A function to be minimized. It should return a scalar result.
gr	A function to return the gradient. Should include "Interval" as an argument, even if not used. If it is NULL, a finite-difference approximation will be used.
	Further arguments to be passed to fn and gr.
method	The method to be used. See 'Details'.
Interval	The epsilon difference to be used by gr.
maxit	The maximum number of iterations. Defaults to 100.
tol	Relative convergence tolerance. The algorithm stops if it is unable to reduce the value by a factor of reltol * (abs(val) + reltol) at a step. Defaults to 1e-8.
full	Boolean argument for returning all results, or only the basics (see Value). Defaults to FALSE.
verbose	Boolean argument for printing update status. Defaults to FALSE.

Details

As with the optim function, ... must be matched exactly. Also as with optim, optimg performs minimization. All the methods implemented are the "steepest" variation of the original methods. This means that tuning parameters are optimized at each step automatically. This makes the algorithms slower, but also more effective, especially for complex models.

The default method (unless the length of par is equal to 1; in which case, the default is "ADAM") is an implementation of the Steepest 2-Group Gradient Descent algorithm. This algorithm is a variation of the Steepest Gradient Descent method which calculates different step sizes for two groups of gradients: those within a standard deviation (below or above), and those beyond a standard deviation (below or above).

Method "ADAM" is the Adaptive Moment Estimation method. This method computes adaptive learning rates for each parameter. Adam stores both exponentially decaying average of past squared gradients, as well as measures of momentum.

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Value

If full = FALSE, a list with components:

par The best set of parameters found.

value The value of fn corresponding to par.

counts A scalar giving the number of iterations before convergence is reached.

convergence An integer code. 0 indicates successful completion. 1 indicates that the iteration

limit maxit had been reached.

If full = TRUE, apart from the above, components regarding the changes of par, gradient, value, and auxiliary parameters will also be returned.

Examples

```
### Fit a simple linear regression with RMSE as cost function
require(optimg)
# Predictor
x <- seq(-3,3,len=100)
# Criterion
y \leftarrow rnorm(100, 2 + \{1.2*x\}, 1)
# RMSE cost function
fn <- function(par, X) {</pre>
  mu \leftarrow par[1] + \{par[2] * X\}
  rmse <- sqrt(mean({y-mu}^2))</pre>
  return(rmse)
}
# Compare optimization methods
optim(c(0,0),fn,X=x,method="Nelder-Mead")
optim(c(0,0),fn,X=x,method="BFGS")
optim(c(0,0),fn,X=x,method="CG")
optim(c(0,0),fn,X=x,method="L-BFGS-B")
optimg(c(0,0),fn,X=x,method="ADAM")
optimg(c(0,0),fn,X=x,method="STGD")
```

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