Package ‘n1qn1’

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Title Port of the 'Scilab' 'n1qn1' and 'qnbd' Modules for (Un)constrained BFGS Optimization

Version 6.0.1-7

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Description Provides 'Scilab' 'n1qn1', or Quasi-Newton BFGS `qn` without constraints and 'qnbd' or Quasi-Newton BFGS with constraints. This takes more memory than traditional L-BFGS. The n1qn1 routine is useful since it allows prespecification of a Hessian. If the Hessian is near enough the truth in optimization it can speed up the optimization problem. Both algorithms are described in the 'Scilab' optimization documentation located at <https://www.scilab.org/sites/default/files/optimization_in_scilab.pdf>.

URL https://github.com/nlmixrdevelopment/n1qn1

BugReports https://github.com/nlmixrdevelopment/n1qn1/issues

Depends R (>= 3.2)

Imports Rcpp (>= 0.12.3)

Suggests testthat, covr

License CeCILL-2

NeedsCompilation yes

LinkingTo RcppArmadillo (>= 0.5.600.2.0), Rcpp (>= 0.12.3)

LazyData true

RoxygenNote 7.0.2

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**Description**

This is an R port of the n1qn1 optimization procedure in scilab.

**Usage**

```r
n1qn1(call_eval, call_grad, vars, environment = parent.frame(1), ...,
  epsilon = .Machine$double.eps, max_iterations = 100, nsim = 100,
  imp = 0, invisible = NULL, zm = NULL, restart = FALSE,
  assign = FALSE, print.functions = FALSE)
```

**Arguments**

- `call_eval`: Objective function
- `call_grad`: Gradient Function
- `vars`: Initial starting point for line search
- `environment`: Environment where `call_eval/call_grad` are evaluated.
- `...`: Ignored additional parameters.
- `epsilon`: Precision of estimate
- `max_iterations`: Number of iterations
- `nsim`: Number of function evaluations
- `imp`: Verbosity of messages.
- `invisible`: boolean to control if the output of the minimizer is suppressed.
- `zm`: Prior Hessian (in compressed format; This format is output in c.hess).
- `restart`: Is this an estimation restart?
- `assign`: Assign hessian to c.hess in environment environment? (Default FALSE)
- `print.functions`: Boolean to control if the function value and parameter estimates are echoed every time a function is called.
Value

The return value is a list with the following elements:

- **value** The value at the minimized function.
- **par** The parameter value that minimized the function.
- **H** The estimated Hessian at the final parameter estimate.
- **c.hess** Compressed Hessian for saving curvature.
- **n.fn** Number of function evaluations
- **n.gr** Number of gradient evaluations

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Examples

```r
## Rosenbrock's banana function
n=3; p=100

fr = function(x)
{
  f=1.0
  for(i in 2:n) {
    f=f+p*(x[i]-x[i-1]**2)**2+(1.0-x[i])**2
  }
  f
}

grr = function(x)
{
  g = double(n)
  g[1]=-4.0*p*(x[2]-x[1]**2)*x[1]
  if(n>2) {
    for(i in 2:(n-1)) {
      g[i]=2.0*p*(x[i]-x[i-1]**2)-4.0*p*(x[i+1]-x[i]**2)*x[i]-2.0*(1.0-x[i])
    }
  }
  g[n]=2.0*p*(x[n]-x[n-1]**2)-2.0*(1.0-x[n])
  g
}

x = c(1.02,1.02,1.02)
eps=1e-3
n=length(x); niter=100L; nsim=100L; imp=3L;
nzm=as.integer(n*(n+13L)/2L)
zm=double(nzm)

(op1 <- n1qn1(fr, grr, x, imp=3))
```
## Note there are 40 function calls and 40 gradient calls in the above optimization

## Now assume we know something about the Hessian:

```r
c.hess <- c(797.861115,
            -393.801473,
            -2.795134,
            991.271179,
            -395.382900,
            200.024349)
c.hess <- c(c.hess, rep(0, 24 - length(c.hess)))

(op2 <- n1qn1(fr, gr, x, imp=3, zm=c.hess))
```

## Note with this knowledge, there were only 29 function/gradient calls

```r
(op3 <- n1qn1(fr, gr, x, imp=3, zm=op1$c.hess))
```

## The number of function evaluations is still reduced because the Hessian
## is closer to what it should be than the initial guess.

## With certain optimization procedures this can be helpful in reducing the
## Optimization time.

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### qnbd

**qnbd optimization**

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**Description**

This is an R port of the qnbd which is a BFGS-B optimization procedure in scilab. (R has L-BFGS-B in optim).

**Usage**

```r
qnbd(par, fn, gr, lower = -Inf, upper = Inf,
     environment = parent.frame(1),
     zero = sqrt(.Machine$double.eps/7e-07), maxFn = 10000L,
     maxIt = 10000L, epsf = sqrt(.Machine$double.eps),
     epsg = sqrt(.Machine$double.eps), epsx = sqrt(.Machine$double.eps),
     print.functions = FALSE)
```

**Arguments**

- **par**: Initial parameter estimate
- **fn**: Function
- **gr**: Gradient
- **lower**: Lower Bound for optimization
Upper bound for optimization

Environment where call_eval/call_grad are evaluated.

Tolerance for Zero

Maximum function evaluations

Maximum iterations

Function eps for exiting

Gradient eps for exiting

Parameter eps for exiting

Boolean to control if the function value and parameter estimates are echoed every time a function is called.

Examples

```r
## Rosenbrock's banana function
n=3; p=100

fr = function(x)
{
  f=1.0
  for(i in 2:n) {
    f=f+p*(x[i]-x[i-1]**2)**2+(1.0-x[i])**2
  }
  f
}

grr = function(x)
{
  g = double(n)
  g[1]=-4.0*p*(x[2]-x[1]**2)*x[1]
  if(n>2) {
    for(i in 2:(n-1)) {
      g[i]=2.0*p*(x[i]-x[i-1]**2)-4.0*p*(x[i+1]-x[i]**2)*x[i]-2.0*(1.0-x[i])
    }
  }
  g[n]=2.0*p*(x[n]-x[n-1]**2)-2.0*(1.0-x[n])
  g
}

x = c(1.02,1.02,1.02)

op1 <- qnbd(x, fr, grr)
```
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