

Package: n1qn1 (via r-universe)

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Title Port of the 'Scilab' 'n1qn1' Module for Unconstrained BFGS Optimization

Version 6.0.1-12

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Description Provides 'Scilab' 'n1qn1'. 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. The algorithm is described in the 'Scilab' optimization documentation located at <https://www.scilab.org/sites/default/files/optimization_in_scilab.pdf>. This version uses manually modified code from 'f2c' to make this a C only binary.

URL <https://github.com/nlmixr2/n1qn1c>

BugReports <https://github.com/nlmixr2/n1qn1c/issues>

Depends R (>= 3.2)

Imports Rcpp (>= 0.12.3)

Suggests testthat, covr

License CeCILL-2

Biarch true

NeedsCompilation yes

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

Encoding UTF-8

RoxygenNote 7.3.2

Repository <https://nlmixr2.r-universe.dev>

RemoteUrl <https://github.com/nlmixr2/n1qn1c>

RemoteRef HEAD

RemoteSha 1a6bbf0634a16b660d5e336b8a646bddea1ad16d

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.n1qn1ptr	<i>This gives the function pointers in the n1qn1 library</i>
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Description

Using this will allow C-level linking by function pointers instead of abi.

Usage

```
.n1qn1ptr()
```

Value

list of pointers to the n1qn1 functions

Author(s)

Matthew L. Fidler

Examples

```
.n1qn1ptr()
```

n1qn1	<i>n1qn1 optimization</i>
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Description

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

Usage

```
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

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

Author(s)

C. Lemarechal, Stephen L. Campbell, Jean-Philippe Chancelier, Ramine Nikoukhah, Wenping Wang & Matthew L. Fidler

Examples

```

## 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:
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, grr, x, imp=3, zm=c.hess))

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

(op3 <- n1qn1(fr, grr, 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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