

Refactoring the nls() function in R

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2021-7-17

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Abstract

This article reports the particular activities of our Google Summer of Code project “Improvements to nls()” that relate to R code for that function. A companion document “Variety in Nonlinear Least Squares Codes” presents an overview of methods for the problem.

The existing function and its shortcomings

`nls()` is the tool in base R (the distributed software package from <https://cran.r-project.org>) for estimating nonlinear statistical models. The function was developed mainly in the 1980s and 1990s by Doug Bates et al., initially for S (https://en.wikipedia.org/wiki/S_%28programming_language%29). The ideas spring primarily from the book by Bates and Watts (1988).

With a long history, it is not surprising that code has become untidy and overly patched. It is, to our mind,

essentially unmaintainable. Moreover, it has deficiencies that can and should be fixed. Let us review some of the issues. We will then propose corrective actions, some of which we have carried out.

Issue: Convergence and termination tests

`nls()`

The default settings of `nls` generally fail on artificial “zero-residual” data problems.

The `nls` function uses a relative-offset convergence criterion that compares the numerical imprecision at the current parameter estimates to the residual sum-of-squares. This performs well on data of the form

$$y = f(x, \theta) + \text{eps}$$

(with $\text{var}(\text{eps}) > 0$). It fails to indicate convergence on data of the form

$$y = f(x, \theta)$$

because the criterion amounts to comparing two components of the round-off error. To avoid a zero-divide in computing the convergence testing value, a positive constant `scaleOffset` should be added to the denominator sum-of-squares; it is set in control; this does not yet apply to algorithm = “port.”

Example of a small-residual problem

```
rm(list=ls())
t <- -10:10
y <- 100/(1+.1*exp(-0.51*t))
lform<-y~a/(1+b*exp(-c*t))
ldata<-data.frame(t=t, y=y)
plot(t,y)
lstartbad<-c(a=1, b=1, c=1)
lstart2<-c(a=100, b=10, c=1)
nlsr::nlxb(lform, data=ldata, start=lstart2)
nls(lform, data=ldata, start=lstart2, trace=TRUE)
# Fix with scaleOffset
nls(lform, data=ldata, start=lstart2, trace=TRUE, control=list(scaleOffset=1.0))
sessionInfo()
```

Edited output of running this function follows:

```
> rm(list=ls())
> t <- -10:10
> y <- 100/(1+.1*exp(-0.51*t))
> lform<-y~a/(1+b*exp(-c*t))
> ldata<-data.frame(t=t, y=y)
> plot(t,y)
> lstart2<-c(a=100, b=10, c=1)
> nlsr::nlxb(lform, data=ldata, start=lstart2)
nlsr object: x
residual sumsquares = 1.007e-19 on 21 observations
      after 13      Jacobian and 19 function evaluations
      name      coeff      SE      tstat      pval      gradient      JSingval
a          100      2.679e-11  3.732e+12  1.863e-216  -6.425e-11      626.6
b           0.1      3.78e-13  2.646e+11  9.125e-196  -3.393e-08      112.3
```

[illegible]

```

Error in nls(lform, data = ldata, start = lstart2, trace = TRUE) :
  number of iterations exceeded maximum of 50

> nls(lform, data=ldata, start=lstart2, trace=TRUE, control=list(scaleOffset=1.0))
40346.      (1.08e+00): par = (100 10 1)
11622.      (2.91e+00): par = (101.47 0.49449 0.71685)
5638.0      (9.23e+00): par = (102.23 0.38062 0.52792)
642.08      (5.17e+00): par = (102.16 0.22422 0.41935)
97.712      (2.31e+00): par = (100.7 0.14774 0.45239)
22.250      (1.11e+00): par = (99.803 0.093868 0.50492)
0.025789    (3.79e-02): par = (100.01 0.10017 0.50916)
6.0571e-08  (5.80e-05): par = (100 0.1 0.51)
4.7017e-19  (1.62e-10): par = (100 0.1 0.51)
Nonlinear regression model
  model: y ~ a/(1 + b * exp(-c * t))
  data: ldata
      a      b      c
100.00  0.10  0.51
residual sum-of-squares: 4.7e-19

Number of iterations to convergence: 8
Achieved convergence tolerance: 1.62e-10

> sessionInfo()
R version 4.1.0 (2021-05-18)
Platform: x86_64-pc-linux-gnu (64-bit)
Running under: Linux Mint 20.2

```

Issue: more general termination tests

roffset smallstest jacobians, residuals

Issue: Failure when Jacobian is computationally singular

This is the infamous “singular gradient” termination. In some cases it is due to failure of the simple finite difference approximation of the Jacobian in the `numericDeriv()` function that is a part of `nls()`. `nlsr` has used analytic derivatives, and we can import this functionality to the `nls()` code.

However, the more common source of the issue is that the Jacobian is very close to singular for some values of the model parameters. In such cases we need to find an alternative algorithm to the Gauss-Newton iteration of `nls()`. The most common work-around is the Levenberg-Marquardt stabilization (Marquardt (1963), Levenberg1944, jn77ima). Versions of this have been implemented in packages `minpack.lm` and `nlsr`.

Issue: Jacobian computation

analytic in `nlsr`, `numericDeriv()`, others

Issue: Subsetting

`nls()` accepts an argument `subset`. Unfortunately, this acts through the mediation of `model.frame` and is not clearly obvious in the source code files `/src/library/stats/R/nls.R` and `/src/library/stats/src/nls.C`.

- implementation via weights
- implementation via `model.frame`
- other concerns

Issue: na.action

`na.action` is an argument to the `nls()` function, but it does not appear in obviously in the source code ...

Issue: model

`model` is an argument to the `nls()` function, but it does not appear in obviously in the source code ...

Issue: sources of data

`nls()` can be called without specifying the `data` argument. In this case, it will search in the available environments (i.e., workspaces) for suitable data objects. We do NOT like this approach. R allows users to leave many objects in the default (`.GlobalEnv`) workspace. Moreover, users have to actively suppress saving this workspace (`.RData`) on exit, and any such file in the path when R is launched will be loaded.

Nevertheless, to provide compatible behaviour with `nls()`, we will need to ensure that equivalent behaviour is guaranteed.

Issue: missing start vector and self-starting models

Nonlinear estimation algorithms are almost all iterative and need a set of starting parameters. `nls()` offers a special class of modeling formulae called **selfStart** models. There are a number of these in base R (??list) and others in R packages such as (?? list or examples). Unfortunately, the structure of the programming of these is such that the methods by which initial parameters are computed is entangled with the particularities of the `nls()` code. Though there is a `getInitial()` function, this is not easy to use to simply compute the initial parameter estimates.

?? weird output in testing

```
> ls()
[1] "ldata"      "lform"      "lstart2"    "lstartbad" "t"          "y"
> apar <- getInitial(y~SSlogis(t, Asym, xmid, scal), data=ldata)
Error in nls(y ~ 1/(1 + exp((xmid - x)/scal)), data = xy, start = list(xmid = aux[[1L]], :
  number of iterations exceeded maximum of 50
```

In the event that a selfStart model is not available, `nls()` sets all the starting parameters to 1. This is, in our view, tolerable, but could possibly be improved by using a set of values that are slightly different e.g., in the case of a model

$$y \sim a * \exp(-b * x) + c * \exp(-d * x)$$

it would be useful to have b and d values different so the Jacobian is not singular. Thus some sort of sequence like 1.0, 1.1, 1.2, 1.3 for the four parameters might be better and it can be provided quite simply instead of all 1s.

Issue: results of running nls()

The output of `nls()` is an object of class “nls” which has the following structure:

?? put in an example and document it.

Concerns with content of the nls object

The nls object contains some elements that are awkward to produce by other algorithms. Moreover, some information that would be useful is not presented obviously (??examples - convergence/termination info, `Jsingvals`)

Issue: code structure

The `nls()` code is structured in a way that inhibits both maintenance and improvement. In particular, the iterative setup is such that introduction of Marquardt stabilization is not easily available.

To obtain performance, a lot of the code is in C with consequent calls and returns that complicate the code. Over time, R has become much more efficient on modern computers, and the need to use compiled C and Fortran is less critical. Moreover, the burden for maintenance could be much reduced by moving code entirely to R.

Issue: code documentation

`setPars()` – explain weaknesses. Only used by `profile.nls()`

The paucity of documentation is exacerbated by the mixed R/C/Fortran code base.

Goals of our effort

What we want to accomplish.

Code rationalization and documentation

- all in R
- try to obtain cleaner structure
- try to explain what we do, either in comments or separate maintainer documentation

Provide tests

- to ensure workalike properties
- to test individual functions to ensure they work across the range of calling mechanisms
- for “silly” inputs to try to see if exceptions are caught

Output of the project

?? see Working Doc etc.

References

Bates, D. M., and D. G. Watts. 1988. *Nonlinear Regression Analysis and Its Applications*. Wiley.

Marquardt, Donald W. 1963. “An Algorithm for Least-Squares Estimation of Nonlinear Parameters.” *SIAM Journal on Applied Mathematics* 11 (2): 431–41.