# Refactoring the nls() function in R

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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

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

(with var(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 \sim a/(1+b*exp(-c*t))
ldata<-data.frame(t=t, y=y)</pre>
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))
> 1 form <-y^a/(1+b*exp(-c*t))
> ldata<-data.frame(t=t, y=y)</pre>
> 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
                   coeff
                                  SE
                                            tstat
                                                                               JSingval
  name
                                                       pval
                                                                  gradient
                      100
                              2.679e-11 3.732e+12 1.863e-216
                                                                 -6.425e-11
                                                                                    626.6
а
b
                      0.1
                               3.78e-13 2.646e+11 9.125e-196
                                                                 -3.393e-08
                                                                                    112.3
```

```
6.9e-13 7.391e+11 8.494e-204
                    0.51
                                                                 1.503e-08
                                                                                  2.791
# Note that this has succeeded. The test in nlsr recognizes small residual problems.
> nls(lform, data=ldata, start=lstart2, trace=TRUE)
          (1.08e+00): par = (100 10 1)
11622.
          (2.93e+00): par = (101.47 \ 0.49449 \ 0.71685)
5638.0
         (1.08e+01): par = (102.23 \ 0.38062 \ 0.52792)
642.08
         (1.04e+01): par = (102.16 \ 0.22422 \ 0.41935)
          (1.79e+01): par = (100.7 \ 0.14774 \ 0.45239)
97.712
22.250
          (1.78e+02): par = (99.803 \ 0.093868 \ 0.50492)
0.025789 (1.33e+03): par = (100.01 0.10017 0.50916)
6.0571e-08 (7.96e+05): par = (100 0.1 0.51)
4.7017e-19 (1.86e+04): par = (100 0.1 0.51)
1.2440e-27 (5.71e-01): par = (100 0.1 0.51)
```

```
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)
            (9.23e+00): par = (102.23 \ 0.38062 \ 0.52792)
5638.0
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)
            (5.80e-05): par = (100 \ 0.1 \ 0.51)
6.0571e-08
4.7017e-19 (1.62e-10): par = (100 0.1 0.51)
Nonlinear regression model
  model: y \sim a/(1 + b * exp(-c * t))
   data: ldata
            h
     a
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 smallsstest 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

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

$$y\tilde{a} * exp(-b * x) + c * exp(-d * x)$$

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.

Following is an email to Dr. Heather Turner from John Nash.

I'm afraid that I don't know the purpose of the recursive call either. I know that I wrote the code to for the response, covariates, etc., but I don't recall anything like a recursive call being necessary.

If the R sources were in a git repository I might try to use `git blame` to find out when and by whom to but they are in an SVN repository, I think, and I haven't used it for a long, long time.

I don't think I will be of much help. My R skills have atrophied to the point where I wouldn't even kn exploring what is happening in the first call as opposed to the recursive call.

On Tue, Jun 29, 2021 at 11:50 AM John Nash <Nashjc@uottawa.ca <mailto:Nashjc@uottawa.ca>> wrote:

Thanks.

```
https://gitlab.com/nashjc/improvenls/-/blob/master/Croucher-expandednlsnoc.R <a href="https://gitlab.com/nashjc/improvenls/-/blob/master/Croucher-expandednlsnoc.R">https://gitlab.com/nashjc/improvenls/-/blob/master/Croucher-expandednlsnoc.R</a>
```

This has the test problem and the expanded code. Around line 367 is where we are scratching our heads. The function code (from nlsModel()) is in the commented lines below the call. This is

```
# > setPars
# function(newPars) {
# setPars(newPars)
# resid <<- .swts * (lhs - (rhs <<- getRHS())) # envir = thisEnv {2 x}
# dev <<- sum(resid^2) # envir = thisEnv
# if(length(gr <- attr(rhs, "gradient")) == 1L) gr <- c(gr)
# QR <<- qr(.swts * gr) # envir = thisEnv
# (QR$rank < min(dim(QR$qr))) # to catch the singular gradient matrix
# }</pre>
```

I'm anticipating that we will be able to set up a (possibly inefficient) code with documentation that will be easier to follow and test, then gradually figure out how to make it more efficient.

```
The equivalent from minpack.lm is
setPars = function(newPars) {
    setPars(newPars)
```

In both there is the recursive call, which must have a purpose I don't understand.

```
Cheers, JN
On 2021-06-29 12:33 p.m., Douglas Bates wrote:
> *Attention : courriel externe | external email*
> Thanks for contacting me, John. Can you point me to a file in the gitlab.com <a href="http://gitlab.com">http://gitlab.com</a>
<http://gitlab.com <http://gitlab.com>> repository that
> contains the definition of setPars?
> (By the way, it is probably best to use the email address dmbates@gmail.com <mailto:dmbates@gmail
<mailto:dmbates@gmail.com <mailto:dmbates@gmail.com>> for me. If email
> goes to bates@stat.wisc.edu <mailto:bates@stat.wisc.edu> <mailto:bates@stat.wisc.edu <mailto:bate
it should get forwarded to the gmail.com <a href="http://gmail.com">http://gmail.com">http://gmail.com</a>
> address but sometimes gmail decides that such mail looks suspicious and puts it in the spam folde
> a long time I used bates@stat.wisc.edu <mailto:bates@stat.wisc.edu <mailto:bates@stat.wisc.edu
<mailto:bates@stat.wisc.edu>> as my "From:" address because it had been my address
> since the 80's but even gmail got suspicious of mail from that address that did not appear to ori
wisc.edu <a href="http://wisc.edu">http://wisc.edu">
> <http://wisc.edu <http://wisc.edu>> domain.)
```

# Goals of our effort

What we want to accomplish.

#### Code rationalization and documentation

We want

- to provide a packaged version of nls() (call it nlsalt) all in R that matches the version in base R and what is packaged in nlspkg as described in the "PkgFromRbase" document
- try to obtain cleaner structure for the overall nls() infrastructure. By this we mean a re-factoring of the routines so they are better suited to maintenance of both the existing nls() methods and features as well as the new features we would like to add.
- try to explain what we do, either in comments or separate maintainer documentation. Since we are complaining about the lack of explanatory material for the current code, we feel it incumbent on us to provide such material for our own work, and if possible for the existing code.

#### Provide tests

- to ensure work-alike 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

### A test runner program

??? this is for the quick testing of sets of problems — documented in TestsDoc.

# Output of the project

?? see Working Doc etc.

- formal reports
- informal reports
- problem sets
- code and documentation

# 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.