

nlmixr: an open-source package for pharmacometric modelling in R

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Aims

nlmixr is an open-source R package under development that builds on both RxODE¹, an R package for simulation of nonlinear mixed effect models using ordinary differential equations (ODEs), and the nlme² package in R, for parameter estimation in nonlinear mixed effect models. nlmixr greatly expands the utility of nlme by providing an efficient and versatile way to specify pharmacometric models and dosing scenarios, with rapid execution due to compilation in C++. NONMEM^{®3} with first-order conditional estimation with interaction was used as a comparator to test nlmixr.

Methods

Richly sampled profiles were simulated for 4 different dose levels (10, 30, 60 and 120 mg) of 30 subjects each as single dose (over 72h), multiple dose (4 daily doses), single and multiple dose combined, and steady state dosing, for a range of test models: 1- and 2-compartment disposition, with and without 1st order absorption, with either linear or Michaelis-Menten (MM) clearance (MM without steady state dosing). This provided a total of 42 test cases. All inter-individual variabilities (IIVs) were set at 30%, residual error at 20% and overlapping PK parameters were the same for all models. A similar set of models was previously used to compare NONMEM and Monolix⁴. Estimates of population parameters, standard errors for fixed-effect parameters, and run times were compared both for closed-form solutions and using ODEs.

Additionally, a sparse data estimation situation was investigated where 500 datasets of 600 subjects each (150 per dose) were generated consisting of 4 random time point samples in 24 hours per subject, using a first-order absorption, 1-compartment disposition, linear elimination model.

Example code

```
library(nlmixr)
datr<-read.csv("BOLUS_1CPT.csv", header=TRUE)
datr$EVID<-ifelse(datr$EVID==1,101,datr$EVID)
specs<-list(fixed=1CL+1V-1,random=pdDiag(form=1CL+1V-1),start=c(1CL=1.6,1V=4.5))

#Closed-form:
fit<-nlme_lin_cmt(datr,par_model=specs,ncmt=1,oral=FALSE,weight=varPower(fixed=c(1)))

#ODE:
ode <- "d/dt(centr) = -(CL/V)*centr;"
mypar <- function(1CL, 1V )
{CL = exp(1CL)
 V = exp(1V)}
fitODE<-nlme_ode(datr,model=ode,par_model=specs,par_trans=mypar,response="centr",
response_scaler="V", weight=varPower(fixed=c(1)),
control=nlmeControl(pnlTol=1))
```

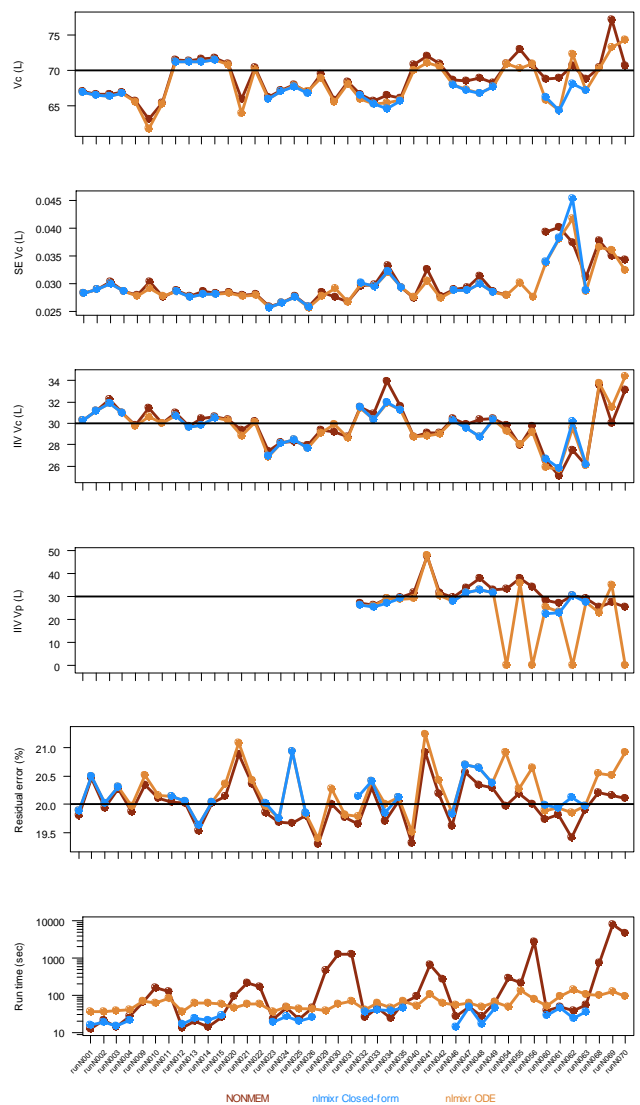


Figure 1. Theta, SE, and IIV estimates for Vc, IIV estimates for Vp, residual error, and log run times comparing NONMEM (red lines, closed form where possible), nlmixr using ODEs (orange lines) and closed-form nlmixr (blue lines). Horizontal lines: values used for simulation.

Results

Theta parameter estimates were comparable across estimation methods applied to rich data. Figure 1 provides results for central volume of distribution (Vc) as illustration because it is the single parameter present in all models. Standard error estimates were obtained for all nlmixr models, but not all NONMEM models. IIV estimates were regularly estimated close to 0% for ill-defined model parameters like peripheral volume (Vp) in nlmixr, whereas NONMEM provided estimates closer to the original simulation values. In comparison to NONMEM, nlmixr was always faster for ODEs (MM-models) and comparable for closed-form models.

The sparse data analyses indicated a bias in values for clearance (CL) and volume (Vc) both for NONMEM and for nlmixr, but bias was much more pronounced for nlmixr (see Figure 2). IIV for absorption constant (Ka) was estimated close to zero for 91.1% of the analyses for nlmixr and only 2.2% for NONMEM.

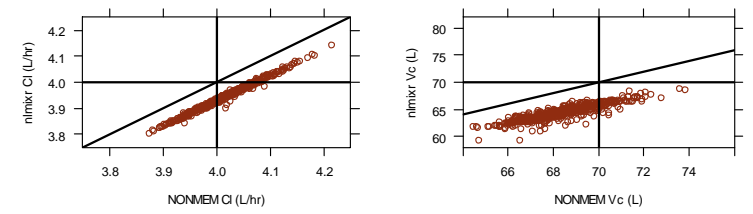


Figure 2 Population estimates for CL (left) and Vc (right) using NONMEM and nlmixr. Horizontal and vertical lines: simulated value. Diagonal line: line of identity.

Conclusions

These findings suggest that nlmixr provides a viable open-source parameter estimation procedure for nonlinear mixed effects pharmacometric models within the R environment. Implementation of alternatives to the nlme estimation routine may be warranted for analysis of sparse data.

References

¹Wang W et al. A Tutorial on RxODE. CPT:PSP (2016) 5, 3–10.
²Pinheiro J et al. (2016). nlme: Linear and Nonlinear Mixed Effects Models.
³Beal SL et al. 1989-2011. NONMEM Users Guides. Icon Development Solutions, USA.
⁴Laveille C et al. PAGE 17 (2008) Abstr 1356 [www.pagemeeting.org/?abstract=1356]

