

Computing Satterthwaite degrees of freedom for glmer models

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February 16, 2026

1 Introduction

The following is based on the SAS documentation at https://documentation.sas.com/doc/en/statug/15.2/statug_glimmix_details39.htm. Also critical was the paper “Fitting Linear Mixed-Effects Models Using lme4” by Bates et al (Journal of Statistical Software, 67(1), 1-48, 2015).

2 Methods

2.1 lmer models

2.1.1 The 1 degree of freedom case

Consider the t-statistic

$$t = \frac{\ell' \hat{\beta}}{\sqrt{\ell' C \ell}} \quad (1)$$

where ℓ is a $p \times 1$ matrix that defines the contrast of interest, $\hat{\beta}$ is the fixed effects coefficient vector estimate (of length p), and C is the variance-covariance matrix of the fixed effects. C depends on σ and the random effects parameters θ (of length q). We can compute the Satterthwaite degrees of freedom for this statistic as

$$\nu = \frac{2(\ell' C \ell)^2}{g' A g} \quad (2)$$

where A is the variance-covariance matrix of the random effects parameter estimates, $\hat{\theta}$, and g is the first derivative of $\ell' C \ell$ with respect to θ , evaluated at $\hat{\theta}$.

Given a lmer fitted object, `rslt`, we can extract

```

 $\hat{\theta}$  = getME(rslt,"theta")
 $\hat{\beta}$  = fixef(rslt)
 $C$  = vcov(rslt)
 $A$  = MASS::ginv(rslt@optinfo$derivs$Hessian/2)[1:q,1:q]

```

where q is the number of random effects parameters (not including σ). Note that this code gives the so-called “theta” formulation of the random effects parameters (instead of the variance formulation). The difficulty is getting g (or equivalently, $dC/d\theta$). To do this we use the `grad` function (important: use right-sided derivatives) from the `numDeriv` package by

calling

```
g = grad(func_lmer,theta,obj=rslt,lvec=lvec)}
```

The key function, `func_lmer`, returns an updated value of $\ell' C \ell$ for a given value of theta. It does this by extracting information from the fitted object (`rslt`) and recomputing C as a function of θ , following information provided in the Bates et al article (particularly critical was equation 54 and section 3.6). See the code for `func_lmer`.

2.1.2 The multi-degree of freedom case

Let ℓ be an $r \times p$ matrix defining the r contrasts of interest so that the hypothesis of interest is

$$H_0 : L\hat{\beta} = 0 \quad (3)$$

Define the F statistic as

$$F = \hat{\beta}' L' (LCL')^{-1} L \hat{\beta}$$

F is distributed as according to an F-distribution with r and df degrees of freedom where df is computed as follows.

Compute the spectral decomposition (R function `eigen`) of $LCL' = U'DU$. Define b_j as the j 'th row of UL (an $r \times p$ matrix). Then compute g_j , the gradient of $b_j C b_j'$ with respect to θ (so g_j is a vector of length q). We use the `jacobian` function from the `numDeriv` package (important: use right-sided derivatives), along with the `func_lmer` function to compute g_j .

Next, compute

$$\nu_j = \frac{2(D_j)^2}{g_j' A g_j}$$

for $j = 1 \dots r$, and let

$$E = \sum_{j=1}^r \frac{\nu_j}{\nu_j - 2} I(\nu_j > 2)$$

Then

$$df = \frac{E}{E - r}$$

if $E > r$ and 0 otherwise.

3 Discussion

Acknowledgements: This research was supported by NIH grant AI29168.