

ImerTest R-package for automated mixed ANOVA modelling

Alexandra Kuznetsova ¹ Rune H.B. Christensen ¹ Per Bruun Brockhoff ¹

¹DTU Compute, Statistical section, Technical University of Denmark

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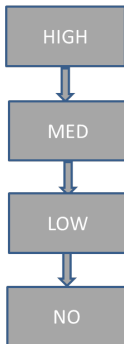
- Simple Mixed Effects models
- Automated analysis/elimination of mixed effects models using lmerTest
- Example 1. Sensory profile with multi-way product structure
- Example 2. External Preference Mapping with Consumer background

Linear mixed effects models: complexity for a user

 lme4

 lmerTest

 SensMixed



Use ImerTest to handle even more complex settings

- Unbalanced sensory profile data (e.g. missing observations)
- Incomplete consumer preference data
- 2- (or higher)way product structure in sensory profile data
- 2- (or higher)way product structure in consumer preference data (Conjoint)
- Extending Conjoint to include Consumer background/design factors/covariates
- Complex blocking, product replication, product batch structures in as well sensory as consumer preference data
- A mixed model approach for performing external preference mapping
- Extending mixed model external preference mapping to include product and consumer background/design factors/covariates

The lmerTest package - functions

- step (automated analysis of both random and fixed parts - finds the best simplest model)
- rand (analysis of the random part of a mixed model, LRT (likelihood ratio test))
- anova (Type I, II and III ANOVA tables with Satterthwaite's approximation to degrees of freedom)
- summary (t -tests for fixed effects with Satterthwaite's approximation to degrees of freedom)
- lsmeans, difflsmeans (least squares means and differences of least square means with confidence intervals) - post-hoc analysis
- plot - plots the post-hoc analysis

Types I - Type III ANOVA tables

Type I ANOVA

- produces sequential sums of squares
- the hypotheses are functions of cell counts
- depends on the order the effects are entered in the model

Type II ANOVA

- the hypotheses are functions of cell counts
- the hypotheses do NOT depend on the order the effects are entered

Type III ANOVA

- the hypotheses are NOT functions of cell counts
- the hypotheses do NOT depend on the order the effects are entered

Types I - III contrast matrices (SAS Technical Report R-101 et. al (1978))
are implemented in **ImerTest**

In balanced situations all types produce the same output

Example 1, Sensory profile with multi-way product structure

TVbo data (Bang and Olufsen, Sensometrics, As, 2006) *Sensory profile with multi-way product structure*

12 Products

- 3 TV sets
- 4 Pictures

2 replicates

8 Assessors

15 Sensory attributes

Example 2. Attach TVbo data

Tell **R** to use lmerTest and to use TVbo data

```
> library(lmerTest)
```

```
> data(TVbo)
```


Example 1. TVbo data

Summarize the data

```
> str(TVbo)
```

```
'data.frame':      192 obs. of  5 variables:
 $ Assessor      : Factor w/ 8 levels "1","2","3","4",...: 1 1 1
 $ TVset         : Factor w/ 3 levels "TV1","TV2","TV3": 3 2 1
 $ Repeat        : Factor w/ 2 levels "0","1": 1 1 1 2 2 2 1 1
 $ Picture       : Factor w/ 4 levels "1","2","3","4": 1 1 1 1
 $ Coloursaturation: num  10.4 9.9 7 9.8 10.6 7.5 7.1 9.9 5 10 ..
```

- **Response variable:** Sharpnessofmovement
- **Fixed effects:** TVset, Picture
- **Random effects:** Assessor, Repeat

```
> tv <- lme4::lmer(Sharpnessofmovement ~ TVset*Picture +  
+                 (1|Assessor) +(1|Assessor:TVset) +  
+                 (1|Assessor:Picture), data = TVbo)  
> anova(tv)
```

Analysis of Variance Table

	Df	Sum Sq	Mean Sq	F value
TVset	2	1.765	0.8825	0.2437
Picture	3	51.857	17.2857	4.7735
TVset:Picture	6	90.767	15.1279	4.1777

```

> library(lmerTest)
> tv <- lmer(Sharpnessofmovement ~ TVset*Picture +
+           (1|Assessor) +(1|Assessor:TVset) +
+           (1|Assessor:Picture), data = TVbo)
> anova(tv)

```

Analysis of Variance Table of type III with Satterthwaite approximation for degrees of freedom

	Sum Sq	Mean Sq	NumDF	DenDF	F.value	Pr(>F)
TVset	1.765	0.8825	2	14	0.2437	0.7869818
Picture	51.857	17.2857	3	21	4.7735	0.0108785 *
TVset:Picture	90.767	15.1279	6	138	4.1777	0.0006845 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Example 1. Specification of the full model. Step 1

Specification of the effects in **R**:

- **Fixed effects**: TVset*Picture will give two main effects and an interaction
- **Random effects**: (1|Assessor), 1|Assessor:TVset), (1|Assessor:TVset:Picture) 1|Repeat), 1|Repeat:TVset), 1|Repeat:Picture), 1|Repeat:TVset:Picture)

Specify the "full" model using lmer syntaxis

```
> M <- lmer(Sharpnessofmovement ~ TVset*Picture +  
+          (1|Assessor:TVset) + (1|Assessor:Picture) +  
+          (1|Assessor:Picture:TVset) + (1|Repeat) +  
+          (1|Repeat:Picture) +  
+          (1|Repeat:TVset) + (1|Repeat:TVset:Picture) +  
+          (1|Assessor) , data=TVbo)
```

Example 1. the step function

Run the `step` function and plot the post-hoc: performs **Step 2** and **Step 3** of the automated analysis

```
> s_TV <- step(M)
```

```
> plot(s_TV)
```

Example 1. Elimination process of random and fixed effects. Step 2 and Step 3

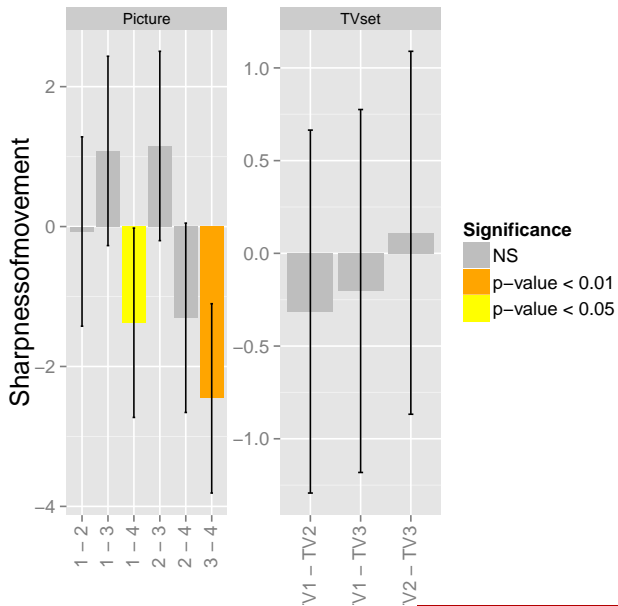
	Chi.sq	Chi.DF	elim.num	p.value
Assessor:Picture:TVset	0.00	1	1	1.000
Repeat:Picture	0.00	1	2	1.000
Repeat	0.00	1	3	1.000
Repeat:TVset	0.00	1	4	1.000
Repeat:TVset:Picture	0.00	1	5	1.000
Assessor:TVset	2.79	1	kept	0.095
Assessor:Picture	12.35	1	kept	<0.001
Assessor	7.47	1	kept	0.006

Table: Automated elimination process of random effects of model M

	DenDF	F.value	elim.num	Pr(>F)
TVset	14	0.24	kept	0.787
Picture	21	4.77	kept	0.011
TVset:Picture	138	4.18	kept	<0.001

Table: Automated elimination process of fixed effects of model M

Example 1. Plot post-hoc



Example 2, External Preference Mapping with consumer background

carrots data (Brockhoff, DTU course 02429) *External Preference Mapping with consumer background*

12 carrots varieties

103 Consumers

Consumer background info

- Homesize (2)

Product background info

- Sensory profile PC scores 1 and 2 (sens1, sens2)

Example 2. carrots data

Summarize the data

```
str(carrots)
```

```
'data.frame':      1236 obs. of  5 variables:
 $ Consumer   : Factor w/ 103 levels "168","169","171",...: 1 1 1 1
 $ Homesize   : Factor w/  2 levels "1","3": 2 2 2 2 2 2 2 2 2 2 ..
 $ Preference : int    4 5 4 7 5 6 5 6 5 6 ...
 $ sens1      : num   6.683 6.441 -1.948 0.481 2.181 ...
 $ sens2      : num  -2.54 3.54 -4.29 -1.33 -4.67 ...
```

- **Response variable:** Preference
- **Fixed effects:** sens1, sens2, Homesize
- **Random effects:** Consumer, product (*For External Preference mapping we consider product as random effect!*)

Example 2. Specification of the model

Specification of the effects in **R**:

- **Fixed effects**: sens1 sens2 Homesize + interactions
- **Random effects**: (1 + sens1 + sens2 | product), (1 | Consumer)

Specify the "full" model using lmer syntaxis

```
M <- lmer(Preference ~ sens2*sens1*Homesize +  
          (1 | product) + (1 + sens1 + sens2 | Consumer),  
          data = carrots)
```

$$(b_0, b_1, b_2) \sim N\left(0, \begin{pmatrix} \sigma_0^2 & \sigma_{01} & \sigma_{02} \\ \sigma_{01} & \sigma_1^2 & \sigma_{12} \\ \sigma_{02} & \sigma_{12} & \sigma_2^2 \end{pmatrix}\right), \quad c \sim N(0, \sigma_{\text{product}}^2), \quad \epsilon_{ijk} \sim N(0, \sigma^2) \quad (1)$$

Example 2. the step function

Run the step function: performs **Step 2** and **Step 3** of the automated analysis

```
t <- step(M)
```

Example 2. Elimination process of random effects. Step 2

	Chi.sq	Chi.DF	elim.num	p.value
sens1:Consumer	2.08	3	1	0.556
product	17.23	1	kept	<0.001
sens2:Consumer	8.06	2	kept	0.018

Table: Automated elimination process of random effects of model M

Example 2. Elimination process of fixed effects. Step 3

	DenDF	F.value	elim.num	Pr(>F)
sens2:sens1:Homesize	1015	1.33	1	0.249
sens2:sens1	7	0.11	2	0.744
sens1:Homesize	1016	0.18	3	0.673
sens1	8	0.52	4	0.489
sens2:Homesize	101	1.04	5	0.311
sens2	12	17.48	kept	0.001
Homesize	100	5.63	kept	0.020

Table: Automated elimination process of fixed effects of model M

Example 2. summary

Reduced model:

$$y_{ijkl} = b_{0j} + \beta_0 + \alpha_k + (b_{2j} + \beta_2)\text{sens2}_{ij} + c_k + \epsilon_{ijkl} \quad (2)$$

$$(b_0, b_1, b_2) \sim N\left(0, \begin{pmatrix} \sigma_0^2 & \sigma_{02} \\ \sigma_{02} & \sigma_2^2 \end{pmatrix}\right), \quad c \sim N(0, \sigma_c^2), \quad \epsilon_{ijk} \sim N(0, \sigma^2)$$

```
M_red <- lmer(Preference ~ sens2 + Homesize +  
              (1 | product) + (1 + sens2 | Consumer),  
              data = carrots)
```

Example 2. summary

```
summary(M_red)
```

	Estimate	Std. Error	df	t value	Pr(> t)
(Intercept)	4.91	0.09	39	56.35	<0.001
sens2	0.07	0.02	12	4.18	0.001
Homesize3	-0.25	0.11	100	-2.37	0.020

Table: Automated elimination process of random effects of model M

Consumers prefer products characterized by the `sens2` (more sweet products)

To remember!






- **Random effects:** Consumers, Assessors, Replicates/Sessions
- For the **random effects** remember to put interaction effects: between **random** and between **random** and **fixed** effects
- For the **fixed effects:** via "*" you may specify all possible main and interaction effects
- Remember to check whether the variable is factor or not via `str()` function. If not and it should be then use `(as.factor())` function , like e.g.:

```
TVbo$Assessor <- as.factor(TVbo$Assessor)
```


Getting started with lmerTest

- 1 Install lmerTest from CRAN (by e.g. *Install* button in RStudio)
- 2 Tell **R** that you want to use it with `library(lmerTest)`
- 3 Take a look at the **lmerTest-manual.pdf**
- 4 Run the examples from lecture at `lecture_lmerTest_Rcode.R`
- 5 Do the exercises (`lecture_lmerTest_Rcode.R`)

Bibliography I

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Harry T. Lawless, Hildegard Heymann. Sensory Evaluation of Food
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