EncPos

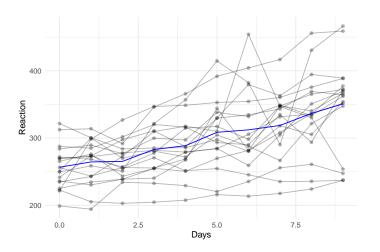
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Motivation: sleepstudy data

- ▶ The average reaction time per day for subjects was evaluated by Gregory et al. (2003) in a sleep deprivation study.
- On day 0 the subjects had their normal amount of sleep and starting that night they were restricted to 3 hours of sleep per night for 9 days, and the reaction time basead on a series of tests was measured on each day for each subject.
- ▶ The data are avaliable at the R package *lme4*.

Motivation: sleepstudy data

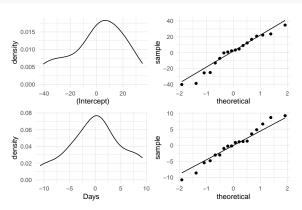


Introduction

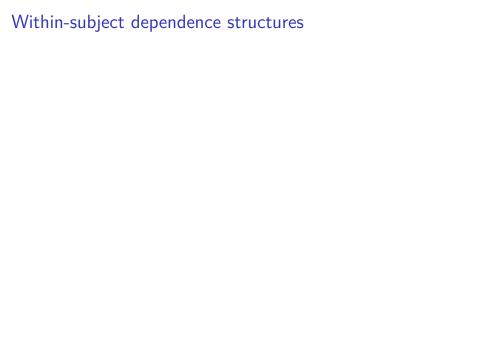
- ▶ Bullet 1
- ▶ Bullet 2
- ▶ Bullet 3

Linear mixed models

Fitting a LMM to the sleepstudy dataset



Model formulation



Tools for model evaluation

The R package skewlmm

- ► The package skewlmm implemets an EM-type algorithm in R using S3 class, containing methods for estimating and predicting the SMSN-LMM.
- It has an user-friendly interface with generic R functions print, summary, plot, fitted, residuals and predict implemented.
- ► The main functions in the package are the smsn.lmm() and smn.lmm() functions, which estimates the parameter of a SMSN-LMM and a SMN-LMM, respectively.

The R package skewlmm

The basic syntax of these functions is as follows:

where

- data: A data frame containing all the variables to be used in the model.
- formFixed: A two-sided linear formula object describing the fixed effects part of the model.
- groupVar: A character containing the name of the variable which represents the subjects or groups in data.

- ► formRandom: A one-sided linear formula object describing the random effects part of the model.
- depStruct: A character indicating which dependence structure should be used.
- distr: A character indicating which distribution should be used.

- formRandom: A one-sided linear formula object describing the random effects part of the model.
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- distr: A character indicating which distribution should be used.

Some other useful arguments:

- ▶ timeVar: A character containing the name of the variable which represents the time in data.
- initialValues: A named list containing initial parameter values.
- ▶ pAR: The order of the autoregressive process that should be used (if depStruct="ARp").

The functions return an object of the class SMSN and SMN, respectively, containing a list of elements, and the following methods/functions are available to these classes:

- print
- summary
- ▶ fitted
- ▶ plot
- predict
- residuals
- ranef
- ▶ acfresid
- ▶ healy.plot
- ▶ lr.test
 - mahalDist

Example: sleepstudy

```
fit0<-smn.lmm(data=sleepstudy,formFixed = Reaction~Days,
             formRandom = ~Davs.groupVar = "Subject".guiet = T)
fit0
## Linear mixed models with distribution norm and dependency structure CI
## Call:
## smn.lmm(data = sleepstudy, formFixed = Reaction ~ Days, groupVar = "Subject",
       formRandom = ~Days, quiet = T)
##
## Fixed:Reaction ~ Davs
## Random: ~Davs
    Estimated variance (D):
##
              (Intercept)
                              Days
## (Intercept) 560.68002 11.91827
               11.91827 32.52525
## Days
##
## Estimated parameters:
##
       (Intercept) Days sigma2 Dsqrt1 Dsqrt2 Dsqrt3
          251,4051 10,4673 655,4400 23,6752 0,4059 5,6886
##
## s.e.
           7 2257 1 5541 31 3784 10 0856 1 7514 1 4270
##
## Model selection criteria:
## logLik
              ATC
                        BTC
## -875 97 1763 94 1783 098
##
## Number of observations: 180
## Number of groups: 18
```

```
fitskew0<-smsn.lmm(data=sleepstudy,formFixed = Reaction~Days,
             formRandom = "Days, groupVar = "Subject", quiet = T)
fitskew0
## Linear mixed models with distribution sn and dependency structure CI
## Call:
## smsn.lmm(data = sleepstudy, formFixed = Reaction ~ Days, groupVar = "Subject",
      formRandom = ~Davs, quiet = T)
##
##
## Fixed:Reaction ~ Days
## Random: ~Davs
## Estimated variance (D):
##
              (Intercept)
                              Days
## (Intercept) 1432.65701 35.22718
## Davs
                 35.22718 33.76855
##
## Estimated parameters:
##
       (Intercept)
                    Days sigma2 Dsqrt1 Dsqrt2 Dsqrt3 lambda1 lambda2
          251,4073 10,4657 652,6598 37,8418 0,8080 5,7546 -4,2917 -0,2477
         11.6328 2.1610 32.7878 28.9642 5.2063 1.6311 NA
                                                                      NA
##
## Model selection criteria:
```

logLik AIC

##

-875 354 1766 709 1792 253

Number of observations: 180 ## Number of groups: 18

BTC

Changing the distribution

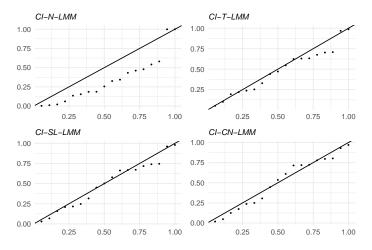
```
fit1<-smn.lmm(data=sleepstudy,formFixed = Reaction~Days,distr = 't',</pre>
              formRandom = ~Days,groupVar = "Subject",quiet = T)
fitskew1<-smsn.lmm(data=sleepstudy,formFixed = Reaction~Days,distr = 'st',
                  formRandom = ~Days, groupVar = "Subject", quiet = T)
fit2<-smn.lmm(data=sleepstudy,formFixed = Reaction~Days,distr = 'sl',</pre>
              formRandom = ~Days,groupVar = "Subject",quiet = T)
fitskew2<-smsn.lmm(data=sleepstudy,formFixed = Reaction~Days,distr = 'ssl',
                  formRandom = ~Days, groupVar = "Subject", quiet = T)
fit3<-smn.lmm(data=sleepstudy,formFixed = Reaction~Days,distr = 'cn',</pre>
              formRandom = ~Days, groupVar = "Subject", quiet = T)
fitskew3<-smsn.lmm(data=sleepstudy,formFixed = Reaction~Days,distr = 'scn',
                  formRandom = ~Days,groupVar = "Subject",quiet = T)
```

Comparing the fitted models

distr	AIC	BIC
norm	1763.9	1783.1
sn	1766.7	1792.3
t	1737.5	1759.8
st	1739.8	1768.6
sl	1736.2	1758.5
ssl	1738.4	1767.2
cn	1733.7	1759.3
scn	1735.9	1767.8

Healy plot

```
grid.arrange(healy.plot(fit0),
    healy.plot(fit1),
    healy.plot(fit2),
    healy.plot(fit3))
```



LR test for the SCN model

lr.test(fitskew3,fit3)

```
##
## Model selection criteria.
##
             logLik
                        AIC
                                 BTC
## fitskew3 -857.926 1735.851 1767.781
## fit3 -858.861 1733.722 1759.266
##
##
      Likelihood-ratio Test
##
## chi-square statistics = 1.871248
## df = 2
## p-value = 0.392341
##
## The null hypothesis that both models represent the
## data equally well is not rejected at level 0.05
```

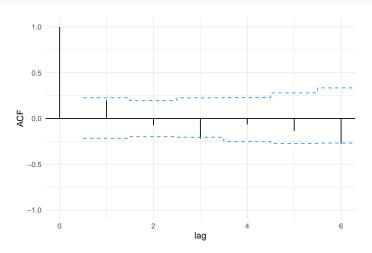
Computing the ACF

acfresid(fit3)

```
ACF n.used
##
      lag
## 1
           1.00000000
                           180
## 2
           0.19793878
                           162
## 3
        2 -0.07329748
                           144
## 4
        3 - 0.21972223
                           126
## 5
        4 -0.06519806
                           108
## 6
        5 -0.13723727
                            90
## 7
        6 - 0.27485055
                            72
##
        7 -0.08778128
                            54
          0.19912222
                            36
##
##
   10
        9
           0.58707921
                            18
```

Plotting the ACF

plot(acfresid(fit3,calcCI = T,maxLag = 6))



Fitting an AR(p)-SMN-LMM

```
#s1.
fit2ar1<-smn.lmm(data=sleepstudy,formFixed = Reaction~Days,</pre>
                 formRandom = ~Days, groupVar = "Subject",
                 distr="sl",depStruct = "ARp",pAR=1,quiet=T)
fit2ar2<-smn.lmm(data=sleepstudy,formFixed = Reaction~Days,
                 formRandom = ~Days, groupVar = "Subject",
                 distr="sl",depStruct = "ARp",pAR=2,quiet=T)
#cn
fit3ar1<-smn.lmm(data=sleepstudy,formFixed = Reaction~Days,
                 formRandom = ~Days,groupVar = "Subject",
                 distr="cn",depStruct = "ARp",pAR=1,quiet=T)
fit3ar2<-smn.lmm(data=sleepstudy,formFixed = Reaction~Days,</pre>
                 formRandom = ~Days,groupVar = "Subject",
                 distr="cn",depStruct = "ARp",pAR=2,quiet=T)
```

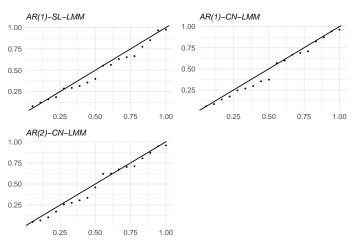
Fitting a DEC-SMN-LMM

Since the data are equally spaced and sorted by time, the use of timeVar in here is optional (the function will use the position if timeVar is not provided).

Comparing the fitted models

distr	depStruct	AIC	BIC
sl	CI	1736.2	1758.5
sl	AR1	1716.9	1742.5
sl	AR2	1717.3	1746.0
sl	DEC	1718.2	1746.9
cn	CI	1733.7	1759.3
cn	AR1	1715.5	1744.3
cn	AR2	1714.9	1746.9
cn	DEC	1716.3	1748.2

Healy plot



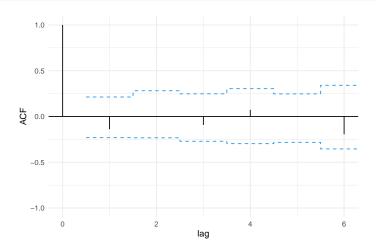
LR test

lr.test(fit3ar1,fit3ar2)

```
##
## Model selection criteria:
            logLik AIC
##
                                 BTC
## fit3ar1 -848.760 1715.520 1744.257
## fit3ar2 -847.463 1714.925 1746.855
##
      Likelihood-ratio Test
##
##
## chi-square statistics = 2.594949
## df = 1
## p-value = 0.1072049
##
## The null hypothesis that both models represent the
## data equally well is not rejected at level 0.05
```

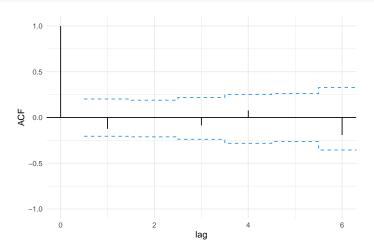
ACF plot for AR(1)-SL-LMM

plot(acfresid(fit2ar1,calcCI = T,maxLag = 6))



ACF plot for AR(1)-CN-LMM

plot(acfresid(fit3ar1,calcCI = T,maxLag = 6))



Fitted models

```
#sl
cbind(fit2ar1$theta,fit2ar1$std.error) %>%
  knitr::kable(digits = 2,col.names = c('estimate','std.error'))
```

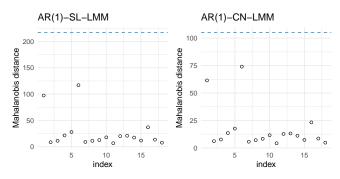
	estimate	std.error
(Intercept)	251.55	7.56
Days	9.81	1.66
sigma2	266.24	63.43
phiAR1	0.57	0.14
Dsqrt1	14.21	10.15
Dsqrt2	2.58	4.16
Dsqrt3	2.17	6.45
nu1	1.52	NA

#cn

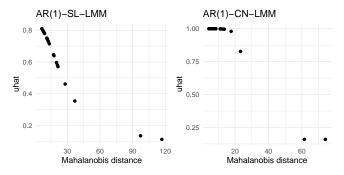
cbind(fit3ar1\$theta,fit3ar1\$std.error) %>%
knitr::kable(digits = 2,col.names = c('estimate','std.error'))

	estimate	std.error
(Intercept)	252.55	7.02
Days	9.99	1.57
sigma2	417.84	78.77
phiAR1	0.56	0.14
Dsqrt1	16.90	11.82
Dsqrt2	2.82	4.76
Dsqrt3	2.89	6.31
nu1	0.12	NA
nu2	0.16	NA

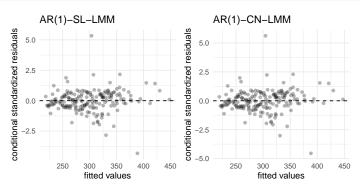
Mahalanobis distance



Mahalanobis distance versus \hat{u}



Plot fit object



Prediction

```
tail(sleepstudy,n=3)
##
      Reaction Days Subject
## 178 343.2199
                        372
## 179 369 1417
               8 372
## 180 364,1236
               9 372
predData <- data.frame(Reaction=NA, Days=10:11, Subject="372")</pre>
predict(fit2ar1,newData = predData)
##
    groupVar Days
                  ypred
## 1
         372 10 377 0998
## 2
         372
               11 389.5745
predict(fit3ar1,newData = predData)
##
    groupVar Days
                   ypred
## 1
         372 10 377 1056
## 2
         372
               11 389.5576
```

References

▶ Gregory Belenky, Nancy J. Wesensten, David R. Thorne, Maria L. Thomas, Helen C. Sing, Daniel P. Redmond, Michael B. Russo and Thomas J. Balkin (2003) Patterns of performance degradation and restoration during sleep restriction and subsequent recovery: a sleep dose-response study. Journal of Sleep Research 12, 1–12.