

Lecture 4 - Correlated and uncorrelated random effects

Witold Wolski

March 12, 2017

Sleepstudy dataset

Lets now consider the sleepstudy dataset from the lme4 package. This is from a report on a study of the effects of sleep deprivation on reaction time for a number of subjects chosen from a population of long-distance truck drivers. These subjects were divided into groups that were allowed only a limited amount of sleep each night. We consider here the group of 18 subjects who were restricted to three hours of sleep per night for the first ten days of the trial. Each subject's reaction time was measured several times on each day of the trial.

```
library(lme4)
```

```
## Loading required package: Matrix
```

```
str(sleepstudy)
```

```
## 'data.frame':   180 obs. of  3 variables:
## $ Reaction: num  250 259 251 321 357 ...
## $ Days      : num   0 1 2 3 4 5 6 7 8 9 ...
## $ Subject   : Factor w/ 18 levels "308","309","310",...: 1 1 1 1 1 1 1 1 1 1 ...
```

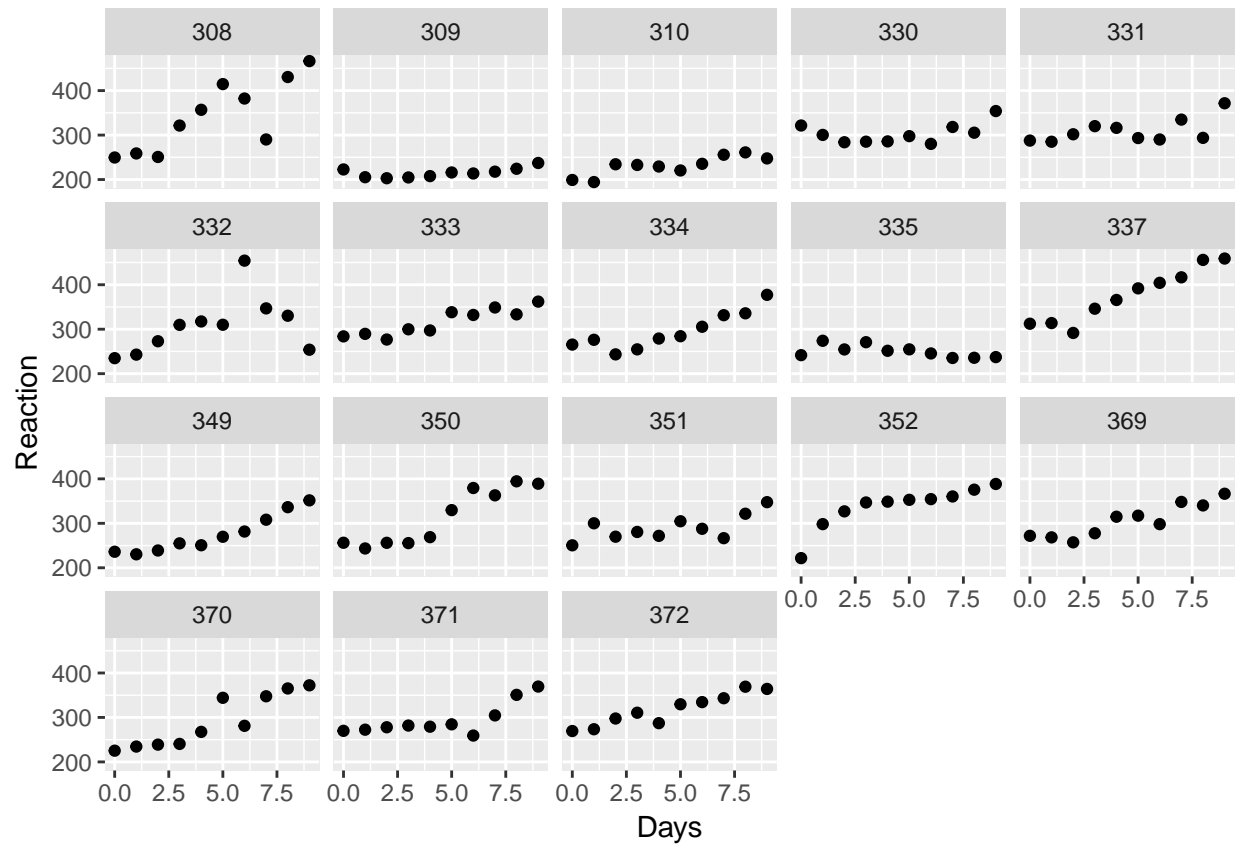
```
head(sleepstudy)
```

```
##   Reaction Days Subject
## 1  249.5600    0     308
## 2  258.7047    1     308
## 3  250.8006    2     308
## 4  321.4398    3     308
## 5  356.8519    4     308
## 6  414.6901    5     308
```

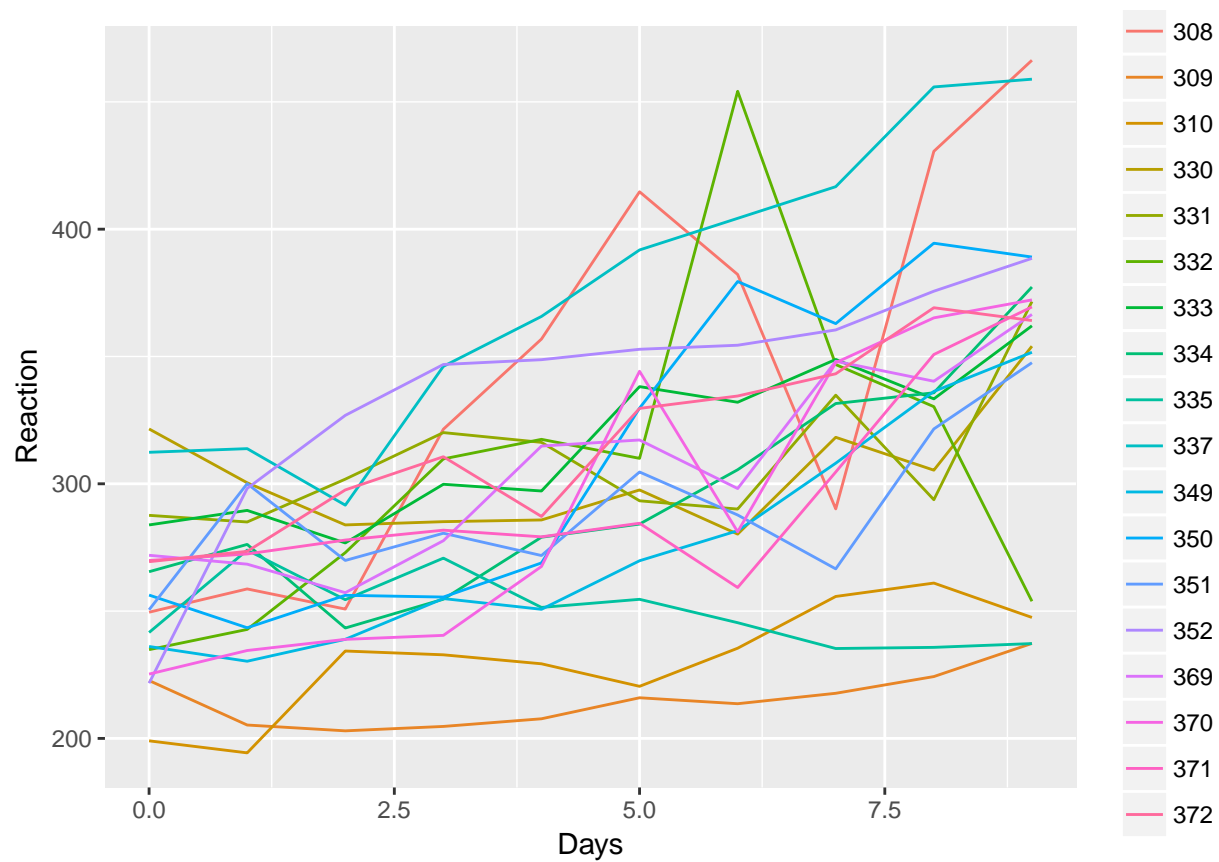
As always, we start by plotting the data

```
library(ggplot2)
```

```
qplot(Days, Reaction, facets=~Subject, data = sleepstudy)
```



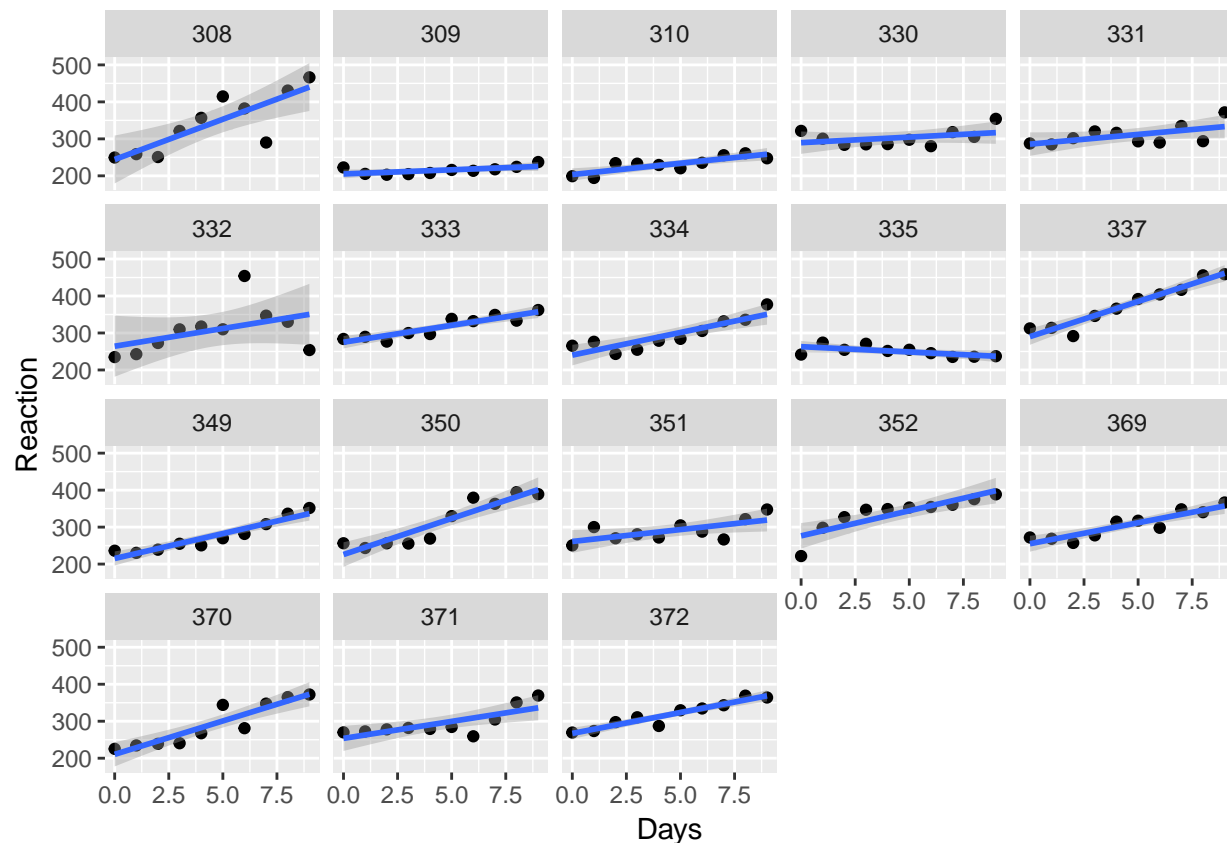
```
ggplot(sleepstudy, aes(x = Days , y = Reaction)) + geom_line(aes(colour=Subject))
```



qplot has an option to also draw on a fitted regression line to each person as well.

```
qplot(Days, Reaction, facets=~Subject, data = sleepstudy,
      geom=c('point', 'smooth'), method='lm')
```

```
## Warning: Ignoring unknown parameters: method
```



Model 1: Correlated random effects

```
fm06 <- lmer(Reaction ~ 1 + Days + (1 + Days | Subject), sleepstudy)
summary(fm06)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: Reaction ~ 1 + Days + (1 + Days | Subject)
## Data: sleepstudy
##
## REML criterion at convergence: 1743.6
##
## Scaled residuals:
##    Min      1Q  Median      3Q      Max
## -3.9536 -0.4634  0.0231  0.4634  5.1793
##
## Random effects:
## Groups Name Variance Std.Dev. Corr
## Subject (Intercept) 612.09 24.740
## Days 35.07 5.922 0.07
## Residual 654.94 25.592
## Number of obs: 180, groups: Subject, 18
##
## Fixed effects:
## Estimate Std. Error t value
## (Intercept) 251.405 6.825 36.84
```

```
## Days          10.467      1.546      6.77
##
## Correlation of Fixed Effects:
##      (Intr)
## Days -0.138

fm06b <- lmer(Reaction ~ 1 + Days + (1 + Days | Subject) + (1|Days), sleepstudy)
summary(fm06b)

## Linear mixed model fit by REML ['lmerMod']
## Formula: Reaction ~ 1 + Days + (1 + Days | Subject) + (1 | Days)
## Data: sleepstudy
##
## REML criterion at convergence: 1743.6
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.9536 -0.4634  0.0231  0.4634  5.1793
##
## Random effects:
## Groups Name Variance Std.Dev. Corr
## Subject (Intercept) 612.09 24.740
##          Days       35.07  5.922  0.07
## Days (Intercept) 0.00 0.000
## Residual       654.94 25.592
## Number of obs: 180, groups: Subject, 18; Days, 10
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept) 251.405      6.825 36.84
## Days        10.467      1.546 6.77
##
## Correlation of Fixed Effects:
##      (Intr)
## Days -0.138
```

- What is the typical initial reaction time? answer $251.41ms$
- How much does reaction time typically increase per day? answer $10.47ms/day$
- What is typical subject-subject variation in the initial reaction time? $24.740ms$
- What is the typical subject-subject variation in the slope? $5.9ms$
- What would approximate 95% confidence intervals be for the slope and the intercept across all subjects? $251.41ms \pm 1.96 \cdot 24.740ms$
- What is the typical within subject variation? $\hat{\sigma} = 25.592ms$
- Is there a strong relationship between a subjects initial reaction time and how strongly affected they are by sleep deprivation? NO since $\rho = 0.07$.

This model has used correlated random effects for the same subject, i.e., there is a correlation between the random subject intercept and the subject specific gradient. Mathematically, we have fit the model

$$y_{ij} = \alpha + a_i + (\beta + b_i)Days_j + \epsilon_{ij}$$

Where

$$\begin{pmatrix} a_i \\ b_i \end{pmatrix} \sim N(0, \Sigma)$$

with

$$\Sigma = \begin{pmatrix} \sigma_a^2 & \sigma_a \sigma_b \\ \sigma_a \sigma_b & \sigma_b^2 \end{pmatrix}$$

and

$$\epsilon_{ij} = N(0, \sigma^2)$$

The output above gives us estimates of the fixed effects (α, β) , and the random effect variances Σ and σ^2 .

$$\alpha = 251.41, \beta = 10.47, \sigma_a = 24.740, \sigma_b = 5.922, \rho = 0.07$$

Model 2: Uncorrelated random effects

If you want to have random effects independent specify them in 2 separate blocks.

```
head(sleepstudy)
```

```
##   Reaction Days Subject
## 1 249.5600    0    308
## 2 258.7047    1    308
## 3 250.8006    2    308
## 4 321.4398    3    308
## 5 356.8519    4    308
## 6 414.6901    5    308
```

```
fm07 <- lmer(Reaction ~ 1 + (1 | Subject) + Days + (Days-1|Subject), sleepstudy)
summary(fm07)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: Reaction ~ 1 + (1 | Subject) + Days + (Days - 1 | Subject)
##   Data: sleepstudy
##
## REML criterion at convergence: 1743.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.9626 -0.4625  0.0204  0.4653  5.1860
##
## Random effects:
##   Groups      Name      Variance Std.Dev.
##   Subject    (Intercept) 627.57   25.051
##   Subject.1 Days         35.86    5.988
##   Residual                653.58   25.565
## Number of obs: 180, groups: Subject, 18
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept)  251.405      6.885    36.51
## Days         10.467      1.560     6.71
##
## Correlation of Fixed Effects:
##      (Intr)
```

```
## Days -0.184
fm08 <- lmer(Reaction ~ 1 + Days + (1 | Subject) + (Days-1|Subject)+(1|Days), sleepstudy)
summary(fm08)

## Linear mixed model fit by REML ['lmerMod']
## Formula: Reaction ~ 1 + Days + (1 | Subject) + (Days - 1 | Subject) +
##      (1 | Days)
## Data: sleepstudy
##
## REML criterion at convergence: 1743.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.9626 -0.4625  0.0204  0.4653  5.1860
##
## Random effects:
## Groups      Name                Variance Std.Dev.
## Subject    (Intercept)  6.276e+02  2.505e+01
## Subject.1 Days          3.586e+01  5.988e+00
## Days        (Intercept)  5.339e-14  2.311e-07
## Residual                    6.536e+02  2.557e+01
## Number of obs: 180, groups: Subject, 18; Days, 10
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept)  251.405      6.885    36.51
## Days         10.467      1.560     6.71
##
## Correlation of Fixed Effects:
##      (Intr)
## Days -0.184
```

Note that we have explicitly ruled out the intercept term in the random effect part for Days. If we didn't do this, R would have fit an additional (unnecessary) intercept. Note that this has fit the model

$$y_{ij} = \alpha + a_i + (\beta + b_i)Days_j + \epsilon_{ij}$$

where

$$a_i \sim N(0, \sigma_a^2), b_i \sim N(0, \sigma_b^2) \text{ and } \epsilon_{ij} \sim N(0, \sigma^2)$$

with all the random effects independent of each other.

In the correlated random effects model, the correlation was estimated to be small. This suggests that the uncorrelated random effects model is a good choice. We will look later at how to formally test which model is better.