

Analysis I

“Aerosol emission is increased in professional singing”

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Purpose

Linear mixed modeling of experiment I (breathing, speaking, singing).

Analysis is inspired by Winter, B. (2013). Linear models and linear mixed effects models in R with linguistic applications. arXiv:1308.5499

The advantages are

- 1) Using mixed modeling no (pre-)averaging is necessary
- 2) Condition & Gender (& Interfatorial effects) were modeled as fixed effects
- 3) ID was modeled as random effects (intercept and slope for Gender and Condition)

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```
rm(list = ls())
library(ggpubr)

## Loading required package: ggplot2

library(gridExtra)
library(lme4)

## Loading required package: Matrix

## Registered S3 methods overwritten by 'lme4':
##   method                from
##   cooks.distance.influence.merMod car
##   influence.merMod       car
##   dfbeta.influence.merMod car
##   dfbetas.influence.merMod car

library(redres) # https://github.com/goodekat/redres
library(GGally)

## Registered S3 method overwritten by 'GGally':
##   method from
##   +.gg    ggplot2

library(ggResidpanel)
```

Loading data

```
# Set working directory according to your file hierarchy.
orig.data <- read.csv('Data.csv')
```

log-transform of emission rate and replacing zeros by NaN

```
PM.log <- orig.data$PM
PM.log[PM.log == 0] <- NA
orig.data$PM.log = log10(PM.log)
length(which(is.na(PM.log)==T)) # Number of NaN detected
```

```
## [1] 12
```

Selection of relevant data

```
exp.data.I <- dplyr::filter(orig.data,
                           Condition == 'breathing'
                           | Condition == 'speaking'
                           | Condition == 'singing'
                           )
```

Test 0: Deviation from normality and homoscedasticity?

```
I00.model <- lmer(PM~Condition*Gender
                 +(1+Condition|ID)
                 +(1+Gender|ID),
                 exp.data.I,
                 REML=FALSE,
                 control = lmerControl(optimizer = 'optimx',
                                       optCtrl=list(method='nlminb'))
                 )
```

```
## Loading required namespace: optimx
```

```
## boundary (singular) fit: see ?isSingular
```

```
# (intermediate) results:
# 1) Deviation from normality and homoscedasticity was observed
```

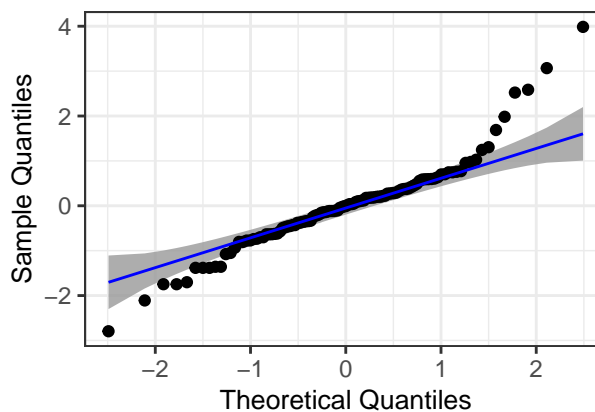
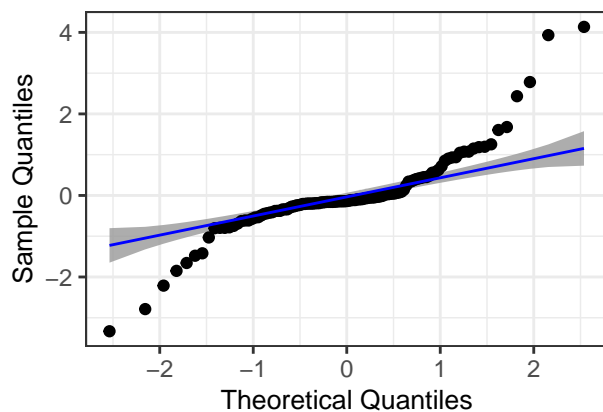
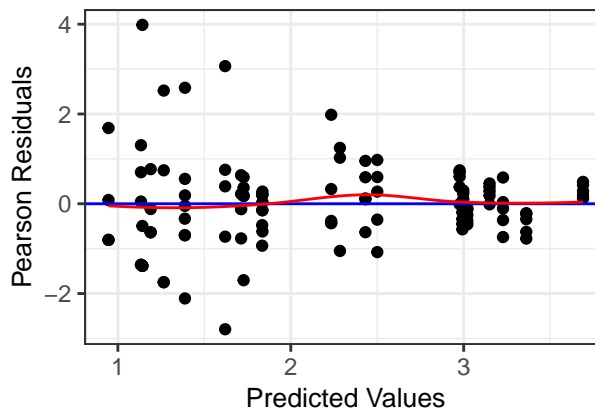
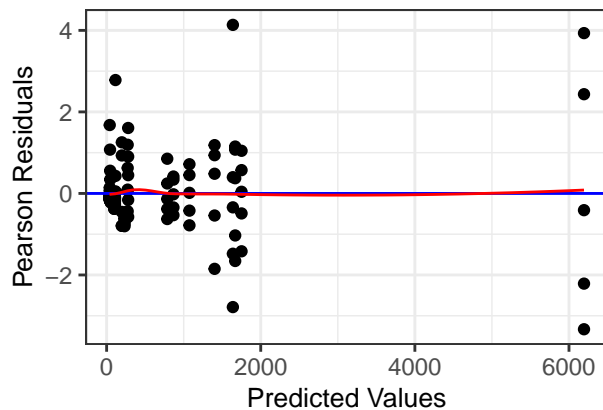
```
I0.model <- lmer(PM.log~Condition*Gender
                +(1+Condition|ID)
                +(1+Gender|ID),
                exp.data.I,
                REML=FALSE,
                control = lmerControl(optimizer = 'optimx',
                                      optCtrl=list(method='nlminb'))
                )
```

```
## boundary (singular) fit: see ?isSingular
```

Comparing model with and without log-transform of P_M

```
resid_compare(models = list(I00.model,
                           I0.model
                           ),
              plots = c("resid", "qq"),
              smoother = TRUE,
              qqbands = TRUE,
              title.opt = FALSE)
```

```
## `geom_smooth()` using formula 'y ~ x'
## `geom_smooth()` using formula 'y ~ x'
```



Pair plot of the data

```
exp.data.I %>% ggpairs(.,
                      mapping = ggplot2::aes(colour=Gender),
                      legend = 1,
                      column = c('ID', 'Condition', 'Gender', 'PM.log'),
                      lower = list(continuous = wrap("smooth", alpha = 0.3, size=0.1)),
                      diag = list(discrete="barDiag",
                                  continuous = wrap("densityDiag", alpha=0.3))
                      )+ theme(legend.position = "bottom")
```

```
## Warning: Removed 7 rows containing non-finite values (stat_boxplot).
```



```

      REML=FALSE,
      control = lmerControl(optimizer = 'optimx',
                            optCtrl=list(method='nlminb'))
    )

## boundary (singular) fit: see ?isSingular
anova(I1.null, I1.model)

## Data: exp.data.I
## Models:
## I1.null: PM.log ~ Condition + Gender + (1 + Condition | ID) + (1 + Gender |
## I1.null: ID)
## I1.model: PM.log ~ Condition * Gender + (1 + Condition | ID) + (1 + Gender |
## I1.model: ID)
##          npar   AIC    BIC  logLik deviance  Chisq Df Pr(>Chisq)
## I1.null   14 131.49 169.67 -51.742   103.48
## I1.model  16 135.41 179.05 -51.704   103.41 0.0768  2    0.9624

# (intermediate) results:
# 1) interaction between Condition and Gender is NOT significant
# 2) Condition:Gender is omitted for now
#####

```

Test 2: Is Condition significant?

```

I2.model <- I1.null

I2.null <- lmer(PM.log~Gender
               +(1+Condition|ID)
               +(1+Gender|ID),
               exp.data.I,
               REML=FALSE,
               control = lmerControl(optimizer = 'optimx',
                                     optCtrl=list(method='nlminb'))
               )

## boundary (singular) fit: see ?isSingular
anova(I2.null, I2.model)

## Data: exp.data.I
## Models:
## I2.null: PM.log ~ Gender + (1 + Condition | ID) + (1 + Gender | ID)
## I2.model: PM.log ~ Condition + Gender + (1 + Condition | ID) + (1 + Gender |
## I2.model: ID)
##          npar   AIC    BIC  logLik deviance  Chisq Df Pr(>Chisq)
## I2.null   12 165.28 198.01 -70.641   141.28
## I2.model  14 131.49 169.67 -51.742   103.48 37.797  2    6.2e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

summary(I2.model)

## Linear mixed model fit by maximum likelihood ['lmerMod']

```

```

## Formula: PM.log ~ Condition + Gender + (1 + Condition | ID) + (1 + Gender |
## ID)
## Data: exp.data.I
## Control: lmerControl(optimizer = "optimx", optCtrl = list(method = "nlminb"))
##
## AIC      BIC    logLik deviance df.resid
## 131.5    169.7   -51.7   103.5    99
##
## Scaled residuals:
##   Min       1Q   Median       3Q      Max
## -2.7757 -0.4807 -0.0064  0.3992  3.9641
##
## Random effects:
##  Groups   Name                Variance Std.Dev. Corr
##  ID       (Intercept)          0.07513  0.2741
##           Conditionsinging  0.06588  0.2567  -1.00
##           Conditionspeaking 0.51619  0.7185  -1.00  1.00
##  ID.1     (Intercept)          0.04476  0.2116
##           Gendermale          0.04476  0.2116  -1.00
## Residual                   0.11490  0.3390
## Number of obs: 113, groups: ID, 8
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept)    1.5760    0.1592   9.900
## Conditionsinging  1.7740    0.1211  14.651
## Conditionspeaking  0.5230    0.2664   1.963
## Gendermale     -0.3453    0.1246  -2.772
##
## Correlation of Fixed Effects:
##              (Intr) Cndtnsn Cndtnsp
## Condtnsngng -0.644
## Condtnspkng -0.666  0.825
## Gendermale  -0.680  0.008  0.004
## convergence code: 0
## boundary (singular) fit: see ?isSingular

```

```

# (intermediate) results:
# 1) Condition IS significant. Condition affected log10(PM) ( $\chi^2(2)=37.797, p=6.2e-9$ ),
#     increasing it by a factor of 0.5230 P/s+0.2664 (standard errors)
#     from breathing to speaking and by a factor of 1.7740+0.1211 from breathing to
#     singing
#####

```

Test 3: Is Gender significant?

```

I3.model <- I1.null

I3.null <- lmer(PM.log~Condition
               +(1+Condition|ID)
               +(1+Gender|ID),
               exp.data.I,

```

```

        REML=FALSE,
        control = lmerControl(optimizer = 'optimx',
                              optCtrl=list(method='nlminb'))
    )

## boundary (singular) fit: see ?isSingular

anova(I3.null, I3.model)

## Data: exp.data.I
## Models:
## I3.null: PM.log ~ Condition + (1 + Condition | ID) + (1 + Gender | ID)
## I3.model: PM.log ~ Condition + Gender + (1 + Condition | ID) + (1 + Gender |
## I3.model:      ID)
##          npar  AIC    BIC logLik deviance Chisq Df Pr(>Chisq)
## I3.null   13 133.79 169.24 -53.894   107.79
## I3.model  14 131.49 169.67 -51.742   103.48 4.3035  1    0.03803 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# (intermediate) results:
# 1) Gender IS significant. Gender affected log10(PM) ( $\chi^2(1)=4.3035, p=0.03803$ ),
#     lowering it by a factor of -0.3453 P/s+-0.1246 (standard errors)
#     from female to male
#####

```

By-subject analysis

```

coef(I3.model)

## $ID
##      (Intercept) Conditionsinging Conditionspeaking Gendermale
## S1  0.9427102      2.070445      1.3528669 -0.2347422
## S2  2.4091803      1.383835     -0.5690541 -0.6500869
## S3  1.6412443      1.743387      0.4373846 -0.1351186
## S4  1.1836732      1.957624      1.0370665 -0.3612540
## S5  1.4754273      1.821023      0.6547005 -0.3452990
## S6  2.3424326      1.415087     -0.4815761 -0.3452997
## S7  1.6277395      1.749710      0.4550837 -0.3453004
## S8  0.9852099      2.050546      1.2971678 -0.3453025
##
## attr(,"class")
## [1] "coef.mer"

# 1) all subjects (except for S2 and S6 - see Fig. 2 in manuscript)
#     show an increase of emitted Particles from breathing to speaking
# 2) all subjects
#     show an increase of emitted Particles from breathing to singing
# 3) all subjects show, that females emitted more particles than males
# 4) the highest ratio was shown by S2

```