

# bm2\_hw7

Yixiao Sun

2024-03-31

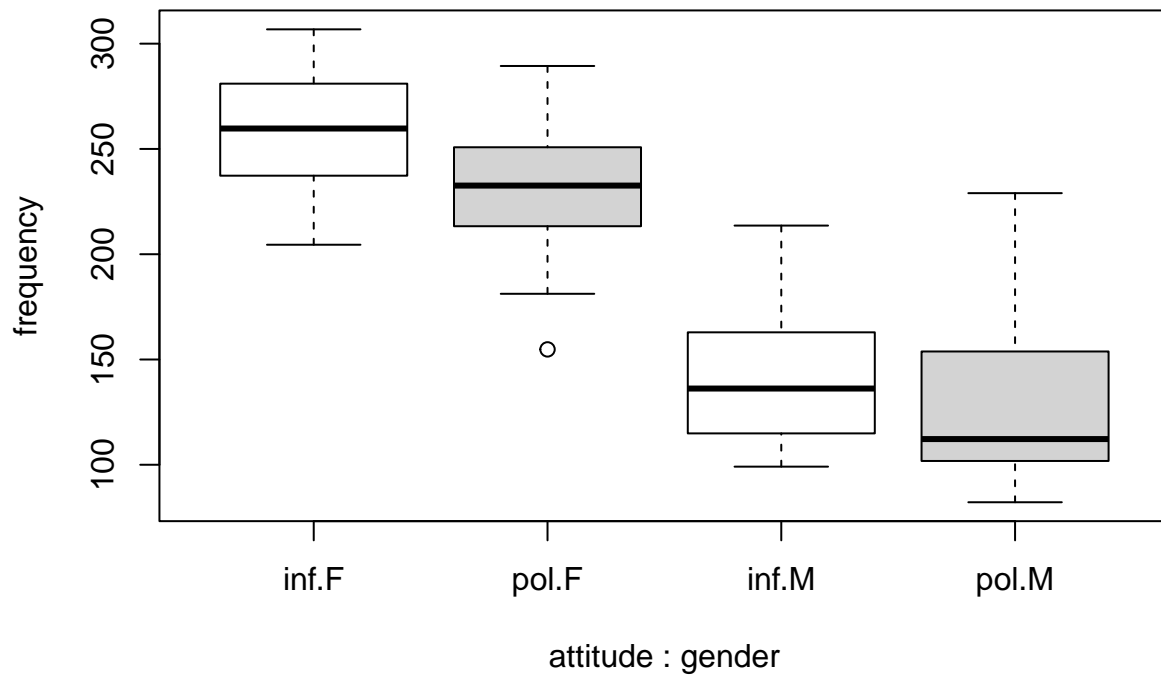
```
set.seed(1)
library(lme4)
```

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

```
rm(list = ls())
data <- read.csv('~/.Desktop/P8130_Biostatistical Method/bw2_hw7/HW7-politeness_data.csv')
summary(data)
```

```
##      subject          gender      scenario  attitude
## Length:84      Length:84      Min.   :1  Length:84
## Class :character Class :character 1st Qu.:2  Class :character
## Mode  :character Mode  :character Median :4  Mode  :character
##                                     Mean   :4
##                                     3rd Qu.:6
##                                     Max.   :7
##      frequency
## Min.   : 82.2
## 1st Qu.:126.8
## Median :201.8
## Mean   :192.6
## 3rd Qu.:247.4
## Max.   :306.8
```

```
boxplot(frequency ~ attitude*gender, col=c("white","lightgray"),data= data)
```



```
m1 <- lmer(frequency ~ gender + attitude + (1|subject), REML = TRUE, data = data)
summary(m1)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: frequency ~ gender + attitude + (1 | subject)
## Data: data
##
## REML criterion at convergence: 796.1
##
## Scaled residuals:
##    Min      1Q  Median      3Q      Max
## -2.3564 -0.5658 -0.2012  0.4618  3.2998
##
## Random effects:
## Groups Name Variance Std.Dev.
## subject (Intercept) 598.2 24.46
## Residual 847.7 29.12
## Number of obs: 84, groups: subject, 6
##
## Fixed effects:
## Estimate Std. Error t value
## (Intercept) 256.987 15.155 16.957
## genderM -108.798 20.956 -5.192
## attitudepol -20.002 6.353 -3.148
##
```

```
## Correlation of Fixed Effects:
##           (Intr) gendrM
## genderM    -0.691
## attitudepol -0.210  0.000
```

```
summary(ranef(m1)$subject)
```

```
## (Intercept)
## Min.      :-32.700
## 1st Qu.: -9.331
## Median :  4.072
## Mean      :  0.000
## 3rd Qu.:  8.813
## Max.      : 27.960
```

```
ranef(m1)
```

```
## $subject
## (Intercept)
## F1 -13.575832
## F2  10.170523
## F3   3.405309
## M3  27.960290
## M4   4.739325
## M7 -32.699615
##
## with conditional variances for "subject"
```

```
residuals(m1)
```

```
##           1           2           3           4           5           6
## -10.1086918 -38.9110728  61.6913082  16.2889272 -19.5086918  43.4889272
##           7           8           9          10          11          12
##  27.3913082  33.3889272   8.4913082   8.9889272 -42.2086918 -12.7110728
##          13          14          15          16          17          18
## -26.9110728 -68.6086918 -10.6898328 -23.0922138  -3.5898328  -9.3922138
##          19          20          21          22          23          24
##  26.6101672   5.6077862  35.0101672  46.4077862  -7.7898328  -7.8922138
##          25          26          27          28          29          30
## -13.8898328  18.4077862   4.0077862 -54.8898328 -22.2262301 -29.3286110
##          31          32          33          34          35          36
##  96.0737699 -38.0286110 -20.7262301  60.6713890  60.4737699   9.9713890
##          37          38          39          40          41          42
## -31.1262301 -26.0286110 -22.9262301 -16.7286110  -6.9286110  -6.4262301
##          43          44          45          46          47          48
##  -9.3872897 -16.3896707 -13.2872897 -11.1896707  -9.5872897  -5.2896707
##          49          50          51          52          53          54
##   1.6127103   4.5103293  -1.7872897 -12.5896707  13.3127103  -7.2896707
##          55          56          57          58          59          60
##   8.9103293  12.1127103 -14.4550468 -35.8574277  -0.8550468  -7.4574277
##          61          62          63          64          65          66
##  42.2449532  34.6425723  -3.9550468  29.0425723  30.5449532  27.0425723
```

```
##           67           68           69           70           71           72
## -39.1550468 -41.2574277 13.8425723 -19.9550468 -2.3471945 12.6504246
##           73           74           75           76           77           78
## -13.7471945 23.5504246 4.0528055 9.9504246 51.3528055 14.7504246
##           79           80           81           82           83           84
## 4.5528055 -19.6495754 -9.4471945 -18.1495754 -15.0495754 -2.8471945
```

```
m2 <- lmer(frequency ~ gender + attitude + gender*attitude+ (1|subject),
REML = F, data = data)
summary(m2)
```

```
## Linear mixed model fit by maximum likelihood ['lmerMod']
## Formula: frequency ~ gender + attitude + gender * attitude + (1 | subject)
## Data: data
##
##      AIC      BIC    logLik deviance df.resid
##   826.3    840.8   -407.1    814.3      78
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.2856 -0.5246 -0.1719  0.4929  3.2294
##
## Random effects:
## Groups Name Variance Std.Dev.
## subject (Intercept) 380.4 19.50
## Residual 822.1 28.67
## Number of obs: 84, groups: subject, 6
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept) 260.686 12.883 20.235
## genderM -116.195 18.219 -6.378
## attitudepol -27.400 8.848 -3.097
## genderM:attitudepol 14.795 12.514 1.182
##
## Correlation of Fixed Effects:
##              (Intr) gendrM atttdp
## genderM -0.707
## attitudepol -0.343 0.243
## gendrM:tttdp 0.243 -0.343 -0.707
```

```
ranef(m2)
```

```
## $subject
## (Intercept)
## F1 -12.950999
## F2 9.702421
## F3 3.248578
## M3 26.673406
## M4 4.521196
## M7 -31.194602
##
## with conditional variances for "subject"
```

```
TS1 <- deviance(m1) - deviance(m2)
```

```
## Warning in deviance.merMod(m1): deviance() is deprecated for REML fits; use  
## REMLcrit for the REML criterion or deviance(.,REML=FALSE) for deviance  
## calculated at the REML fit
```

```
p1 <- 1-pchisq(TS1, 1)
```

```
m3 <- lmer(frequency ~ gender + attitude + (1+attitude|subject), data = data)
```

```
## boundary (singular) fit: see help('isSingular')
```

```
summary(m3)
```

```
## Linear mixed model fit by REML ['lmerMod']  
## Formula: frequency ~ gender + attitude + (1 + attitude | subject)  
## Data: data  
##  
## REML criterion at convergence: 796.1  
##  
## Scaled residuals:  
##      Min       1Q   Median       3Q      Max   
## -2.3482 -0.5751 -0.2125  0.4770  3.2953   
##  
## Random effects:  
## Groups Name Variance Std.Dev. Corr  
## subject (Intercept) 578.7865 24.0580  
##          attitudepol  0.6511  0.8069  1.00  
## Residual          847.5471 29.1127  
## Number of obs: 84, groups: subject, 6  
##  
## Fixed effects:  
##              Estimate Std. Error t value  
## (Intercept)  257.588     15.040  17.127  
## genderM      -110.000     20.933  -5.255  
## attitudepol  -20.002      6.361  -3.144  
##  
## Correlation of Fixed Effects:  
##              (Intr) gendrM  
## genderM      -0.696  
## attitudepol -0.177  0.000  
## optimizer (nloptwrap) convergence code: 0 (OK)  
## boundary (singular) fit: see help('isSingular')
```

```
ranef(m3)
```

```
## $subject  
##      (Intercept) attitudepol  
## F1 -13.956088 -0.46807765  
## F2  9.436867  0.31650606
```

```
## F3      2.744934  0.09206321
## M3     28.055212  0.94095263
## M4      5.301848  0.17782035
## M7    -31.582772 -1.05926460
##
## with conditional variances for "subject"

m4 <- lmer(frequency ~ gender + attitude + (1|subject)+(1|scenario), data = data)
summary(m4)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: frequency ~ gender + attitude + (1 | subject) + (1 | scenario)
## Data: data
##
## REML criterion at convergence: 784.1
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.2690 -0.6331 -0.0878  0.5204  3.5326
##
## Random effects:
## Groups Name Variance Std.Dev.
## scenario (Intercept) 224.5 14.98
## subject (Intercept) 613.2 24.76
## Residual 637.8 25.25
## Number of obs: 84, groups: scenario, 7; subject, 6
##
## Fixed effects:
## Estimate Std. Error t value
## (Intercept) 256.987 16.101 15.961
## genderM -108.798 20.956 -5.192
## attitudepol -20.002 5.511 -3.630
##
## Correlation of Fixed Effects:
## (Intr) gendrM
## genderM -0.651
## attitudepol -0.171 0.000
```

```
ranef(m4)
```

```
## $scenario
## (Intercept)
## 1 -13.485886
## 2 6.741980
## 3 11.600172
## 4 20.885315
## 5 -1.835659
## 6 -12.940097
## 7 -10.965825
##
## $subject
## (Intercept)
## F1 -13.916125
```

```
## F2    10.425458
## F3     3.490667
## M3    28.661144
## M4     4.858122
## M7   -33.519266
##
## with conditional variances for "scenario" "subject"
```