

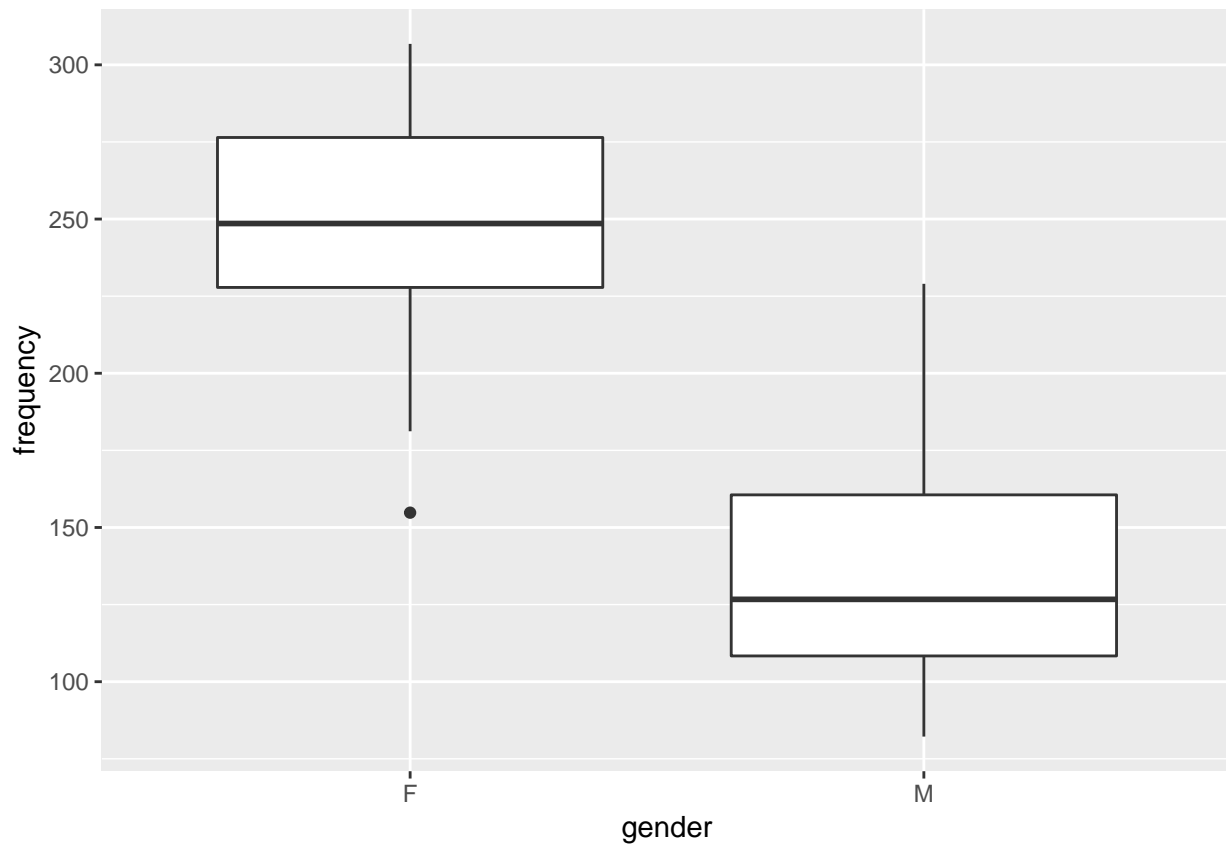
BM_HW7

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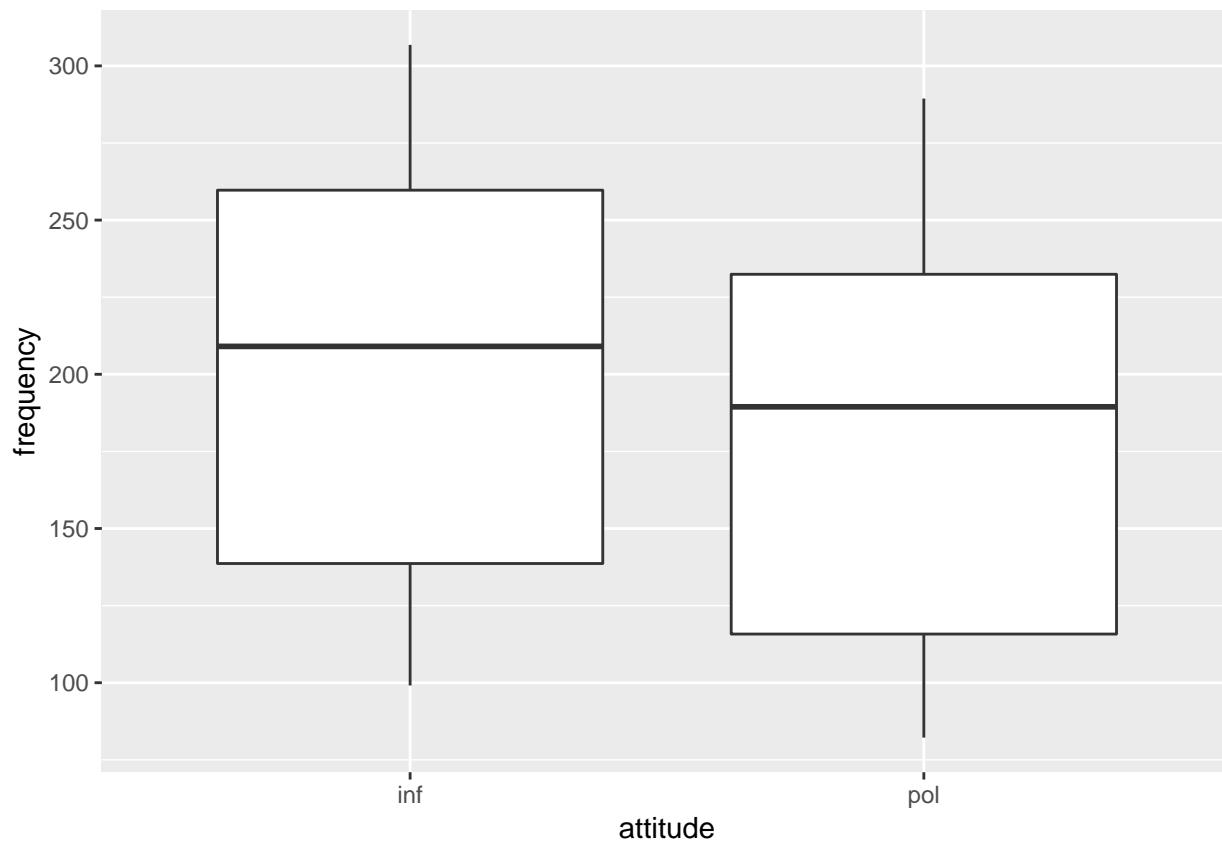
4/15/2019

1.

```
ggplot(df) + geom_boxplot(aes(x = gender, y = frequency))
```



```
ggplot(df) + geom_boxplot(aes(x = attitude, y = frequency))
```



Based on the plot, female have higher frequency of pitcher compared to male. Inf attitude tends to have higher frequency of pitcher compared to plr attitude.

2.

```
# fit a random intercept model
lmm1 = lme(frequency ~ gender + attitude, random = ~1 | subject, data = df, method = 'REML')
summary(lmm1)
```

```
## Linear mixed-effects model fit by REML
## Data: df
##      AIC      BIC    logLik
## 806.0805 818.0527 -398.0402
##
## Random effects:
## Formula: ~1 | subject
##      (Intercept) Residual
## StdDev:    24.45803 29.11537
##
## Fixed effects: frequency ~ gender + attitude
##              Value Std.Error DF   t-value p-value
## (Intercept)  256.98690 15.154986 77 16.957251  0.0000
## genderM      -108.79762 20.956235  4 -5.191659  0.0066
## attitudepol  -20.00238  6.353495 77 -3.148248  0.0023
## Correlation:
```

```
##          (Intr) gendrM
## genderM      -0.691
## attitudepol -0.210  0.000
##
## Standardized Within-Group Residuals:
##      Min      Q1      Med      Q3      Max
## -2.3564422 -0.5658319 -0.2011979  0.4617895  3.2997610
##
## Number of Observations: 84
## Number of Groups: 6
```

```
VarCorr(lmm1)
```

```
## subject = pdLogChol(1)
##          Variance StdDev
## (Intercept) 598.1953 24.45803
## Residual    847.7049 29.11537
```

covariance matrix for a subject Y_i

$$\sigma_b = 24.45803$$

$$\sigma = 29.11537$$

$$\text{cov}(Y_i) = \begin{bmatrix} 1445.9 & 598.2 & \dots & 598.2 \\ 598.2 & 1445.9 & \dots & 598.2 \\ \vdots & \vdots & \ddots & \vdots \\ 598.2 & 598.2 & \dots & 1445.9 \end{bmatrix}_{14 \times 14}$$

covariance matrix for the REML estimates of fixed effects

```
vcov(lmm1)
```

```
##          (Intercept)      genderM      attitudepol
## (Intercept)   229.67362 -2.195819e+02 -2.018345e+01
## genderM       -219.58189  4.391638e+02  6.451438e-15
## attitudepol   -20.18345  6.451438e-15  4.036690e+01
```

BLUPs

```
random.effects(lmm1)
```

```
##          (Intercept)
## F1   -13.575831
## F2    10.170522
## F3     3.405309
## M3    27.960288
## M4     4.739325
## M7   -32.699613
```

Residuals

```
lmm1$residuals[, 2]
```

```
##           1           2           3           4           5           6
## -10.1086926 -38.9110735  61.6913074  16.2889265 -19.5086926  43.4889265
##           7           8           9          10          11          12
##  27.3913074  33.3889265   8.4913074   8.9889265 -42.2086926 -12.7110735
##          13          14          15          16          17          18
## -26.9110735 -68.6086926 -10.6898326 -23.0922136  -3.5898326  -9.3922136
##          19          20          21          22          23          24
##  26.6101674   5.6077864  35.0101674  46.4077864  -7.7898326  -7.8922136
##          25          26          27          28          29          30
## -13.8898326  18.4077864   4.0077864 -54.8898326 -22.2262298 -29.3286108
##          31          32          33          34          35          36
##  96.0737702 -38.0286108 -20.7262298  60.6713892  60.4737702   9.9713892
##          37          38          39          40          41          42
## -31.1262298 -26.0286108 -22.9262298 -16.7286108  -6.9286108  -6.4262298
##          43          44          45          46          47          48
##  -9.3872916 -16.3896725 -13.2872916 -11.1896725  -9.5872916  -5.2896725
##          49          50          51          52          53          54
##   1.6127084   4.5103275  -1.7872916 -12.5896725  13.3127084  -7.2896725
##          55          56          57          58          59          60
##   8.9103275  12.1127084 -14.4550462 -35.8574271  -0.8550462  -7.4574271
##          61          62          63          64          65          66
##  42.2449538  34.6425729  -3.9550462  29.0425729  30.5449538  27.0425729
##          67          68          69          70          71          72
## -39.1550462 -41.2574271  13.8425729 -19.9550462  -2.3471929  12.6504261
##          73          74          75          76          77          78
## -13.7471929  23.5504261   4.0528071   9.9504261  51.3528071  14.7504261
##          79          80          81          82          83          84
##   4.5528071 -19.6495739  -9.4471929 -18.1495739 -15.0495739  -2.8471929
```

3

```
lmm2 = lme(frequency ~ gender * attitude, random = ~ 1|subject, data = df, method = 'REML')
summary(lmm2)
```

```
## Linear mixed-effects model fit by REML
## Data: df
##      AIC      BIC    logLik
##  799.8018 814.094 -393.9009
##
## Random effects:
## Formula: ~1 | subject
##      (Intercept) Residual
## StdDev:    24.46382 29.04716
##
## Fixed effects: frequency ~ gender * attitude
##              Value Std.Error DF   t-value p-value
## (Intercept)    260.68571 15.481307 76 16.838740  0.0000
## genderM        -116.19524 21.893875  4  -5.307203  0.0061
```

```
## attitudepol          -27.40000  8.964149 76 -3.056620  0.0031
## genderM:attitudepol  14.79524 12.677221 76  1.167073  0.2468
## Correlation:
##              (Intr) gendrM atttdp
## genderM      -0.707
## attitudepol  -0.290  0.205
## genderM:attitudepol  0.205 -0.290 -0.707
##
## Standardized Within-Group Residuals:
##      Min      Q1      Med      Q3      Max
## -2.2344163 -0.5454437 -0.1646159  0.4697182  3.1800944
##
## Number of Observations: 84
## Number of Groups: 6
# compare two models

lmm_1 = lme(frequency ~ gender + attitude, random = ~1 | subject, data = df, method = 'ML')
lmm_2 = lme(frequency ~ gender * attitude, random = ~1 | subject, data = df, method = 'ML')

anova(lmm_2, lmm_1)

##      Model df      AIC      BIC    logLik    Test  L.Ratio p-value
## lmm_2      1  6 826.2508 840.8357 -407.1254
## lmm_1      2  5 825.6363 837.7904 -407.8182 1 vs 2 1.385523  0.2392
```

p-value is 0.2392 which is greater than the significant level. We cannot reject the null hypothesis that the small model fit data well. Therefore, we have not evidence to show that the interaction term is significantly associated with pitch.

4

```
lmm3 = lme(frequency ~ gender + attitude, random = ~1 + attitude | subject, data = df, method = 'REML')
summary(lmm3)

## Linear mixed-effects model fit by REML
## Data: df
##      AIC      BIC    logLik
## 810.0805 826.8416 -398.0402
##
## Random effects:
## Formula: ~1 + attitude | subject
## Structure: General positive-definite, Log-Cholesky parametrization
##      StdDev      Corr
## (Intercept) 24.458032213 (Intr)
## attitudepol  0.003285569  0
## Residual    29.115372269
##
## Fixed effects: frequency ~ gender + attitude
##      Value Std.Error DF   t-value p-value
## (Intercept) 256.98691 15.154987 77 16.957250  0.0000
## genderM     -108.79762 20.956235  4 -5.191659  0.0066
## attitudepol -20.00238  6.353495 77 -3.148248  0.0023
```

```
## Correlation:
##          (Intr) gendrM
## genderM    -0.691
## attitudepol -0.210  0.000
##
## Standardized Within-Group Residuals:
##          Min          Q1          Med          Q3          Max
## -2.3564422 -0.5658319 -0.2011979  0.4617896  3.2997610
##
## Number of Observations: 84
## Number of Groups: 6
```

```
VarCorr(lmm3)
```

```
## subject = pdLogChol(1 + attitude)
##          Variance      StdDev      Corr
## (Intercept) 5.981953e+02 24.458032213 (Intr)
## attitudepol 1.079496e-05 0.003285569 0
## Residual     8.477049e+02 29.115372269
```

model with random intercept and slope

$$Y_{ij} = \beta_1 + \beta_2 * I\{gender = male\} + \beta_3 * I\{attitude = pol\} + b_{1i} + b_{2i} * I\{attitude = pol\} + \epsilon_{ij}$$

$$b_i | N(0, G) \quad \epsilon_i | N(0, \sigma^2 I)$$

Attitude of inf:

$$var(Y_{ij} = \sigma_{b_{1i}}^2 + \sigma^2 \quad cov(Y_{ij} = cov(b_{1i} + \epsilon_{ij}, b_{1i} + \epsilon_{ik}) = \sigma_{b_{1i}}^2$$

$$cov(Y_i) = \begin{bmatrix} \sigma_{b_{1i}}^2 + \sigma^2 & \sigma_{b_{1i}}^2 & \cdots & \sigma_{b_{1i}}^2 \\ \sigma_{b_{1i}}^2 & \sigma_{b_{1i}}^2 + \sigma^2 & \cdots & \sigma_{b_{1i}}^2 \\ \vdots & \vdots & \ddots & \vdots \\ \sigma_{b_{1i}}^2 & \sigma_{b_{1i}}^2 & \cdots & \sigma_{b_{1i}}^2 + \sigma^2 \end{bmatrix}_{14 \times 14}$$

Attitude of pol

$$cov(Y_i) = \begin{bmatrix} \sigma_{b_{2i}}^2 + \sigma_{b_{1i}}^2 + \sigma^2 & \sigma_{b_{2i}}^2 + \sigma_{b_{1i}}^2 & \cdots & \sigma_{b_{2i}}^2 + \sigma_{b_{1i}}^2 \\ \sigma_{b_{2i}}^2 + \sigma_{b_{1i}}^2 & \sigma_{b_{2i}}^2 + \sigma_{b_{1i}}^2 + \sigma^2 & \cdots & \sigma_{b_{2i}}^2 + \sigma_{b_{1i}}^2 \\ \vdots & \vdots & \ddots & \vdots \\ \sigma_{b_{2i}}^2 + \sigma_{b_{1i}}^2 & \sigma_{b_{2i}}^2 + \sigma_{b_{1i}}^2 & \cdots & \sigma_{b_{2i}}^2 + \sigma_{b_{1i}}^2 + \sigma^2 \end{bmatrix}_{14 \times 14}$$

$$cov(Y_{ij} - in, Y_{ij} - pol) = \sigma_{b_{1i}}^2 + \sigma_{b_{2i}}^2 \text{ approximate to } \sigma_{b_{1i}}$$

Therefore, the covariance matrix for subject Y_i is

$$cov(Y_i) = \begin{bmatrix} 1445.9 & 598.2 & \cdots & 598.2 \\ 598.2 & 1445.9 & \cdots & 598.2 \\ \vdots & \vdots & \ddots & \vdots \\ 598.2 & 598.2 & \cdots & 1445.9 \end{bmatrix}_{14 \times 14}$$

Which is approximate to compound symmetry.

```
fixed.effects(lmm3)
```

```
## (Intercept)      genderM attitudepol  
##    256.98691  -108.79762   -20.00238
```

```
random.effects(lmm3)
```

```
##      (Intercept)      attitudepol  
## F1  -13.575831 -8.408891e-07  
## F2   10.170522  1.499413e-07  
## F3    3.405308 -2.981919e-07  
## M3   27.960288  1.009764e-06  
## M4    4.739325  7.794162e-07  
## M7  -32.699612 -8.000404e-07
```

Fixed effect of first female is 257.0 and random effect is -0.0000008408891. BLUP of first female is -13.575831