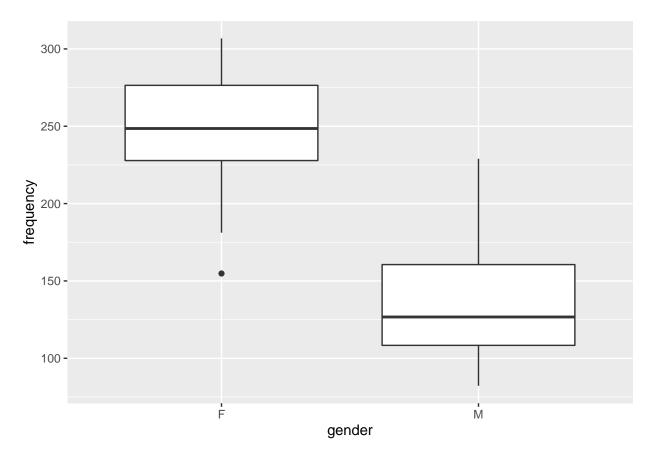
P8131 HW7

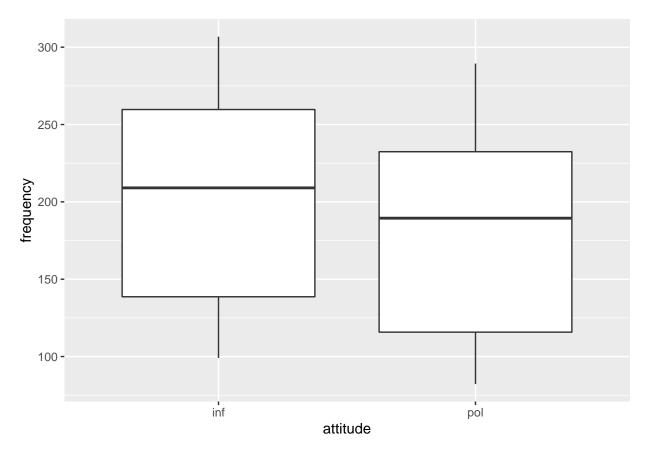
1. The relationship between pitch and politeness

(a) Exploratory analysis: provide boxplots to show the relation between gender/attitude and pitch (ignoring different scenarios).

```
df <- read_csv("HW7-politeness_data.csv", col_names = TRUE)
# gender v.s. frequency
df %>%
    ggplot(aes(x = gender, y = frequency)) +
    geom_boxplot()
```



```
# attitudes v.s. frequency
df %>%
  ggplot(aes(x = attitude, y = frequency)) +
  geom_boxplot()
```



The distributions of pitch (Hz) by gender are different. With female subjects, the pitch has a average of 250 Hz, and with male subjects the average pitch is only about 125 Hz. The distributions of pitch (Hz) by attitude are also different. Informal registers tend to have a higher average pitch compares with that of formal registers.

(d) Fit a mixed effects model with random intercepts for different subjects (gender and attitude being the fixed effects).

Fit the model with random intercepts

```
LMM.rI <- lme (frequency ~ gender + attitude, random = ~1 | subject, data = df, method ='REML')
summary (LMM.rI)
## 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
              24.45803 29.11537
## StdDev:
## Fixed effects: frequency ~ gender + attitude
```

```
##
                   Value Std.Error DF t-value p-value
## (Intercept) 256.98690 15.154986 77 16.957251 0.0000
             -108.79762 20.956235 4 -5.191659 0.0066
## attitudepol -20.00238 6.353495 77 -3.148248 0.0023
   Correlation:
##
              (Intr) gendrM
              -0.691
## genderM
## attitudepol -0.210 0.000
##
## Standardized Within-Group Residuals:
         Min
                     Q1
                               Med
                                                     Max
## -2.3564422 -0.5658319 -0.2011979 0.4617895 3.2997610
## Number of Observations: 84
## Number of Groups: 6
```

Covariance matrix for a subject Y_i

```
VarCorr(LMM.rI) # covariance estimates for random effects and variance for residuals
```

Therefore variance of residuals is $\sigma^2 = 847.7049$, and the subject-specific variance for random effects is $\sigma_b^2 = 598.1953$. And there are 14 observations for each subject. Therefore $\sigma^2 + \sigma_b^2 = 1445.9$ So the covariance matrix becomes:

```
\begin{bmatrix} 1445.9 & 598.1953 & \dots & 598.1953 \\ 598.1953 & 1445.9 & \dots & 598.1953 \\ 598.1953 & 598.1953 & \dots & 598.1953 \\ \dots & & & & & \\ 598.1953 & 598.1953 & \dots & 598.1953 \end{bmatrix}_{14.14}
```

Covariance matrix for the estimates of fixed effects

```
vcov(LMM.rI) # covariance for fixed effects estimates (inv fisher info)

## (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

#
#fixed.effects(LMM.rI) # fixed effects coeff
```

BLUP for Subject i and Residuals

```
# ordered random effects, BLUP (in this case, just b_i)
random.effects(LMM.rI)
##
      (Intercept)
## F1
      -13.575831
## F2
        10.170522
## F3
         3.405309
## M3
        27.960288
## M4
         4.739325
## M7
      -32.699613
# fixed+random residuals
```

fixed subject ## 1 -23.6845238 -10.1086926 ## 2 -52.4869048 -38.9110735 ## 3 48.1154762 61.6913074 16.2889265 ## 4 2.7130952 ## 5 -33.0845238 -19.5086926 ## 6 29.9130952 43.4889265 ## 7 13.8154762 27.3913074 ## 8 19.8130952 33.3889265 ## 9 -5.0845238 8.4913074 ## 10 -4.5869048 8.9889265 ## 11 -55.7845238 -42.2086926 ## 12 -26.2869048 -12.7110735 ## 13 -40.4869048 -26.9110735 ## 14 -82.1845238 -68.6086926 -7.2845238 -10.6898326 ## 16 -19.6869048 -23.0922136 ## 17 -0.1845238 -3.5898326 ## 18 -5.9869048 -9.3922136 ## 19 30.0154762 26.6101674 9.0130952 ## 20 5.6077864 ## 21 38.4154762 35.0101674 ## 22 49.8130952 46.4077864 ## 23 -4.3845238 -7.7898326 -7.8922136 ## 24 -4.4869048 ## 25 -10.4845238 -13.8898326 ## 26 21.8130952 18.4077864 ## 27 7.4130952 4.0077864 ## 28 -51.4845238 -54.8898326 ## 29 -17.4869048 -22.2262298 ## 30 -24.5892857 -29.3286108 ## 31 100.8130952 96.0737702 ## 32 -33.2892857 -38.0286108 ## 33 -15.9869048 -20.7262298

65.4107143 60.6713892

65.2130952 60.4737702

34 ## 35

LMM.rI\$residuals

```
## 36 14.7107143 9.9713892
## 37 -26.3869048 -31.1262298
## 38 -21.2892857 -26.0286108
## 39 -18.1869048 -22.9262298
## 40 -11.9892857 -16.7286108
## 41 -2.1892857 -6.9286108
## 42 -1.6869048 -6.4262298
## 43 -42.0869048 -9.3872916
## 44 -49.0892857 -16.3896725
## 45 -45.9869048 -13.2872916
## 46 -43.8892857 -11.1896725
## 47 -42.2869048 -9.5872916
## 48 -37.9892857 -5.2896725
                                        1.6127084
## 49 -31.0869048
## 50 -28.1892857
                                        4.5103275
## 51 -34.4869048 -1.7872916
## 52 -45.2892857 -12.5896725
                                      13.3127084
## 53 -19.3869048
## 54 -39.9892857
                                       -7.2896725
## 55 -23.7892857
                                         8.9103275
## 56 -20.5869048 12.1127084
             -4.2845238 -14.4550462
## 58 -25.6869048 -35.8574271
## 59
                9.3154762
                                       -0.8550462
## 60
                2.7130952 -7.4574271
## 61
             52.4154762 42.2449538
              44.8130952 34.6425729
## 62
## 63
                6.2154762
                                       -3.9550462
              39.2130952 29.0425729
## 64
## 65
              40.7154762 30.5449538
## 66
              37.2130952 27.0425729
## 67 -28.9845238 -39.1550462
## 68 -31.0869048 -41.2574271
              24.0130952 13.8425729
## 69
## 70
              -9.7845238 -19.9550462
## 71
              25.6130952 -2.3471929
             40.6107143 12.6504261
## 73
              14.2130952 -13.7471929
              51.5107143 23.5504261
## 74
## 75
              32.0130952
                                         4.0528071
## 76
              37.9107143
                                          9.9504261
              79.3130952 51.3528071
## 77
## 78
              42.7107143 14.7504261
## 79
              32.5130952
                                        4.5528071
## 80
                8.3107143 -19.6495739
              18.5130952 -9.4471929
## 81
## 82
                9.8107143 -18.1495739
## 83
             12.9107143 -15.0495739
## 84 25.1130952 -2.8471929
## attr(,"std")
       [1] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537
     [9] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537
## [17] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.1
## [25] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537
```

```
## [33] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537
## [41] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.1
## [49] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537
## [57] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537
## [65] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.1
## [73] 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.11537 29.1
## [81] 29.11537 29.11537 29.11537 29.11537
```

(c) Fit a mixed effects model with intercepts for different subjects

 H_0 : Model 2 (larger model with the interaction term) is no better than Model 1 (smaller model)

 H_1 : Model 2 (larger model with the interaction term) has better performance than Model 1 (smaller model)

```
# do NOT use REML for likelihood ratio
LMM.2 <- lme(frequency ~ gender + attitude + gender*attitude, random = ~1 | subject, data = df, method
summary(LMM.2)
## Linear mixed-effects model fit by maximum likelihood
##
    Data: df
##
          AIC
                   BIC
                          logLik
     826.2508 840.8357 -407.1254
##
##
## Random effects:
   Formula: ~1 | subject
           (Intercept) Residual
##
              19.50493 28.67234
## StdDev:
## Fixed effects: frequency ~ gender + attitude + gender * attitude
                            Value Std.Error DF
                                                 t-value p-value
## (Intercept)
                        260.68571 13.200754 76 19.747790 0.0000
## genderM
                       -116.19524 18.668685 4 -6.224072
                                                         0.0034
## attitudepol
                        -27.40000 9.066991 76 -3.021951
                                                          0.0034
## genderM:attitudepol
                        14.79524 12.822662 76 1.153835 0.2522
   Correlation:
##
                       (Intr) gendrM atttdp
## genderM
                       -0.707
                       -0.343 0.243
## attitudepol
## genderM:attitudepol 0.243 -0.343 -0.707
## Standardized Within-Group Residuals:
##
          Min
                      Q1
                                Med
                                            QЗ
                                                      Max
## -2.2856421 -0.5245601 -0.1718554 0.4929026 3.2293520
##
## Number of Observations: 84
## Number of Groups: 6
LMM.1 <- lme(frequency ~ gender + attitude, random = ~1 | subject, data = df, method = 'ML')
# Compare
anova(LMM.1, LMM.2)
        Model df
                       AIC
                                BIC
                                       logLik
                                                Test L.Ratio p-value
            1 5 825.6363 837.7904 -407.8182
## LMM.1
             2 6 826.2508 840.8357 -407.1254 1 vs 2 1.385523 0.2392
## LMM.2
```

The p-value of the test is 0.2392, so we fail to reject the null hypothesis. We conclude that the interaction term is not significantly associated with pitch.

(d) Fit a mixed effects model with random intercepts for different subjects and scenrio (gender and attitude being the fixed effects).

Fit the model with random intercepts

```
# grouped data
LMM.3 <- lme4::lmer(frequency ~ gender + attitude + (1|subject) + (1|scenario),
                    data = df, REML = TRUE)
summary(LMM.3)
## Linear mixed model fit by REML ['lmerMod']
## Formula: frequency ~ gender + attitude + (1 | subject) + (1 | scenario)
##
      Data: df
##
## REML criterion at convergence: 784.1
## Scaled residuals:
               1Q Median
##
## -2.2690 -0.6331 -0.0878 0.5204 3.5326
##
## Random effects:
  Groups
           Name
                         Variance Std.Dev.
  scenario (Intercept) 224.5
                                  14.98
                                  24.76
   subject (Intercept) 613.2
## 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
              -0.651
## genderM
## attitudepol -0.171 0.000
```

Covariance matrix for a subject Y_i

The variance of residuals is $\sigma^2 = 637.8$, and the subject-specific variance for random effects of group scenario is $\sigma_{b_{1i}}^2 = 224.5$, and that of group subject $\sigma_{b_{2i}}^2 = 613.2$. And there are 14 observations for each subject. Therefore $Var[Y_i] = E[(Y_i - \mu)^2] = E[(b_{1i} + b_{2i} + \epsilon_i)^2] = \sigma^2 + \sigma_{b_{1i}}^2 + \sigma_{b_{2i}}^2 = 1475.5$, $Cov[Y_{ij}, Y_{ik}] = 837.7$

So the covariance matrix becomes:

```
\begin{bmatrix} 1475.5 & 837.7 & 837.7 & \dots & 837.7 \\ 837.7 & 1475.5 & 837.7 & \dots & 837.7 \\ 837.7 & 837.7 & \dots & \dots & \dots \\ \dots & \dots & 837.7 & 837.7 & 1475.5 \end{bmatrix}_{14\cdot 14}
```

Interpretation of attitude coefficient

With the gender of the subject known, formal registers have a lower average pitch of $20.002~\mathrm{Hz}$ comparing with the average pitch of informal register.