

36-763 Homework 5

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December 10, 2013

1

```
> ratings=read.csv("ratings.csv",header=T)
```

(a)

Analysis of Variance Table

Model 1: Classical ~ Harmony + Instrument

Model 2: Classical ~ Voice + Harmony + Instrument

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	2487	13193				
2	2485	13108	2	85.64	8.1181	0.0003061 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Analysis of Variance Table

Model 1: Classical ~ Voice + Instrument

Model 2: Classical ~ Voice + Harmony + Instrument

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	2488	13381				
2	2485	13108	3	273.65	17.293	4.107e-11 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Analysis of Variance Table

Model 1: Classical ~ Voice + Harmony

Model 2: Classical ~ Voice + Harmony + Instrument

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	2487	17235				
2	2485	13108	2	4127.6	391.26	< 2.2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
Call:
lm(formula = Classical ~ Voice + Harmony + Instrument, data = ratings)
```

Residuals:

Min	1Q	Median	3Q	Max
-6.8718	-1.7137	-0.0297	1.7576	11.4766

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	4.34016	0.12987	33.420	< 2e-16 ***
Voicepar3rd	-0.41247	0.11271	-3.660	0.000258 ***
Voicepar5th	-0.37058	0.11264	-3.290	0.001016 **
HarmonyI-V-IV	-0.03108	0.13008	-0.239	0.811168
HarmonyI-V-VI	0.76909	0.13008	5.913	3.83e-09 ***
HarmonyIV-I-V	0.05007	0.12997	0.385	0.700092
Instrumentpiano	1.37359	0.11298	12.158	< 2e-16 ***
Instrumentstring	3.13312	0.11230	27.899	< 2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.297 on 2485 degrees of freedom

(27 observations deleted due to missingness)

Multiple R-squared: 0.255, Adjusted R-squared: 0.2529

F-statistic: 121.5 on 7 and 2485 DF, p-value: < 2.2e-16

All three variables are highly significant in linear regression model. Yet we see that in the factor variables *Harmony* and *Voice* only one of the levels has significant different mean from other levels. I re-coded these factor variables to examine the effect of the level of interest alone.

```
> ratings$Voice1=(ratings$Voice=="contrary")
> ratings$Harmony1=(ratings$Harmony=="I-V-VI")
> reg1a2=lm(Classical~Voice1+Harmony1+Instrument,data=ratings)
> summary(reg1a2)
```

Call:

```
lm(formula = Classical ~ Voice1 + Harmony1 + Instrument, data = ratings)
```

Residuals:

Min	1Q	Median	3Q	Max
-6.851	-1.720	-0.088	1.758	11.520

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	3.95492	0.08995	43.970	< 2e-16 ***
Voice1TRUE	0.39166	0.09753	4.016	6.10e-05 ***

```

Harmony1TRUE      0.76283    0.10625    7.180 9.17e-13 ***
Instrumentpiano    1.37369    0.11292   12.165 < 2e-16 ***
Instrumentstring    3.13310    0.11225   27.913 < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

Residual standard error: 2.296 on 2488 degrees of freedom
(27 observations deleted due to missingness)
Multiple R-squared:  0.2549,    Adjusted R-squared:  0.2537
F-statistic: 212.8 on 4 and 2488 DF,  p-value: < 2.2e-16

```

```
> BIC(reg1a2)
```

```
[1] 11259.92
```

```
> BIC(reg1a1)
```

```
[1] 11282.84
```

The new model has a lower BIC. And all levels interested in are significant.

(b)

(i)

$$Y_j = \alpha_{0i[j]} + X_j\beta_1 + \epsilon_j, \epsilon_j \sim N(0, \sigma^2)$$

$$\alpha_{0i} = \beta_0 + \eta_i, \eta_i \sim N(0, \tau^2)$$

Where Y_j is the classical rating for a particular entryl. $i[j]$ is the corresponding subject id for entry j . α_i is the random intercept for subject i. X_j is the vector of *Harmony*, *Voice*, *Instrument*(or their corresponding re-coded variables) for entry j .

(ii)

```
> reg1b1=lmer(Classical ~ Voice+Harmony+Instrument+(1|Subject),data=ratings)
> BIC(reg1b1)
```

```
[1] 10549.73
```

```
> BIC(reg1a1)
```

```
[1] 11282.84
```

```
> display(reg1b1)
```

```
lmer(formula = Classical ~ Voice + Harmony + Instrument + (1 |
  Subject), data = ratings)
```

	coef.est	coef.se
(Intercept)	4.34	0.19
Voicepar3rd	-0.42	0.09
Voicepar5th	-0.37	0.09
HarmonyI-V-IV	-0.03	0.11
HarmonyI-V-VI	0.77	0.11
HarmonyIV-I-V	0.05	0.11
Instrumentpiano	1.38	0.09
Instrumentstring	3.13	0.09

Error terms:

Groups	Name	Std.Dev.
Subject	(Intercept)	1.30
	Residual	1.89

```
number of obs: 2493, groups: Subject, 70
AIC = 10491.5, DIC = 10426.2
deviance = 10448.9
```

The multilevel model has a much lower BIC compared to the linear regression model.

```
> exactRLRT(reg1b1)
```

simulated finite sample distribution of RLRT.

(p-value based on 10000 simulated values)

data:

```
RLRT = 763.3759, p-value < 2.2e-16
```

The RLRT method also agrees with the multilevel model.

(iii)

```
> reg1b2 = lmer(Classical ~ Instrument + Voice + (1 | Subject),
+               data = ratings)
> anova(reg1b2, reg1b1)
```

Data: ratings

Models:

```
reg1b2: Classical ~ Instrument + Voice + (1 | Subject)
```

```
reg1b1: Classical ~ Voice + Harmony + Instrument + (1 | Subject)
```

	Df	AIC	BIC	logLik	deviance	Chisq	Chi	Df	Pr(>Chisq)
reg1b2	7	10539	10580	-5262.4	10525				

```
reg1b1 10 10469 10527 -5224.4    10449 75.931      3  2.288e-16 ***
---
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Hence *Harmony* is significant and should be included.

```
> reg1b3 = lmer(Classical ~ Harmony + Voice + (1 | Subject),
+               data = ratings)
> anova(reg1b3,reg1b1)
```

Data: ratings

Models:

```
reg1b3: Classical ~ Harmony + Voice + (1 | Subject)
reg1b1: Classical ~ Voice + Harmony + Instrument + (1 | Subject)
      Df    AIC    BIC logLik deviance Chisq Chi Df Pr(>Chisq)
reg1b3  8 11408 11455 -5696.2    11392
reg1b1 10 10469 10527 -5224.4    10449 943.59      2 < 2.2e-16 ***
---
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Hence *Instrument* is significant and should be included.

```
> reg1b4 = lmer(Classical ~ Harmony + Instrument + (1 | Subject),
+               data = ratings)
> anova(reg1b4,reg1b1)
```

Data: ratings

Models:

```
reg1b4: Classical ~ Harmony + Instrument + (1 | Subject)
reg1b1: Classical ~ Voice + Harmony + Instrument + (1 | Subject)
      Df    AIC    BIC logLik deviance Chisq Chi Df Pr(>Chisq)
reg1b4  8 10489 10536 -5236.6    10473
reg1b1 10 10469 10527 -5224.4    10449 24.24      2  5.45e-06 ***
---
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Hence *Voice* is significant and should be included.

(c)

(i)

```
> reg1c1=lmer(Classical ~ Harmony + Instrument + Voice + (1|Subject:Harmony) +
+             (1 | Subject:Instrument)+ (1|Subject:Voice), data = ratings)
```

Model	AIC	BIC	DIC
reg1a1	11467.34	11519.82	
reg1b1	10755.46	10813.78	10690.9
reg1c1	10307.76	10377.75	10247.9

reg1c1 turns out to be the best model.

(ii)

```
> display(reg1c1)
```

```
lmer(formula = Classical ~ Harmony + Instrument + Voice + (1 |  
      Subject:Harmony) + (1 | Subject:Instrument) + (1 | Subject:Voice),  
      data = ratings)
```

	coef.est	coef.se
(Intercept)	4.34	0.21
HarmonyI-V-IV	-0.03	0.14
HarmonyI-V-VI	0.77	0.14
HarmonyIV-I-V	0.06	0.14
Instrumentpiano	1.36	0.26
Instrumentstring	3.13	0.26
Voicepar3rd	-0.41	0.08
Voicepar5th	-0.37	0.08

Error terms:

Groups	Name	Std.Dev.
Subject:Harmony	(Intercept)	0.67
Subject:Voice	(Intercept)	0.17
Subject:Instrument	(Intercept)	1.48
Residual		1.56

number of obs: 2493, groups: Subject:Harmony, 280; Subject:Voice, 210; Subject:Instrument, 203
AIC = 10075.5, DIC = 10015.5
deviance = 10033.5

From the summary table for `reg1c1`, we see that all three variables have levels with parameter estimate greater than twice their standard error. Hence all three variables (or at least the levels interested in) are significant. The variance of *Instrument* is comparable to that of the residual while the other two are much smaller. This might indicate other variables should be included in the model.

(iii)

$$Y_j = X_j\beta + \sum_{k=1}^3 \alpha_{h_k[i[j]]} + \epsilon_j, \epsilon_j \sim N(0, \sigma^2)$$

$$\alpha_{h_k} = \mu_{0k} + \eta_{h_k}, \eta_{h_k} \sim N(0, \tau_k), k = 1, 2, 3$$

Y_j is the classical rating for j th entry. $i[j]$ is the corresponding subject id for entry j . α_{h_k} is the random intercept for subject i combined with a level of *Harmony* (or *Voice* or *Instrument*). X_j is the vector of *Harmony*, *Voice*, *Instrument* for entry j .

2

(a)

The best model is reg1c1 in 1. After testing the significance of all variables using ANOVA procedure, I ended up with this model including three more variables.

```
> reg2a=update(reg1c1, . ~. + X16.minus.17+
+           X1990s2000s.minus.1960s1970s+ GuitarPlay)
> display(reg2a)
```

```
lmer(formula = Classical ~ Harmony + Instrument + Voice + (1 |
  Subject:Harmony) + (1 | Subject:Instrument) + (1 | Subject:Voice) +
  X16.minus.17 + X1990s2000s.minus.1960s1970s + GuitarPlay,
  data = ratings)
```

	coef.est	coef.se
(Intercept)	4.33	0.26
HarmonyI-V-IV	-0.03	0.15
HarmonyI-V-VI	0.76	0.15
HarmonyIV-I-V	0.08	0.15
Instrumentpiano	1.48	0.26
Instrumentstring	3.26	0.26
Voicepar3rd	-0.42	0.09
Voicepar5th	-0.37	0.09
X16.minus.17	-0.10	0.04
X1990s2000s.minus.1960s1970s	0.00	0.06
GuitarPlay	0.09	0.08

Error terms:

Groups	Name	Std.Dev.
Subject:Harmony	(Intercept)	0.67
Subject:Voice	(Intercept)	0.18
Subject:Instrument	(Intercept)	1.39
Residual		1.57

number of obs: 2325, groups: Subject:Harmony, 260; Subject:Voice, 195; Subject:Instrument, 195

AIC = 9420, DIC = 9331.2

deviance = 9360.6

```
> BIC(reg2a)
```

```
[1] 9506.232
```

(b)

```
> reg2b0 = lmer(Classical ~ (1 | Subject:Harmony) + (1 | Subject:Instrument) +
+           (1 | Subject:Voice) + Harmony + Instrument + Voice + X16.minus.17 +
```

```

+           X1990s2000s.minus.1960s1970s + GuitarPlay,data=ratings)
> reg2bh = lmer( Classical ~ (1 | Subject:Harmony) + Harmony + Instrument +
+           Voice + X16.minus.17 + X1990s2000s.minus.1960s1970s +
+           GuitarPlay, data = ratings)
> reg2bh1 = update(reg2b0, . ~ . - (1 | Subject:Harmony))
> exactRLRT(reg2bh, m0 = reg2bh1, mA = reg2b0)

```

simulated finite sample distribution of RLRT.

(p-value based on 10000 simulated values)

data:

RLRT = 94.5898, p-value < 2.2e-16

Hence we keep the random effect for *Harmony*.

```

> reg2bi = lmer( Classical ~ (1 | Subject:Instrument) + Harmony + Instrument +
+           Voice + X16.minus.17 + X1990s2000s.minus.1960s1970s +
+           GuitarPlay, data = ratings)
> reg2bi1 = update(reg2b0, . ~ . - (1 | Subject:Instrument))
> exactRLRT(reg2bi, m0 = reg2bi1, mA = reg2b0)

```

simulated finite sample distribution of RLRT.

(p-value based on 10000 simulated values)

data:

RLRT = 498.568, p-value < 2.2e-16

Hence we keep the random effect for *Instrument*.

```

> reg2bv = lmer( Classical ~ (1 | Subject:Voice) + Harmony + Instrument +
+           Voice + X16.minus.17 + X1990s2000s.minus.1960s1970s +
+           GuitarPlay, data = ratings)
> reg2bv1 = update(reg2b0, . ~ . - (1 | Subject:Voice))
> exactRLRT(reg2bv, m0 = reg2bv1, mA = reg2b0)

```

simulated finite sample distribution of RLRT.

(p-value based on 10000 simulated values)

data:

RLRT = 1.3049, p-value = 0.1195

Hence we don't keep the random effect for *Voice*.

```
> reg2b=lmer( Classical ~ (1 | Subject:Harmony) +(1 | Subject:Instrument) + Harmony +
+           Voice + X16.minus.17 + X1990s2000s.minus.1960s1970s +
+           GuitarPlay, data = ratings)
> display(reg2b)
```

```
lmer(formula = Classical ~ (1 | Subject:Harmony) + (1 | Subject:Instrument) +
      Harmony + Instrument + Voice + X16.minus.17 + X1990s2000s.minus.1960s1970s +
      GuitarPlay, data = ratings)
```

	coef.est	coef.se
(Intercept)	4.33	0.26
HarmonyI-V-IV	-0.03	0.15
HarmonyI-V-VI	0.76	0.15
HarmonyIV-I-V	0.08	0.15
Instrumentpiano	1.47	0.26
Instrumentstring	3.26	0.26
Voicepar3rd	-0.42	0.08
Voicepar5th	-0.36	0.08
X16.minus.17	-0.10	0.04
X1990s2000s.minus.1960s1970s	0.00	0.06
GuitarPlay	0.09	0.08

Error terms:

Groups	Name	Std.Dev.
Subject:Harmony	(Intercept)	0.67
Subject:Instrument	(Intercept)	1.40
Residual		1.58

```
number of obs: 2325, groups: Subject:Harmony, 260; Subject:Instrument, 195
AIC = 9419.3, DIC = 9332
deviance = 9361.6
```

(c)

The parameter for each level of fixed effects is the change compared to baseline case of being in that certain level. The random effects are the corresponding group mean given subject and the factor variable(harmony or instrument).

3

```
> table(ratings$Selfdeclare)
```

1	2	3	4	5	6
576	936	468	432	72	36

```
> ratings$Selfdeclare1=(ratings$Selfdeclare>2)
> reg3=lmer(Classical ~ Selfdeclare1+Selfdeclare1:Harmony+Selfdeclare1:X16.minus.17 +
+           Selfdeclare1:X1990s2000s.minus.1960s1970s + Selfdeclare1:GuitarPlay+
+           Selfdeclare1:Instrument+Selfdeclare1:Voice+ X16.minus.17 +
+           X1990s2000s.minus.1960s1970s + GuitarPlay+Instrument+Harmony+
+           Voice+(1|Subject:Harmony)+(1|Subject:Instrument),data=ratings)
> BIC(reg3)
```

```
[1] 9557.078
```

```
> display(reg3)
```

```
lmer(formula = Classical ~ Selfdeclare1 + Selfdeclare1:Harmony +
      Selfdeclare1:X16.minus.17 + Selfdeclare1:X1990s2000s.minus.1960s1970s +
      Selfdeclare1:GuitarPlay + Selfdeclare1:Instrument + Selfdeclare1:Voice +
      X16.minus.17 + X1990s2000s.minus.1960s1970s + GuitarPlay +
      Instrument + Harmony + Voice + (1 | Subject:Harmony) + (1 |
      Subject:Instrument), data = ratings)
```

	coef.est	coef.se
(Intercept)	4.13	0.31
Selfdeclare1TRUE	0.61	0.61
X16.minus.17	-0.02	0.05
X1990s2000s.minus.1960s1970s	0.01	0.07
GuitarPlay	1.26	0.38
Instrumentpiano	1.64	0.32
Instrumentstring	3.46	0.32
HarmonyI-V-IV	-0.04	0.18
HarmonyI-V-VI	0.31	0.18
HarmonyIV-I-V	0.03	0.18
Voicepar3rd	-0.40	0.10
Voicepar5th	-0.29	0.10
Selfdeclare1TRUE:HarmonyI-V-IV	0.03	0.29
Selfdeclare1TRUE:HarmonyI-V-VI	1.18	0.29
Selfdeclare1TRUE:HarmonyIV-I-V	0.14	0.29
Selfdeclare1TRUE:X16.minus.17	-0.22	0.09
Selfdeclare1TRUE:X1990s2000s.minus.1960s1970s	-0.05	0.14
Selfdeclare1TRUE:GuitarPlay	-1.18	0.39
Selfdeclare1TRUE:Instrumentpiano	-0.44	0.52
Selfdeclare1TRUE:Instrumentstring	-0.51	0.52
Selfdeclare1TRUE:Voicepar3rd	-0.05	0.17
Selfdeclare1TRUE:Voicepar5th	-0.19	0.17

```
Error terms:
```

Groups	Name	Std.Dev.
Subject:Harmony	(Intercept)	0.61

```

Subject:Instrument (Intercept) 1.37
Residual                      1.58
---
number of obs: 2325, groups: Subject:Harmony, 260; Subject:Instrument, 195
AIC = 9413.3, DIC = 9278.7
deviance = 9321.0

```

The variable *selfdeclare1* (equal to 1 if self-declare is greater than 2 and 0 otherwise) isn't significant itself, yet several interaction terms are significant. This implies that self-identification as a musician does change people's pattern of classical music controlling for the other variables in this model.

4

(a)

```

> reg4a1=lm(Popular ~ Voice+Harmony+Instrument,data=ratings)
> reg4av=lm(Popular ~ Harmony+Instrument,data=ratings)
> reg4ai=lm(Popular ~ Voice+Harmony,data=ratings)
> reg4ah=lm(Popular ~ Voice+Instrument,data=ratings)
> anova(reg4a1,reg4av)

```

Analysis of Variance Table

Model 1: Popular ~ Voice + Harmony + Instrument

Model 2: Popular ~ Harmony + Instrument

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	2485	12656				
2	2487	12672	-2	-15.263	1.4984	0.2237

```

> anova(reg4a1,reg4ai)

```

Analysis of Variance Table

Model 1: Popular ~ Voice + Harmony + Instrument

Model 2: Popular ~ Voice + Harmony

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	2485	12656				
2	2487	15580	-2	-2923.9	287.05	< 2.2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

> anova(reg4a1,reg4ah)

```

Analysis of Variance Table

Model 1: Popular ~ Voice + Harmony + Instrument

Model 2: Popular ~ Voice + Instrument

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	2485	12656				
2	2488	12688	-3	-31.092	2.0349	0.1069

Only instrument is significant here and we leave other fixed effects out.

need to leave experimental factors in the model

```
> reg4a2=lmer(Popular ~ Instrument+(1|Subject:Voice)+(1|Subject:Harmony)+(1|Subject:Instrument), data = ratings)
> display(reg4a2)
```

```
lmer(formula = Popular ~ Instrument + (1 | Subject:Voice) + (1 | Subject:Harmony) + (1 | Subject:Instrument), data = ratings)
              coef.est coef.se
(Intercept)      6.57    0.18
Instrumentpiano  -0.95    0.25
Instrumentstring -2.61    0.25
```

Error terms:

Groups	Name	Std.Dev.
Subject:Harmony	(Intercept)	0.65
Subject:Instrument	(Intercept)	1.41
Subject:Voice	(Intercept)	0.20
Residual		1.58

number of obs: 2493, groups: Subject:Harmony, 280; Subject:Instrument, 210; Subject:Voice, 203
AIC = 10083.9, DIC = 10060.4
deviance = 10065.2

```
> reg4a2v=update(reg4a2, . ~ .-(1|Subject:Voice))
> reg4a2i=update(reg4a2, . ~ .-(1|Subject:Instrument))
> reg4a2h=update(reg4a2, . ~ .-(1|Subject:Harmony))
> anova(reg4a2,reg4a2v)
```

Data: ratings

Models:

reg4a2v: Popular ~ Instrument + (1 | Subject:Harmony) + (1 | Subject:Instrument)

reg4a2: Popular ~ Instrument + (1 | Subject:Voice) + (1 | Subject:Harmony) +

reg4a2: (1 | Subject:Instrument)

	Df	AIC	BIC	logLik	deviance	Chisq	Chi	Df	Pr(>Chisq)
reg4a2v	6	10079	10114	-5033.5	10067				
reg4a2	7	10079	10120	-5032.6	10065	1.8017		1	0.1795

```
> anova(reg4a2,reg4a2i)
```

Data: ratings

Models:

```
reg4a2i: Popular ~ Instrument + (1 | Subject:Voice) + (1 | Subject:Harmony)
reg4a2: Popular ~ Instrument + (1 | Subject:Voice) + (1 | Subject:Harmony) +
reg4a2:      (1 | Subject:Instrument)
```

	Df	AIC	BIC	logLik	deviance	Chisq	Chi	Df	Pr(>Chisq)
reg4a2i	6	10571	10606	-5279.3	10559				
reg4a2	7	10079	10120	-5032.6	10065	493.39		1	< 2.2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
> anova(reg4a2,reg4a2h)
```

Data: ratings

Models:

```
reg4a2h: Popular ~ Instrument + (1 | Subject:Voice) + (1 | Subject:Instrument)
reg4a2: Popular ~ Instrument + (1 | Subject:Voice) + (1 | Subject:Harmony) +
reg4a2:      (1 | Subject:Instrument)
```

	Df	AIC	BIC	logLik	deviance	Chisq	Chi	Df	Pr(>Chisq)
reg4a2h	6	10169	10204	-5078.5	10157				
reg4a2	7	10079	10120	-5032.6	10065	91.913		1	< 2.2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
> reg4a0=lmer(Popular ~ Instrument+(1|Subject:Harmony)+(1|Subject:Instrument),data=rat
> display(reg4a0)
```

```
lmer(formula = Popular ~ Instrument + (1 | Subject:Harmony) +
      (1 | Subject:Instrument), data = ratings)
```

	coef.est	coef.se
(Intercept)	6.57	0.18
Instrumentpiano	-0.95	0.25
Instrumentstring	-2.61	0.25

Error terms:

Groups	Name	Std.Dev.
Subject:Harmony	(Intercept)	0.65
Subject:Instrument	(Intercept)	1.42
Residual		1.59

number of obs: 2493, groups: Subject:Harmony, 280; Subject:Instrument, 210
AIC = 10083.7, DIC = 10062.3
deviance = 10067.0

The random effects of *Instrument* and *Harmony* are significant, which we would leave in.

(b)

```
> reg4b0=lmer(Popular ~ Instrument+(1|Subject:Harmony)+(1|Subject:Instrument),data=rat
```

By adding single variables and comparing BIC, I ended up with the following model:

```
> reg4b1=lmer(Popular ~ (1 | Subject:Harmony) + (1 | Subject:Instrument)+
+           Harmony + Instrument + Voice +OMSI+
+           X1990s2000s.minus.1960s1970s,data = ratings)
> display(reg4b1)
```

need to be
sure no cases
were dropped
because of
NA's, which
would change
the sample
size and
invalidate BIC
comparisons.

```
lmer(formula = Popular ~ (1 | Subject:Harmony) + (1 | Subject:Instrument) +
      Harmony + Instrument + Voice + OMSI + X1990s2000s.minus.1960s1970s,
      data = ratings)
```

	coef.est	coef.se
(Intercept)	6.31	0.27
HarmonyI-V-IV	-0.03	0.15
HarmonyI-V-VI	-0.29	0.15
HarmonyIV-I-V	-0.23	0.15
Instrumentpiano	-0.99	0.26
Instrumentstring	-2.63	0.26
Voicepar3rd	0.19	0.08
Voicepar5th	0.18	0.08
OMSI	0.00	0.00
X1990s2000s.minus.1960s1970s	0.01	0.06

Error terms:

Groups	Name	Std.Dev.
Subject:Harmony	(Intercept)	0.65
Subject:Instrument	(Intercept)	1.39
Residual		1.59

number of obs: 2325, groups: Subject:Harmony, 260; Subject:Instrument, 195

AIC = 9454.8, DIC = 9358.2

deviance = 9393.5

One level of harmony is significant. Instrument is significant and voice is significant in the model.

(c)

```
> reg4c=lmer(Popular ~ (1 | Subject:Harmony) + (1 | Subject:Instrument)+
+           Selfdeclare1:Harmony + Selfdeclare1:Instrument + Selfdeclare1:Voice +
+           Selfdeclare1 +Selfdeclare1:OMSI+Selfdeclare1:X1990s2000s.minus.1960s1970s
+           + Harmony + Instrument + Voice + Selfdeclare1 +OMSI+
+           X1990s2000s.minus.1960s1970s      ,data = ratings)
> display(reg4c)
```

```
lmer(formula = Popular ~ (1 | Subject:Harmony) + (1 | Subject:Instrument) +
  Selfdeclare1:Harmony + Selfdeclare1:Instrument + Selfdeclare1:Voice +
  Selfdeclare1 + Selfdeclare1:OMSI + Selfdeclare1:X1990s2000s.minus.1960s1970s +
  Harmony + Instrument + Voice + Selfdeclare1 + MSI + X1990s2000s.minus.1960s1970s,
  data = ratings)
```

	coef.est	coef.se
(Intercept)	6.44	0.38
Selfdeclare1TRUE	-0.18	0.63
HarmonyI-V-IV	-0.10	0.18
HarmonyI-V-VI	0.00	0.18
HarmonyIV-I-V	-0.22	0.18
Instrumentpiano	-1.15	0.33
Instrumentstring	-2.90	0.33
Voicepar3rd	0.22	0.10
Voicepar5th	0.12	0.10
OMSI	0.00	0.00
X1990s2000s.minus.1960s1970s	0.00	0.07
Selfdeclare1TRUE:HarmonyI-V-IV	0.18	0.30
Selfdeclare1TRUE:HarmonyI-V-VI	-0.77	0.30
Selfdeclare1TRUE:HarmonyIV-I-V	-0.02	0.30
Selfdeclare1TRUE:Instrumentpiano	0.42	0.53
Selfdeclare1TRUE:Instrumentstring	0.70	0.53
Selfdeclare1TRUE:Voicepar3rd	-0.09	0.17
Selfdeclare1TRUE:Voicepar5th	0.16	0.17
Selfdeclare1TRUE:OMSI	0.00	0.00
Selfdeclare1TRUE:X1990s2000s.minus.1960s1970s	0.03	0.13

Error terms:

Groups	Name	Std.Dev.
Subject:Harmony	(Intercept)	0.62
Subject:Instrument	(Intercept)	1.41
Residual		1.59

number of obs: 2325, groups: Subject:Harmony, 260; Subject:Instrument, 195
 AIC = 9477.2, DIC = 9322.8
 deviance = 9377.0

Still, selfdeclare itself isn't significant yet quite some interaction terms are significant. Meaning change in features of the music have different influence on a person's rating of the popularity of that piece of music whether or not this person identify himself as a musician.

5

5.1 Using classical as dependent variable

I started with multiple linear regressions and included all three fixed effects in it. Analysis of variance confirms that all three effects are significant, while from the regression table (in (1)) we see that certain levels of the variables (conveniently the levels mentioned in his main hypotheses) are significant. Recoding these levels in the linear regression model leads to a model with better BIC. I then added the random intercept for each person, which turned out to be helpful to the model. Then I included the "personal biases" in the model and found out that all three variables combined with person should be included. The variance component isn't a standard variance model since the residual variance is still higher than two random effects. We hence need more covariates in the model. With further anova testing I included *GuitarPlay*, *X16.minus.17*, *X1990s2000s.minus.1960s1970s* in the model. And judging from the variance components, variance of the residual is roughly comparable to that of random effects.

Also, after recoding the self-declare variable into two levels with roughly equal subjects, we find that the variable itself isn't significant yet some of its interactions with other variables are. which indicates that self-identified as musician produces another layer of bias in rating a piece of music as classical. To be specific, the interaction of self-declare above 2 and Harmony level "IV-V-VI", *X16.minus.17* and *PlayGuitar* are significant.

5.2 Using popular as dependent variable

As before, I started with linear models and found only instrument is significant. I then included the personal random intercept and found the model improved in that its BIC fell. Then the three "personal bias" random effects are tested and found instrument and harmony significant. Variance of the residual is comparable to one of the random effects and bigger than the other one, implying need for more variables. By further testing I added *OMIS*, *X1990s2000s.minus.1960s1970s* to the model.

The recoded *selfdeclare* variable is still not significant itself. Yet its interaction with *Harmony* level "I-V-VI" is significant.

4: 16

5: 20

36