

Experiment 4: Main Analyses

Bethany Gardner

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Setup

Load data and select columns used in model. See data/exp4_data_about.txt for more details.

```
d <- read.csv("../data/exp4_data.csv", stringsAsFactors=TRUE) %>%
  rename("Participant"="SubjID", "Item"="Name") %>%
  select(Participant, Condition, GenderRating, Item, Male, Female, Other)
str(d)
```

```
## 'data.frame': 8771 obs. of 7 variables:
## $ Participant : Factor w/ 1253 levels "R_00dmdQaotbTidXz",...: 1001 1001 1001 1001 1001 1001 1001 1001 23
## $ Condition : Factor w/ 3 levels "first","full",...: 1 1 1 1 1 1 1 1 1 ...
## $ GenderRating: num 6.24 2.61 6.82 5.34 1.28 4.39 3.87 5.22 1.24 5.86 ...
## $ Item : Factor w/ 63 levels "Ashley Cook",...: 1 18 21 22 25 28 50 5 7 15 ...
## $ Male : int 0 1 0 0 1 1 1 1 1 0 ...
## $ Female : int 1 0 1 1 0 0 0 0 0 1 ...
## $ Other : int 0 0 0 0 0 0 0 0 0 0 ...
```

Center gender rating for names: Original scale from 1 to 7, with 1 as most masculine and 7 as most feminine. Mean-centered with higher still as more feminine.

```
d %>% mutate(GenderRatingCentered=scale(d$GenderRating, scale=FALSE))
```

Set contrasts for name conditions, now weighted to account for uneven sample sizes. This uses Scott Fraundorf's function for weighted contrasts. (The psycholing package version doesn't support doing 2v1 comparisons, only 1v1.) Condition1 is Last vs First+Full. Condition2 is First vs Full.

```
source("centerfactor.R")
contrasts(d$Condition) <- centerfactor(d$Condition, c("last","first"))
contrasts(d$Condition)
```

```
##           [,1]      [,2]
## first  0.3312051 -0.497605746
## full   0.3312051  0.502394254
## last   -0.6687949  0.002394254
```

Data Summary

Responses by condition.

```
d <- d %>% mutate(ResponseAll=case_when(
  Male==1 ~ "Male",
  Female==1 ~ "Female",
  Other==1 ~ "Other"))

d.count_responses <- d %>% group_by(Condition, ResponseAll) %>%
  summarise(n=n()) %>%
  pivot_wider(names_from=ResponseAll,
    values_from=n) %>%
  mutate(Female_MaleOther = Female / (Male+Other),
    Female_Male = Female / Male)

kable(d.count_responses)
```

Condition	Female	Male	Other	Female_MaleOther	Female_Male
first	1381	1511	62	0.8779402	0.9139643
full	1380	1416	116	0.9007833	0.9745763
last	1292	1529	84	0.8009919	0.8449967

- First name condition has second-most (slightly) *female* responses
- Full name condition has most *female* responses
- Last name condition has fewest *female* responses

Main Model

Because Experiment 4 always introduces the character with a full name, then manipulates the name form in the subsequent 3 references, the main analysis is 1 model, as opposed to the 2 for Experiments 1 and 2.

Effects of Name Condition (first name, last name, full name) and first name Gender Rating (centered, + fem, -masc) on the likelihood of *female* responses, as opposed to *male* and *other* responses. Participant and Item are included as random intercepts, with items defined as the unique first, last and first + last name combinations. Condition1 is the contrast between last and first+full. Condition2 is the contrast between first and full.

```
m.main <- glmer(Female ~ Condition * GenderRatingCentered +
  (1|Participant) + (1|Item),
  data=d, family=binomial)
```

```
m.main_tidy <- tidy(m.main)
summary(m.main)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: Female ~ Condition * GenderRatingCentered + (1 | Participant) +
## (1 | Item)
## Data: d
##
##      AIC      BIC   logLik deviance df.resid
##  9145.4   9202.1  -4564.7   9129.4     8763
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.4531 -0.5754 -0.2627  0.5724  5.4528
##
## Random effects:
## Groups      Name      Variance Std.Dev.
## Participant (Intercept) 0.2014   0.4488
## Item          (Intercept) 0.3598   0.5999
## Number of obs: 8771, groups: Participant, 1253; Item, 63
##
## Fixed effects:
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -0.25605    0.08160  -3.138 0.001703 **
## Condition1       0.12636    0.06170   2.048 0.040558 *
## Condition2       0.06836    0.07245   0.944 0.345422
## GenderRatingCentered 0.76407    0.04590  16.648 < 2e-16 ***
## Condition1:GenderRatingCentered 0.13147    0.03451   3.809 0.000139 ***
## Condition2:GenderRatingCentered -0.10288    0.04204  -2.447 0.014403 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) Cndtn1 Cndtn2 GndrRC C1:GRC
## Condition1    0.012
## Condition2  -0.012 -0.016
## GndrRtngCnt -0.028  0.002  0.011
## Cndtn1:GnRC  0.001 -0.121  0.016  0.035
## Cndtn2:GnRC  0.011  0.016 -0.112 -0.030 -0.046
```

- Less likely to recall character as female overall
- More likely to recall character as female in the First and Full Name conditions than in the Last Name condition
- More likely to recall character as female as first names become more feminine

Double check the directions of the interactions:

L v F+F Interaction

Dummy code to get the gender rating effect for just the First and Full Name conditions.

```
d$FFdummy = as.numeric(d$Condition)
d$FFdummy[d$FFdummy == 1] <- 0
d$FFdummy[d$FFdummy == 2] <- 0
d$FFdummy[d$FFdummy == 3] <- 1
with(d, tapply(FFdummy, list(Condition), mean))
```

```
## first full last
##      0      0      1
```

```
m.genderRating.FF <- glmer(Female ~ FFdummy * GenderRatingCentered +
  (1|Participant) + (1|Item),
  data=d, family=binomial)
m.genderRating.FF_tidy <- tidy(m.genderRating.FF)
summary(m.genderRating.FF)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial (logit)
## Formula: Female ~ FFdummy * GenderRatingCentered + (1 | Participant) +
## (1 | Item)
## Data: d
##
##      AIC      BIC    logLik deviance df.resid
##  9147.7   9190.2  -4567.8   9135.7     8765
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.2676 -0.5751 -0.2669  0.5734  4.9543
##
## Random effects:
## Groups      Name      Variance Std.Dev.
## Participant (Intercept) 0.2005   0.4477
## Item          (Intercept) 0.3603   0.6002
## Number of obs: 8771, groups: Participant, 1253; Item, 63
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -0.21216    0.08437  -2.515  0.011918 *
## FFdummy         -0.12796    0.06164  -2.076  0.037901 *
## GenderRatingCentered  0.80513    0.04766  16.893 < 2e-16 ***
## FFdummy:GenderRatingCentered -0.12944    0.03446  -3.756  0.000172 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) FFdmmy GndrRC
## FFdummy      -0.253
## GndrRtngCnt -0.032  0.026
## FFdmmy:GnRC  0.027 -0.120 -0.272
```

Then dummy code to get the gender rating effect just in the Last Name condition.

```

d$Ldummy = as.numeric(d$Condition)
d$Ldummy[d$Ldummy == 1] <- 1
d$Ldummy[d$Ldummy == 2] <- 1
d$Ldummy[d$Ldummy == 3] <- 0
with(d, tapply(Ldummy, list(Condition), mean))

## first full last
##      1      1      0

m.genderRating.L <- glmer(Female ~ Ldummy * GenderRatingCentered +
  (1|Participant) + (1|Item), data=d, family=binomial)
m.genderRating.L_tidy <- tidy(m.genderRating.L)
summary(m.genderRating.L)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial (logit)
## Formula: Female ~ Ldummy * GenderRatingCentered + (1 | Participant) +
## (1 | Item)
## Data: d
##
##      AIC      BIC    logLik deviance df.resid
##  9147.7   9190.2  -4567.8   9135.7     8765
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.2676 -0.5751 -0.2669  0.5734  4.9543
##
## Random effects:
## Groups      Name      Variance Std.Dev.
## Participant (Intercept) 0.2005   0.4477
## Item          (Intercept) 0.3603   0.6002
## Number of obs: 8771, groups: Participant, 1253; Item, 63
##
## Fixed effects:
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      -0.34013    0.09101  -3.737 0.000186 ***
## Ldummy              0.12797    0.06164   2.076 0.037901 *
## GenderRatingCentered  0.67569    0.05066  13.337 < 2e-16 ***
## Ldummy:GenderRatingCentered 0.12944    0.03446   3.757 0.000172 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) Ldummy GndrRC
## Ldummy      -0.442
## GndrRtngCnt -0.049  0.057
## Ldmmy:GndRC  0.056 -0.120 -0.425

m.genderRating.FF_tidy %>%
  filter(term=="GenderRatingCentered") %>%
  select(estimate) %>% as.numeric()

```

```
## [1] 0.8051347
```

```
m.genderRating.L_tidy %>%  
  filter(term=="GenderRatingCentered") %>%  
  select(estimate) %>% as.numeric()
```

```
## [1] 0.6756906
```

Interaction indicates Gender Rating has a larger effect in the First and Full Name conditions (0.81) than in the Last Name condition (0.67). This makes sense because the gendered first name is repeated all 4x in the First and Full name conditions, but only once in the Last Name condition.

F v F Interaction

Dummy code to get the gender rating effect for just the First Name condition.

```
d$FirstDummy = as.numeric(d$Condition)  
d$FirstDummy[d$FirstDummy == 1] <- 0  
d$FirstDummy[d$FirstDummy == 2] <- 1  
d$FirstDummy[d$FirstDummy == 3] <- 1  
with(d, tapply(FirstDummy, list(Condition), mean))
```

```
## first full last  
##      0      1      1
```

```
m.genderRating.First <- glmer(Female ~ FirstDummy * GenderRatingCentered +  
  (1|Participant) + (1|Item), data=d, family=binomial)  
m.genderRating.First_tidy <- tidy(m.genderRating.First)  
summary(m.genderRating.First)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace  
##   Approximation) [glmerMod]  
##   Family: binomial ( logit )  
##   Formula: Female ~ FirstDummy * GenderRatingCentered + (1 | Participant) +  
##           (1 | Item)  
##   Data: d  
##  
##           AIC          BIC    logLik deviance df.resid  
##    9151.6    9194.0 -4569.8   9139.6      8765  
##  
## Scaled residuals:  
##      Min       1Q   Median       3Q      Max  
## -3.4660 -0.5784 -0.2629   0.5803   5.4716  
##  
## Random effects:  
##   Groups      Name      Variance Std.Dev.  
## Participant (Intercept) 0.2059   0.4538  
## Item          (Intercept) 0.3592   0.5994  
## Number of obs: 8771, groups: Participant, 1253; Item, 63  
##  
## Fixed effects:
```

```
##                                Estimate Std. Error z value Pr(>|z|)
## (Intercept)                   -0.24885    0.09228  -2.697    0.007 **
## FirstDummy                    -0.01315    0.06304  -0.209    0.835
## GenderRatingCentered          0.85944    0.05280  16.277 < 2e-16 ***
## FirstDummy:GenderRatingCentered -0.14454    0.03661  -3.948 7.88e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##      (Intr) FrstDm GndrRC
## FirstDummy  -0.468
## GndrRtngCnt -0.055  0.065
## FrstDmm:GRC  0.064 -0.125 -0.497
```

Dummy code to get the gender rating effect for just the Full Name condition.

```
d$FullDummy = as.numeric(d$Condition)
d$FullDummy[d$FullDummy == 1] <- 1
d$FullDummy[d$FullDummy == 2] <- 0
d$FullDummy[d$FullDummy == 3] <- 1
with(d, tapply(FullDummy, list(Condition), mean))
```

```
## first full last
##      1      0      1
```

```
m.genderRating.Full <- glmer(Female ~ FullDummy * GenderRatingCentered +
  (1|Participant) + (1|Item), data=d, family=binomial)
m.genderRating.Full_tidy <- tidy(m.genderRating.Full)
summary(m.genderRating.Full)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: Female ~ FullDummy * GenderRatingCentered + (1 | Participant) +
## (1 | Item)
## Data: d
##
##      AIC      BIC    logLik deviance df.resid
##  9164.2   9206.7 -4576.1   9152.2     8765
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.0410 -0.5746 -0.2710  0.5694  4.7906
##
## Random effects:
## Groups      Name      Variance Std.Dev.
## Participant (Intercept) 0.2015   0.4489
## Item          (Intercept) 0.3602   0.6001
## Number of obs: 8771, groups: Participant, 1253; Item, 63
##
## Fixed effects:
##                                Estimate Std. Error z value Pr(>|z|)
```

```
## (Intercept) -0.18022 0.09138 -1.972 0.0486 *
## FullDummy -0.11477 0.06200 -1.851 0.0642 .
## GenderRatingCentered 0.75588 0.05139 14.709 <2e-16 ***
## FullDummy:GenderRatingCentered 0.00606 0.03511 0.173 0.8630
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
## (Intr) FllDmm GndrRC
## FullDummy -0.450
## GndrRtngCnt -0.035 0.038
## FllDmmy:GRC 0.038 -0.104 -0.451
```

```
m.genderRating.First_tidy %>%
  filter(term=="GenderRatingCentered") %>%
  select(estimate) %>% as.numeric()
```

```
## [1] 0.8594438
```

```
m.genderRating.Full_tidy %>%
  filter(term=="GenderRatingCentered") %>%
  select(estimate) %>% as.numeric()
```

```
## [1] 0.7558761
```

The effect of name gender rating is larger in the First Name condition (0.86) than in the Full Name condition (0.76).

Convert to Odds Ratios

Intercept

```
m.main_intercept <- m.main_tidy %>% filter(term=="(Intercept)") %>%
  select(estimate) %>% as.numeric()

exp(m.main_intercept)
```

```
## [1] 0.774106
```

```
exp(-m.main_intercept)
```

```
## [1] 1.291813
```

0.77x less likely to recall as female overall. Easier to interpret: 1.29x more likely to recall as male overall.

Condition: Last vs First+Full

```
m.main_LFF <- m.main_tidy %>%
  filter(term=="Condition1") %>%
  select(estimate) %>% as.numeric()
exp(m.main_LFF)
```



```
## [1] 1.134694
```

1.13x more likely to recall as female in First + Full compared to Last.

Condition: Last Only

Dummy code with Last Name as 0, so that intercept is the Last Name condition only.

```
d %<>% mutate(Condition_Last=case_when(  
  Condition=="first" ~ 1,  
  Condition=="full" ~ 1,  
  Condition=="last" ~ 0))  
d$Condition_Last %<>% as.factor()
```

```
m.last <- glmer(Female ~ Condition_Last + (1|Participant) + (1|Item),  
  data=d, family=binomial)  
m.last_tidy <- tidy(m.last)
```

```
m.lastonly <- m.last_tidy %>%  
  filter(term=="(Intercept)") %>%  
  select(estimate) %>% as.numeric()  
  
exp(m.lastonly)
```

```
## [1] 0.695725
```

```
exp(-m.lastonly)
```

```
## [1] 1.437349
```

0.17x times less likely to recall as female in the Last Name condition -> 5.72x more likely to recall as male in the Last Name condition (trending).

Condition: First and Full Only

Dummy code with First and Full Name as 0, so that intercept is average for these two conditions.

```
d %<>% mutate(Condition_FF=case_when(  
  Condition=="first" ~ 0,  
  Condition=="full" ~ 0,  
  Condition=="last" ~ 1))  
d$Condition_FF %<>% as.factor()
```

```
m.ff <- glmer(Female ~ Condition_FF + (1|Participant) + (1|Item),  
  data=d, family=binomial)  
m.ff_tidy <- tidy(m.ff)
```

```
m.cond_ff <- m.ff_tidy %>%  
  filter(term=="(Intercept)") %>%  
  select(estimate) %>% as.numeric()  
  
exp(m.cond_ff)
```

```
## [1] 0.8151591
```

```
exp(-m.cond_ff)
```

```
## [1] 1.226754
```

0.82x less likely to recall as female in First and Full Name conditions → 1.23x more likely to recall as male in First and Full Name conditions