

Experiment 3: Main Analyses

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Contents

| | |
|--|----------|
| Setup | 1 |
| Data Summary | 2 |
| Model 1: With <i>Other</i> Responses | 3 |
| Odds Ratios: Intercept | 4 |
| Odds Ratios: Last vs First+Full | 5 |
| Odds Ratios: Last Only | 5 |
| Odds Ratios: First and Full Only | 6 |
| Model 2: Without <i>Other</i> Responses | 8 |
| Odds Ratios: Intercept | 9 |
| Odds Ratios: Last vs First+Full | 9 |
| Odds Ratios: Last Only | 10 |
| Odds Ratios: First and Full Only | 11 |

Setup

Variable names:

- Experiment: exp3
- Type
 - d = data
 - m = model
 - est = log odds estimate from model
 - OR = odds ratio converted from est
- Analysis
 - count = sums of response types
 - all = including *other* responses

- noOther = excluding *other* responses
- Subset
 - FF = First and Full Name conditions only
 - Last = Last Name condition only

Load data and select columns used in model. See data/exp3_data_about.txt for more details.

```
exp3_d <- read.csv("../data/exp3_data.csv",
                  stringsAsFactors=TRUE) %>%
  rename("Participant"="SubjID", "Item"="Name") %>%
  select(Participant, Condition, GenderRating,
         Item, He, She, Other)
str(exp3_d)
```

```
## 'data.frame': 8904 obs. of 7 variables:
## $ Participant : Factor w/ 1272 levels "R_020U0b05Lb0EtX3",...: 216 216 216 216 216 216 216 41 41 41
## $ Condition : Factor w/ 3 levels "first","full",...: 1 1 1 1 1 1 1 1 1 ...
## $ GenderRating: num 5.22 1.24 5.86 3.75 6.78 4.34 2.41 6.24 2.61 6.82 ...
## $ Item : Factor w/ 63 levels "Ashley Cook",...: 6 9 13 43 47 52 62 2 16 20 ...
## $ He : int 0 1 0 0 0 0 1 0 1 0 ...
## $ She : int 0 0 1 0 1 1 0 0 0 1 ...
## $ Other : int 1 0 0 1 0 0 0 1 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.

```
exp3_d %<>% mutate(GenderRatingCentered=
  scale(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(exp3_d$Condition) <- centerfactor(
  exp3_d$Condition, c("last","first"))
contrasts(exp3_d$Condition)
```

```
##           [,1]      [,2]
## first  0.4009434 -0.48113208
## full   0.4009434  0.51886792
## last  -0.5990566  0.01886792
```

Data Summary

Responses by condition.

```
exp3_d %<>% mutate(ResponseAll=case_when(
  He==1 ~ "He",
  She==1 ~ "She",
  Other==1 ~ "Other"))

exp3_d_count <- exp3_d %>%
  group_by(Condition, ResponseAll) %>%
  summarise(n=n()) %>%
  pivot_wider(names_from=ResponseAll,
              values_from=n) %>%
  mutate(She_HeOther = She / (He+Other),
         She_He = She / He)

kable(exp3_d_count, digits=3)
```

| Condition | He | Other | She | She_HeOther | She_He |
|-----------|------|-------|------|-------------|--------|
| first | 992 | 902 | 941 | 0.497 | 0.949 |
| full | 899 | 752 | 848 | 0.514 | 0.943 |
| last | 1378 | 1113 | 1079 | 0.433 | 0.783 |

Model 1: With *Other* Responses

Effects of Condition (first name, last name, full name) and Gender Rating on the likelihood of a *she* response, as opposed to a *he* or *other* response. Participant and Item are included as random intercepts, with items defined as the unique first, last and first + last name combinations. Because the condition manipulations were fully between-subject and between-item, fitting a random slope model was not possible.

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

Condition1 is the contrast between last and first+full. Condition2 is the contrast between first and full.

```
exp3_m_all <- glmer(
  She ~ Condition * GenderRatingCentered +
    (1|Participant) + (1|Item),
  data=exp3_d, family=binomial)
summary(exp3_m_all)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: She ~ Condition * GenderRatingCentered + (1 | Participant) +
## (1 | Item)
## Data: exp3_d
##
##      AIC      BIC    logLik deviance df.resid
##  7825.8   7882.5  -3904.9   7809.8     8896
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.0250 -0.4836 -0.1394  0.5355  9.7282
```

```
##
## Random effects:
## Groups      Name      Variance Std.Dev.
## Participant (Intercept) 0.7931  0.8905
## Item        (Intercept) 0.4209  0.6488
## Number of obs: 8904, groups: Participant, 1272; Item, 63
##
## Fixed effects:
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -1.52419    0.10101 -15.090  <2e-16 ***
## Condition1      0.15325    0.09155   1.674   0.0941 .
## Condition2      0.09120    0.11596   0.786   0.4316
## GenderRatingCentered 1.14844    0.06039  19.017  <2e-16 ***
## Condition1:GenderRatingCentered 0.10499    0.04875   2.153   0.0313 *
## Condition2:GenderRatingCentered -0.05627    0.06294  -0.894   0.3713
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) Cndtn1 Cndtn2 GndrRC C1:GRC
## Condition1    0.000
## Condition2  -0.015  0.023
## GndrRtngCnt -0.287 -0.004  0.016
## Cndtn1:GnRC -0.009 -0.495  0.000  0.025
## Cndtn2:GnRC  0.016  0.000 -0.488 -0.023  0.009
```

- Fewer *she* responses overall
- Last Name vs First+Full Names condition effect only trending
- More *she* responses as first names become more feminine
- Larger effect of first name gender in First+Full Name conditions than in Last Name conditions, which makes sense because there are 4 repetitions of the gendered first name, as opposed to only 1.

Odds Ratios: Intercept

```
exp3_est_all_intercept <- exp3_m_all %>%
  tidy() %>%
  filter(term=="(Intercept)") %>%
  select(estimate) %>% as.numeric()

exp(exp3_est_all_intercept)
```

```
## [1] 0.2177981
```

```
exp(-exp3_est_all_intercept)
```

```
## [1] 4.591408
```

```
#Save this for the table comparing all 4 experiments
exp3_OR_all_I <- exp(-exp3_est_all_intercept) %>% round(2)
```

0.22x less likely to use *she* overall. Easier to interpret: 4.59x more likely to use *he* and *other* overall, $p < .001$

Odds Ratios: Last vs First+Full

```
exp3_est_all_LFF <- exp3_m_all %>%
  tidy() %>%
  filter(term=="Condition1") %>%
  select(estimate) %>% as.numeric()
exp(exp3_est_all_LFF)
```

```
## [1] 1.165616
```

```
#Save this for the table comparing all 4 experiments
exp3_OR_all_LFF <- exp(exp3_est_all_LFF) %>% round(2)
```

1.17x more likely to use *she* than *he* and *other* in First + Full compared to Last, 0.09

Odds Ratios: Last Only

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

```
exp3_d %<>% mutate(Condition_Last=case_when(
  Condition=="first" ~ 1,
  Condition=="full" ~ 1,
  Condition=="last" ~ 0))
exp3_d$Condition_Last %<>% as.factor()
```

```
exp3_m_all_L <- glmer(
  She ~ Condition_Last + (1|Participant) + (1|Item),
  data=exp3_d, family=binomial)
summary(exp3_m_all_L)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: She ~ Condition_Last + (1 | Participant) + (1 | Item)
## Data: exp3_d
##
##      AIC      BIC   logLik deviance df.resid
##  7962.3   7990.7  -3977.1   7954.3     8900
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.9131 -0.4946 -0.1440  0.5311  8.8113
##
```

```
## Random effects:
## Groups      Name      Variance Std.Dev.
## Participant (Intercept) 0.7738  0.8796
## Item        (Intercept) 5.3393  2.3107
## Number of obs: 8904, groups: Participant, 1272; Item, 63
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -1.74419    0.30132  -5.788  7.1e-09 ***
## Condition_Last1 0.24968    0.07806   3.198  0.00138 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr)
## Condtn_Lst1 -0.160
```

```
exp3_est_all_L <- exp3_m_all_L %>%
  tidy() %>%
  filter(term=="(Intercept)") %>%
  select(estimate) %>% as.numeric()

exp(exp3_est_all_L)
```

```
## [1] 0.1747868
```

```
exp(-exp3_est_all_L)
```

```
## [1] 5.721256
```

```
#Save this for the table comparing all 4 experiments
exp3_OR_all_L <- exp(-exp3_est_all_L) %>%
  round(2)
```

0.18x times less likely to use *she* than *he* and *other* in the Last Name condition → 5.72x more likely to use *he* and *other* in the Last Name condition, $p < .001$

Odds Ratios: First and Full Only

Dummy code with First and Full Name as 0, so the intercept is the combination of those two.

```
exp3_d %<>% mutate(Condition_FF=case_when(
  Condition=="first" ~ 0,
  Condition=="full" ~ 0,
  Condition=="last" ~ 1))
exp3_d$Condition_FF %<>% as.factor()
```

```
exp3_m_all_FF <- glmer(
  She ~ Condition_FF + (1|Participant) + (1|Item),
  data=exp3_d, family=binomial)
summary(exp3_m_all_FF)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: She ~ Condition_FF + (1 | Participant) + (1 | Item)
## Data: exp3_d
##
##      AIC      BIC   logLik deviance df.resid
##  7962.3   7990.7 -3977.1   7954.3     8900
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.9131 -0.4946 -0.1440  0.5311  8.8113
##
## Random effects:
## Groups      Name      Variance Std.Dev.
## Participant (Intercept) 0.7738   0.8796
## Item         (Intercept) 5.3393   2.3107
## Number of obs: 8904, groups: Participant, 1272; Item, 63
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -1.49451    0.29912  -4.996 5.84e-07 ***
## Condition_FF1 -0.24967    0.07807  -3.198 0.00138 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr)
## Conditn_FF1 -0.100
```

```
exp3_est_all_FF <- exp3_m_all_FF %>%
  tidy() %>%
  filter(term=="(Intercept)") %>%
  select(estimate) %>% as.numeric()

exp(exp3_est_all_FF)
```

```
## [1] 0.2243583
```

```
exp(-exp3_est_all_FF)
```

```
## [1] 4.457156
```

```
#Save this for the table comparing all 4 experiments
exp3_OR_all_FF <- exp(-exp3_est_all_FF) %>%
  round(2)
```

0.22x times less likely to use *she* than *he* and *other* in the First and Full Name conditions → 4.46x more likely to use *he* and *other* in the First and Full Name conditions, $p < .001$

Model 2: Without *Other* Responses

The sentence completion prompt for Experiment 3 is more open-ended than in Experiment 1. So, we get a much higher proportion of *other* responses (31% vs 7%), which I didn't anticipate.

```
sum(exp3_d$Other)
```

```
## [1] 2767
```

```
sum(exp3_d$Other)/length(exp3_d$Other)
```

```
## [1] 0.3107592
```

```
exp3_d_noOther <- exp3_d %>% filter(Other==0)
```

So, rerun the main model predicting the likelihood of *she* responses vs *he* responses, with *other* responses excluded.

```
exp3_m_noOther <- glmer(  
  She ~ Condition * GenderRatingCentered +  
    (1|Participant) + (1|Item),  
  data=exp3_d_noOther, family=binomial)  
summary(exp3_m_noOther)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace  
## Approximation) [glmerMod]  
## Family: binomial ( logit )  
## Formula: She ~ Condition * GenderRatingCentered + (1 | Participant) +  
## (1 | Item)  
## Data: exp3_d_noOther  
##  
##      AIC      BIC    logLik deviance df.resid  
##  4209.0   4262.8  -2096.5   4193.0     6129  
##  
## Scaled residuals:  
##      Min       1Q   Median       3Q      Max  
## -9.0292 -0.3424 -0.0521  0.2952 12.5649  
##  
## Random effects:  
## Groups      Name      Variance Std.Dev.  
## Participant (Intercept) 0.5394   0.7345  
## Item          (Intercept) 0.6807   0.8251  
## Number of obs: 6137, groups: Participant, 1223; Item, 63  
##  
## Fixed effects:  
##  
##              Estimate Std. Error z value Pr(>|z|)  
## (Intercept)    -0.42366    0.12377  -3.423 0.000619 ***  
## Condition1      0.25702    0.09784   2.627 0.008616 **  
## Condition2     -0.01455    0.12816  -0.114 0.909584  
## GenderRatingCentered 1.67709    0.08371 20.034 < 2e-16 ***  
## Condition1:GenderRatingCentered 0.41953    0.07691   5.455 4.9e-08 ***
```



```
## Condition2:GenderRatingCentered -0.14907    0.11205  -1.330 0.183394
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##          (Intr) Cndtn1 Cndtn2 GndrRC C1:GRC
## Condition1  0.053
## Condition2 -0.020  0.005
## GndrRtngCnt -0.155 -0.005  0.005
## Cndtn1:GnRC -0.007 -0.210  0.003  0.201
## Cndtn2:GnRC  0.005  0.004 -0.182 -0.061 -0.053
```

These results are more similar to what we predicted from the previous experiments:

- Fewer *she* responses overall
- Fewer *she* responses in the Last Name condition as compared to the First + Full Name conditions (although we wouldn't predict as large a difference as in Exp1, because here there is one instance of the first name in the Last Name condition)
- More *she* responses as first names become more feminine
- Larger effect of first name gender in First+Full Name conditions than in Last Name conditions (which makes sense because there are 4 repetitions of the gendered first name, as opposed to only 1.)

But, to keep the analyses consistent between experiments and avoid post-hoc decision weirdness, both versions are reported.

Odds Ratios: Intercept

```
exp3_est_noOther_intercept <- exp3_m_noOther %>%
  tidy() %>%
  filter(term=="(Intercept)") %>%
  select(estimate) %>% as.numeric()

exp(exp3_est_noOther_intercept)
```

```
## [1] 0.6546471
```

```
exp(-exp3_est_noOther_intercept)
```

```
## [1] 1.527541
```

```
#Save this for the table comparing all 4 experiments
exp3_OR_noOther_I <- exp(-exp3_est_noOther_intercept) %>%
  round(2)
```

0.65x less likely to use *she* than *he* overall. Easier to interpret: 1.53x more likely to use *he* than *she* overall, $p < .001$

Odds Ratios: Last vs First+Full

```
exp3_est_noOther_LFF <- exp3_m_noOther %>%
  tidy() %>%
  filter(term=="Condition1") %>%
  select(estimate) %>% as.numeric()
exp(exp3_est_noOther_LFF)
```

```
## [1] 1.293077
```

```
#Save this for the table comparing all 4 experiments
exp3_OR_noOther_LFF <- exp(-exp3_est_noOther_LFF) %>%
  round(2)
```

1.29x more likely to use *she* than *he* in First+Full than in Last -> 1.29x more likely to use *he* than *she* in Last than in First+Full, $p < .001$

Odds Ratios: Last Only

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

```
exp3_d_noOther %<>% mutate(Condition_Last=case_when(
  Condition=="first" ~ 1,
  Condition=="full" ~ 1,
  Condition=="last" ~ 0))
exp3_d_noOther$Condition_Last %<>% as.factor()
```

```
exp3_m_noOther_L <- glmer(
  She ~ Condition_Last + (1|Participant) + (1|Item),
  data=exp3_d_noOther, family=binomial)
summary(exp3_m_noOther_L)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: She ~ Condition_Last + (1 | Participant) + (1 | Item)
## Data: exp3_d_noOther
##
##      AIC      BIC    logLik deviance df.resid
##  4383.5   4410.4  -2187.8   4375.5     6133
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -7.4256 -0.3377 -0.0653  0.2875 10.2132
##
## Random effects:
##  Groups      Name      Variance Std.Dev.
## Participant (Intercept)  0.4906  0.7004
## Item          (Intercept) 10.1950  3.1930
## Number of obs: 6137, groups: Participant, 1223; Item, 63
##
## Fixed effects:
```

```
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -0.67727    0.41228  -1.643    0.1
## Condition_Last1 0.37418    0.09174   4.079 4.53e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr)
## Condtn_Lst1 -0.135
```

```
exp3_est_noOther_L <- exp3_m_noOther_L %>%
  tidy() %>%
  filter(term=="(Intercept)") %>%
  select(estimate) %>% as.numeric()

exp(exp3_est_noOther_L)
```

```
## [1] 0.5080018
```

```
exp(-exp3_est_noOther_L)
```

```
## [1] 1.968497
```

```
#Save this for the table comparing all 4 experiments
exp3_OR_noOther_L <- exp(-exp3_est_noOther_L) %>%
  round(2)
```

0.51x times less likely to use *she* than *he* in the Last Name condition → 1.97x more likely to use *he* than *she* in the Last Name condition, $p=.10$

Odds Ratios: First and Full Only

Dummy code with First and Full Name as 0, so the intercept is the combination of those two.

```
exp3_m_noOther_FF <- glmer(
  She ~ Condition_FF + (1|Participant) + (1|Item),
  data=exp3_d_noOther, family=binomial)
summary(exp3_m_noOther_FF)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: She ~ Condition_FF + (1 | Participant) + (1 | Item)
## Data: exp3_d_noOther
##
##      AIC      BIC    logLik deviance df.resid
##  4383.5   4410.4  -2187.8   4375.5     6133
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
```

```
## -7.4256 -0.3377 -0.0653 0.2875 10.2132
##
## Random effects:
## Groups Name Variance Std.Dev.
## Participant (Intercept) 0.4906 0.7004
## Item (Intercept) 10.1949 3.1929
## Number of obs: 6137, groups: Participant, 1223; Item, 63
##
## Fixed effects:
## Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.30308 0.41003 -0.739 0.46
## Condition_FF1 -0.37418 0.09174 -4.079 4.53e-05 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
## (Intr)
## Conditn_FF1 -0.088
```

```
exp3_est_noOther_FF <- exp3_m_noOther_FF %>%
  tidy() %>%
  filter(term=="(Intercept)") %>%
  select(estimate) %>% as.numeric()

exp(exp3_est_noOther_FF)
```

```
## [1] 0.7385373
```

```
exp(-exp3_est_noOther_FF)
```

```
## [1] 1.354028
```

```
#Save this for the table comparing all 4 experiments
exp3_OR_noOther_FF <- exp(-exp3_est_noOther_FF) %>%
  round(2)
```

0.74x times less likely to use *she* than *he* and *other* in the First and Full Name conditions → 1.35x more likely to use *he* and *other* in the First and Full Name conditions, $p=.46$