Study 4

April 10, 2025

Items

Read Data	. 2
Variable Names	. 4
Demographics	. 4
Primary Analysis	. 6
Robustness	. 8
Figure 6 Code	. 13
System of Equations	. 16

Read Data

```
# Set this to TRUE if you have API access, FALSE if using CSV
USE API <- FALSE
if(USE_API) {
  ## Pull directly from Qualtrics API
  d0 <- fetch survey(surveyID='SV 1BJC5tq7iCTDLpk',</pre>
                       label = T,
                       convert = F,
                       start_date = "2024-03-27",
                       end_date ="2024-05-15",
                       force_request = T)
} else {
  # Read the processed data directly from CSV
  d0 <- read.csv('Study4.csv', check.names = F)</pre>
# Define the categories
women <- c('Anne Frank', 'Tina Fey', 'Jackie Kennedy', 'Helen Keller', 'Barbra

    Streisand')

pages <- c('Tina Fey', 'Keith Richards', 'Andre Agassi', 'Henry Winkler', 'Willie</pre>
→ Nelson', 'Michael J. Fox', 'John Stamos')
year <- c('Steve Jobs', 'J. Robert Oppenheimer', 'Mark Twain', 'Anthony Bourdain',

→ 'Andrew Jackson', 'Tina Fey', 'Johnny Cash', 'Jackie Kennedy', 'Theodore
→ Roosevelt', 'Che Guevara', 'Tennessee Williams', 'Keith Richards', 'Napoleon
→ Bonaparte', 'Andre Agassi', 'Henry Winkler', 'Robin Williams', 'Willie Nelson',
\hookrightarrow 'Michael J. Fox', 'John Stamos')
entertainer <- c('Anthony Bourdain', 'Tina Fey', 'Johnny Cash', 'Jim Carroll', 'Keith</pre>
Aichards', 'Henry Winkler', 'Robin Williams', 'Willie Nelson', 'Michael J. Fox',
→ 'John Stamos', 'Barbra Streisand')
if(USE_API) {
  d0 <- d0 |>
    filter(!is.na(`choice-7`)) |>
    mutate(
      gender feedback = as.numeric(grep1("feature a female protagonist", feedbackItem1) |
                      grepl("feature a female protagonist", feedbackItem2) |
                      grepl("feature a female protagonist", feedbackItem3)),
      pages_shown = as.numeric(grepl("are over 500 pages", feedbackItem1) |
                      grepl("are over 500 pages", feedbackItem2) |
                      grepl("are over 500 pages", feedbackItem3)),
      year shown = as.numeric(grep1("were written in the past 25 years", feedbackItem1)
                   grepl("were written in the past 25 years", feedbackItem2) |
                   grepl("were written in the past 25 years", feedbackItem3)),
      entertainer_shown = as.numeric(grep1("feature an entertainer", feedbackItem1) |
                     grepl("feature an entertainer", feedbackItem2) |
                     grepl("feature an entertainer", feedbackItem3)),
      female_pick = case_when(`choice-7` %in% women ~ 1,
                               TRUE \sim 0),
      pages_pick = case_when(`choice-7` %in% pages ~ 1,
                               TRUE \sim 0),
      year_pick = case_when(`choice-7` %in% year ~ 1,
```

Variable Names

Variable	Description		
gender_feedback	Binary indicator of whether a participant was randomly assigned		
	to gender feedback condition.		
female_pick	Binary indicator of whether a participant selected a biopic with a		
	female protagonist for their seventh selection.		
budget_shown	Binary indicator of whether a participant was randomly assigned		
	to receive budget feedback.		
budget_pick	Binary indicator of whether a participant selected a film with a		
	big budget for their seventh selection.		
year_shown	Binary indicator of whether a participant was randomly assigned		
	to receive film year feedback.		
year_pick	Binary indicator of whether a participant selected a film released		
	after 2010 for their seventh selection.		
poli_shown	Binary indicator of whether a participant was randomly assigned		
	to receive political leader feedback		
poli_pick	Binary indicator of whether a participant selected a film with a		
	political protagonist for their seventh selection.		
base_gender	Count of the number of films with female protagonists selected in		
	the initial six films.		
choice-1 to choice-7	The selected biographies		
gender	Self-selected gender.		
race	Self-selected race.		
age	Self-entered age.		
gender_code	Dummy code for gender (male $= 1$).		
race_code	Dummy code for race (white $= 1$).		

Demographics

```
Percentage gender
## 1 Another gender not listed here:
                                      0.36
## 2
                                Man 52.52
## 3
                         Non-binary
                                      2.16
## 4
                              Woman 44.96
##
                           Percentage Race
## 1 American Indian or Alaskan Native 0.36
## 2
             Asian / Pacific Islander 42.81
## 3
            Black or African American 10.79
                    Hispanic / Latinx 4.68
## 4
## 5
                    White / Caucasian 41.37
## Mean (age): 25.11
## SD (age): 10.87
## Mean (num of initial women selected): 1.51
## SD (num of initial women selected): 1.15
```

```
## Percentage (initial women selected): 0.2516667
## SD (initial women selected): 0.1916667
## # A tibble: 2 x 2
##
   gender_feedback mean
             <int> <dbl>
## 1
                  0 0.230
## 2
                  1 0.273
##
## Welch Two Sample t-test
##
## data: base_gender/6 by gender_feedback
## t = -1.9672, df = 288.06, p-value = 0.05012
\#\# alternative hypothesis: true difference in means between group 0 and group 1 is not equal to 0
## 95 percent confidence interval:
## -8.611628e-02 2.356926e-05
## sample estimates:
## mean in group 0 mean in group 1
       0.2295806
                        0.2726269
```

Primary Analysis

```
# primary model
r1 <- lm(female_pick ~ gender_feedback, data=d0)
# robust standard errors
robust_summary(r1)
##
## Call:
## lm(formula = female_pick ~ gender_feedback, data = d0)
## Residuals:
      Min
              1Q Median
                             3Q
## -0.3510 -0.3510 -0.2185 0.6490 0.7815
## Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                  ## gender_feedback 0.13245
                             0.05172 2.561 0.0109 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.4479 on 300 degrees of freedom
## Multiple R-squared: 0.02153,
                                 Adjusted R-squared: 0.01827
## F-statistic: 6.602 on 1 and 300 DF, p-value: 0.01067
robust_confint(r1)
                      2.5 %
##
                               97.5 %
## (Intercept)
                 0.15192039 0.2851657
## gender_feedback 0.03067135 0.2342293
# robustness test
# primary model
r_robust <- lm(female_pick ~ gender_feedback + base_gender, data=d0)
# robust standard errors
robust_summary(r_robust)
##
## lm(formula = female_pick ~ gender_feedback + base_gender, data = d0)
## Residuals:
      Min
               1Q Median
                              3Q
## -0.3558 -0.3470 -0.2197 0.6442 0.7892
## Coefficients:
```

```
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.222608    0.043876    5.074 6.87e-07 ***
## gender_feedback    0.133213    0.052968    2.515    0.0124 *
## base_gender    -0.002951    0.023299    -0.127    0.8993
## ---
## Signif. codes:    0 '****' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error:    0.4486 on 299 degrees of freedom
## Multiple R-squared:    0.02159,    Adjusted R-squared:    0.01504
## F-statistic:    3.299 on 2 and 299 DF,    p-value:    0.03828
```

robust_confint(r_robust)

```
## 2.5 % 97.5 %

## (Intercept) 0.13626369 0.30895301

## gender_feedback 0.02897536 0.23744980

## base_gender -0.04880113 0.04289862
```

Robustness

```
## which feedback was shown with gender, remove constant due to ollinearity
r2 <- lm(female_pick ~ gender_feedback + pages_shown + year_shown + entertainer_shown -
\rightarrow 1, data=d0)
# Display the robust_summary with robust standard errors
robust_summary(r2)
##
## Call:
## lm(formula = female_pick ~ gender_feedback + pages_shown + year_shown +
      entertainer_shown - 1, data = d0)
##
## Residuals:
      Min
               10 Median
                              30
                                     Max
## -0.3770 -0.3276 -0.2185 0.6230 0.7815
## Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
                   ## gender_feedback
                  0.09472
                               0.05614 1.687 0.0926 .
## pages_shown
## year_shown
                    0.04526 0.05628 0.804 0.4219
## entertainer_shown 0.07856
                               0.06592 1.192 0.2343
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.4491 on 298 degrees of freedom
## Multiple R-squared: 0.301, Adjusted R-squared: 0.2917
## F-statistic: 32.09 on 4 and 298 DF, p-value: < 2.2e-16
robust_confint(r2)
                         2.5 %
                                  97.5 %
##
                   0.11097645 0.2965564
## gender_feedback
## pages shown
                   -0.01576071 0.2052072
                   -0.06550212 0.1560227
## year shown
## entertainer shown -0.05116184 0.2082808
## robust to demographic controls
r3 <- lm(female_pick ~ gender_feedback + gender_code + race_code + age, data=d0)
# Display the robust_summary with robust standard errors
robust_summary(r3)
##
## Call:
## lm(formula = female_pick ~ gender_feedback + gender_code + race_code +
      age, data = d0)
##
```

```
##
## Residuals:
      Min
               1Q Median
## -0.5613 -0.3016 -0.1894 0.5678 0.9084
## Coefficients:
                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                   0.157807
                              0.090183 1.750
                                                 0.0815 .
## gender_feedback 0.135678
                              0.057372 2.365
                                                 0.0189 *
## gender_code -0.085584
                            0.058297 -1.468
                                                 0.1434
## race_code
                  -0.079007
                              0.058126 -1.359
                                                 0.1754
                   0.004687
                              0.003045 1.539
                                                 0.1251
## age
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.436 on 234 degrees of freedom
     (63 observations deleted due to missingness)
## Multiple R-squared: 0.05085,
                                   Adjusted R-squared: 0.03463
## F-statistic: 3.134 on 4 and 234 DF, p-value: 0.01548
robust_confint(r3)
##
                         2.5 %
                                   97.5 %
## (Intercept)
                  -0.019868533 0.33548194
## gender_feedback 0.022645502 0.24871006
## gender_code
                 -0.200438565 0.02927120
## race_code
                 -0.193524563 0.03551136
## age
                  -0.001312344 0.01068566
## logistic regression
# Fit the logistic regression model
r4 <- glm(female_pick ~ gender_feedback, family = binomial, data=d0)
# Odds ratio
tidy_r4 <- tidy(r4, exponentiate = TRUE, conf.int = T)</pre>
print(tidy_r4)
## # A tibble: 2 x 7
   term
                    estimate std.error statistic p.value conf.low conf.high
##
    <chr>
                              <dbl> <dbl>
                                                    <dbl> <dbl>
                                                                       <dbl>
                       <dbl>
                                           -6.47 9.77e-11
                                 0.197
                                                                      0.406
## 1 (Intercept)
                       0.280
                                                             0.187
## 2 gender_feedback
                       1.93
                                 0.260
                                            2.53 1.13e- 2
                                                             1.17
                                                                      3.24
summary(r4)
##
## Call:
## glm(formula = female_pick ~ gender_feedback, family = binomial,
##
      data = d0)
##
## Coefficients:
```

```
## Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.2742     0.1969 -6.471 9.77e-11 ***
## gender_feedback     0.6595     0.2605     2.532     0.0113 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 360.83 on 301 degrees of freedom
## Residual deviance: 354.28 on 300 degrees of freedom
## AIC: 358.28
##
## Number of Fisher Scoring iterations: 4
```

```
##
## Call:
## lm(formula = pages_pick ~ pages_shown, data = d0)
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -0.1680 -0.1680 -0.1680 -0.1552 0.8448
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.15517
                          0.04838
                                    3.208 0.00148 **
## pages_shown 0.01286
                          0.05402
                                    0.238 0.81198
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3729 on 300 degrees of freedom
## Multiple R-squared: 0.0001858, Adjusted R-squared: -0.003147
## F-statistic: 0.05574 on 1 and 300 DF, p-value: 0.8135
##
                    2.5 %
                             97.5 %
## (Intercept) 0.05997303 0.2503718
## pages_shown -0.09344118 0.1191619
##
## Call:
## lm(formula = year_pick ~ year_shown, data = d0)
##
## Residuals:
      Min
               1Q Median
                               3Q
## -0.7178 -0.6721 0.2822 0.2822 0.3279
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
                          0.06111 10.999
## (Intercept) 0.67213
                                            <2e-16 ***
## year_shown
              0.04571
                          0.06769
                                    0.675
                                               0.5
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4555 on 300 degrees of freedom
## Multiple R-squared: 0.001631, Adjusted R-squared: -0.001697
## F-statistic: 0.4901 on 1 and 300 DF, p-value: 0.4844
##
                    2.5 %
                             97.5 %
## (Intercept) 0.55187855 0.7923837
## year_shown -0.08749005 0.1789124
##
## Call:
## lm(formula = entertainer_pick ~ entertainer_shown, data = d0)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -0.3370 -0.3370 -0.3370 0.6630 0.7188
```

```
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                ## entertainer_shown 0.05579
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.4719 on 300 degrees of freedom
## Multiple R-squared: 0.001331, Adjusted R-squared: -0.001998
## F-statistic: 0.3999 on 1 and 300 DF, p-value: 0.5276
##
                     2.5 %
                             97.5 %
## (Intercept)
                 0.1197947 0.4427053
## entertainer_shown -0.1153753 0.2269494
```

Figure 6 Code

```
dfemale_plot <- d0 |>
  dplyr::select(gender_feedback, female_pick) |>
  dplyr::group_by(gender_feedback) |>
  dplyr::summarise(
   n = n(),
   freq = mean(female_pick),
   sd = sd(female pick) * 100,
   se = (sd(female_pick) / sqrt(n())) * 100
  dplyr::mutate(gender_feedback = case_when(gender_feedback==1 ~ "\"Treatment\"",
                          TRUE ~ "\"Control\"")) |>
 rename(Condition = gender_feedback)
##### entertainer
dentertainer_plot <- d0 |>
  dplyr::select(entertainer_shown, entertainer_pick) |>
  dplyr::group_by(entertainer_shown) |>
  dplyr::summarise(
   n = n(),
   freq = mean(entertainer_pick),
   sd = sd(entertainer_pick) * 100,
   se = (sd(entertainer_pick) / sqrt(n())) * 100
  dplyr::mutate(entertainer_shown = case_when(entertainer_shown==1 ~ "\"Treatment\"",
                          TRUE ~ "\"Control\"")) |>
  rename(Condition = entertainer_shown)
##### year
dyear_plot <- d0 |>
  dplyr::select(year_shown, year_pick) |>
  dplyr::group_by(year_shown) |>
 dplyr::summarise(
   n = n(),
   freq = mean(year_pick),
   sd = sd(year_pick) * 100,
   se = (sd(year_pick) / sqrt(n())) * 100
  dplyr::mutate(year_shown = case_when(year_shown==1 ~ "\"Treatment\"",
                          TRUE ~ "\"Control\"")) |>
 rename(Condition = year_shown)
#### pages
dpages_plot <- d0 |>
 dplyr::select(pages_shown, pages_pick) |>
```

```
dplyr::group_by(pages_shown) |>
 dplyr::summarise(
   n = n(),
   freq = mean(pages_pick),
   sd = sd(pages_pick) * 100,
   se = (sd(pages_pick) / sqrt(n())) * 100
 ) |>
 dplyr::mutate(pages_shown = case_when(pages_shown==1 ~ "\"Treatment\"",
                         TRUE ~ "\"Control\"")) |>
 rename(Condition = pages_shown)
## Combine plots
df_combined <- dplyr::bind_rows(</pre>
 dentertainer_plot %>% dplyr::mutate(Category = "\nFeatured an\nEntertainer"),
 dyear_plot %>% dplyr::mutate(Category = "\nWere Written in\nthe Past 25 Years"),
 dpages_plot %>% dplyr::mutate(Category = "\nWere Over \n500 Pages"),
 dfemale_plot %>% dplyr::mutate(Category = "\nFeatured a\nWoman")
 .id = "id") %>%
 mutate(Category = factor(Category, levels = c('\nFeatured an\nEntertainer', '\nWere
  → Written in nthe Past 25 Years', '\nWere Over \n500 Pages', '\nFeatured a\nWoman')))
p_combined \leftarrow ggplot(df_combined, aes(x = Condition, y = freq*100, fill = Condition)) +
 geom_bar(stat="identity", width = 0.85, position = position_dodge(width = 0.7)) +
 geom_text(data = df_combined %>% filter(Category %in% c('\nWere Over \n500 Pages')),

→ aes(label=paste0(sprintf("%.1f", freq*100),"%")),
  → position=position_dodge(width=0.7), vjust=5, size = 5, color = "white", family =
  → "Times New Roman") +
 geom text(data = df combined %>% filter(!Category %in% c('\nWere Over \n500 Pages')),

→ aes(label=paste0(sprintf("%.1f", freq*100),"%")),
  → position=position_dodge(width=0.7), vjust=5, size = 5, color = "white", family =
  → "Times New Roman") +
 geom_errorbar(aes(ymin=freq*100-se, ymax=freq*100+se), width = .1, position =
  → position_dodge(width = 0.7)) +
 facet_wrap(~factor(Category, c('\nFeatured an\nEntertainer', '\nWere Written in\nthe
  → Past 25 Years', '\nWere Over \n500 Pages', '\nFeatured a\nWoman')), nrow = 1,

    strip.position = "bottom") +

 geom_segment(data = df_combined %>% filter(Condition == "\"Treatment\""),
              aes(x = 1, xend = 2, y = freq*100 + se + 5, yend = freq*100 + se + 5),
              inherit.aes = FALSE) +
 geom text(data = df combined %>% filter(Category %in% c('\nWere Over \n500 Pages',
  → '\nFeatured an\nEntertainer', '\nWere Written in\nthe Past 25 Years') & Condition
  aes(x = 1.5, xend = 1.5, y = freq*100 + se + 7, yend = freq*100 + se + 7,
            \rightarrow label = "n.s."),
           inherit.aes = FALSE, vjust = 0, family = "Times New Roman", size = 6) +
 geom_text(data = df_combined %% filter(Category %in% c('\nFeatured a\nWoman') &
  aes(x = 1.5, xend = 1.5, y = freq*100 + se + 5, yend = freq*100 + se + 5,
            → label = "***"),
           inherit.aes = FALSE, vjust = 0, family = "Times New Roman", size = 6) +
 theme_bw() +
 scale_fill_manual(values = c("#990000", "#011F5B"), labels = c("No feedback provided",
 → "Feedback provided"), "Feedback") +
```

```
scale_y_continuous(labels = function(x) paste0(x,"%"), limits = c(0,100)) +
  scale_x_discrete(labels = c("\"Control\"" = "Not\nShown", "\"Treatment\"" = "Shown")) +
  labs(x = "Feedback on % of biographies that...", y = "% of Final Biographies with the
  → Target Attribute",
       title = "The Effect of Getting Feedback on Your Biography Selections") +
  theme(text = element_text(family = "Times New Roman"),
       plot.caption = element text(face = "italic", family = "Times New Roman"),
        legend.position = c(0.5, 0.95),
        legend.title = element_blank(),
        legend.direction = "horizontal",
       legend.text = element_text(size = 16, family = "Times New Roman"),
        legend.key.size = unit(7, 'mm'),
        legend.background = element_rect(fill = "white"),
        panel.grid.minor = element_blank(),
       panel.grid = element_blank(),
        panel.border = element_rect(fill= NA, color = "white"),
       plot.background = element_rect(fill = "white"),
       panel.background = element rect(fill = "white"),
       axis.title.x = element_text(face="bold", size = 16, vjust = 19, family = "Times
        → New Roman"),
       plot.title = element_blank(),
       axis.title.y = element_text(size = 16, color = "black", family = "Times New
        → Roman"),
        axis.text.x = element_blank(),
       axis.ticks = element blank(),
       axis.text.y = element_text(size = 16, color = "black", family = "Times New
        → Roman"),
       strip.text = element_text(size = 16, color = "black", family = "Times New
        → Roman"),
       strip.background = element_rect(colour = "white", fill = "white"))
# Save the plot with Times New Roman font
ggsave("../Manuscript_Figures/Figure-6.pdf", plot = p_combined, width = 10, height = 8,

    units = "in", device = cairo_pdf, family = "Times New Roman")
```

System of Equations

##		Test	Wald.Coefficient	P_Value
##	1	Female Feedback - Pages	30.233954	5.686358e-08
##	2	Female Feedback - Year	2.391022	1.225646e-01
##	3	Female Feedback - Entertainer	1.752781	1.860351e-01