

# Results Reproduction: Study 3

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## Introduction

YourMorals.org is a website that the authors of the Graham et al paper created to allow people to utilize social media to explore their morals. Participants self-select to become part of the research studies that are ran on the platform.

One of the datasets was used in the paper. This dataset contains responses from over 8,000 individuals on both the *What would you do for a million dollars* survey and the entirety of the 30-item MFQ. The MFQ will be explored in a different set of notes. This notebook is dedicated to reproducing the findings of Study 3 of the study, which will be described later.

To clean the dataset, R scripts were ran and the code can be found under `MilDolMan.R`. The dataset used in this analysis is a subset of the original that the authors posted on the Harvard Dataverse website.

The goal of the third study in the paper is to explore the differences between liberals and conservatives through another perspective: moral tradeoffs. Here, participants were given a series of acts that are deemed immoral under each of the foundations. Their task is to judge the acts and decide if they would do it *so long as* people pay them x amount of money such that the scoring looks as follows

Scoring (Values in dollars)

1 = 0 (I'd do it for free)

2 = 10

3 = 100

4 = 1,000

5 = 10,000

6 = 100,000

7 = a million dollars

8 = never for any amount of money

Before we begin with reproducing the figure and analyses, I will load the required dataset and packages.

```
morals <- read.csv("~/Desktop/Working/Moral-Psychology/YourMorals/YM-MilDol.csv",  
  header = TRUE, na.strings = c("", " ", "NA"))
```

```
library(car)  
library(dplyr)  
library(psych)  
library(ggplot2)  
library(GGally)  
library("ggpubr")  
library("reshape2")  
library(scales)  
library(lsr)
```

## Figure 4 - Moral Tradeoffs

In Figure 4, the graph displays items for each foundation averaged as a function of each level of political ideology. To reproduce this graph, I begin by creating averages across each foundation.

```
### Harm ###  
morals$Harm <- rowMeans(morals[, c("dogkick", "endangered", "overweight",  
  "anthill", "palm")], na.rm = TRUE)  
  
### Fairness ###  
morals$Fairness <- rowMeans(morals[, c("cards", "stealpoor",  
  "apartment", "ballots", "racepledge")], na.rm = TRUE)  
  
### Ingroup ###  
morals$Ingroup <- rowMeans(morals[, c("sportsbet", "flagburn",  
  "talkradio", "familyshun", "citizenrenounce", "leaveclub")],  
  na.rm = TRUE)  
  
### Authority ###  
morals$Authority <- rowMeans(morals[, c("parentcurse", "founderscurse",  
  "handgesture", "rottentomato", "fatherslap")], na.rm = TRUE)  
  
### Purity ###  
morals$Purity <- rowMeans(morals[, c("soulseal", "eatdog", "tail",
```

```
"molesterblood", "stageanimal")], na.rm = TRUE)
```

Next, I clean the political ideology variable. Here, we need to remove the answers where participants indicate that they don't know or belong in a third party ideology other than those that are considered mainstream.

```
table(morals$politics_new)
```

```
##
##           Conservative      Don't know/not political
##           511                233
##           Liberal           Libertarian
##           2554              1034
## Moderate/middle-of-the-road      Other
##           845                304
##           Slightly Conservative      Slightly Liberal
##           442                1100
##           Very Conservative          Very Liberal
##           139                1010
```

```
morals$ideo <- as.character(as.factor(morals$politics_new))
```

```
morals$ideo <- recode(morals$ideo, ` Moderate/middle-of-the-road` = "Moderate")
```

```
morals <- morals[!(morals$ideo == " Don't know/not political"),
  ]
```

```
morals <- morals[!(morals$ideo == " Libertarian"), ]
```

```
morals <- morals[!(morals$ideo == " Other"), ]
```

```
morals$ideo <- as.factor(as.character(morals$ideo))
```

```
library(forcats)
```

```
morals$ideo <- fct_explicit_na(morals$ideo, na_level = "NA")
```

```
table(morals$ideo)
```

```
##
##           Conservative      Liberal      Slightly Conservative
##           511                2554                442
##           Slightly Liberal      Very Conservative      Very Liberal
##           1100                139                1010
##           Moderate            NA
##           845                21
```

```
morals$ideo <- factor(morals$ideo, levels = c(" Very Liberal",
  " Liberal", " Slightly Liberal", "Moderate", " Slightly Conservative",
  " Conservative", " Very Conservative"))
```

```
table(morals$ideo)
```

```
##
##          Very Liberal          Liberal          Slightly Liberal
##          1010          2554          1100
##          Moderate  Slightly Conservative          Conservative
##          845          442          511
##          Very Conservative
##          139
```

To create the graph, we need to generate averages for each foundation based on each level of political ideology.

```
Harm <- aggregate(Harm ~ ideo, morals, mean, na.rm = TRUE)
Fairness <- aggregate(Fairness ~ ideo, morals, mean, na.rm = TRUE)
Ingroup <- aggregate(Ingroup ~ ideo, morals, mean, na.rm = TRUE)
Authority <- aggregate(Authority ~ ideo, morals, mean, na.rm = TRUE)
Purity <- aggregate(Purity ~ ideo, morals, mean, na.rm = TRUE)
```

The above step generates multiple data frames that would need to be combined in order to get one succinct, graphable data frame. This frame would then need to be “reshaped” so that the variables of interest such as ideology, moral foundations, and average values, would fit in it’s own variable.

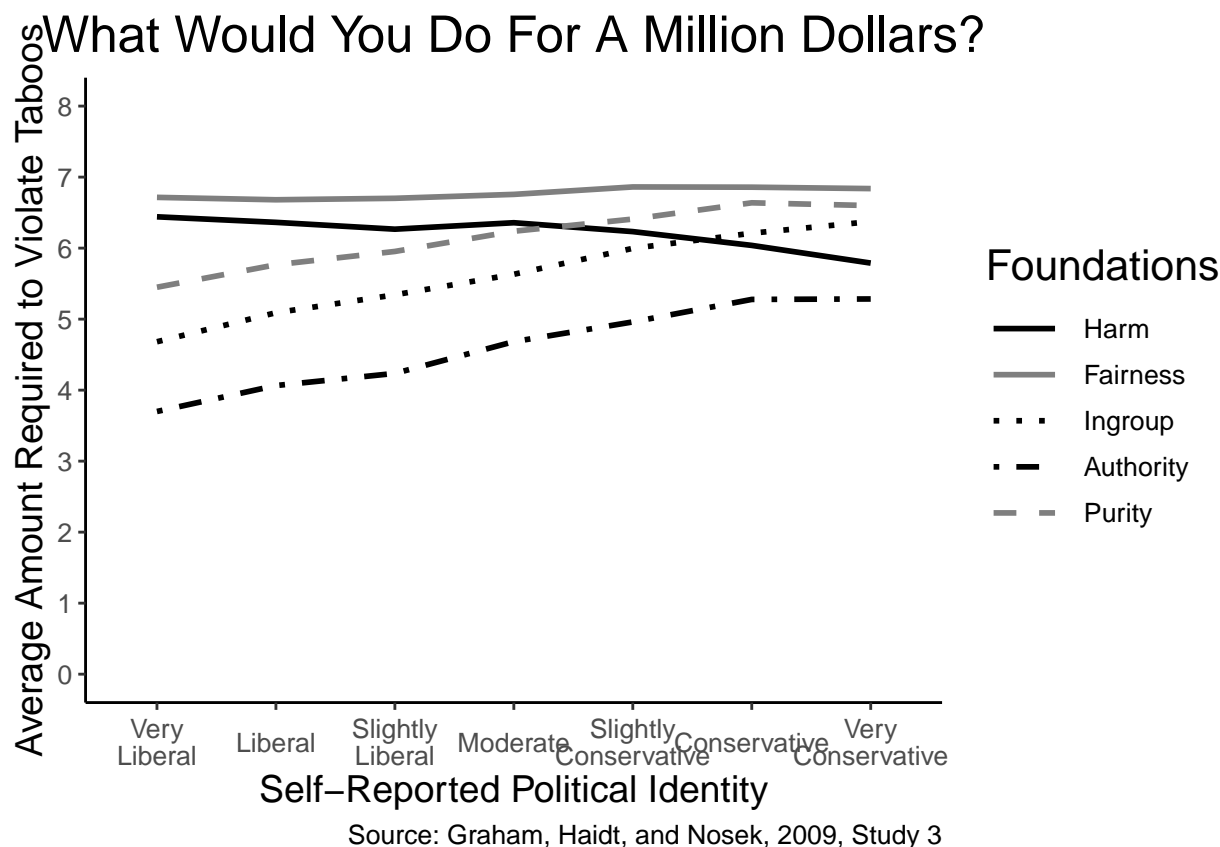
```
moral <- merge(Harm, Fairness, by.x = "ideo", by.y = "ideo",
  all.x = TRUE, all.y = TRUE)
moral <- merge(moral, Ingroup, by.x = "ideo", by.y = "ideo",
  all.x = TRUE, all.y = TRUE)
moral <- merge(moral, Authority, by.x = "ideo", by.y = "ideo",
  all.x = TRUE, all.y = TRUE)
moral <- merge(moral, Purity, by.x = "ideo", by.y = "ideo", all.x = TRUE,
  all.y = TRUE)

mfq <- reshape2::melt(moral, id.var = "ideo")
```

Finally, we generate the graph.

```
ggplot(mfq, aes(x = ideo, y = value, group = variable)) + geom_line(aes(linetype = variable,
  color = variable), size = 1) + theme_classic() + scale_linetype_manual("Foundations",
  breaks = c("Harm", "Fairness", "Ingroup", "Authority", "Purity"),
  values = c(Harm = "solid", Fairness = "solid", Ingroup = "dotted",
    Authority = "dotdash", Purity = "dashed")) + scale_color_manual("Foundations",
  breaks = c("Harm", "Fairness", "Ingroup", "Authority", "Purity"),
  values = c(Harm = "black", Fairness = "grey50", Ingroup = "black",
    Authority = "black", Purity = "grey50")) + ggtitle("What Would You Do For A Million Dollars")
  xlab("Self-Reported Political Identity") + ylab("Average Amount Required to Violate Moral Foundations")
```

```
labs(caption = "Source: Graham, Haidt, and Nosek, 2009, Study 3") +
theme(text = element_text(size = 12, colour = "black"), axis.title = element_text(s
  colour = "black"), title = element_text(size = 16, colour = "black"),
plot.caption = element_text(size = 10, color = "black"),
axis.text.x = element_text(angle = 0, hjust = 0.5, vjust = 0.5),
plot.title = element_text(hjust = 0.5), legend.key.width = unit(2,
  "line")) + scale_x_discrete(labels = wrap_format(10)) +
scale_y_continuous(breaks = seq(0, 8, 1), limits = c(0, 8))
```



## Repeated Measures GLM

To compare the liberals and conservatives, the authors conduct a repeated measures GLM.

We begin by creating a combined individualising and binding subscale score.

```
# Individualizing and binding foundation scores
morals$indiv <- rowMeans(morals[, c("dogkick", "endangered",
  "overweight", "anthill", "palm", "cards", "stealpoor", "apartment",
  "ballots", "racepledge")], na.rm = TRUE)
morals$bind <- rowMeans(morals[, c("sportsbet", "flagburn", "talkradio",
  "familyshun", "citizenrenounce", "leaveclub", "parentcurse",
```

```
"founderscurse", "handgesture", "rottentomato", "fatherslap",
"soulseal", "eatdog", "tail", "molesterblood", "stageanimal")],
na.rm = TRUE)
```

Next, I generate a difference score between the individualizing and binding foundations.

```
morals$diffscore <- morals$indiv - morals$bind
```

Now, I run the model and print out a table summarizing the results and include an  $\eta^2$  statistic.

```
diff.model <- lm(diffscore ~ politics, data = morals)
summary(diff.model)
```

```
##
## Call:
## lm(formula = diffscore ~ politics, data = morals)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.9207 -0.6586 -0.0430  0.6190  4.4126
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  2.169271   0.025449   85.24  <2e-16 ***
## politics    -0.290951   0.007771  -37.44  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9788 on 6576 degrees of freedom
## (44 observations deleted due to missingness)
## Multiple R-squared:  0.1757, Adjusted R-squared:  0.1756
## F-statistic: 1402 on 1 and 6576 DF, p-value: < 2.2e-16
```

```
etaSquared(diff.model)
```

```
##              eta.sq eta.sq.part
## politics 0.1757032  0.1757032
```

The results are interpreted as follows:

The F-statistic: 1402 on 1 and 6576 DF, p-value: < 2.2e-16 reflects the moderation of politics in the model. To find the difference between the scales as is, we square the t-value next to the (Intercept) row and use that p-value

The results are as follows: - Aggregate difference between individualizing and binding foundations:  $F(1, 6576) = 7265.85$ ,  $p < .001$  - Moderation by politics:  $F(1, 6576) = 1402$ ,  $p < .001$ ,  $\eta^2 = .175$

To see the distribution of the scores, I generate a scatterplot with the linear model fitted. the political ideology variable is across the x-axis and it is represented by 1 = Very Liberal to 7 = Very Conservative