

Moral Foundations Sacredness Scale: Measuring Morality

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1/6/2020

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Introduction

The [Measuring Morality Dataset](#) contains questionnaires collected by researchers at Duke University.

From this dataset, I am interested in the Moral Foundations Sacredness Scale, adopted from the Moral Tradeoffs questionnaire that was used in Study 3 of Graham, Haidt and Nosek's paper. While this is a shortened version of the one used in the original paper, it can still be insightful for the purposes of a replication to the results of the original study.

Before the analysis, I set up the process by loading some packages that will come in handy.

```
# Load packages
library(tidyverse)
library(psych)
library(ggplot2)
library(GGally)
library("ggpubr")
library("reshape2")
library(scales)
library(lsr)
```

Moral Foundations Sacredness Scale

Clean Data

I begin by loading the data available [here](#).

```
morals <- read.csv("~/Desktop/Working/Moral-Psychology/MMorality/mfss.csv",  
  header = TRUE)
```

I create an average score for each foundation by averaging an individual's response to each of the questions that belong in each foundation.

```
### Harm ###  
morals$Harm <- rowMeans(morals[, c("dogkick", "overweight", "palm")],  
  na.rm = TRUE)  
  
### Fairness ###  
morals$Fairness <- rowMeans(morals[, c("cards", "ballots", "racepledge")],  
  na.rm = TRUE)  
  
### Ingroup ###  
morals$Ingroup <- rowMeans(morals[, c("flagburn", "talkradio",  
  "familyshun")], na.rm = TRUE)  
  
### Authority ###  
morals$Authority <- rowMeans(morals[, c("parentcurse", "handgesture",  
  "rottentomato")], na.rm = TRUE)  
  
### Purity ###  
morals$Purity <- rowMeans(morals[, c("soulseal", "molesterblood",  
  "stageanimal")], na.rm = TRUE)
```

For the purposes of the graph, I create an ordered factor variable for the political ideology question.

```
morals$ideology <- as.character(as.numeric(morals$ideo7))  
  
morals$ideology <- recode(morals$ideology, `1` = "Extremely Liberal")  
morals$ideology <- recode(morals$ideology, `2` = "Liberal")  
morals$ideology <- recode(morals$ideology, `3` = "Slightly Liberal")  
morals$ideology <- recode(morals$ideology, `4` = "Moderate")  
morals$ideology <- recode(morals$ideology, `5` = "Slightly Conservative")  
morals$ideology <- recode(morals$ideology, `6` = "Conservative")  
morals$ideology <- recode(morals$ideology, `7` = "Extremely Conservative")  
# Rid implicit NAs for the ideology variable  
library(forcats)
```

```

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

# Convert to Factor
morals$ideology <- as.factor(morals$ideology)
morals$ideology <- factor(morals$ideology, levels = c("Extremely Liberal",
  "Liberal", "Slightly Liberal", "Moderate", "Slightly Conservative",
  "Conservative", "Extremely Conservative"))

# Remove NA
library(forcats)
morals$ideology <- fct_explicit_na(morals$ideology, na_level = "NA")

# Remove NA from ideology
morals <- morals[!(morals$ideology == "NA"), ]

table(morals$ideology)

##
##      Extremely Liberal      Liberal      Slightly Liberal
##              43              192              166
##      Moderate Slightly Conservative      Conservative
##              533              199              314
## Extremely Conservative      NA
##              54              0

```

Descriptive Statistics Plot

To create the plot, I generate average scores for each foundation as a function of each level of political ideology.

```

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

```

To generate a data frame that is usable to graph, I merge each of the data frames that were created above to one single frame.

```

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

```

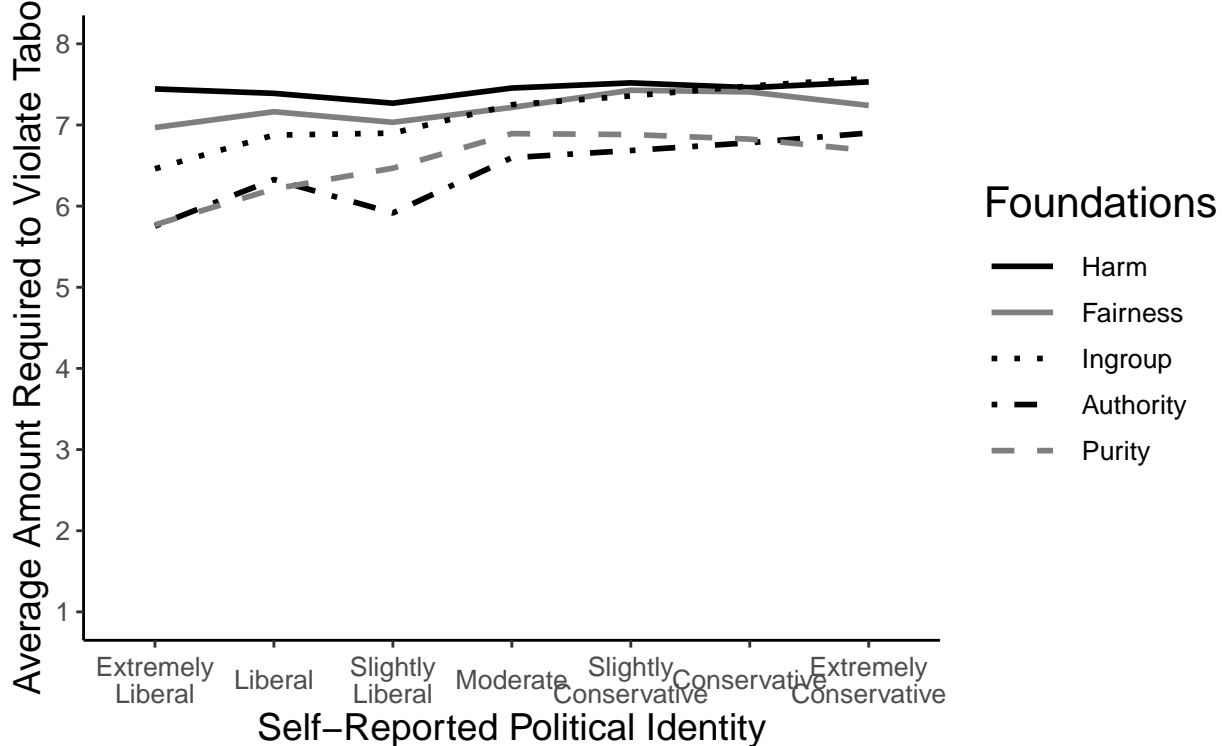
```
moral <- merge(moral, Purity, by.x = "ideology", by.y = "ideology",
  all.x = TRUE, all.y = TRUE)

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

Now, I graph.

```
ggplot(mfq, aes(x = ideology, 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") +
  labs(caption = "Source: Graham, Haidt, and Nosek, 2009, Study 3") +
  theme(text = element_text(size = 12, colour = "black"), axis.title = element_text(
    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(1, 8, 1), limits = c(1, 8))
```

What Would You Do For A Million Dollars?



Source: Graham, Haidt, and Nosek, 2009, Study 3

Cronbach's Alpha

I use the `psych` package to calculate the Cronbach's Alpha for the questions in each foundation

```
# Harm
Harm <- morals %>% select(c("dogkick", "overweight", "palm"))
psych::alpha(Harm)

##
## Reliability analysis
## Call: psych::alpha(x = Harm)
##
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean  sd median_r
##      0.66      0.67    0.58      0.41 2.1 0.014  7.4 1.1      0.42
##
##   lower alpha upper      95% confidence boundaries
## 0.63 0.66 0.69
##
## Reliability if an item is dropped:
##           raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## dogkick      0.59      0.60    0.43      0.43 1.5  0.021  NA  0.43
## overweight   0.54      0.55    0.38      0.38 1.2  0.023  NA  0.38
```

```

## palm          0.56      0.59      0.42      0.42 1.4      0.021      NA      0.42
##
## Item statistics
##          n raw.r std.r r.cor r.drop mean  sd
## dogkick   1475  0.71  0.77  0.57  0.47  7.7 1.1
## overweight 1481  0.83  0.79  0.62  0.51  7.2 1.7
## palm      1481  0.78  0.78  0.59  0.48  7.4 1.4
##
## Non missing response frequency for each item
##          1      2      3      4      5      6      7      8 miss
## dogkick   0.01 0.00 0.01 0.02 0.02 0.02 0.04 0.88 0.02
## overweight 0.02 0.01 0.03 0.04 0.05 0.04 0.07 0.74 0.01
## palm      0.03 0.00 0.01 0.02 0.02 0.05 0.09 0.79 0.01

# Fairness
Fairness <- morals %>% select(c("cards", "ballots", "racepledge"))
psych::alpha(Fairness)

##
## Reliability analysis
## Call: psych::alpha(x = Fairness)
##
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean  sd median_r
##      0.72      0.73      0.64      0.47 2.7 0.012  7.3 1.3      0.48
##
##   lower alpha upper      95% confidence boundaries
## 0.7 0.72 0.74
##
## Reliability if an item is dropped:
##          raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## cards          0.65      0.65      0.49      0.49 1.9   0.018   NA  0.49
## ballots        0.61      0.62      0.45      0.45 1.6   0.019   NA  0.45
## racepledge     0.64      0.65      0.48      0.48 1.9   0.018   NA  0.48
##
## Item statistics
##          n raw.r std.r r.cor r.drop mean  sd
## cards    1481  0.84  0.80  0.63  0.54  6.8 1.9
## ballots  1479  0.79  0.82  0.67  0.57  7.5 1.4
## racepledge 1481  0.78  0.80  0.64  0.54  7.4 1.4
##
## Non missing response frequency for each item
##          1      2      3      4      5      6      7      8 miss
## cards     0.03 0.01 0.04 0.06 0.07 0.06 0.09 0.63 0.01
## ballots   0.02 0.00 0.01 0.02 0.03 0.04 0.06 0.82 0.01
## racepledge 0.03 0.00 0.01 0.02 0.02 0.05 0.09 0.79 0.01

```

```
# Ingroup
```

```
Ingroup <- morals %>% select(c("flagburn", "talkradio", "familyshun"))  
psych::alpha(Ingroup)
```

```
##  
## Reliability analysis  
## Call: psych::alpha(x = Ingroup)  
##  
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean  sd median_r  
##     0.64     0.64    0.56     0.37 1.8 0.015  7.2 1.3     0.32  
##  
##   lower alpha upper      95% confidence boundaries  
## 0.61 0.64 0.67  
##  
## Reliability if an item is dropped:  
##           raw_alpha std.alpha G6(smc) average_r  S/N alpha se var.r med.r  
## flagburn      0.48      0.49   0.32      0.32 0.96   0.026   NA 0.32  
## talkradio      0.41      0.43   0.27      0.27 0.74   0.029   NA 0.27  
## familyshun     0.68      0.69   0.52      0.52 2.20   0.016   NA 0.52  
##  
## Item statistics  
##           n raw.r std.r r.cor r.drop mean  sd  
## flagburn  1476 0.83 0.78 0.63 0.50 7.1 1.9  
## talkradio  1480 0.82 0.81 0.67 0.55 7.1 1.7  
## familyshun 1483 0.64 0.70 0.42 0.34 7.5 1.3  
##  
## Non missing response frequency for each item  
##           1 2 3 4 5 6 7 8 miss  
## flagburn  0.04 0.01 0.02 0.05 0.03 0.04 0.08 0.72 0.02  
## talkradio  0.02 0.01 0.02 0.05 0.05 0.06 0.09 0.70 0.01  
## familyshun 0.02 0.00 0.00 0.01 0.02 0.04 0.13 0.76 0.01
```

```
# Authority
```

```
Authority <- morals %>% select(c("parentcurse", "handgesture",  
  "rottentomato"))  
psych::alpha(Authority)
```

```
##  
## Reliability analysis  
## Call: psych::alpha(x = Authority)  
##  
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean  sd median_r  
##     0.64     0.65    0.56     0.38 1.8 0.015  6.5 1.7     0.36  
##  
##   lower alpha upper      95% confidence boundaries
```

```
## 0.61 0.64 0.66
##
## Reliability if an item is dropped:
##      raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## parentcurse      0.63      0.63      0.46      0.46 1.73      0.019      NA 0.46
## handgesture      0.43      0.48      0.32      0.32 0.93      0.025      NA 0.32
## rottentomato      0.48      0.53      0.36      0.36 1.11      0.024      NA 0.36
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean sd
## parentcurse 1477 0.63 0.73 0.48 0.39 7.4 1.5
## handgesture 1481 0.82 0.79 0.63 0.51 6.2 2.4
## rottentomato 1480 0.82 0.78 0.60 0.49 6.0 2.6
##
## Non missing response frequency for each item
##      1 2 3 4 5 6 7 8 miss
## parentcurse 0.03 0.01 0.01 0.02 0.04 0.04 0.07 0.79 0.02
## handgesture 0.12 0.02 0.04 0.06 0.07 0.07 0.12 0.51 0.01
## rottentomato 0.14 0.02 0.04 0.06 0.08 0.06 0.09 0.51 0.01

# Purity
Purity <- morals %>% select(c("soulsell", "molesterblood", "stageanimal"))
psych::alpha(Purity)

##
## Reliability analysis
## Call: psych::alpha(x = Purity)
##
##      raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
##      0.43      0.51      0.42      0.26 1.1 0.023 6.7 1.5      0.22
##
## lower alpha upper      95% confidence boundaries
## 0.39 0.43 0.48
##
## Reliability if an item is dropped:
##      raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## soulsell      0.30      0.36      0.22      0.22 0.57      0.028      NA 0.22
## molesterblood 0.53      0.54      0.37      0.37 1.15      0.024      NA 0.37
## stageanimal   0.29      0.33      0.20      0.20 0.49      0.031      NA 0.20
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean sd
## soulsell 1476 0.63 0.73 0.51 0.31 7.3 1.7
## molesterblood 1480 0.82 0.66 0.35 0.25 5.5 3.0
## stageanimal 1483 0.62 0.74 0.54 0.34 7.2 1.4
```



```
##
## Non missing response frequency for each item
##      1      2      3      4      5      6      7      8 miss
## soulsell      0.03 0.01 0.02 0.03 0.02 0.03 0.06 0.81 0.02
## molesterblood 0.26 0.01 0.02 0.04 0.05 0.04 0.08 0.50 0.01
## stageanimal   0.01 0.00 0.01 0.04 0.06 0.06 0.13 0.69 0.01
```

Repeated Measures GLM

To compare the aggregate individualizing and binding moral foundation, I generate the score below that represents an average of the responses to the questions under each category.

```
### Individualizing and Binding items ###
morals$indiv <- rowMeans(morals[, c("dogkick", "overweight",
  "palm", "cards", "ballots", "racepledge")], na.rm = TRUE)
morals$bind <- rowMeans(morals[, c("flagburn", "talkradio", "familyshun",
  "parentcurse", "handgesture", "rottentomato", "soulsell",
  "molesterblood", "stageanimal")], na.rm = TRUE)
```

I then create a difference score that represents the difference between the individual and binding foundation scores. I conduct this analysis by running a linear regression with political ideology as a moderating variable.

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

diff.model <- lm(diffscore ~ ideo7, data = morals)
summary(diff.model)
```

```
##
## Call:
## lm(formula = diffscore ~ ideo7, data = morals)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.9667 -0.5501 -0.2167  0.4434  5.1264
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.98147    0.07812  12.564 < 2e-16 ***
## ideo7         -0.10786    0.01753  -6.153 9.75e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9865 on 1486 degrees of freedom
## (13 observations deleted due to missingness)
```

```
## Multiple R-squared:  0.02485,    Adjusted R-squared:  0.02419
## F-statistic: 37.86 on 1 and 1486 DF,  p-value: 9.745e-10
```

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

```
##           eta.sq eta.sq.part
## ideo7 0.02484611  0.02484611
```

The reported results are as follows:

- Aggregate difference between Individualizing and binding foundation: $F(1, 1486) = 157.854$, $p < .001$
- Moderation by Politics: $F(1, 1486) = 37.86$, $p < .001$, $\eta^2 = .025$