

Can Exposure to Moral Foundations Affect Our Reactions to Policy Proposals?

W241 Experiments and Causality (submitted December X, 2019)

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December 10, 2019

Abstract

TBD

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1 Background

[[TBD]]

2 Data

[[TBD]]

```
# Pilot
# results_panel1_raw = read.xlsx("./data/pilot/survey_results_pilot_panel1.xlsx") %>% filter(StartDate
# results_panel2_raw = read.xlsx("./data/pilot/survey_results_pilot_panel2.xlsx") %>% filter(StartDate
# participant_detail_panel1 = read.csv("./data/pilot/participant_detail_pilot_panel1.csv", stringsAsFac
# participant_detail_panel2 = read.csv("./data/pilot/participant_detail_pilot_panel2.csv", stringsAsFac

# Study
results_panel1_raw = read.xlsx("./data/study/MF Framing Pilot - Full Recruitment - Panel 1_November 20,
results_panel2_raw = read.xlsx("./data/study/MF Framing Pilot - Full Recruitment - Panel 2_November 20,
participant_detail_panel1 = read.csv("./data/study/prolific_export_5dd4a350108b6748b25b5de1.csv", string
participant_detail_panel2 = read.csv("./data/study/prolific_export_5dd4a34135582248315dfdca.csv", string
```

2.0.1 Data Cleaning

[[TBD]]¹

```
# Stack panel data
results_stacked = bind_rows(results_panel1_raw %>% mutate(panel = 1)
                           , results_panel2_raw %>% mutate(panel = 2)) %>%
  merge(bind_rows(participant_detail_panel1, participant_detail_panel2) %>% select(-session_id, -status
        , by.x = "PROLIFIC_PID"
        , by.y = "participant_id"
        , all.x = TRUE)

# Adjust all variable names to remove '-' and '.' + lowercase
names(results_stacked) = tolower(gsub(x = names(results_stacked), pattern = "\\-|\\.", replacement = "_

# Discrete variables as factors (manual ordering for plotting)
ideology_levels = c("Very Liberal", "Lean Liberal", "Liberal", "Moderate", "Conservative", "Lean Conser
ubi_group_levels = c("Promoter", "Passive", "Detractor")
ubi_familiarity_levels = c("Extremely familiar", "Very familiar", "Moderately familiar", "Slightly fami

results_full = results_stacked %>%
  # Define arms and nodes
  mutate(arm = case_when(grepl('a', fc_b_1, ignore.case = TRUE) ~ "purity_base"
                        , grepl('a', fc_c_1, ignore.case = TRUE) ~ "purity_extension"
                        , grepl('a', fc_d_1, ignore.case = TRUE) ~ "fairness_base"
                        , grepl('a', fc_e_1, ignore.case = TRUE) ~ "fairness_extension"
                        , TRUE ~ "control") %>% factor()
        , node = paste0(arm, "_panel_", panel)
        # Combine reaction vars from different arms
        , purity_q1_self = case_when(grepl('a', fc_b_1, ignore.case = TRUE) ~ fc_b_1
```

¹[[Example footnote]]

```

, TRUE ~ fc_c_1)
, purity_q2_repulsed = case_when(grepl('a', fc_b_2, ignore.case = TRUE) ~ fc_b_2
, TRUE ~ fc_c_2)
, purity_q3_injustice = case_when(grepl('a', fc_b_3, ignore.case = TRUE) ~ fc_b_3
, TRUE ~ fc_c_3)
, purity_q4_relieved = fc_c_4
, fairness_q1_self = case_when(grepl('a', fc_d_1, ignore.case = TRUE) ~ fc_d_1
, TRUE ~ fc_e_1)
, fairness_q2_pain = case_when(grepl('a', fc_d_2, ignore.case = TRUE) ~ fc_d_2
, TRUE ~ fc_e_2)
, fairness_q3_injustice = case_when(grepl('a', fc_d_3, ignore.case = TRUE) ~ fc_d_3
, TRUE ~ fc_e_3)
, fairness_q4_relieved = fc_e_4
, open_text_reaction = q3_fc2
# Factor variables
, ideology = factor(polispect, levels = ideology_levels)
, ideology_bin = case_when(is.na(ideology) ~ "missing"
, ideology == "Very Liberal" ~ "liberal"
, ideology == "Lean Liberal" ~ "liberal"
, ideology == "Liberal" ~ "liberal"
, ideology == "Very Conservative" ~ "conservative"
, ideology == "Lean Conservative" ~ "conservative"
, ideology == "Conservative" ~ "conservative"
, TRUE ~ "moderate")
, ubi_group = factor(ubi_2_nps_group, levels = ubi_group_levels)
, ubi_familiarity = factor(ubi_f, levels = ubi_familiarity_levels)
, ubi_familiarity_bin = case_when(is.na(ubi_f) ~ "missing"
, ubi_f == "Not familiar at all" ~ "no"
, TRUE ~ "yes")

# Numeric variables
, ubi_number = as.numeric(ubi_2))

results_clean = results_full %>%
  select(prolific_pid, panel, arm, node
, ideology, ideology_bin, age, gender, urban, employment_status, student_status
, purity_q1_self, purity_q2_repulsed, purity_q3_injustice, purity_q4_relieved
, fairness_q1_self, fairness_q2_pain, fairness_q3_injustice, fairness_q4_relieved
, open_text_reaction
, ubi_number, ubi_group, ubi_familiarity)

```

2.1 Exploratory Analysis

[[TBD]]

2.1.1 Study Setup

```

arm_counts_bypanel = results_clean %>%
  group_by(arm, panel, node) %>%
  summarise(count = n())
arm_counts_all = results_clean %>%
  group_by(arm) %>%

```

```

summarise(count = n())

# nodes = data.frame(
#   id = 1:12
#   , group = c("liberal", "conservative", arm_counts$arm)
#   # , label = c("liberal", "conservative", arm_counts$node)
#   , label = c("All Liberals", "All Conservatives"
#   , "Control: Liberal", "Control: Conservative"
#   , "Fairness Base: Liberal", "Fairness Base: Conservative"
#   , "Fairness Extension: Liberal", "Fairness Extension: Conservative"
#   , "Purity Base: Liberal", "Purity Base: Conservative"
#   , "Purity Extension: Liberal", "Purity Extension: Conservative")
#   , value = c(sum(arm_counts$count[arm_counts$panel==1]), sum(arm_counts$count[arm_counts$panel==2])),
#   , shape = rep("box", 12)
#   # color, shape
# )
#
# edges = data.frame(
#   from = c(1, 2, 1, 2, 1, 2, 5, 6, 9, 10)
#   , to = c(3, 4, 5, 6, 9, 10, 7, 8, 11, 12)
#   , dashes = c(rep(FALSE, 6), rep(TRUE, 4))
# )

nodes = data.frame(
  id = 1:7
  , group = c("ideology", "ideology"
  , "fairness", "fairness"
  , "purity", "purity"
  , "control"
  )
  # , label = c("liberal", "conservative", arm_counts$node)
  , label = c("All Liberals", "All Conservatives"
  , "Fairness Base"
  , "Fairness Extension"
  , "Purity Base"
  , "Purity Extension"
  , "Control"
  )
  , level = c(1, 1, 2, 3, 2, 3, 2)
  # , mass = c(sum(arm_counts_bypanel$count[arm_counts_bypanel$panel==1])
  #           , sum(arm_counts_bypanel$count[arm_counts_bypanel$panel==2]), arm_counts_all$count)
  , shape = rep("box", 7)
)

edges = data.frame(
  from = c(1, 2, 1, 2, 1, 2, 3, 5)
  , to = c(3, 3, 5, 5, 7, 7, 4, 6)
  , dashes = c(rep(FALSE, 6), rep(TRUE, 2))
)

visNetwork(nodes, edges) %>%
  visEdges(arrows = "to") %>%
  visHierarchicalLayout(direction = "LR")

```

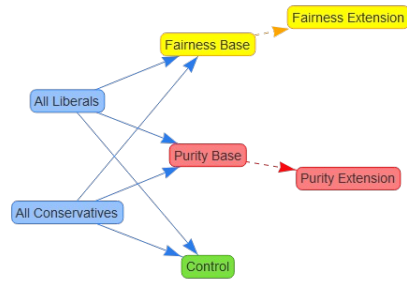


Figure 1: Study Setup

2.1.2 Demographics

```

grpstackbar_plot = ggplot() +
  facet_grid( ~ panel) +
  scale_fill_brewer(type = "div", palette = 5, direction = -1, aesthetics = "fill") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))

# Ideology
plot_ideology = grpstackbar_plot +
  geom_bar(data = results_clean %>% group_by(ideology, panel) %>% summarise(count = n())
    , aes(x = ideology, y = count, fill = ideology)
    , stat = "identity") +
  theme(legend.position = "left", axis.text.x=element_blank())

# Age
plot_age = grpstackbar_plot +
  geom_bar(data = results_clean %>% group_by(age, ideology, panel) %>% summarise(count = n())
    , aes(x = age, y = count, fill = ideology)
    , stat = "identity", show.legend = FALSE)

# Gender
plot_gender = grpstackbar_plot +
  geom_bar(data = results_clean %>% group_by(gender, ideology, panel) %>% summarise(count = n())
    , aes(x = gender, y = count, fill = ideology)
    , stat = "identity", show.legend = FALSE)

# Urban

```

```

plot_urban = grpstackbar_plot +
  geom_bar(data = results_clean %>% group_by(urban, ideology, panel) %>% summarise(count = n())
    , aes(x = urban, y = count, fill = ideology)
    , stat = "identity", show.legend = FALSE)

grid.arrange(plot_ideology, plot_age
  , plot_gender, plot_urban
  , nrow = 2)

```

Example reference to r cell *Figure 2* shows *[[TBD]]*

2.1.3 Reactions

```

response_levels = c("A great deal", "A lot", "A moderate amount", "A little", "None at all")

results_response = results_clean %>%
  select(panel, arm
    , purity_q1_self, purity_q2_repulsed, purity_q3_injustice, purity_q4_relieved
    , fairness_q1_self, fairness_q2_pain, fairness_q3_injustice, fairness_q4_relieved) %>%
  gather(prompt, value, -panel, -arm) %>%
  filter(!is.na(value)) %>%
  group_by(panel, arm, prompt, value) %>% summarise(count = n()) %>%
  mutate(response = factor(value, levels = response_levels))
  # spread(value, count) %>%
  # arrange(panel, arm, prompt) %>%
  # select("panel", "arm", "prompt", "A great deal", "A lot", "A moderate amount", "A little", "None at all")

ggplot(data = results_response
  , aes(x = prompt, y = response, fill = count)) +
  geom_tile() +
  facet_grid(rows = vars(arm), cols = vars(panel)) +
  scale_fill_distiller(direction = 1) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))

```

2.1.4 Outcome

```

# Histogram of familiarity
plot_familiarity = ggplot(data = results_clean %>% group_by(ubi_familiarity, ideology, panel) %>% summarise(count = n())) +
  aes(x = ubi_familiarity, y = count, fill = ideology)) +
  geom_bar(stat="identity", show.legend = FALSE) +
  facet_grid( ~ panel) +
  scale_fill_brewer(type = "div", palette = 5, direction = -1, aesthetics = "fill") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))

# Heat map of number UBI like
plot_ubi = ggplot(data = results_clean %>% group_by(ubi_familiarity, ideology) %>% summarise(ubi_number_avg = mean(ubi_number))) +
  aes(x = ubi_familiarity, y = ideology, fill = ubi_number_avg)) +
  geom_tile() +
  scale_fill_distiller(direction = 1) +

```

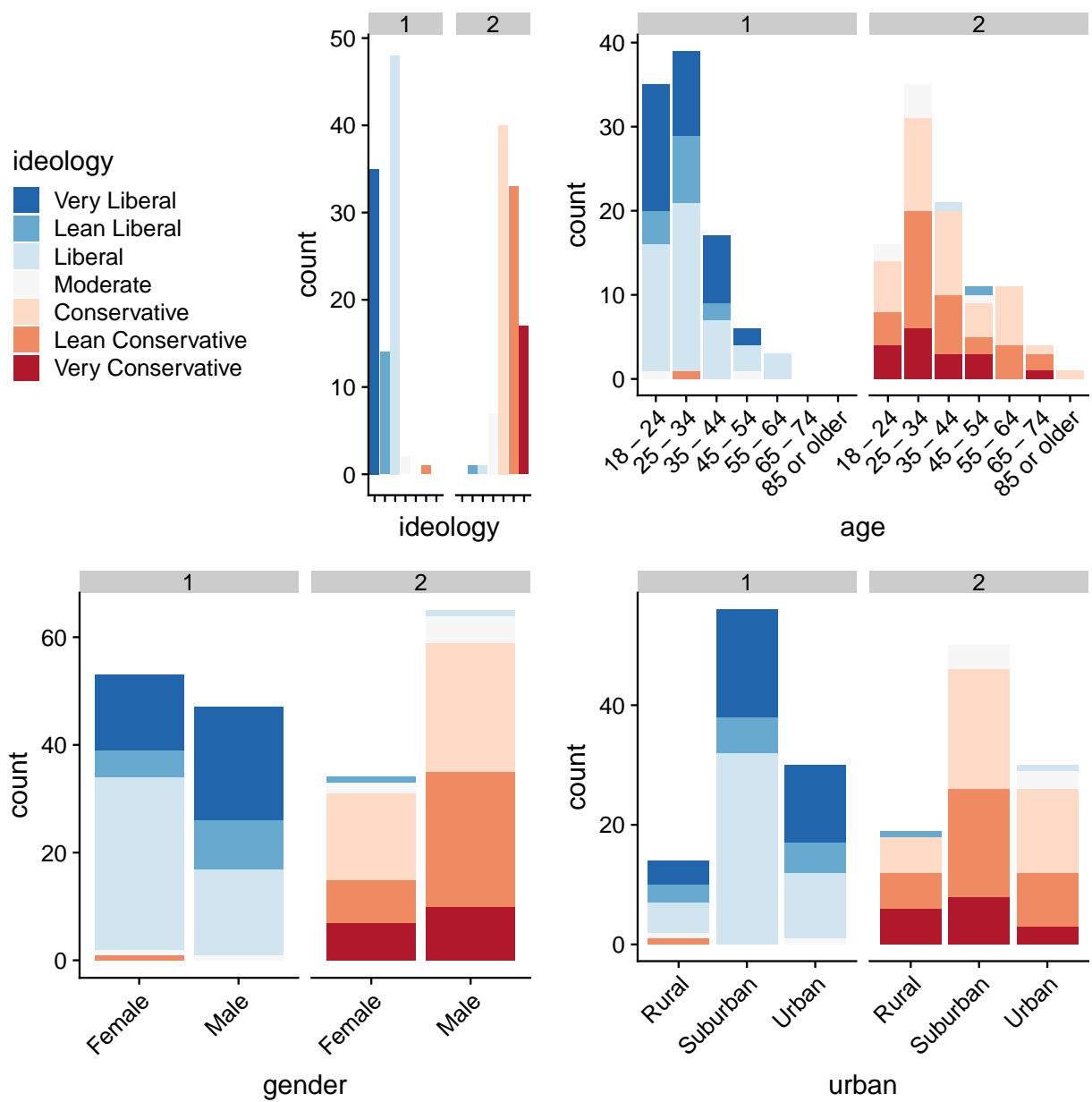


Figure 2: Demographics

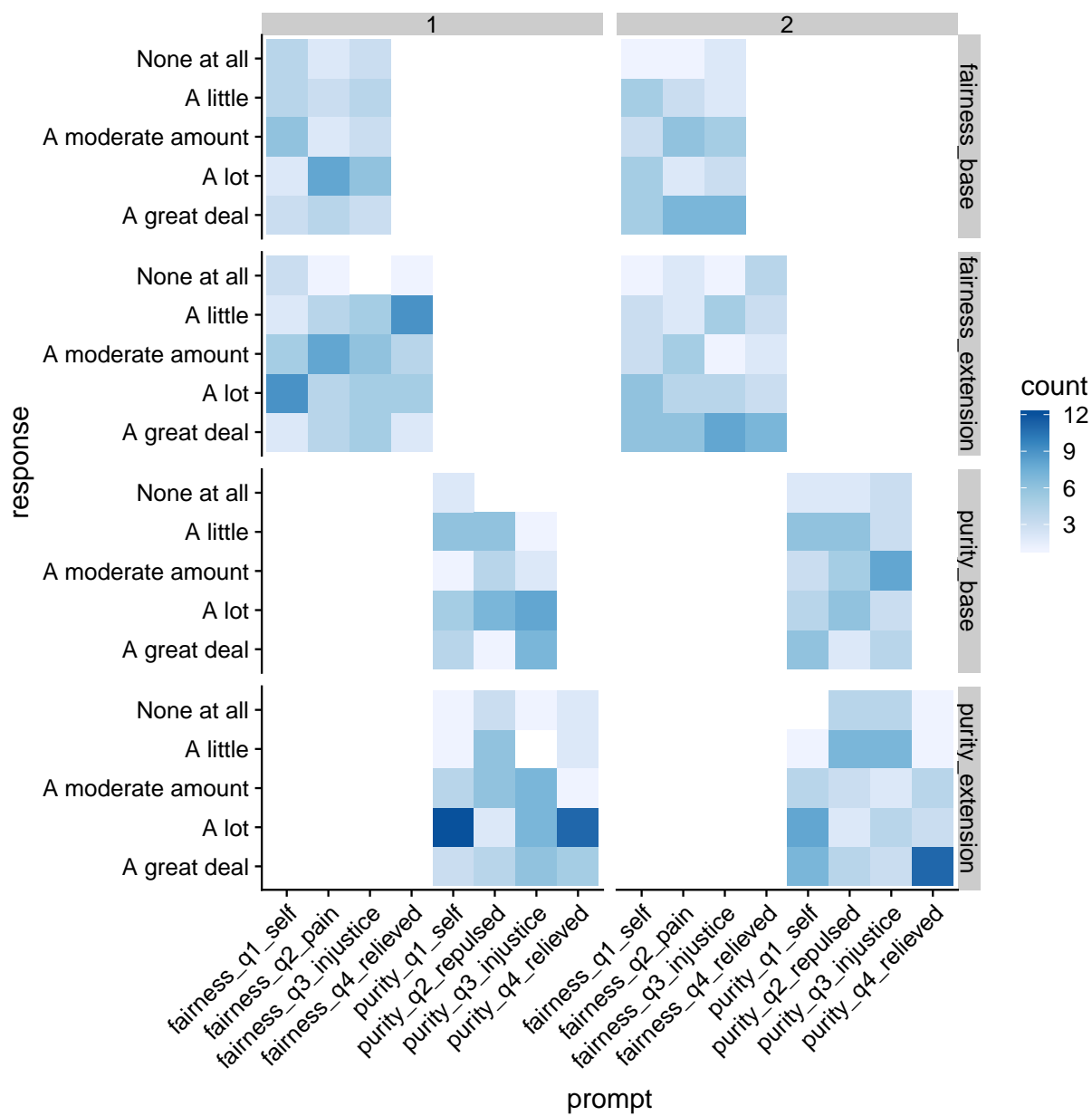


Figure 3: Reactions


```
theme(axis.text.x = element_text(angle = 45, hjust = 1)
      , legend.position = "right")

grid.arrange(plot_familiarity, plot_ubi
              , nrow = 2)
```

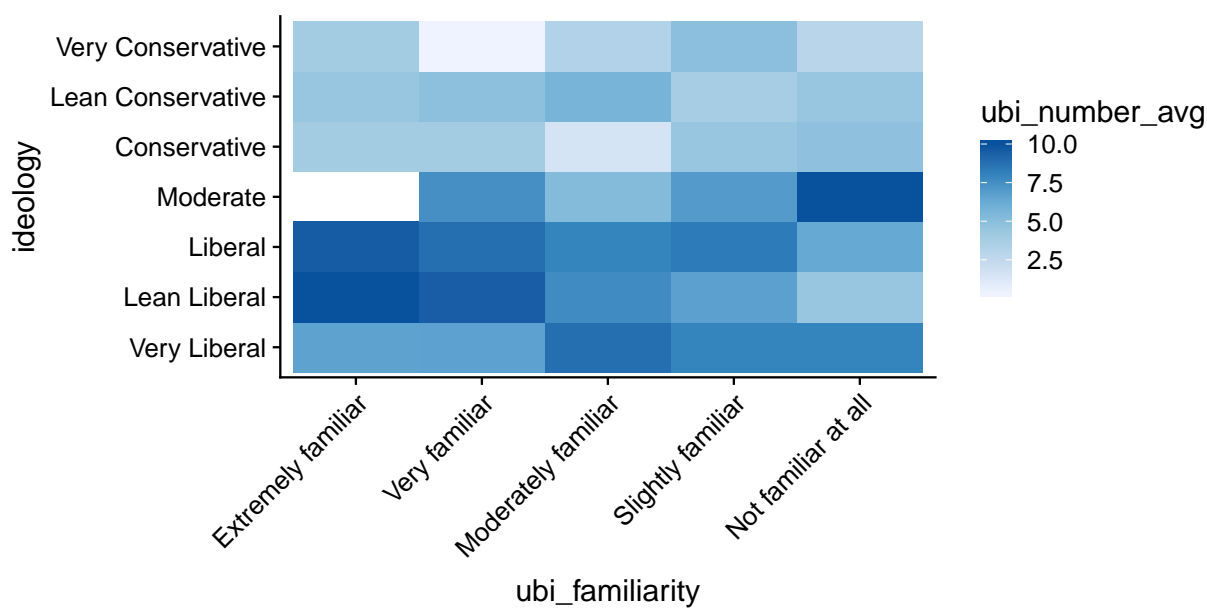
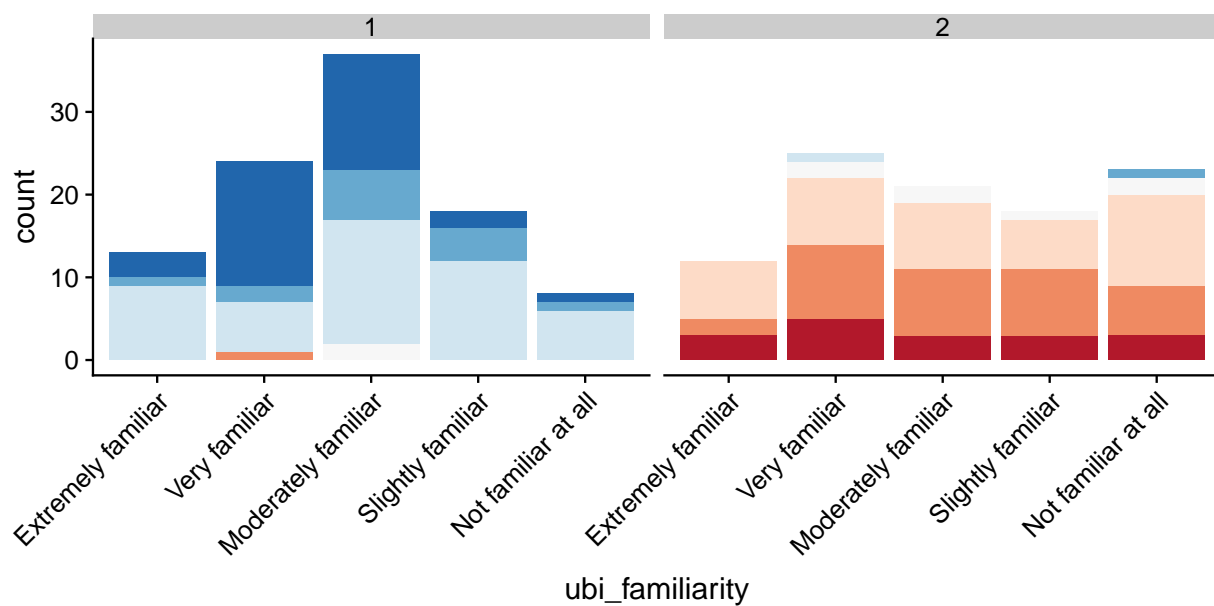


Figure 4: Outcomes

3 Methodology

3.1 Model 1

Independent variable

Dependent variable

Model specification

[[TBD]]. (see ??)

```
my_lm_calcs = function(lm_in, clusters_in){  
  # Robust  
  vcov_robust = vcovHC(lm_in)  
  se_robust = sqrt(diag(vcov_robust))  
  # Cluster  
  if(length(clusters_in) > 1){  
    vcov_cluster = cluster.vcov(lm_in, clusters_in)  
    se_cluster = sqrt(diag(vcov_cluster))  
  } else {  
    vcov_cluster = NA  
    se_cluster = NA  
  }  
  # Output  
  lm_out = list(lm = lm_in  
                , vcov_robust = vcov_robust  
                , se_robust = se_robust  
                , vcov_cluster = vcov_cluster  
                , se_cluster = se_cluster  
                )  
  return(lm_out)  
}
```

```
# Summaries  
all_data = data.table(results_full)  
all_data[, ideology_bin := case_when(is.na(ideology) ~ "missing",  
                                     ideology=="Very Liberal" ~ "liberal",  
                                     ideology=="Lean Liberal" ~ "liberal",  
                                     ideology=="Liberal" ~ "liberal",  
                                     ideology=="Very Conservative" ~ "conservative",  
                                     ideology=="Lean Conservative" ~ "conservative",  
                                     ideology=="Conservative" ~ "conservative",  
                                     TRUE ~ "moderate"), ]  
all_data[, story := case_when(is.na(arm) ~ "missing",  
                              arm=="control" ~ "",  
                              arm=="purity_base" ~ "homeless",  
                              arm=="purity_extension" ~ "homeless-cleaned",  
                              arm=="fairness_base" ~ "bullying",  
                              arm=="fairness_extension" ~ "bullying-addressed",  
                              TRUE ~ "others"), ]  
all_data[, FemaleReader := ifelse(gender!="Male", 1, 0), ]  
all_data[, ubi_familiarity_dummy := ifelse(ubi_familiarity_bin!="yes", 1, 0), ]  
all_data[, ubi := as.numeric(ubi_2), ]  
all_data[, .(ubi=mean(as.numeric(ubi_2)), .N), keyby=list(ideology_bin, story, FemaleReader) ]
```

##	ideology_bin	story	FemaleReader	ubi	N
## 1:	conservative		0	4.642857	14
## 2:	conservative		1	0.500000	4
## 3:	conservative	bullying	0	4.363636	11
## 4:	conservative	bullying	1	1.857143	7
## 5:	conservative	bullying-addressed	0	4.500000	10
## 6:	conservative	bullying-addressed	1	4.750000	8
## 7:	conservative	homeless	0	5.416667	12
## 8:	conservative	homeless	1	1.714286	7
## 9:	conservative	homeless-cleaned	0	4.666667	12
## 10:	conservative	homeless-cleaned	1	3.166667	6
## 11:	liberal		0	9.142857	7
## 12:	liberal		1	7.500000	14
## 13:	liberal	bullying	0	7.700000	10
## 14:	liberal	bullying	1	8.000000	10
## 15:	liberal	bullying-addressed	0	8.100000	10
## 16:	liberal	bullying-addressed	1	6.111111	9
## 17:	liberal	homeless	0	7.909091	11
## 18:	liberal	homeless	1	9.428571	7
## 19:	liberal	homeless-cleaned	0	7.444444	9
## 20:	liberal	homeless-cleaned	1	8.666667	12
## 21:	moderate		0	7.000000	1
## 22:	moderate		1	8.000000	1
## 23:	moderate	bullying-addressed	0	4.333333	3
## 24:	moderate	homeless	0	7.500000	2
## 25:	moderate	homeless-cleaned	1	10.000000	2
##	ideology_bin	story	FemaleReader	ubi	N

```
# augment model with results from coefteest and coefci using
# the supplied variance/covariance matrix
```

```
apply_robust_estimates = function(model, vcovmatrix) {
  model$robust.coefteest = coefteest(model, vcov. = vcovmatrix)
  model$robust.estimate = model$robust.coefteest[, 'Estimate']
  model$robust.std.error = model$robust.coefteest[, 'Std. Error']
  model$robust.statistic = model$robust.coefteest[, 't value']
  model$robust.p.value = model$robust.coefteest[, 'Pr(>|t|)']
  model$robust.coefci = coefci(model, vcov. = vcovmatrix)
  model$robust.ci.low = model$robust.coefci[, "2.5 %"]
  model$robust.ci.high = model$robust.coefci[, "97.5 %"]
  model
}
```

```
# liberals - homeless story
```

```
hom_lib_mod<-all_data[ ideology_bin=='liberal' & story != 'bullying' & story != 'bullying-addressed', ]
hom_lib_mod<-apply_robust_estimates(hom_lib_mod, vcovHC(hom_lib_mod))
```

```
# conservatives - homeless story
```

```
hom_con_mod<-all_data[ ideology_bin=='conservative' & story != 'bullying' & story != 'bullying-addressed', ]
hom_con_mod<-apply_robust_estimates(hom_con_mod, vcovHC(hom_con_mod))
```

```
# liberals - jacket story
```

```
jac_lib_mod<-all_data[ ideology_bin=='liberal' & story != 'homeless' & story != 'homeless-cleaned', ]
jac_lib_mod<-apply_robust_estimates(jac_lib_mod, vcovHC(jac_lib_mod))
```

```
# conservatives - jacket story
```

```
jac_con_mod<-all_data[ ideology_bin=='conservative' & story != 'homeless' & story != 'homeless-cleaned', ]
jac_con_mod<-apply_robust_estimates(jac_con_mod, vcovHC(jac_con_mod))
```

```
stargazer(hom_lib_mod, hom_con_mod,
```

```

type = 'text',
se = list(sqrt(diag(vcovHC(hom_lib_mod))),
          sqrt(diag(vcovHC(hom_con_mod)))),
header=F,
report=('v*c*sp'),
add.lines=list(c("Ideology", "Liberal", "Conservative")))

```

```

##
## =====
##                               Dependent variable:
##                               -----
##                               ubi
##                               (1)          (2)
## -----
## storyhomeless                -1.234**      0.774
##                               (0.616)      (1.605)
##                               p = 0.046      p = 0.630
##
## storyhomeless-cleaned        -1.698*       0.024
##                               (1.003)      (1.401)
##                               p = 0.091      p = 0.987
##
## FemaleReader                 -1.643**      -4.143***
##                               (0.813)      (1.038)
##                               p = 0.044      p = 0.0001
##
## storyhomeless:FemaleReader    3.162***      0.440
##                               (0.976)      (1.902)
##                               p = 0.002      p = 0.817
##
## storyhomeless-cleaned:FemaleReader 2.865**      2.643
##                               (1.271)      (1.774)
##                               p = 0.025      p = 0.137
##
## Constant                     9.143***      4.643***
##                               (0.436)      (0.983)
##                               p = 0.000      p = 0.00001
##
## -----
## Ideology                     Liberal        Conservative
## Observations                  60             55
## R2                           0.154          0.188
## Adjusted R2                   0.075          0.105
## Residual Std. Error          1.815 (df = 54)  3.320 (df = 49)
## F Statistic                   1.962* (df = 5; 54) 2.267* (df = 5; 49)
## =====
## Note:                         *p<0.1; **p<0.05; ***p<0.01

```

```

stargazer(jac_lib_mod, jac_con_mod,
type = 'text',
se = list(sqrt(diag(vcovHC(jac_lib_mod))),
          sqrt(diag(vcovHC(jac_con_mod)))),
header=F,

```

```
report=('v*c*sp'),
add.lines=list(c("Ideology", "Liberal", "Conservative"))))
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               ubi
##                               (1)          (2)
## -----
## storybullying                -1.443      -0.279
##                               (1.097)      (1.446)
##                               p = 0.189      p = 0.847
##
## storybullying-addressed       -1.043*      -0.143
##                               (0.612)      (1.673)
##                               p = 0.089      p = 0.932
##
## FemaleReader                 -1.643**      -4.143***
##                               (0.813)      (1.038)
##                               p = 0.044      p = 0.0001
##
## storybullying:FemaleReader     1.943        1.636
##                               (1.538)      (2.001)
##                               p = 0.207      p = 0.414
##
## storybullying-addressed:FemaleReader -0.346        4.393*
##                               (1.218)      (2.286)
##                               p = 0.777      p = 0.055
##
## Constant                     9.143***      4.643***
##                               (0.436)      (0.983)
##                               p = 0.000      p = 0.00001
##
## -----
## Ideology                     Liberal        Conservative
## Observations                  60             54
## R2                           0.126          0.136
## Adjusted R2                   0.045          0.046
## Residual Std. Error           2.278 (df = 54)  3.546 (df = 48)
## F Statistic                   1.555 (df = 5; 54) 1.510 (df = 5; 48)
## =====
## Note:                        *p<0.1; **p<0.05; ***p<0.01
```

```
# augment model with results from coeftest and coefci using
# the supplied variance/covariance matrix
# liberals - homeless story
hom_lib_mod_2<-all_data[ ideology_bin=='liberal' & story != 'bullying' & story != 'bullying-addressed',
hom_lib_mod_2<-apply_robust_estimates(hom_lib_mod_2, vcovHC(hom_lib_mod_2))
# conservatives - homeless story
hom_con_mod_2<-all_data[ ideology_bin=='conservative' & story != 'bullying' & story != 'bullying-addressed',
hom_con_mod_2<-apply_robust_estimates(hom_con_mod_2, vcovHC(hom_con_mod_2))
# liberals - jacket story
```

```

jac_lib_mod_2<-all_data[ ideology_bin=='liberal' & story != 'homeless' & story != 'homeless-cleaned', ]
jac_lib_mod_2<-apply_robust_estimates(jac_lib_mod_2, vcovHC(jac_lib_mod_2))
# conservatives - jacket story
jac_con_mod_2<-all_data[ ideology_bin=='conservative' & story != 'homeless' & story != 'homeless-cleaned', ]
jac_con_mod_2<-apply_robust_estimates(jac_con_mod_2, vcovHC(jac_con_mod_2))
stargazer(hom_lib_mod_2, hom_con_mod_2,
          type = 'text',
          se = list(sqrt(diag(vcovHC(hom_lib_mod_2))),
                    sqrt(diag(vcovHC(hom_con_mod_2)))),
          header=F,
          report=('v*c*sp'),
          add.lines=list(c("Ideology", "Liberal", "Conservative"))))

```

```

##
## =====
##                                     Dependent variable:
##                                     -----
##                                     ubi
##                                     (1)          (2)
## -----
## storyhomeless                    0.196          0.633
##                                     (1.374)
##                                     p = 0.645
##
## storyhomeless-cleaned            0.111          1.049
##                                     (1.269)
##                                     p = 0.409
##
## ubi_familiarity_dummy            -2.000          1.492
##                                     (2.321)
##                                     p = 0.521
##
## storyhomeless:ubi_familiarity_dummy  1.471          -0.433
##                                     (7.504)
##                                     p = 0.954
##
## storyhomeless-cleaned:ubi_familiarity_dummy -0.111          -2.349
##                                     (3.322)
##                                     p = 0.480
##
## Constant                        8.333          3.308***
##                                     (0.982)
##                                     p = 0.001
## -----
## Ideology                        Liberal          Conservative
## Observations                     60              55
## R2                               0.115            0.021
## Adjusted R2                     0.033            -0.078
## Residual Std. Error             1.856 (df = 54)    3.644 (df = 49)
## F Statistic                     1.403 (df = 5; 54)  0.215 (df = 5; 49)
## =====
## Note:                            *p<0.1; **p<0.05; ***p<0.01

```

```
stargazer(jac_lib_mod_2, jac_con_mod_2,
          type = 'text',
          se = list(sqrt(diag(vcovHC(jac_lib_mod_2))),
                    sqrt(diag(vcovHC(jac_con_mod_2)))),
          header=F,
          report=('v*c*sp'),
          add.lines=list(c("Ideology", "Liberal", "Conservative")))
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               ubi
##                               (1)          (2)
## -----
## storybullying                -0.228          -0.522
##                               (1.357)
##                               p = 0.701
##
## storybullying-addressed      -1.111          1.615
##                               (1.549)
##                               p = 0.298
##
## ubi_familiarity_dummy        -2.000          1.492
##                               (2.321)
##                               p = 0.521
##
## storybullying:ubi_familiarity_dummy -3.105          1.222
##                               (3.152)
##                               p = 0.699
##
## storybullying-addressed:ubi_familiarity_dummy 0.778          -2.615
##                               (3.162)
##                               p = 0.409
##
## Constant                    8.333          3.308***
##                               (0.982)
##                               p = 0.001
## -----
## Ideology                    Liberal          Conservative
## Observations                60              54
## R2                          0.140           0.071
## Adjusted R2                 0.061           -0.025
## Residual Std. Error        2.259 (df = 54)  3.676 (df = 48)
## F Statistic                 1.765 (df = 5; 54) 0.738 (df = 5; 48)
## =====
## Note:                        *p<0.1; **p<0.05; ***p<0.01
```

```
# Exploratory
```

```
modell1_liberal = my_lm_calcs(lm_in = lm(ubi_number ~ arm, data = results_clean %>% filter(ideology_bin == "Liberal")),
                             data = results_clean %>% filter(ideology_bin == "Liberal"))
modell1_conservative = my_lm_calcs(lm_in = lm(ubi_number ~ arm, data = results_clean %>% filter(ideology_bin == "Conservative")),
                                   data = results_clean %>% filter(ideology_bin == "Conservative"))
```



```

stargazer(model1_liberal$lm, model1_conservative$lm
, type = stargazer_type, header = F
, se = list(model1_liberal$se_robust, model1_conservative$se_robust)
, title = "Moral Foundations Prelim Regression Specifications"
, column.labels = c("Liberals", "Conservative")
, notes = "HC Robust Standard Errors"
, report = ('v*c*sp')
)

```

Table 1: Moral Foundations Prelim Regression Specifications

	<i>Dependent variable:</i>	
	ubi_number	
	Liberals	Conservative
	(1)	(2)
armfairness_base	-0.198 (0.794) p = 0.804	-0.333 (1.208) p = 0.783
armfairness_extension	-0.890 (0.691) p = 0.198	0.889 (1.289) p = 0.491
armpurity_base	0.452 (0.600) p = 0.451	0.330 (1.282) p = 0.797
armpurity_extension	0.095 (0.662) p = 0.886	0.444 (1.138) p = 0.697
Constant	8.048*** (0.497) p = 0.000	3.722*** (0.868) p = 0.00002
Observations	99	91
R ²	0.042	0.013
Adjusted R ²	0.001	-0.032
Residual Std. Error	2.117 (df = 94)	3.624 (df = 86)
F Statistic	1.034 (df = 4; 94)	0.293 (df = 4; 86)

Note: *p<0.1; **p<0.05; ***p<0.01
HC Robust Standard Errors

```

model_1 = lm(ubi_number ~ arm+ideology_bin, data = results_full)

stargazer(model_1, type = stargazer_type)

```

% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu
% Date and time: Sun, Nov 24, 2019 - 1:17:02 PM

Table 2:

	<i>Dependent variable:</i>
	ubi_number
armfairness_base	-0.296 (0.662)
armfairness_extension	-0.253 (0.652)
armpurity_base	0.353 (0.656)
armpurity_extension	0.366 (0.648)
ideology_binliberal	3.954*** (0.426)
ideology_binmoderate	2.973*** (1.031)
Constant	3.951*** (0.514)
Observations	199
R ²	0.316
Adjusted R ²	0.295
Residual Std. Error	2.932 (df = 192)
F Statistic	14.805*** (df = 6; 192)
<i>Note:</i>	*p<0.1; **p<0.05; ***p<0.01

```

modell1_liberal_gender = my_lm_calcs(lm_in = lm(ubi_number ~ arm*gender, data = results_clean %>% filter(
modell1_conservative_gender = my_lm_calcs(lm_in = lm(ubi_number ~ arm*gender, data = results_clean %>% filter(

stargazer(modell1_liberal_gender$lm, modell1_conservative_gender$lm
, type = stargazer_type, header = F
, se = list(modell1_liberal_gender$se_robust, modell1_conservative_gender$se_robust)
, title = "Moral Foundations Prelim Regression Specifications"
, column.labels = c("Liberals", "Conservative")
, notes = "HC Robust Standard Errors"
, report = ('v*c*sp')
)

```

[[Example Table]]

Model	Specification	Interpretation	Figure
Model 1	$crmrte \sim prbarr$	$\Delta crmrte = \beta_1 \Delta prbarr$??

Stargazer

Table 3: Moral Foundations Prelim Regression Specifications

	<i>Dependent variable:</i>	
	ubi_number	
	Liberals	Conservative
	(1)	(2)
armfairness_base	0.500 (1.078) p = 0.643	1.357 (1.383) p = 0.327
armfairness_extension	-1.389 (1.053) p = 0.188	4.250*** (1.558) p = 0.007
armpurity_base	1.929** (0.757) p = 0.011	1.214 (1.020) p = 0.234
armpurity_extension	1.167 (0.780) p = 0.135	2.667** (1.088) p = 0.015
genderMale	1.643** (0.813) p = 0.044	4.143*** (1.038) p = 0.0001
armfairness_base:genderMale	-1.943 (1.538) p = 0.207	-1.636 (2.001) p = 0.414
armfairness_extension:genderMale	0.346 (1.218) p = 0.777	-4.393* (2.286) p = 0.055
armpurity_base:genderMale	-3.162*** (0.976) p = 0.002	-0.440 (1.902) p = 0.817
armpurity_extension:genderMale	-2.865** (1.271) p = 0.025	-2.643 (1.774) p = 0.137
Constant	7.500*** (0.686) p = 0.000	0.500 (0.333) p = 0.134
Observations	99	91
R ²	0.154	0.145
Adjusted R ²	0.069	0.050
Residual Std. Error	2.044 (df = 89)	3.477 (df = 81)
F Statistic	1.806* (df = 9; 89)	1.521 (df = 9; 81)

*Note:**p<0.1; **p<0.05; ***p<0.01
HC Robust Standard Errors

4 Results

[[TBD]]

5 Conclusion

[[TBD]]

6 Discussion

[[TBD]]

6.1 Limitations

[[TBD]]

7 Technical Appendix

7.1 Data Dictionary

Variable Name	Variable	Values	Source	Notes
county	County ID	Odd Integers		Few missing

7.2 Exploratory Data Analysis

Additional steps taken not included in the body of the report

[[TBD]]