

Morality Analysis

Load the data

```
## Load the data
library(readxl)

file_path <- "/Users/daryani/Desktop/data.xlsx"
data <- read_excel(file_path)

Sys.setenv(RGL_USE_NULL = "TRUE")
options(rgl.useNULL = TRUE)
# Packages
library(dplyr)
```

Attaching package: 'dplyr'

The following objects are masked from 'package:stats':

filter, lag

The following objects are masked from 'package:base':

intersect, setdiff, setequal, union

```
library(car)          # Levene tests, Type-III ANCOVA tables
```

Loading required package: carData

```
Attaching package: 'car'
```

```
The following object is masked from 'package:dplyr':
```

```
recode
```

```
library(heplots)      # Box's M, multivariate SSPE/HE utilities
```

```
Loading required package: broom
```

```
library(MVN)          # Multivariate normality + Mahalanobis outliers
library(emmeans)       # Estimated marginal means + contrasts
```

```
Welcome to emmeans.
```

```
Caution: You lose important information if you filter this package's results.  
See '? untidy'
```

```
library(effectsize)  # Effect sizes (partial eta^2, CI)
```

Ensure factors are factors

```
# Robust recode to canonical labels
data <- data %>%
  mutate(
    group_clean = case_when(
      group %in% c(1, "1", "Prolife", "prolife") ~ "Prolife",
      group %in% c(2, "2", "Prochoice", "prochoice") ~ "Prochoice",
      TRUE ~ NA_character_
    ),
    condition_clean = case_when(
      condition %in% c(1, "1", "Outgroup", "outgroup", "Actual", "Actual Perception") ~ "Outgroup",
      condition %in% c(2, "2", "Metaperception", "metaperception", "Meta") ~ "Metaperception",
      condition %in% c(3, "3", "Ingroup", "ingroup") ~ "Ingroup",
      TRUE ~ NA_character_
    ),
    group      = factor(group_clean, levels = c("Prolife", "Prochoice")),
    condition = factor(condition_clean, levels = c("Outgroup", "Metaperception", "Ingroup"))
  )
```

```

) %>%
select(-group_clean, -condition_clean)

# Sanity check: you should now see real levels, not NA
print(with(data, table(group, condition, useNA = "ifany")))

```

		condition		
group		Outgroup	Metaperception	Ingroup
Prolife		89	85	89
Prochoice		96	97	96

Create Composite Scores

```

item_cols <- c(
  "care_1", "care_2", "equality_1", "equality_2", "proportionality_1", "proportionality_2",
  "loyalty_1", "loyalty_2", "authority_1", "authority_2", "purity_1", "purity_2"
)
missing <- setdiff(item_cols, names(data))
if (length(missing)) stop("Missing columns: ", paste(missing, collapse = ", "))

data <- data %>%
  mutate(across(all_of(item_cols), ~ suppressWarnings(as.numeric(.)))) %>%
  mutate(
    care_mean      = rowMeans(pick(care_1, care_2), na.rm = TRUE),
    equality_mean = rowMeans(pick(equality_1, equality_2), na.rm = TRUE),
    proportionality_mean = rowMeans(pick(proportionality_1, proportionality_2), na.rm = TRUE),
    individualizing = rowMeans(cbind(care_mean, equality_mean), na.rm = TRUE),
    loyalty_mean   = rowMeans(pick(loyalty_1, loyalty_2), na.rm = TRUE),
    authority_mean = rowMeans(pick(authority_1, authority_2), na.rm = TRUE),
    purity_mean    = rowMeans(pick(purity_1, purity_2), na.rm = TRUE),
    binding        = rowMeans(cbind(loyalty_mean, authority_mean, purity_mean, proportionality_mean))
  )

```

Simple Descriptives

```

desc <- data %>%
  group_by(group, condition) %>%
  summarise(

```

```

n          = dplyr::n(),
mean_ind  = mean(individualizing, na.rm = TRUE),
sd_ind    = sd(individualizing, na.rm = TRUE),
mean_bind = mean(binding, na.rm = TRUE),
sd_bind   = sd(binding, na.rm = TRUE),
.groups   = "drop"
)

# Print PDF-friendly
print(as.data.frame(desc), row.names = FALSE)

      group      condition  n mean_ind    sd_ind mean_bind    sd_bind
Prolife     Outgroup 89 3.078652 0.9650096 2.029494 1.0121812
Prolife Metaperception 85 3.064706 0.7371447 3.573529 0.8654440
Prolife     Ingroun 89 3.508427 0.7644902 3.556180 0.8209809
Prochoice   Outgroup 96 2.119792 0.7880515 3.334635 0.9404112
Prochoice Metaperception 97 3.113402 1.0819426 1.472938 0.2773026
Prochoice   Ingroun 96 3.841146 0.8112043 2.098958 0.8269548

# Quick wide views
ind_wide <- tidyr::pivot_wider(desc |> select(group, condition, mean_ind),
                                names_from = condition, values_from = mean_ind)
bind_wide <- tidyr::pivot_wider(desc |> select(group, condition, mean_bind),
                                names_from = condition, values_from = mean_bind)
cat("\nIndividualizing (means):\n"); print(as.data.frame(ind_wide), row.names = FALSE)

```

Individualizing (means):

	group	Outgroup	Metaperception	Ingroun
Prolife	3.078652	3.064706	3.508427	
Prochoice	2.119792	3.113402	3.841146	

```
cat("\nBinding (means):\n");           print(as.data.frame(bind_wide), row.names = FALSE)
```

Binding (means):

	group	Outgroup	Metaperception	Ingroun
Prolife	2.029494	3.573529	3.556180	
Prochoice	3.334635	1.472938	2.098958	

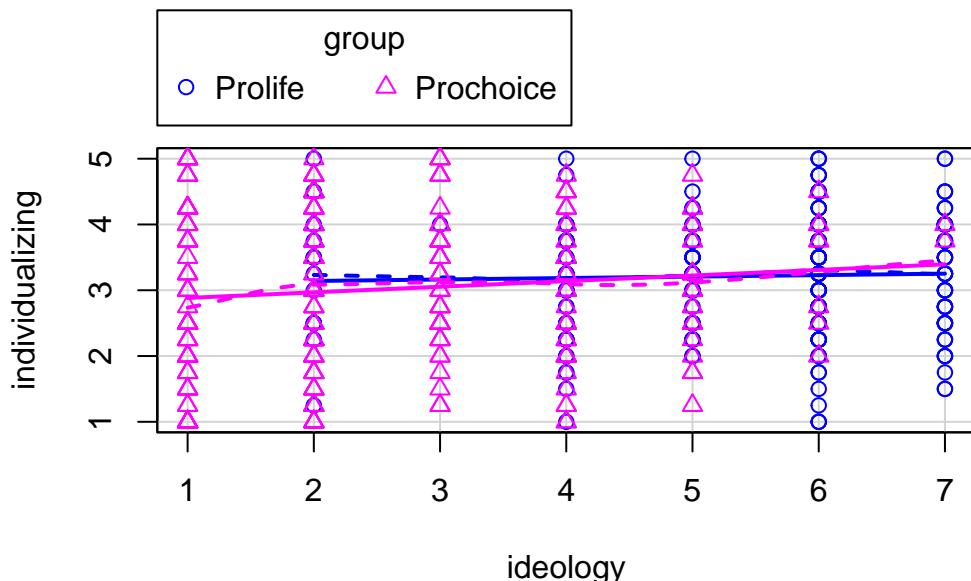
Equality of covariance matrices

```
# Note: Box's M is sensitive to non-normality; interpret alongside Pillai later  
boxM(cbind(individualizing, binding) ~ group * condition, data = data)
```

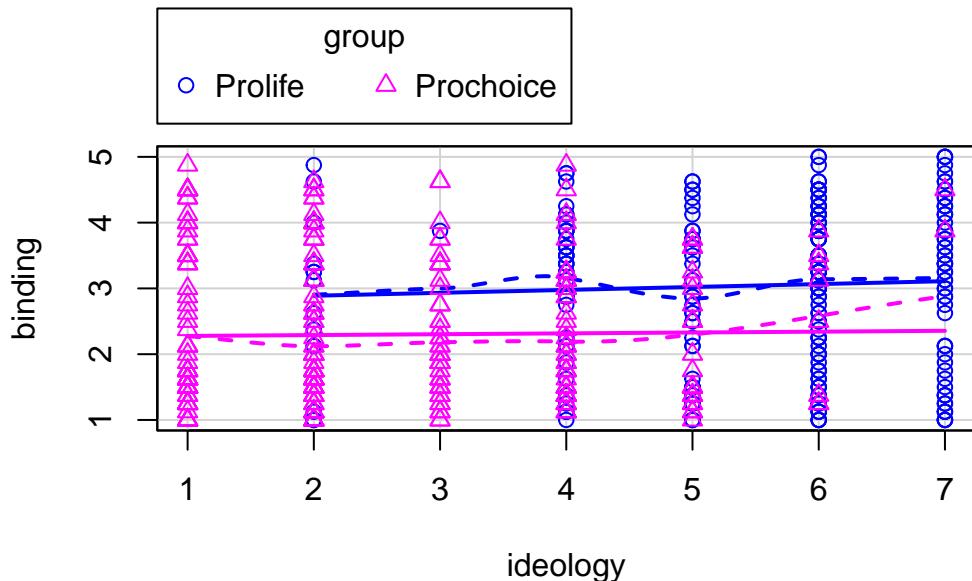
```
Box's M-test for Homogeneity of Covariance Matrices  
  
data: Y  
Chi-Sq (approx.) = 175.02, df = 15, p-value < 2.2e-16
```

Linearity of DV–covariate relations (quick visual check)

```
# Scatterplots of DV vs ideology, stratified by factors (optional visual check)  
car:::scatterplot(individualizing ~ ideology | group*condition, data = data)
```



```
car:::scatterplot(binding ~ ideology | group*condition, data = data)
```



Homogeneity of regression slopes (covariate × factors)

```
# For each DV, check whether ideology interacts with group/condition
m_ind <- lm(individualizing ~ ideology * group * condition, data = data)
m_bin <- lm(binding ~ ideology * group * condition, data = data)

anova(m_ind) # Look at ideology:group and ideology:condition (and 3-way)
```

Analysis of Variance Table

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
ideology	1	8.64	8.642	11.7160	0.0006666	***
group	1	0.05	0.053	0.0720	0.7885106	
condition	2	110.72	55.360	75.0555	< 2.2e-16	***
ideology:group	1	1.77	1.773	2.4031	0.1216785	
ideology:condition	2	34.94	17.471	23.6868	1.379e-10	***
group:condition	2	11.54	5.772	7.8261	0.0004462	***
ideology:group:condition	2	6.75	3.376	4.5774	0.0106839	*
Residuals	540	398.30	0.738			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1						

```
anova(m_bin)
```

Analysis of Variance Table

Response: binding

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
ideology	1	47.78	47.784	71.1514	3.039e-16	***
group	1	29.55	29.546	43.9947	8.010e-11	***
condition	2	10.84	5.421	8.0717	0.0003515	***
ideology:group	1	0.26	0.264	0.3938	0.5305967	
ideology:condition	2	183.42	91.710	136.5567	< 2.2e-16	***
group:condition	2	122.52	61.258	91.2142	< 2.2e-16	***
ideology:group:condition	2	0.10	0.048	0.0708	0.9316518	
Residuals	540	362.66	0.672			

Signif. codes:	0	'***'	0.001	'**'	0.01	'*'
	0.05	'. '	0.1	' '	1	

Fit the MANCOVA (report both Pillai and Wilks)

```
# Multivariate model
mancova_model <- manova(cbind(individualizing, binding) ~ group * condition + ideology,
                           data = data)

summary(mancova_model, test = "Pillai") # robust
```

	Df	Pillai	approx F	num Df	den Df	Pr(>F)	
group	1	0.17239	56.658	2	544	<2e-16	***
condition	2	0.24342	37.762	4	1090	<2e-16	***
ideology	1	0.00579	1.585	2	544	0.2059	
group:condition	2	0.52545	97.104	4	1090	<2e-16	***
Residuals	545						

Signif. codes:	0	'***'	0.001	'**'	0.01	'*'	0.05 '. '
	0.1	' '	1				

```
summary(mancova_model, test = "Wilks") # conventional
```

	Df	Wilks	approx F	num Df	den Df	Pr(>F)	
group	1	0.82761	56.658	2	544	<2e-16	***

```

condition          2  0.76243   39.508      4    1088 <2e-16 ***
ideology          1  0.99421    1.585      2     544  0.2059
group:condition  2  0.48066   120.329      4    1088 <2e-16 ***
Residuals        545

---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Univariate ANCOVAs with Type-III tests + effect sizes

```

# Fit per-DV models with Type-III SS
fit_ind <- lm(individualizing ~ group * condition + ideology, data = data)
fit_bin <- lm(binding           ~ group * condition + ideology, data = data)

Anova(fit_ind, type = 3)

```

Anova Table (Type III tests)

```

Response: individualizing
            Sum Sq Df F value Pr(>F)
(Intercept) 197.85  1 262.8672 < 2.2e-16 ***
group        23.12  1 30.7135 4.663e-08 ***
condition    10.77  2  7.1523 0.0008588 ***
ideology      2.72  1  3.6130 0.0578573 .
group:condition 43.11  2 28.6398 1.490e-12 ***
Residuals    410.20 545
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
Anova(fit_bin, type = 3)
```

Anova Table (Type III tests)

```

Response: binding
            Sum Sq Df F value Pr(>F)
(Intercept) 99.60   1 147.2381 <2e-16 ***
group       60.79   1  89.8664 <2e-16 ***
condition  138.34   2 102.2486 <2e-16 ***
ideology      0.06   1   0.0931 0.7604
group:condition 300.27   2 221.9296 <2e-16 ***
Residuals   368.69 545

```

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Partial eta-squared + 95% CI for key effects
eta_squared(Anova(fit_ind, type = 3), partial = TRUE, ci = 0.95)
```

Type 3 ANOVAs only give sensible and informative results when covariates are mean-centered and factors are coded with orthogonal contrasts (such as those produced by `contr.sum`, `contr.poly`, or `contr.helmert`, but **not** by the default `contr.treatment`).

```
# Effect Size for ANOVA (Type III)
```

Parameter	Eta2 (partial)	95% CI
group	0.05	[0.03, 1.00]
condition	0.03	[0.01, 1.00]
ideology	6.59e-03	[0.00, 1.00]
group:condition	0.10	[0.06, 1.00]

- One-sided CIs: upper bound fixed at [1.00].

```
eta_squared(Anova(fit_bin, type = 3), partial = TRUE, ci = 0.95)
```

Type 3 ANOVAs only give sensible and informative results when covariates are mean-centered and factors are coded with orthogonal contrasts (such as those produced by `contr.sum`, `contr.poly`, or `contr.helmert`, but **not** by the default `contr.treatment`).

```
# Effect Size for ANOVA (Type III)
```

Parameter	Eta2 (partial)	95% CI
group	0.14	[0.10, 1.00]
condition	0.27	[0.22, 1.00]
ideology	1.71e-04	[0.00, 1.00]
group:condition	0.45	[0.40, 1.00]

- One-sided CIs: upper bound fixed at [1.00].

Planned Contrasts for Ingroup Perception and Outgroup Perception

```
# EMMs for each DV, adjusted for ideology
emm_ind <- emmeans(fit_ind, ~ group * condition)
emm_bin <- emmeans(fit_bin, ~ group * condition)

# Inspect order (for sanity)
emm_ind
```

group	condition	emmean	SE	df	lower.CL	upper.CL
Prolife	Outgroup	3.01	0.0993	545	2.81	3.20
Prochoice	Outgroup	2.19	0.0958	545	2.00	2.38
Prolife	Metaperception	2.99	0.1020	545	2.79	3.19
Prochoice	Metaperception	3.19	0.0961	545	3.00	3.38
Prolife	Ingroup	3.43	0.1020	545	3.23	3.63
Prochoice	Ingroup	3.91	0.0954	545	3.72	4.10

Confidence level used: 0.95

```
emm_bin
```

group	condition	emmean	SE	df	lower.CL	upper.CL
Prolife	Outgroup	2.02	0.0942	545	1.83	2.20
Prochoice	Outgroup	3.35	0.0908	545	3.17	3.52
Prolife	Metaperception	3.56	0.0970	545	3.37	3.75
Prochoice	Metaperception	1.48	0.0911	545	1.31	1.66
Prolife	Ingroup	3.54	0.0965	545	3.35	3.73
Prochoice	Ingroup	2.11	0.0904	545	1.93	2.29

Confidence level used: 0.95

```
# EMMs (already fit)
emm_ind <- emmeans(fit_ind, ~ group * condition)

# Get the row labels in emm_ind's order (e.g., "Prolife.Ingroup", etc.)
labs <- with(as.data.frame(emm_ind),
            paste(group, condition, sep = "."))

# Helper to create a full-length contrast weight vector safely
make_w <- function(pos_names, neg_names, labels = labs) {
```

```

w <- setNames(rep(0, length(labels)), labels)
w[pos_names] <- w[pos_names] + 1
w[neg_names] <- w[neg_names] - 1
w
}

# H1: Ingroup (one group) vs Outgroup (other group)
w1 <- make_w("Prolife.Ingroup", "Prochoice.Outgroup")
w2 <- make_w("Prochoice.Ingroup", "Prolife.Outgroup")

h1_ind <- contrast(
  emm_ind,
  method = list(
    "Prolife Ingroup vs Prochoice Outgroup" = w1,
    "Prochoice Ingroup vs Prolife Outgroup" = w2
  ),
  adjust = "none"
)

summary(h1_ind, infer = c(TRUE, TRUE))

```

contrast	estimate	SE	df	lower.CL	upper.CL
Prolife Ingroup vs Prochoice Outgroup	1.236	0.151	545	0.940	1.53
Prochoice Ingroup vs Prolife Outgroup	0.901	0.147	545	0.612	1.19
t.ratio p.value					
8.202 <.0001					
6.129 <.0001					

Confidence level used: 0.95

```
eff_size(h1_ind, sigma = sigma(fit_ind), edf = df.residual(fit_ind))
```

```

contrast
Prolife Ingroup vs Prochoice Outgroup - Prochoice Ingroup vs Prolife Outgroup
effect.size   SE  df lower.CL upper.CL
      0.386 0.273 545     -0.15     0.923

sigma used for effect sizes: 0.8676
Confidence level used: 0.95

```

```

emm_bin <- emmeans(fit_bin, ~ group * condition)
labs_b <- with(as.data.frame(emm_bin), paste(group, condition, sep = "."))
make_wb <- function(pos_names, neg_names, labels = labs_b) {
  w <- setNames(rep(0, length(labels)), labels)
  w[pos_names] <- w[pos_names] + 1
  w[neg_names] <- w[neg_names] - 1
  w
}

h1_bin <- contrast(
  emm_bin,
  method = list(
    "Prolife Ingroup vs Prochoice Outgroup" = make_wb("Prolife.Ingroup", "Prochoice.Outgroup"),
    "Prochoice Ingroup vs Prolife Outgroup" = make_wb("Prochoice.Ingroup", "Prolife.Outgroup")
  ),
  adjust = "none"
)
summary(h1_bin, infer = c(TRUE, TRUE))

```

contrast	estimate	SE	df	lower.CL	upper.CL
Prolife Ingroup vs Prochoice Outgroup	0.1984	0.143	545	-0.0823	0.479
Prochoice Ingroup vs Prolife Outgroup	0.0906	0.139	545	-0.1833	0.364
t.ratio p.value					
1.388 0.1657					
0.650 0.5162					

Confidence level used: 0.95

```
eff_size(h1_bin, sigma = sigma(fit_bin), edf = df.residual(fit_bin))
```

contrast	estimate	SE	df	lower.CL	upper.CL
Prolife Ingroup vs Prochoice Outgroup - Prochoice Ingroup vs Prolife Outgroup	0.131	0.273	545	-0.405	0.667
effect.size					
sigma used for effect sizes: 0.8225					
Confidence level used: 0.95					

Planned contrast for Outgroup Perception and Metaperception

```
# EMMs for each DV, adjusted for ideology
emm_ind <- emmeans(fit_ind, ~ group * condition)
emm_bin <- emmeans(fit_bin, ~ group * condition)

# Inspect order (sanity check)
emm_ind
```

group	condition	emmean	SE	df	lower.CL	upper.CL
Prolife	Outgroup	3.01	0.0993	545	2.81	3.20
Prochoice	Outgroup	2.19	0.0958	545	2.00	2.38
Prolife	Metaperception	2.99	0.1020	545	2.79	3.19
Prochoice	Metaperception	3.19	0.0961	545	3.00	3.38
Prolife	Ingroup	3.43	0.1020	545	3.23	3.63
Prochoice	Ingroup	3.91	0.0954	545	3.72	4.10

Confidence level used: 0.95

```
emm_bin
```

group	condition	emmean	SE	df	lower.CL	upper.CL
Prolife	Outgroup	2.02	0.0942	545	1.83	2.20
Prochoice	Outgroup	3.35	0.0908	545	3.17	3.52
Prolife	Metaperception	3.56	0.0970	545	3.37	3.75
Prochoice	Metaperception	1.48	0.0911	545	1.31	1.66
Prolife	Ingroup	3.54	0.0965	545	3.35	3.73
Prochoice	Ingroup	2.11	0.0904	545	1.93	2.29

Confidence level used: 0.95

```
# ----- Helper (same as before) -----
labs <- with(as.data.frame(emm_ind), paste(group, condition, sep = "."))
make_w <- function(pos_names, neg_names, labels = labs) {
  w <- setNames(rep(0, length(labels)), labels)
  w[pos_names] <- w[pos_names] + 1
  w[neg_names] <- w[neg_names] - 1
  w
}
```

```

# =====
# H2: Outgroup (one group) vs Metaperception (other group)
# =====
# Individualizing
w3_ind <- make_w("Prolife.Outgroup", "Prochoice.Metaperception")
w4_ind <- make_w("Prochoice.Outgroup", "Prolife.Metaperception")

h2_ind <- contrast(
  emm_ind,
  method = list(
    "Prolife Outgroup vs Prochoice Metaperception" = w3_ind,
    "Prochoice Outgroup vs Prolife Metaperception" = w4_ind
  ),
  adjust = "none"
)

summary(h2_ind, infer = c(TRUE, TRUE)) # estimates, SEs, CIs, p

```

contrast	estimate	SE	df	lower.CL	upper.CL
Prolife Outgroup vs Prochoice Metaperception	-0.179	0.148	545	-0.47	0.112
Prochoice Outgroup vs Prolife Metaperception	-0.799	0.150	545	-1.09	-0.504
t.ratio					
	-1.208	0.2276			
p.value					<.0001

Confidence level used: 0.95

```
eff_size(h2_ind, sigma = sigma(fit_ind), edf = df.residual(fit_ind)) # standardized effects
```

contrast	effect.size	SE	df	lower.CL	upper.CL
Prolife Outgroup vs Prochoice Metaperception - Prochoice Outgroup vs Prolife Metaperception	0.714	0.274	545	0.176	1.25

sigma used for effect sizes: 0.8676

Confidence level used: 0.95

```
# ----- Binding -----
labs_b <- with(as.data.frame(emm_bin), paste(group, condition, sep = "."))
make_wb <- function(pos_names, neg_names, labels = labs_b) {
```

```

w <- setNames(rep(0, length(labels)), labels)
w[pos_names] <- w[pos_names] + 1
w[neg_names] <- w[neg_names] - 1
w
}

w3_bin <- make_wb("Prolife.Outgroup", "Prochoice.Metaperception")
w4_bin <- make_wb("Prochoice.Outgroup", "Prolife.Metaperception")

h2_bin <- contrast(
  emm_bin,
  method = list(
    "Prolife Outgroup vs Prochoice Metaperception" = w3_bin,
    "Prochoice Outgroup vs Prolife Metaperception" = w4_bin
  ),
  adjust = "none"
)

summary(h2_bin, infer = c(TRUE, TRUE))

```

contrast	estimate	SE	df	lower.CL	upper.CL
Prolife Outgroup vs Prochoice Metaperception	0.535	0.141	545	0.258	0.8107
Prochoice Outgroup vs Prolife Metaperception	-0.217	0.143	545	-0.497	3.803
					0.0002
					0.0633
					-1.520
					0.1291

Confidence level used: 0.95

```
eff_size(h2_bin, sigma = sigma(fit_bin), edf = df.residual(fit_bin))
```

contrast	estimate	SE	df	lower.CL	upper.CL
Prolife Outgroup vs Prochoice Metaperception - Prochoice Outgroup vs Prolife Metaperception	0.913	0.274	545	0.376	1.45

sigma used for effect sizes: 0.8225

Confidence level used: 0.95

Graph

```
# =====
# Ideology-adjusted scores + violin plots
# =====
library(dplyr)
library(tidyr)
library(ggplot2)
library(forcats)

# 0) Ensure ideology is numeric
if (!is.numeric(data$ideology)) {
  data <- data %>% mutate(ideology = suppressWarnings(as.numeric(ideology)))
}

# 1) Fit ANCOVA-style models to get a single slope for ideology (no interactions)
#     This mirrors your main analysis specification.
fit_ind <- lm(individualizing ~ ideology + group * condition, data = data)
fit_bind <- lm(binding ~ ideology + group * condition, data = data)

b_ind <- coef(fit_ind)[["ideology"]]
b_bind <- coef(fit_bind)[["ideology"]]
xbar <- mean(data$ideology, na.rm = TRUE)

# 2) Compute ideology-adjusted scores (unit-preserving adjustment to the covariate mean)
#     y_adj = y - b_hat * (x - xbar)
data <- data %>%
  mutate(
    individualizing_adj = individualizing - b_ind * (ideology - xbar),
    binding_adj = binding - b_bind * (ideology - xbar)
  )

# 3) Tidy to long format for plotting
plot_df_adj <- data %>%
  filter(!is.na(group), !is.na(condition)) %>%
  mutate(
    # Pretty labels (optional)
    group_label = fct_recode(group, "Pro-life" = "Prolife", "Pro-choice" = "Prochoice"),
    condition_label = fct_recode(
      condition,
      "Outgroup Perception" = "Outgroup",
      "Metaperception" = "Metaperception",
```

```

    "Ingroup Perception" = "Ingroup"
),
condition_label = factor(condition_label,
                        levels = c("Ingroup Perception", "Outgroup Perception", "Metaperception"),
) %>%
select(group_label, condition_label, individualizing_adj, binding_adj) %>%
pivot_longer(
  cols = c(individualizing_adj, binding_adj),
  names_to = "moral.foundation",
  values_to = "value.adj"
) %>%
mutate(
  foundation_label = if_else(
    moral.foundation == "individualizing_adj",
    "Individualizing Foundations",
    "Binding Foundations"
  )
)

# 4) Summary stats (means and SEs) on adjusted scores
sum_df_adj <- plot_df_adj %>%
  group_by(group_label, condition_label, foundation_label) %>%
  summarise(
    mean = mean(value.adj, na.rm = TRUE),
    sd = sd(value.adj, na.rm = TRUE),
    n = sum(!is.na(value.adj)),
    se = sd / sqrt(n),
    .groups = "drop"
  )

# Optional: inspect adjusted cell means
print(sum_df_adj %>% arrange(foundation_label, group_label, condition_label))

# A tibble: 12 x 7
  group_label condition_label foundation_label     mean      sd      n      se
  <fct>       <fct>           <chr>          <dbl> <dbl> <int> <dbl>
1 Pro-life    Ingroup Perception Binding Foundations 3.54  0.819   89  0.0869
2 Pro-life    Outgroup Perception Binding Foundations 2.02  1.01    89  0.107 
3 Pro-life    Metaperception    Binding Foundations  3.56  0.865   85  0.0939
4 Pro-choice   Ingroup Perception Binding Foundations 2.11  0.825   96  0.0842
5 Pro-choice   Outgroup Perception Binding Foundations 3.35  0.942   96  0.0961
6 Pro-choice   Metaperception   Binding Foundations  1.48  0.278   97  0.0282

```

7	Pro-life	Ingroup Perception	Individualizing Fou~	3.43	0.761	89	0.0807
8	Pro-life	Outgroup Perception	Individualizing Fou~	3.01	0.968	89	0.103
9	Pro-life	Metaperception	Individualizing Fou~	2.99	0.742	85	0.0805
10	Pro-choice	Ingroup Perception	Individualizing Fou~	3.91	0.827	96	0.0844
11	Pro-choice	Outgroup Perception	Individualizing Fou~	2.19	0.762	96	0.0778
12	Pro-choice	Metaperception	Individualizing Fou~	3.19	1.07	97	0.109

```

# 5) Plot: violins of ideology-adjusted scores
ggplot(plot_df_adj, aes(x = group_label, y = value_adj, fill = condition_label)) +
  geom_violin(position = position_dodge(width = 0.8), alpha = 0.7, width = 0.7, trim = TRUE)

  # mean points (adjusted)
  geom_point(
    data = sum_df_adj,
    aes(x = group_label, y = mean, group = condition_label),
    position = position_dodge(width = 0.8),
    size = 2, color = "black", inherit.aes = FALSE
  ) +

  # mean ± SE error bars (adjusted)
  geom_errorbar(
    data = sum_df_adj,
    aes(x = group_label, ymin = mean - se, ymax = mean + se, group = condition_label),
    width = 0.15,
    position = position_dodge(width = 0.8),
    inherit.aes = FALSE
  ) +

  facet_wrap(~ foundation_label, ncol = 2) +
  scale_fill_manual(
    values = c(
      "Ingroup Perception" = "#66c2a5",
      "Outgroup Perception" = "#8da0cb",
      "Metaperception" = "#fc8d62"
    ),
    name = "Condition"
  ) +
  labs(
    title = "Moral Perceptions Across Pro-life and Pro-choice Groups",
    subtitle = "Violins of morality scores with means and mean ± SE",
    x = NULL, y = "Rating (1-5 scale)"
  )
  
```

```

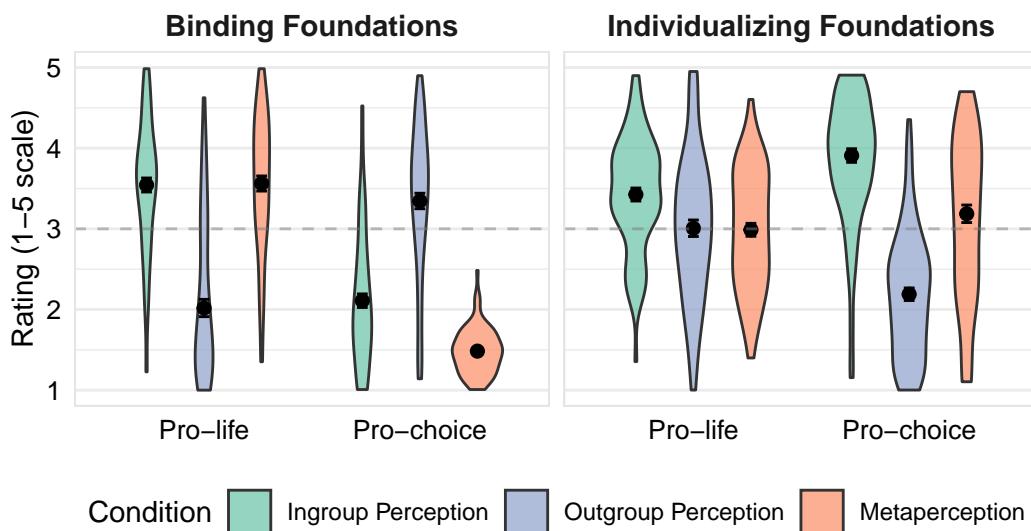
theme_minimal() +
theme(
  legend.position = "bottom",
  plot.title = element_text(hjust = 0.5, face = "bold", size = 14),
  plot.subtitle = element_text(hjust = 0.5, size = 12),
  strip.text = element_text(face = "bold", size = 11),
  panel.grid.major.x = element_blank(),
  panel.border = element_rect(fill = NA, color = "gray90"),
  axis.text = element_text(color = "black", size = 10)
) +
geom_hline(yintercept = 3, linetype = "dashed", color = "gray50", alpha = 0.5) +
scale_y_continuous(limits = c(1, 5), breaks = 1:5)

```

Warning: Removed 26 rows containing non-finite outside the scale range
(`stat_ydensity()`).

Moral Perceptions Across Pro-life and Pro-choice Group

Violins of morality scores with means and mean \pm SE



```

# 6) Save
ggsave(
  filename = "/Users/daryani/Desktop/moral_perception_plot_adjusted.jpg",
  width = 8, height = 6, dpi = 300
)

```

Warning: Removed 26 rows containing non-finite outside the scale range
(`stat_ydensity()`).