Absolute treatment effect by Pointwise meta-analysis - severe stunting

Pearl Ante-Testard

2024-12-16

```
source(here::here("R", "0-config.R"))
##
##
                     0
                          1
##
     DOSE
                  811
                        126
##
     DYADG
                  1024
                         13
##
     DYADM
                  589
                         67
##
     GHANA
                  158
                          1
##
     HAITI
                  287
                         10
##
     JiVitA
                  3651
                        483
##
     MAHAY
                 2588
                        796
##
                 1695
     PROMISBF
                         82
##
     PROMISBF_CS 828
##
     PROMISM
                  926
                         86
##
     PROMISM CS 1726
                       163
##
     RDNS
                 2269 209
##
     SHINE_HIV-
                 3372 302
##
                 4069 492
     WASHB
##
     WASHK
                 6013 581
                 2383 224
##
     ZINC
```

Note: DYADG and DYADM have empty rows for arm_primary – why? These are the studies that had maternal interventions. Removing them from the analysis. Also removed HAITI and GHANA because of very small sample sizes for severe stunting

```
# Remove studies "DYADG", "DYADM", "HAITI", and "GHANA"

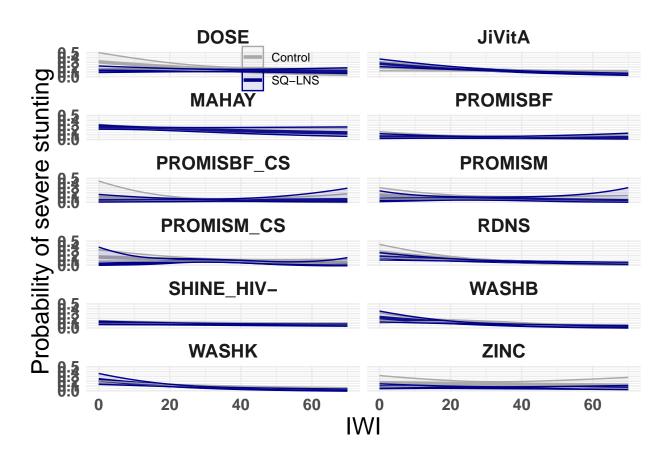
df_analysis_severe_stunted <- subset(df_analysis_severe_stunted, !(study %in% c("DYADG", "DYADM", "HAIT
```

Fit a cubic splines model per study

```
#SS.Comb <- df_analysis %>%
# split(.$study) %>%
# map_dfr(function(df_analysis) {
# study_name <- unique(df_analysis$study)
# if (study_name %in% c(study_name)) {
# model <- gam(severe_stunted ~ arm_primary + iwi * arm_primary + s(iwi, bs = "cr",</pre>
```

```
by = arm_primary),
  #
                   data = df_analysis,
                   family = "qaussian", method = "REML")
   #
   # }
   ## Return a data frame with study_name and model
   # data.frame(study = study_name, model = I(list(model)))
  #})
SS.Comb_severe_stunted <- df_analysis_severe_stunted %>%
  mutate(study = factor(study)) %>%
  split(.$study) %>%
  map_dfr(function(df_analysis_severe_stunted) {
    study_name <- unique(df_analysis_severe_stunted$study)</pre>
    if (study_name %in% c("DOSE")) {
      model <- gam(severe_stunted ~ arms_maternal + iwi * arms_maternal + s(iwi, bs = "cr",
                                           by = arms_maternal),
                  data = df_analysis_severe_stunted,
                  family = binomial("logit"), method = "REML")
   } else {
      model <- gam(severe_stunted ~ arms_maternal + iwi * arms_maternal + s(iwi, bs = "cr",
                                           by = arms_maternal) +
                                          s(cluster, bs = "re", by = dummy),
                  data = df_analysis_severe_stunted,
                  family = binomial("logit"), method = "REML")
   }
    # Return a data frame with study_name and model
    data.frame(study = study_name, model = I(list(model)))
 })
new.data_severe_stunted <- df_analysis_severe_stunted %>%
  select(study, cluster, arms_maternal) %>%
  mutate(dummy = 0) %>%
  distinct() %>%
  expand_grid(iwi = seq(0, 70, by = 1))
predictions.SS.Comb_severe_stunted <- new.data_severe_stunted %>%
  droplevels() %>%
  arrange(desc(study)) %>%
  group_by(study) %>%
  nest() %>%
  full_join(SS.Comb_severe_stunted, by = "study") %>%
  group_by(study)%>%
  do(augment(.$model[[1]], newdata = .$data[[1]], se_fit =T))
# View the resulting dataframe
print(predictions.SS.Comb_severe_stunted)
## # A tibble: 99,755 x 7
## # Groups: study [12]
     study cluster arms maternal dummy iwi .fitted .se.fit
      <chr> <fct> <fct>
                                 <dbl> <dbl>
##
                                                <dbl> <dbl>
```

```
## 1 DOSE ""
                                                        0.283
                    SQ-LNS
                                      0
                                                -1.81
## 2 DOSE
                    SQ-LNS
                                      0
                                                -1.81
                                                        0.276
                                            1
           11 11
## 3 DOSE
                    SQ-LNS
                                      0
                                                -1.81
                                                        0.269
## 4 DOSE
                    SQ-LNS
                                      0
                                                -1.82
                                                       0.262
                                            3
## 5 DOSE
                    SQ-LNS
                                      0
                                                -1.82
                                                       0.255
## 6 DOSE
                    SQ-LNS
                                      0
                                            5
                                               -1.82
                                                       0.248
## 7 DOSE
           11 11
                    SQ-LNS
                                      0
                                               -1.83
                                                       0.242
                                            6
## 8 DOSE
                    SQ-LNS
                                      0
                                            7
                                                -1.83
                                                        0.235
## 9 DOSE
                    SQ-LNS
                                      0
                                            8
                                                -1.84
                                                        0.228
## 10 DOSE
           11 11
                                      0
                                            9 -1.84
                                                        0.222
                    SQ-LNS
## # i 99,745 more rows
#write csv(predictions.SS.Comb, here::here("output", "predictions SSComb.csv"))
#write_csv(predictions.SS.Comb, here::here("output", "predictions_SSComb_nocluster.csv"))
plot_study_severe_stunted <- predictions.SS.Comb_severe_stunted%>%
  ggplot(., aes(iwi,expit(.fitted), color= arms_maternal))+
  geom line(size=1.25)+
  facet_wrap(.~study, ncol = 2, nrow = 6) +
  scale_colour_manual(values = c('darkgrey', 'darkblue')) +
                 scale_fill_manual(values = c('darkgrey', 'darkblue')) +
  geom_ribbon(mapping = aes(ymin= expit(.fitted - 1.96*.se.fit),
                            ymax = expit(.fitted + 1.96*.se.fit),
                            fill= arms_maternal), alpha= 0.1)+
  ylab("Probability of severe stunting") +
  xlab("IWI") +
  labs(color='Treatment',fill='Treatment') +
  theme minimal() +
  theme(legend.position = c(0.35, 0.95),
          #"none",
       legend.title = element_blank(),
        axis.title.y = element_text(size = 18),
       axis.title.x = element text(size = 18),
        strip.text = element_text(face="bold", size=14, hjust = 0.5),
       axis.text.y = element_text(face="bold", size=12),
        axis.text.x = element_text(face="bold", size=12))
print(plot_study_severe_stunted)
```



table(df_analysis_severe_stunted\$study, df_analysis_severe_stunted\$severe_stunted)

```
##
##
                      0
                           1
                         126
##
     DOSE
                   811
##
     JiVitA
                  3651
                         483
##
     MAHAY
                  2588
                         796
##
     PROMISBF
                   1695
                          82
##
     PROMISBF_CS
                   828
                          39
##
     PROMISM
                   926
                          86
##
     PROMISM_CS
                  1726
                         163
##
                         209
     RDNS
                  2269
##
     SHINE_HIV-
                  3372
                         302
##
     WASHB
                  4069
                        492
##
     WASHK
                   6013
                         581
     ZINC
                  2383 224
##
```

Point-wise Meta-analysis

```
# Create a new dataframe for the pointwise meta-analysis
# Since the predicted outcomes are similar with or without the random effect for the cluster variable,
new.data_severe_stunted <- df_analysis_severe_stunted %>%
select(study, cluster, arms_maternal) %>%
```

```
mutate(dummy = 0) %>%
  mutate(cluster = rep(unique(cluster), length.out = 1)) %>%
  distinct() %>%
  expand_grid(iwi = seq(0, 70, by = 1))
predictions.SS.Comb_severe_stunted <- new.data_severe_stunted %>%
  droplevels() %>%
  arrange(desc(study)) %>%
  group_by(study) %>%
  nest() %>%
  full_join(SS.Comb_severe_stunted, by = "study") %>%
  group_by(study)%>%
  do(augment(.$model[[1]], newdata = .$data[[1]], se_fit =T))
point.wise.DF.SS.Comb_severe_stunted = pointwise.ma(predictions.SS.Comb_severe_stunted,
                                      clustering.variable = "study",
                                      combining.variables = c("iwi","arms_maternal"),
                                      predicted.outcome = ".fitted",
                                      predicted.outcome.se = ".se.fit",
                                      predicted.outcome.CI = NULL,
                                      tau.method = "REML")
##
```

```
[1] "1%"
##
[1] "2%"
[1] "3%"
##
[1] "4%"
##
[1] "5%"
##
[1] "6%"
##
[1] "7%"
##
[1] "8%"
##
[1] "9%"
##
[1] "10%"
##
[1] "11%"
##
[1] "12%"
##
[1] "13%"
##
[1] "14%"
##
[1] "15%"
##
```

[1] "16%"

##

[1] "17%"

##

[1] "18%"

##

[1] "19%"

##

[1] "20%"

##

[1] "21%"

##

[1] "22%"

##

[1] "23%"

##

[1] "24%"

##

[1] "25%"

##

[1] "26%"

##

[1] "27%"

##

[1] "28%"

##

[1] "29%"

##

[1] "30%"

##

[1] "31%"

##

[1] "32%"

##

[1] "33%"

##

[1] "34%"

[1]

[1] "35%"

##

[1] "36%" ##

[1] "37%"

##

##

[1] "38%" ##

[1] "39%"

##

[1] "40%"

[1] "41%"

##

[1] "42%"

##

[1] "43%"

##

[1] "44%"

##

[1] "45%"

##

[1] "46%"

##

[1] "47%"

##

[1] "48%"

##

[1] "49%"

##

[1] "50%"

##

[1] "51%"

##

[1] "52%"

##

[1] "53%"

##

[1] "54%"

##

[1] "55%"

##

[1] "56%"

##

[1] "57%"

##

[1] "58%"

##

[1] "59%"

##

[1] "60%"

##

[1] "61%"

##

[1] "62%" ##

[1] "63%"

##

[1] "64%"

##

[1] "65%" ##

[1] "66%"

##

[1] "67%"

##

[1] "68%"

##

[1] "69%"

##

[1] "70%"

##

[1] "71%"

##

[1] "72%"

##

[1] "73%"

##

[1] "74%"

##

[1] "75%"

##

[1] "76%"

##

[1] "77%"

##

[1] "78%"

##

[1] "79%"

##

[1] "80%"

##

[1] "81%"

##

[1] "82%"

##

[1] "83%"

##

[1] "84%"

##

[1] "85%"

##

[1] "86%"

##

[1] "87%"

##

[1] "88%"

##

[1] "89%" ##

[1] "90%"

##

[1] "91%"

##

[1] "92%" ##

[1] "93%"

##

[1] "94%"

##

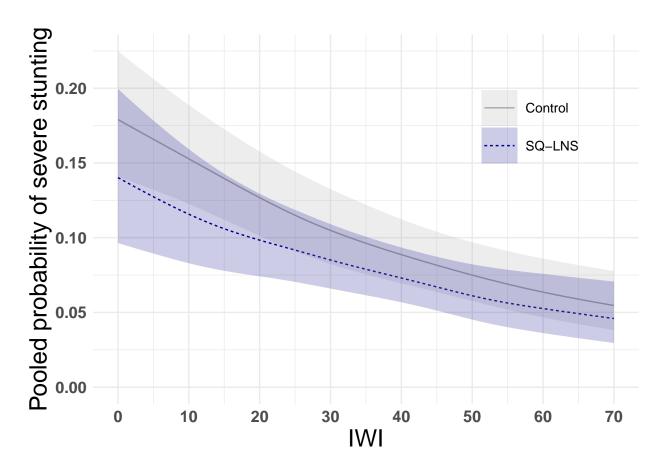
[1] "95%"

##

[1] "96%"

##

```
[1] "97%"
##
[1] "98%"
##
[1] "99%"
##
[1] "100%"
## Backtransform predicted outcomes and their corresponding confidence intervals
point.wise.DF.SS.Comb_severe_stunted$RE.meta =
                                                       expit(point.wise.DF.SS.Comb_severe_stunted$RE.me
point.wise.DF.SS.Comb_severe_stunted$RE.meta.upper = expit(point.wise.DF.SS.Comb_severe_stunted$RE.me
point.wise.DF.SS.Comb_severe_stunted$RE.meta.lower = expit(point.wise.DF.SS.Comb_severe_stunted$RE.me
point.wise.DF.SS.Comb.plot_severe_stunted = point.wise.DF.SS.Comb_severe_stunted%>%
  mutate(Treatment = as.factor(arms_maternal)) %>%
  ggplot(aes(x = iwi, y = RE.meta, fill = Treatment, linetype = Treatment)) +
  geom_line(size=0.5, aes(colour = Treatment))+
  scale_color_manual(values = c('darkgrey', 'darkblue')) +
  scale_fill_manual(values = c('darkgrey', 'darkblue')) +
  geom_ribbon(aes(ymin = RE.meta.lower,ymax=RE.meta.upper),alpha=0.2) +
  #scale_color_jama(name= "Treatment")+
  #scale_linetype_discrete(name ="Treatment")+ylab("") +
  #scale linetype manual(values=c("solid", "dashed")) +
  labs(color='Treatment', fill='Treatment', linetype='Treatment') +
  xlab(bquote('IWI')) +
  ylab(bquote('Pooled probability of severe stunting')) +
  theme minimal() +
  theme(legend.position = c(0.80, 0.75),
        legend.title = element_blank(),
        axis.title.y = element_text(size = 18),
       axis.title.x = element_text(size = 18),
       strip.text = element_text(face="bold", size=14, hjust = 0.5),
       axis.text.y = element_text(face="bold", size=12),
        axis.text.x = element_text(face="bold", size=12),
       legend.key.size = unit(1, "cm"),
        legend.key.width = unit(1, "cm"),
        legend.text=element_text(size=10, hjust = 0)) + ylim(c(0,NA)) +
  scale_x_continuous(breaks = seq(min(point.wise.DF.SS.Comb_severe_stunted$iwi), max(point.wise.DF.SS.C
print(point.wise.DF.SS.Comb.plot severe stunted)
```



```
# Save the plot
#ggsave(here::here("output", "gam_pooled_pointwise_wasted.png"),
# plot = point.wise.DF.SS.Comb.plot, width = 6, height = 6, dpi = 300)
```

Absolute treatment effect

```
point.wise.absolute_diff_SS.Comb_severe_stunted = pointwise.ma(data = absolute_diff_SS.Comb_severe_st
                                                   clustering.variable = "study",
                                                   combining.variables = c("iwi"),
                                                   predicted.outcome = "fit.diff",
                                                   predicted.outcome.se = NULL,
                                                   predicted.outcome.CI = c("diff.lower","diff.upper"),
                                                   tau.method = "REML"
##
[1] "1%"
##
[1] "3%"
##
[1] "4%"
##
[1] "6%"
##
[1] "7%"
##
[1] "8%"
##
[1] "10%"
##
[1] "11%"
##
[1] "13%"
##
[1] "14%"
##
[1] "15%"
##
[1] "17%"
##
[1] "18%"
##
[1] "20%"
##
[1] "21%"
##
[1] "23%"
##
[1] "24%"
##
[1] "25%"
##
[1] "27%"
##
[1] "28%"
##
[1] "30%"
##
```

[1] "31%"

##

[1] "32%"

##

[1] "34%"

##

[1] "35%"

##

[1] "37%"

##

[1] "38%"

##

[1] "39%"

##

[1] "41%"

##

[1] "42%"

##

[1] "44%"

##

[1] "45%"

##

[1] "46%"

##

[1] "48%"

##

[1] "49%"

##

[1] "51%"

##

[1] "52%"

##

[1] "54%"

##

[1] "55%"

##

[1] "56%"

##

[1] "58%"

##

[1] "59%"

##

[1] "61%"

##

[1] "62%"

##_

[1] "63%"

##

[1] "65%"

##

[1] "66%"

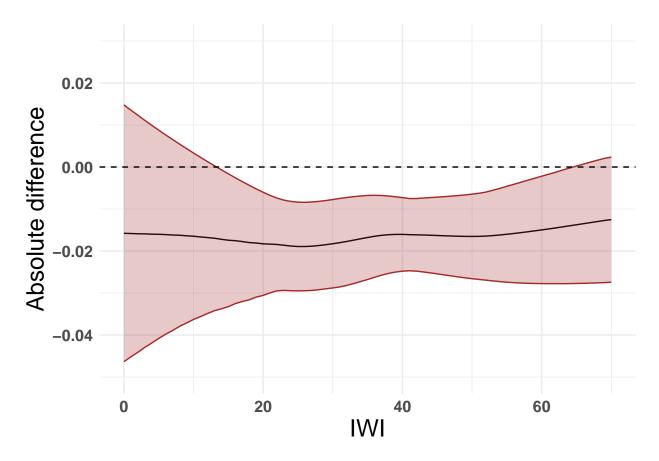
##

[1] "68%"

##

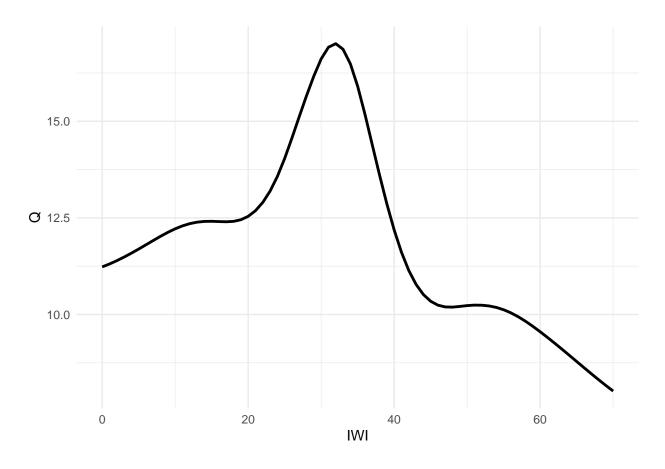
[1] "69%"

```
##
[1] "70%"
##
[1] "72%"
##
[1] "73%"
##
[1] "75%"
##
[1] "76%"
##
[1] "77%"
##
[1] "79%"
##
[1] "80%"
##
[1] "82%"
##
[1] "83%"
##
[1] "85%"
##
[1] "86%"
##
[1] "87%"
##
[1] "89%"
##
[1] "90%"
##
[1] "92%"
##
[1] "93%"
##
[1] "94%"
##
[1] "96%"
##
[1] "97%"
##
[1] "99%"
[1] "100%"
point.wise.absolute_diff_SS.Comb_severe_stunted = point.wise.absolute_diff_SS.Comb_severe_stunted%>%
 mutate(iwi = as.numeric(iwi))
point.wise.DF.SS.Comb.diff.plot_severe_stunted=point.wise.absolute_diff_SS.Comb_severe_stunted%>%
  ggplot(aes(x = iwi,y=RE.meta)) + geom_line(size=0.5)+
  geom_ribbon(mapping = aes(ymin=RE.meta.lower, ymax=RE.meta.upper),alpha=0.25,
              color="brown", fill="brown")+
  geom_hline(yintercept = 0, linetype=2)+ylab("") +
  xlab(bquote('IWI')) +
```



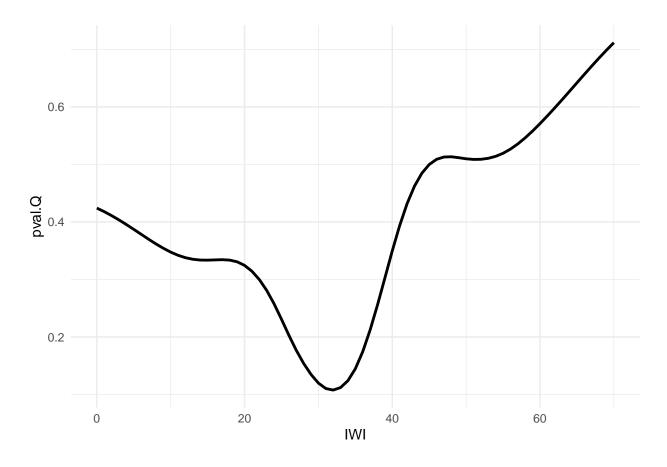
```
point.wise.Q.plot_severe_stunted=point.wise.absolute_diff_SS.Comb_severe_stunted%>%
    ggplot(aes(x = iwi,y=Q)) + geom_line(size=1)+
    xlab(bquote('IWI')) +
    theme_minimal()

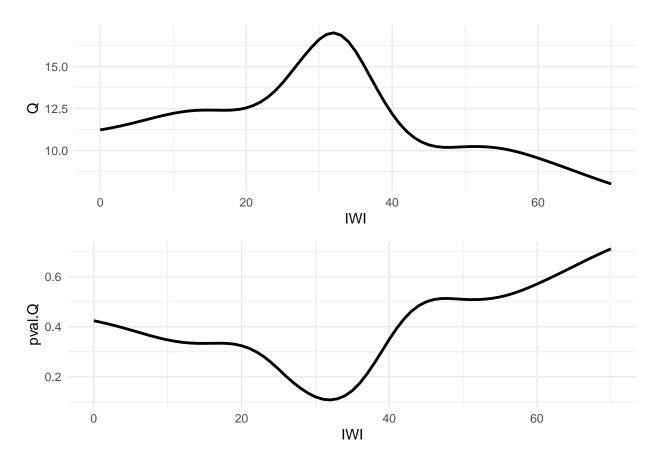
print(point.wise.Q.plot_severe_stunted)
```



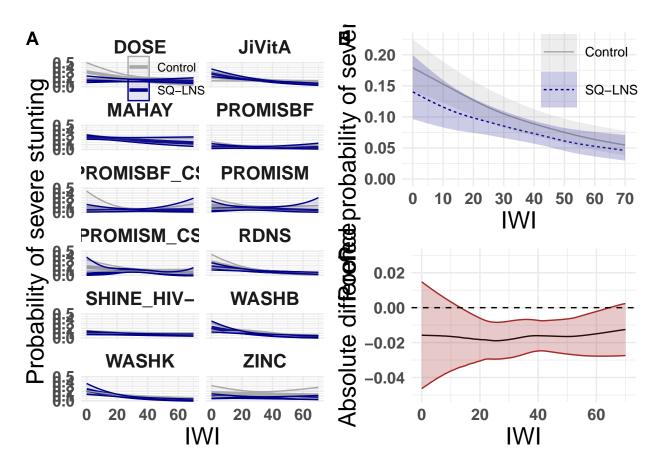
```
point.wise.pvalQ.plot_severe_stunted=point.wise.absolute_diff_SS.Comb_severe_stunted%>%
    ggplot(aes(x = iwi,y=pval.Q)) + geom_line(size=1)+
    xlab(bquote('IWI')) +
    theme_minimal()

print(point.wise.pvalQ.plot_severe_stunted)
```





```
# Save the plot
#ggsave(here::here("output", "qplot_pooled_pointwise_wasted.png"),
# plot = point.wise.Q.plot, width = 6, height = 6, dpi = 300)
```



```
# Save the plot
#ggsave(here::here("output", "gam_pooled_pointwise_severe_stunted_arm_maternal.png"),
# plot = plot_composite_severe_stunted, width = 14, height = 16, dpi = 300)
```

Session info

sessionInfo()

```
## R version 4.3.3 (2024-02-29)
## Platform: aarch64-apple-darwin20 (64-bit)
## Running under: macOS Sonoma 14.5
##
## Matrix products: default
          /System/Library/Frameworks/Accelerate.framework/Versions/A/Frameworks/vecLib.framework/Versi
## LAPACK: /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/lib/libRlapack.dylib; LAPACK v
##
## locale:
  [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
## time zone: America/Los_Angeles
## tzcode source: internal
##
## attached base packages:
                graphics grDevices utils
## [1] stats
                                               datasets methods
                                                                   base
##
```

```
## other attached packages:
    [1] geepack_1.3.10
                                               sandwich_3.1-0
                                                                   {\tt msm\_1.7.1}
                           estimatr_1.0.4
                           zoo 1.8-12
    [5] lmtest 0.9-40
                                               tidymv 3.4.2
                                                                   meta 7.0-0
   [9] metadat_1.2-0
                           mgcv_1.9-1
                                               nlme_3.1-164
                                                                   patchwork_1.2.0
##
## [13] RColorBrewer_1.1-3 cowplot_1.1.3
                                               gridExtra_2.3
                                                                   ggrepel_0.9.5
## [17] ggsci 3.0.1
                           broom 1.0.5
                                               rmarkdown 2.26
                                                                   skimr 2.1.5
## [21] readxl 1.4.3
                           lubridate 1.9.3
                                               forcats 1.0.0
                                                                   stringr 1.5.1
## [25] purrr 1.0.2
                           readr_2.1.5
                                               tidyr_1.3.1
                                                                   tibble_3.2.1
                                                                   table1_1.4.3
## [29] ggplot2_3.5.1
                           tidyverse_2.0.0
                                               haven_2.5.4
                           here_1.0.1
## [33] dplyr_1.1.4
## loaded via a namespace (and not attached):
   [1] tidyselect_1.2.1
                             farver_2.1.2
                                                                      CompQuadForm_1.4.3
                                                 fastmap_1.1.1
    [5] mathjaxr_1.6-0
                             digest_0.6.35
                                                                      lifecycle_1.0.4
                                                 timechange_0.3.0
   [9] survival_3.5-8
                            magrittr_2.0.3
                                                 compiler_4.3.3
                                                                      sass_0.4.9
## [13] rlang_1.1.4
                             tools_4.3.3
                                                 utf8_1.2.4
                                                                      yaml_2.3.8
## [17] knitr_1.45
                             labeling_0.4.3
                                                 xm12_1.3.6
                                                                      repr_1.1.7
## [21] expm 0.999-9
                             withr 3.0.0
                                                 numDeriv 2016.8-1.1
                                                                      grid 4.3.3
## [25] fansi_1.0.6
                             colorspace_2.1-0
                                                 scales_1.3.0
                                                                      MASS_7.3-60.0.1
## [29] tinytex 0.50
                             cli 3.6.2
                                                 mvtnorm 1.2-4
                                                                      metafor 4.6-0
## [33] ragg_1.3.0
                             generics_0.1.3
                                                 rstudioapi_0.15.0
                                                                      tzdb_0.4.0
## [37] cachem 1.0.8
                            minga 1.2.6
                                                 splines 4.3.3
                                                                      cellranger 1.1.0
## [41] base64enc_0.1-3
                                                 boot_1.3-29
                                                                      Matrix_1.6-5
                             vctrs_0.6.5
## [45] jsonlite 1.8.8
                            hms 1.1.3
                                                 Formula 1.2-5
                                                                      systemfonts 1.0.6
## [49] jquerylib 0.1.4
                                                 nloptr_2.0.3
                                                                      stringi_1.8.4
                             glue_1.7.0
## [53] gtable_0.3.5
                             lme4_1.1-35.1
                                                 munsell_0.5.1
                                                                      pillar 1.9.0
## [57] htmltools_0.5.7
                             R6_2.5.1
                                                 textshaping_0.3.7
                                                                      rprojroot_2.0.4
## [61] evaluate_0.23
                             lattice_0.22-5
                                                 highr_0.10
                                                                      backports_1.4.1
## [65] bslib_0.6.1
                             Rcpp_1.0.12
                                                 xfun_0.42
                                                                      pkgconfig_2.0.3
```