Absolute treatment effect by Pointwise meta-analysis - laz

Pearl Ante-Testard

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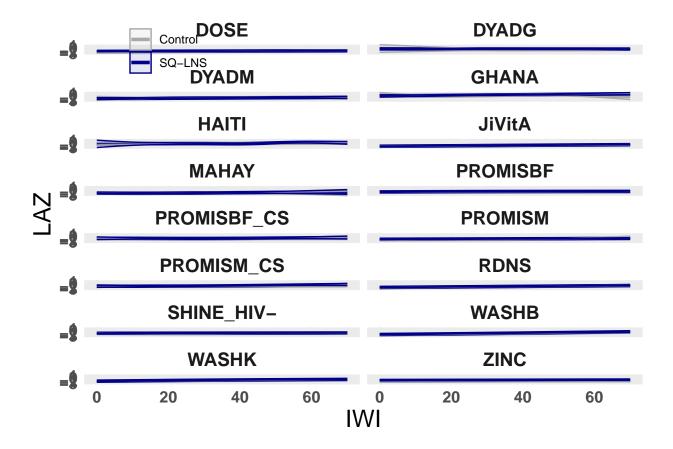
```
source(here::here("R", "0-config.R"))
```

Fit a cubic splines model per study

```
#SS.Comb <- df_analysis %>%
 # split(.$study) %>%
 # map dfr(function(df analysis) {
   study_name <- unique(df_analysis$study)</pre>
    if (study_name %in% c(study_name)) {
       model <- gam(laz_el ~ arm_primary + iwi * arm_primary + s(iwi, bs = "cr",
                                             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_laz <- df_analysis_laz %>%
  split(.$study) %>%
  map_dfr(function(df_analysis_laz) {
   study_name <- unique(df_analysis_laz$study)</pre>
    if (study_name %in% c("DOSE", "GHANA", "HAITI", "DYADG", "DYADM")) {
      model <- gam(laz_el ~ arms_maternal + iwi * arms_maternal + s(iwi, bs = "cr",</pre>
                                            by = arms_maternal),
                  data = df_analysis_laz,
                  family = "gaussian", method = "REML")
   } else {
      model <- gam(laz_el ~ arms_maternal + iwi * arms_maternal + s(iwi, bs = "cr",
                                            by = arms_maternal) +
                                           s(cluster, bs = "re", by = dummy),
                  data = df_analysis_laz,
                  family = "gaussian", method = "REML")
    # Return a data frame with study_name and model
    data.frame(study = study_name, model = I(list(model)))
 })
```

```
new.data_laz <- df_analysis_laz %>%
  select(study, cluster, arms maternal) %>%
  mutate(dummy = 0) %>%
  distinct() %>%
  expand_grid(iwi = seq(0, 70, by = 1))
predictions.SS.Comb_laz <- new.data_laz %>%
  droplevels() %>%
  arrange(desc(study)) %>%
  group_by(study) %>%
  nest() %>%
  full_join(SS.Comb_laz, by = "study") %>%
  group_by(study)%>%
  do(augment(.$model[[1]], newdata = .$data[[1]], se_fit =T))
# View the resulting dataframe
print(predictions.SS.Comb_laz)
## # A tibble: 100,323 x 7
## # Groups:
              study [16]
      study cluster arms maternal dummy iwi .fitted .se.fit
##
      <fct> <fct>
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                                  <dbl> <dbl>
                                                <dbl>
                                                        <dbl>
## 1 DOSE ""
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## 2 DOSE ""
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## 3 DOSE
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                                            2
                                                -1.96 0.0962
## 4 DOSE
                    SQ-LNS
                                      0
                                            3
                                               -1.96 0.0938
## 5 DOSE
           11 11
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                                      0
                                                -1.95 0.0914
## 6 DOSE
                    SQ-LNS
                                      0
                                            5
                                                -1.95 0.0890
## 7 DOSE
           11 11
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                                      0
                                            6
                                                -1.95 0.0866
## 8 DOSE
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                                            7
                                               -1.94 0.0842
                    SQ-LNS
## 9 DOSE
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                                      0
                                            8 -1.94 0.0819
                    SQ-LNS
           11 11
## 10 DOSE
                    SQ-LNS
                                      0
                                            9
                                                -1.93 0.0796
## # i 100,313 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_laz <- predictions.SS.Comb_laz%>%
  ggplot(., aes(iwi,.fitted, color= arms_maternal))+
  geom_line(size=1.25)+
  facet_wrap(.~study, ncol = 2, nrow = 8) +
  scale_colour_manual(values = c('darkgrey', 'darkblue')) +
                 scale_fill_manual(values = c('darkgrey', 'darkblue')) +
  geom_ribbon(mapping = aes(ymin= .fitted - 1.96*.se.fit,
                            ymax = .fitted + 1.96*.se.fit,
                            fill= arms_maternal), alpha= 0.1)+
  ylab("LAZ") +
  xlab("IWI") +
  labs(color='Treatment',fill='Treatment') +
  theme minimal() +
  theme(legend.position = c(0.15, 0.98),
```

```
#"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_laz)
```



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_laz <- df_analysis_laz %>%
    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_laz <- new.data_laz %>%
```

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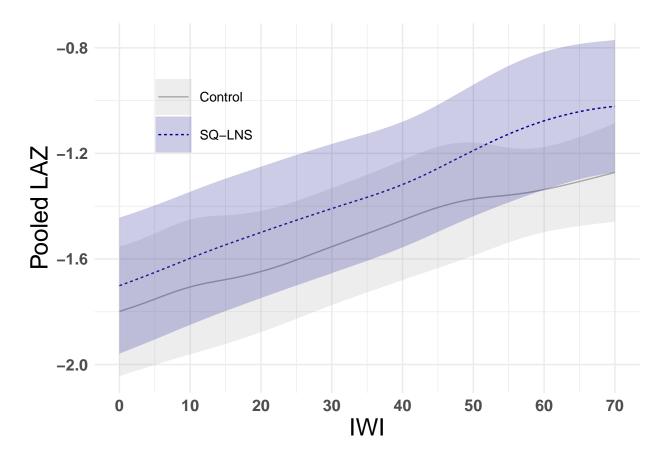
##

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##

```
## Backtransform predicted outcomes and their corresponding confidence intervals
point.wise.DF.SS.Comb_laz$RE.meta = point.wise.DF.SS.Comb_laz$RE.meta
point.wise.DF.SS.Comb_laz$RE.meta.upper = point.wise.DF.SS.Comb_laz$RE.meta.upper
point.wise.DF.SS.Comb_laz$RE.meta.lower = point.wise.DF.SS.Comb_laz$RE.meta.lower
```

```
point.wise.DF.SS.Comb.plot_laz = point.wise.DF.SS.Comb_laz%>%
  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 LAZ')) +
  theme_minimal() +
  theme(legend.position = c(0.20, 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)) +
scale_x_continuous(breaks = seq(min(point.wise.DF.SS.Comb_laz$iwi),
                                                                      max(point.wise.DF.SS.Comb_laz$iwi
print(point.wise.DF.SS.Comb.plot_laz)
```



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

Absolute treatment effect

```
point.wise.absolute_diff_SS.Comb_laz = pointwise.ma(data = absolute_diff_SS.Comb_laz,
                                                   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"
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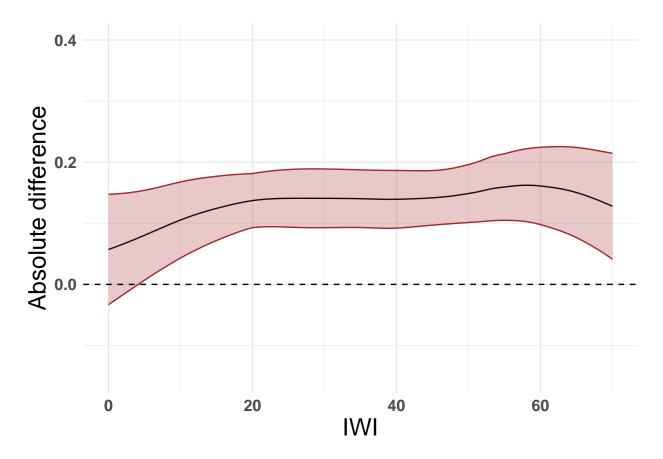
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##

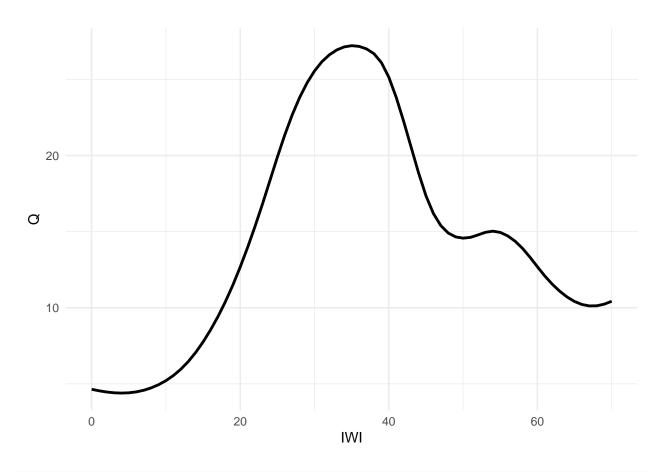
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point.wise.absolute_diff_SS.Comb_laz = point.wise.absolute_diff_SS.Comb_laz%>%
 mutate(iwi = as.numeric(iwi))
point.wise.DF.SS.Comb.diff.plot_laz=point.wise.absolute_diff_SS.Comb_laz%>%
  ggplot(aes(x = iwi,y=RE.meta)) + geom_line(size=0.5)+
  geom_ribbon(mapping = 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')) +
```



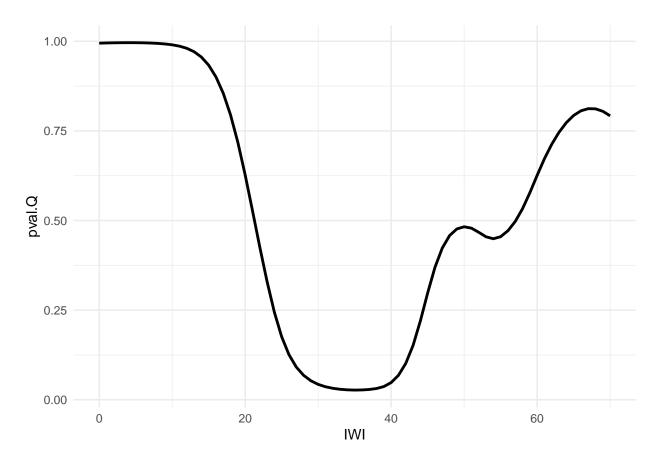
```
point.wise.Q.plot_laz=point.wise.absolute_diff_SS.Comb_laz%>%
    ggplot(aes(x = iwi,y=Q)) + geom_line(size=1)+
    xlab(bquote('IWI')) +
    theme_minimal()

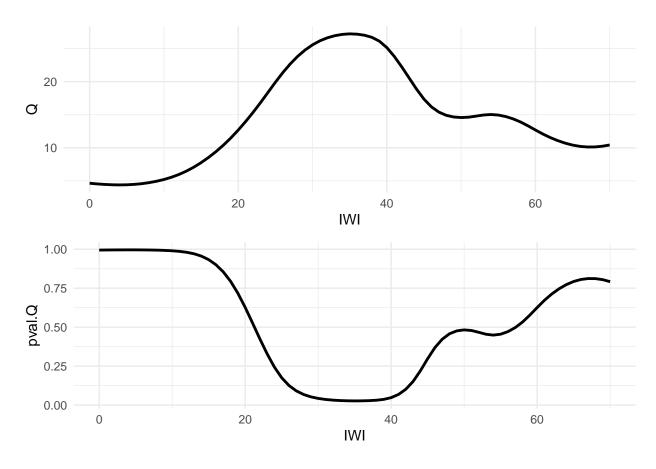
print(point.wise.Q.plot_laz)
```



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

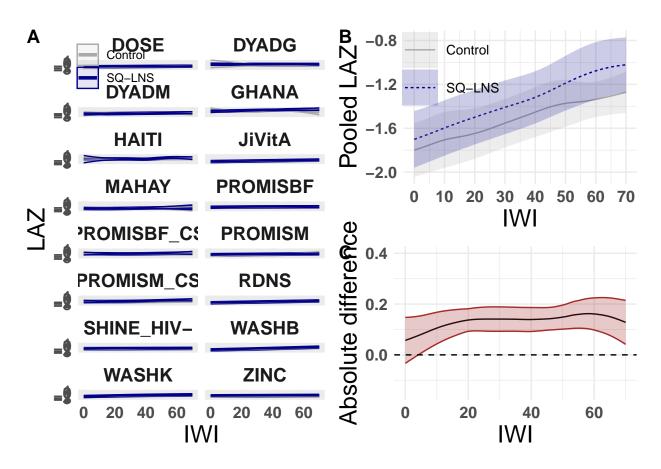
print(point.wise.pvalQ.plot_laz)
```





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

```
plot_comp_effect_laz <- plot_grid(point.wise.DF.SS.Comb.plot_laz, point.wise.DF.SS.Comb.diff.plot_laz, plot_composite_laz <- plot_grid(plot_study_laz, plot_comp_effect_laz, ncol=2, labels = c("A", "", ""))</pre>
```



```
# Save the plot
#ggsave(here::here("output", "gam_pooled_pointwise_laz_arm_maternal.png"),
# plot = plot_composite_laz, 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:
## [1] stats
                graphics grDevices utils
                                               datasets methods
                                                                   base
```

```
## other attached packages:
    [1] geepack_1.3.10
                                               sandwich_3.1-0
                                                                   {\tt msm\_1.7.1}
                           estimatr_1.0.4
    [5] lmtest 0.9-40
                           zoo 1.8-12
                                               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
                                                                   tibble_3.2.1
## [25] purrr 1.0.2
                           readr_2.1.5
                                               tidyr_1.3.1
                                                                   table1_1.4.3
## [29] ggplot2_3.5.1
                           tidyverse_2.0.0
                                               haven_2.5.4
## [33] dplyr_1.1.4
                           here_1.0.1
## 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] cli 3.6.2
                             mvtnorm 1.2-4
                                                 metafor 4.6-0
                                                                      ragg_1.3.0
## [33] generics_0.1.3
                             rstudioapi_0.15.0
                                                 tzdb_0.4.0
                                                                      cachem_1.0.8
## [37] minga 1.2.6
                             splines 4.3.3
                                                 cellranger 1.1.0
                                                                      base64enc 0.1-3
## [41] vctrs_0.6.5
                             boot_1.3-29
                                                 Matrix_1.6-5
                                                                      jsonlite_1.8.8
## [45] hms 1.1.3
                             Formula 1.2-5
                                                 systemfonts 1.0.6
                                                                      jquerylib 0.1.4
## [49] glue_1.7.0
                             nloptr_2.0.3
                                                 stringi_1.8.4
                                                                      gtable_0.3.5
## [53] lme4 1.1-35.1
                             munsell 0.5.1
                                                 pillar 1.9.0
                                                                      htmltools_0.5.7
## [57] R6_2.5.1
                             textshaping_0.3.7
                                                 rprojroot_2.0.4
                                                                      evaluate_0.23
## [61] lattice_0.22-5
                             highr_0.10
                                                 backports_1.4.1
                                                                      bslib_0.6.1
## [65] Rcpp_1.0.12
                             xfun_0.42
                                                 pkgconfig_2.0.3
```