Analysis

```
three_lvl_dat <- readRDS("three_lvl_dat.rds")</pre>
head(three_lvl_dat)
##
     site patient visit X1 X2 X3
                                         Х4
                                                    Х5
                                                               Х6
## 1
                      1 1 0 0 -0.2333860 0.5258316 0.8517787
                                                                    9.713744
                1
## 2
                      2 0 0 0 -1.7007881 -2.3441436 0.2597312 -16.215685
                      3 0 1 0 -0.6558381 -0.1574295 -1.1642077 -5.654015
## 3
        1
                1
## 4
       1
                      4 0 0 0 -0.6304705 -1.4508508 -1.4120594 -15.497097
## 5
                      5 0 0 1 -0.9323276 -0.3451917 -1.2283032 -7.509270
        1
## 6
                      1 1 1 0.3219037 -1.6069905 -0.5368642 -3.596303
H <- n_distinct(three_lvl_dat$site)</pre>
px <- 6
source("../PEAL Engine/PEAL_Engine_RI.R")
##
## Attaching package: 'nlme'
## The following object is masked from 'package:lme4':
##
##
       lmList
## The following object is masked from 'package:dplyr':
##
##
       collapse
# Preprocessing
# Calculate the number of visits per patient per site
visit_count <- three_lvl_dat %>%
  dplyr::group_by(site, patient) %>%
  dplyr::summarise(total_visits = n(), .groups = "drop")
# Reorder data (as a preprocess probably)
rearranged_data <- merge(three_lvl_dat, visit_count, by = c("site", "patient")) %>%
  arrange(site, total_visits, patient) %>%
  mutate(site = factor(site))
```

PEAL

```
# XYZ
Y <- rearranged_data$Y
X <- as.matrix(rearranged_data[, paste0("X", 1:px)])</pre>
X \leftarrow cbind(X)
Z <- list()</pre>
for(i in 1:H){
  count mat = rearranged data %>%
                     filter(site == i) %>%
                     group_by(site, patient) %>%
                     dplyr::summarise(n_hi = n(), .groups = 'drop')
  Z[[i]] <- (generate_Zhv_matrix(count_mat))</pre>
id.site <- rearranged_data$site</pre>
ShXYZ <- peal.get.summary(Y, X, Z, id.site)</pre>
fit.peal = peal.fit.RI(Y = NULL, X = NULL, Z = NULL,
                       id.site = NULL, weights = NULL,
                       pooled = F, reml = T,
                       common.s2 = T,
                       ShXYZ = ShXYZ, # only need summary stats
                       corstr = 'independence',
                       mypar.init = NULL)
## Default mypar.init (var comp) = 1 1 1 1
## Convergence Reached and Hessian PD
## The number of function evaluations used is 33
## Done!
```

LMM

Loading required namespace: optimx

fixed effects

random effects

RS

```
# Create formula for design matrix
# fixed effects part
formula <- ~ X1 + X2 + X3 + X4 + X5 + X6 -1
X <- model.matrix(formula, data = rearranged_data)

# RS part
formula_RS <- ~ X6 - 1
X_RS <- model.matrix(formula_RS, data = rearranged_data)</pre>
```

```
Z <- construct_Z_list(rearranged_data, colnames(X_RS))</pre>
m_h_all = (rearranged_data %>% group_by(site) %>% dplyr::summarise(n_distinct(patient)))[,2]
ShXYZ <- peal.get.summary(Y, X, Z, id.site, m_h_all = m_h_all)
fit.peal = peal.fit.RS(Y = NULL, X = NULL, Z = NULL,
                       id.site = NULL, weights = NULL,
                       pooled = F, reml = T,
                       common.s2 = T,
                       ShXYZ = ShXYZ, # only need summary stats
                       corstr = 'independence',
                       mypar.init = NULL)
## Default mypar.init (var comp) = 1 1 1 1 1 1 1
## Convergence Reached and Hessian PD
## The number of function evaluations used is 45
## Done!
fit.lmer <- lmer(</pre>
  Y \sim -1 + X1 + X2 + X3 + X4 + X5 + X6 +
       (1 | site) +
       (0 + dummy(site, "1")|| site:patient) + (0 + covar_composite_1 || site:patient) +
       (0 + dummy(site, "2")|| site:patient) + (0 + covar_composite_2 || site:patient) +
(0 + dummy(site, "3")|| site:patient) + (0 + covar_composite_3 || site:patient),
  data = rearranged_data %>%
    mutate(
      covar_composite_1 = dummy(site, "1") * X6,
      covar_composite_2 = dummy(site, "2") * X6,
      covar_composite_3 = dummy(site, "3") * X6
    ),
  control = lmerControl(
    optimizer = "optimx",
    optCtrl = list(method = "L-BFGS-B")
  )
lmer.mod_RS_summ <- summary(fit.lmer)</pre>
lmer beta <- fixef(fit.lmer)</pre>
lmer_sigma = c(sqrt(lmer.mod_RS_summ$varcor$site),
                 diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.5)),
                  diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.3)),
                  diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.1)),
                 diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.4)),
                 diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.2)),
                 diag(sqrt(lmer.mod_RS_summ$varcor$site.patient)),
                lmer.mod_RS_summ$sigma)
names(lmer_sigma) <- c("site_RI_sigma", paste0("pt_sigma_RI_bysite", 1:H),paste0("pt_sigma_RS_bysite",</pre>
```

fixed effects

```
peal_beta = c(fit.peal$b)
lmm_beta = fixef(fit.lmer)
cbind(lmm_beta, peal_beta)

## lmm_beta peal_beta
## X1 1.025053   1.025055
## X2 2.104015   2.104018
## X3 2.936898   2.936901
## X4 3.999226   3.999225
## X5 5.029522   5.029522
## X6 6.012435   6.012434
```

random effects

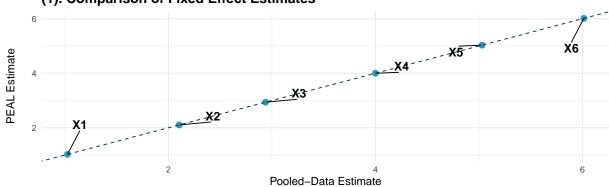
```
summ.lmer <- summary(fit.lmer)</pre>
lmm_sigma = c(sqrt(lmer.mod_RS_summ$varcor$site),
                diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.5)),
                 diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.3)),
                 diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.1)),
                diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.4)),
                diag(sqrt(lmer.mod_RS_summ$varcor$site.patient.2)),
                diag(sqrt(lmer.mod_RS_summ$varcor$site.patient)),
               lmer.mod_RS_summ$sigma)
H = 3
names(lmm_sigma) <- c("site RI", paste0("pt RI by site", 1:H), paste0("pt RS by site", 1:H), "residual")</pre>
peal_sigma = c(sqrt(fit.peal$V), sqrt(fit.peal$s2))
cbind(lmm_sigma, peal_sigma)
##
                  lmm sigma peal sigma
## site RI
                  0.1112932 0.1113393
## pt RI by site1 0.9590522 0.9590176
## pt RI by site2 0.9623781 0.9624495
## pt RI by site3 0.8379339 0.8378841
## pt RS by site1 0.3219149 0.3219251
## pt RS by site2 0.2795292 0.2795442
## pt RS by site3 0.2733985 0.2733912
## residual
                  0.6379580 0.6379587
library(ggplot2)
library(tibble)
library(patchwork)
##
## Attaching package: 'patchwork'
```

```
## The following object is masked from 'package:MASS':
##
##
       area
library(ggrepel)
# Panel A: Beta
df beta <- tibble(</pre>
 dlmm = as.numeric(peal beta),
 lmer = as.numeric(lmm_beta),
 coef = names(lmm_beta)
)
p_beta <- ggplot(df_beta, aes(x = lmer, y = dlmm, label = coef)) +</pre>
  geom_point(color = "#219ebc", size = 3) +
  geom_abline(intercept = 0, slope = 1, linetype = "dashed", color = "#023047") +
 geom_text_repel(size = 5, max.overlaps = Inf, box.padding = 2, fontface = "bold") +
 labs(
   x = "Pooled-Data Estimate", y = "PEAL Estimate",
   title = "(1). Comparison of Fixed Effect Estimates"
 theme_minimal(base_size = 13) + theme(plot.title = element_text(face = "bold"))
# Panel B: Sigma
# Create df_sigma with grouping for color
df_sigma <- tibble(</pre>
 dlmm = as.numeric(peal_sigma),
 lmer = as.numeric(lmm_sigma),
 component = names(lmm_sigma),
 type = c(
   "Site RI", rep("Pt RI", H), rep("Pt RS", H), "Residual"
)
type_colors <- c(</pre>
  "Site RI" = #1f78b4",
                             # Blue
 "Pt RI" = \#33a02c",
                            # Green
 "Pt RS" = "#ff7f00",
                            # Orange
 "Residual" = "#e31a1c"
                            # Red
# Sigma plot with color
p_sigma <- ggplot(df_sigma, aes(x = lmer, y = dlmm, label = component, color = type)) +
  geom_point(size = 3.5) +
  geom_abline(intercept = 0, slope = 1, linetype = "dashed", color = "gray30") +
 geom_text_repel(size = 3.5, max.overlaps = Inf, box.padding = 1.2, fontface = "bold") +
 labs(
   x = "Pooled-Data Estimate",
   y = "PEAL Estimate",
   title = "(2). Comparison of Random Effect SDs and Residual SD",
   color = "Component Type"
  ) +
  scale_color_manual(values = type_colors) +
```

```
theme_minimal(base_size = 13) +
theme(
   legend.position = "none",
   plot.title = element_text(face = "bold")
)

# Combine with patchwork
compare_plot <- p_beta / p_sigma +
   plot_annotation(
   title = ""
)</pre>
```

(1). Comparison of Fixed Effect Estimates



(2). Comparison of Random Effect SDs and Residual SD

