Data analysis script 2: Testing for replication of ERP to DDM concordance in RDoC dataset

Template Rmd

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About

This script does preliminary data analysis comparing the psychometrics of HDDM models vs. raw accuracy vs. NIH Toolbox derived score for the Flanker task of the RDoC study.

v = **drift rate**: "The parameter of primary interest for the present study is the drift rate, v, which is the average rate of approach to a boundary and indexes the quality or strength of evidence extracted from the stimulus. A large value of drift indicates strong decision evidence, meaning the decision process will approach the appropriate boundary quickly, leading to fast and accurate responses." Larger values of v mean faster accumulation of evidence (faster rate), larger t means slower non-decision time processing.

z = **response bias**: "If participants were biased toward one of the two responses (e.g., by increasing the proportion of one response over the other), they would move their starting point closer to that boundary. This produces faster and more probable responses at that boundary since less evidence is needed to reach it." Distance between the the start of the drift process and upper boundary.

a = separation between the two boundaries: "response caution or speed/accuracy tradeoffs. If the boundary separation is relatively small, responses will take less time to reach a boundary, leading to faster responses, but they will also be more likely to reach the wrong boundary due to noise in the process, leading to more errors." Larger a means more separation between the correct and incorrect response boundaries

t = non-decision time: "takes into account the duration of nondecisional processes... such processes may comprise basic encoding processes, the configuration of working memory for a task, and processes of response execution (i.e., motor activity)." (Voss et al., 2013)

Per Allie, larger values of v mean faster accumulation of evidence (faster rate), larger t means slower non-decision time processing, larger a means more separation between the correct and incorrect response boundaries, and z is the distance from the upper boundary to the start of the drift process.

Get Setup

Clear everything & set width

Load Libraries

[1] "/Users/brentrappaport/Documents/temp_files/DDM/work"

Function correlation matrix

```
use %in% c("all", "upper", "lower")
  is.logical(replace_diagonal)
  is.logical(show_significance)
  is.character(replacement)
})
# we need the Hmisc package for this
require(Hmisc)
# retain only numeric and boolean columns
isNumericOrBoolean = vapply(df, function(x) is.numeric(x) | is.logical(x), logical(1))
if (sum(!isNumericOrBoolean) > 0) {
  cat('Dropping non-numeric/-boolean column(s):', paste(names(isNumericOrBoolean)[!isNumericOrBoolean
df = df[isNumericOrBoolean]
# transform input data frame to matrix
x <- as.matrix(df)</pre>
# run correlation analysis using Hmisc package
correlation_matrix <- Hmisc::rcorr(x, type = )</pre>
R <- correlation_matrix$r # Matrix of correlation coeficients
p <- correlation_matrix$P # Matrix of p-value</pre>
# transform correlations to specific character format
Rformatted = formatC(R, format = 'f', digits = digits, decimal.mark = decimal.mark)
# if there are any negative numbers, we want to put a space before the positives to align all
if (sum(R < 0) > 0) {
  Rformatted = ifelse(R > 0, pasteO(' ', Rformatted), Rformatted)
}
# add significance levels if desired
if (show_significance) {
  # define notions for significance levels; spacing is important.
  stars <- ifelse(is.na(p), " ", ifelse(p < .001, "***", ifelse(p < .01, "** ", ifelse(p < .05, "* ^{**}")
  Rformatted = pasteO(Rformatted, stars)
# build a new matrix that includes the formatted correlations and their significance stars
Rnew <- matrix(Rformatted, ncol = ncol(x))</pre>
rownames(Rnew) <- colnames(x)</pre>
colnames(Rnew) <- paste(colnames(x), "", sep =" ")</pre>
# replace undesired values
if (use == 'upper') {
  Rnew[lower.tri(Rnew, diag = replace_diagonal)] <- replacement</pre>
} else if (use == 'lower') {
  Rnew[upper.tri(Rnew, diag = replace_diagonal)] <- replacement</pre>
} else if (replace_diagonal) {
  diag(Rnew) <- replacement</pre>
return(Rnew)
```

```
save_correlation_matrix = function(df, filename, ...) {
  write.csv2(correlation_matrix(df, ...), file = filename)
}
```

Load Data

Remember to immediately rename and remove. Avoid overwriting old data.

```
here::i_am("work/analysis/do01_LDDM_rdoc.Rmd")
## here() starts at /Users/brentrappaport/Documents/temp_files/DDM
LDDM_do1_alt_rdoc <- read.csv(here("./RDoC/DDM_Results/Block_Based/RDoC_Day1_Block11_Alternative_Models
LDDM_do1_rdoc <- read_sav(here("./work/data/RDoC_ERN_0-80ms_Final.sav"))
LDDM_do1_rdoc$ID <- LDDM_do1_rdoc$SubjectID</pre>
LDDM do2 rdoc <- LDDM do1 rdoc %>%
    left_join(LDDM_do1_alt_rdoc, by="ID") %>%
    filter(DQ >= 0.5)
load(file=here("./work/data/LDDM_cleaning04_fullbeh_rdoc.RData"))
load(file=here("./work/data/LDDM_cleaning04_rdoc_calc4.RData"))
## Load Color-Word D-KEFS test (Stroop) data
LDDM_do1_rdoc_iq <- read_sav(here("./work/data/RDoC_WTAR_FINAL 1.28.2019.sav"))</pre>
LDDM_do1_rdoc_iq$ID <- LDDM_do1_rdoc_iq$SubjectID</pre>
# lapply(LDDM_do1_rdoc_iq, attr, "label")
LDDM_do1_neuro <- read_sav(here("./work/data/RDoC_Neuropsych_Cleaning_HL_FINAL_3.19.18.sav"))
LDDM_do1_neuro$ID <- LDDM_do1_neuro$SubjectID</pre>
\# LDDM\_do1\_education \leftarrow read\_sav(here("./work/data/FINAL\_RDoC\_Proband\_Sibling\_ALL\_Questionnaires\_ItemLettles = for all the statements of the statement of the 
## Load data on participant's gender
LDDM_do3_rdoc_gender <- readxl::read_excel(here("./work/data/RDoC_ProbandSiblingCombined All 12.3.xlsx"
LDDM do3 rdoc gender$ID <- LDDM do3 rdoc gender$\SubjectID of RDoC ProbandSIbling AllQuestionnaires Pro
LDDM_do3_rdoc_gender$Gender <- LDDM_do3_rdoc_gender$`Gender of RDoC_ProbandSiblingCombined 12.3.jmp`
LDDM_do3_rdoc_gender_only <- LDDM_do3_rdoc_gender %>%
    select(ID,Gender)
LDDM_do2_neuro <- LDDM_do1_neuro %>%
    left_join(LDDM_do1_rdoc_iq, by="ID")
LDDM_do3_rdoc <- LDDM_do2_rdoc %>%
    left_join(LDDM_do3_rdoc_gender_only, by="ID") %>%
    left_join(LDDM_cleaning04_rdoc_calc4, by="ID") %>%
    left_join(LDDM_do2_neuro, by="ID") %>%
    select(-starts with("Reason"))
# LDDM_do3_rdoc_no_outliers <- LDDM_do2_rdoc_no_outliers %>%
    # left_join(LDDM_cleaning04_rdoc_calc4, by="ID")
```

Remove subjects with bad data

```
all_rdoc <- read_sav(here("./RDoC/All_Enrolled_IDs.sav"))</pre>
all_ern_rdoc<- read_sav(here("./RDoC/RDoC_ERN_0-80ms_Final.sav"))
all_rdoc <- all_rdoc %>%
  mutate(Eligible_use_this = Eligible) %>%
  select(SubjectID, Eligible_use_this, Reason)
all_ern_rdoc <- all_ern_rdoc %>%
  mutate(Has atleast 10 error trials = all ern rdoc$Number of errors included in average>9,
         good_data = all_ern_rdoc$DQ>=0.5 & all_ern_rdoc$DQ<2) %>%
  select(SubjectID, Has_atleast_10_error_trials, good_data)
LDDM_do3_rdoc_with_bad_data <- full_join(LDDM_do3_rdoc, all_rdoc, by="SubjectID") %>%
  left join(all ern rdoc, by="SubjectID")
LDDM_do3_rdoc <- full_join(LDDM_do3_rdoc, all_rdoc, by="SubjectID") %>%
  full_join(all_ern_rdoc, by="SubjectID") %>%
  filter(Eligible_use_this==1) %>%
  filter(Has_atleast_10_error_trials==TRUE) %>%
  filter(good_data==TRUE)
```

Make residualized score

```
electrodes_list <- c("Fz","Cz","FCz")

# Calculate residualized ERN and difference score ERN (for verification of direction)

for (ex in electrodes_list[1:3]){
    eval(parse(text=paste0('LDDM_do3_rdoc$',ex,'_ERN_080 <- stdres(lm(',ex,'3chan_error_500_300 ~ ',ex,
    eval(parse(text=paste0('LDDM_do3_rdoc$',ex,'_ERN_080_diff <- LDDM_do3_rdoc$',ex,'3chan_error_500_30
    eval(parse(text=paste0('LDDM_do3_rdoc$',ex,'_ERN_080_diff <- as.numeric(LDDM_do3_rdoc$',ex,'_ERN_080
    eval(parse(text=paste0('LDDM_do3_rdoc$',ex,'_ERN_080 <- as.numeric(LDDM_do3_rdoc$',ex,'_ERN_080)'))
}

LDDM_do3_rdoc$B11_avtz_D0_v_v <- rowMeans(LDDM_do3_rdoc[,c('B11_avtz_D0_v_v_con','B11_avtz_D0_v_v_incon')]
LDDM_do3_rdoc$B11_avtz_D0_v_v <- rowMeans(LDDM_do3_rdoc[,c('B11_avtz_D0_v_v_con','B11_avtz_D0_v_v_incon')]
LDDM_do3_rdoc$B11_avtz_D0_v_v <- rowMeans(LDDM_do3_rdoc[,c('B11_avtz_D0_v_v_con','B11_avtz_D0_v_v_incon')]</pre>
```

Remove outliers

```
# lapply(LDDM_do3_rdoc, attr, "label")
LDDM_do3_rdoc$NL_time_rev <- -1*LDDM_do3_rdoc$NL_time
LDDM_do3_rdoc$InSw_time_rev <- -1*LDDM_do3_rdoc$InSw_time
LDDM_do3_rdoc$exec_composite <- rowMeans(LDDM_do3_rdoc[,c("FF_cor","Switch_cor","NL_time_rev","InSw_time</pre>
```

Standardize scores

```
var_list <- c("B11_avtz_D0_v_v","B11_avtz_D0_v_v_con","B11_avtz_D0_v_v_incon","B11_avtz_D0_v_a","B11_av</pre>
              "B11_avt_v", "B11_avt_a", "B11_avt_t",
              "B11_avtz_v", "B11_avtz_a", "B11_avtz_t", "B11_avtz_z",
              "B11_avt_D0_v_v", "B11_avt_D0_v_v_con", "B11_avt_D0_v_v_incon", "B11_avt_D0_v_a", "B11_avt_D
              "B11_avtz_D0_v_v","B11_avtz_D0_v_v_con","B11_avtz_D0_v_v_incon","B11_avtz_D0_v_a","B11_av
              "flanker_score_rdoc",
              "FCz_ERN_080", "FCz_ERN_080_diff",
              "Inhib_time", "exec_composite", "Pred_FSIQ", "Motor_time", "accuracy", "accuracy_congruent_log
for (v in var_list){
  print(paste0("LDDM do3 rdoc$",v))
  eval(parse(text=paste0('LDDM_do3_rdoc$',v,'_z <- scale(LDDM_do3_rdoc$',v,', center=T, scale=T)')))</pre>
}
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v_con"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v_incon"
## [1] "LDDM do3 rdoc$B11 avtz DO v a"
## [1] "LDDM do3 rdoc$B11 avtz DO v t"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_z"
## [1] "LDDM_do3_rdoc$B11_avt_v"
## [1] "LDDM_do3_rdoc$B11_avt_a"
## [1] "LDDM_do3_rdoc$B11_avt_t"
## [1] "LDDM_do3_rdoc$B11_avtz_v"
## [1] "LDDM_do3_rdoc$B11_avtz_a"
## [1] "LDDM_do3_rdoc$B11_avtz_t"
## [1] "LDDM_do3_rdoc$B11_avtz_z"
## [1] "LDDM_do3_rdoc$B11_avt_D0_v_v"
## [1] "LDDM_do3_rdoc$B11_avt_D0_v_v_con"
## [1] "LDDM do3 rdoc$B11 avt DO v v incon"
## [1] "LDDM_do3_rdoc$B11_avt_D0_v_a"
## [1] "LDDM_do3_rdoc$B11_avt_D0_v_t"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v_con"
```

```
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v_incon"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_a"
## [1] "LDDM do3 rdoc$B11 avtz DO v t"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_z"
## [1] "LDDM_do3_rdoc$flanker_score_rdoc"
## [1] "LDDM do3 rdoc$FCz ERN 080"
## [1] "LDDM do3 rdoc$FCz ERN 080 diff"
## [1] "LDDM_do3_rdoc$Inhib_time"
## [1] "LDDM_do3_rdoc$exec_composite"
## [1] "LDDM_do3_rdoc$Pred_FSIQ"
## [1] "LDDM_do3_rdoc$Motor_time"
## [1] "LDDM_do3_rdoc$accuracy"
## [1] "LDDM_do3_rdoc$accuracy_congruent_log"
## [1] "LDDM_do3_rdoc$accuracy_incongruent"
for (v in var_list){
  print(paste0("LDDM_do3_rdoc$",v))
  eval(parse(text=paste0('LDDM_do3_rdoc_no_outliers$',v,'_z <- scale(LDDM_do3_rdoc_no_outliers$',v,', c
## [1] "LDDM do3 rdoc$B11 avtz DO v v"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v_con"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v_incon"
## [1] "LDDM do3 rdoc$B11 avtz D0 v a"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_t"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_z"
## [1] "LDDM_do3_rdoc$B11_avt_v"
## [1] "LDDM_do3_rdoc$B11_avt_a"
## [1] "LDDM_do3_rdoc$B11_avt_t"
## [1] "LDDM_do3_rdoc$B11_avtz_v"
## [1] "LDDM_do3_rdoc$B11_avtz_a"
## [1] "LDDM_do3_rdoc$B11_avtz_t"
## [1] "LDDM_do3_rdoc$B11_avtz_z"
## [1] "LDDM_do3_rdoc$B11_avt_D0_v_v"
## [1] "LDDM_do3_rdoc$B11_avt_D0_v_v_con"
## [1] "LDDM do3 rdoc$B11 avt DO v v incon"
## [1] "LDDM do3 rdoc$B11 avt DO v a"
## [1] "LDDM_do3_rdoc$B11_avt_D0_v_t"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v_con"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_v_incon"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_a"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_t"
## [1] "LDDM_do3_rdoc$B11_avtz_D0_v_z"
## [1] "LDDM_do3_rdoc$flanker_score_rdoc"
## [1] "LDDM_do3_rdoc$FCz_ERN_080"
## [1] "LDDM_do3_rdoc$FCz_ERN_080_diff"
## [1] "LDDM_do3_rdoc$Inhib_time"
## [1] "LDDM do3 rdoc$exec composite"
```

[1] "LDDM_do3_rdoc\$Pred_FSIQ"
[1] "LDDM_do3_rdoc\$Motor_time"
[1] "LDDM_do3_rdoc\$accuracy"

[1] "LDDM_do3_rdoc\$accuracy_congruent_log"
[1] "LDDM do3 rdoc\$accuracy incongruent"

```
make_CI <- function(lower, upper) {
  paste0("[", round(lower,2), ", ", round(upper,2), "]")
}</pre>
```

Multiple regressions

```
i=1
mr_ERN_table_rdoc <- as.data.frame(matrix(nrow=6, ncol=4))</pre>
mr_ERN_table_rdoc[,1] <- c("Drift rate", "Drift rate (congruent)", "Drift rate (incongruent)", "Boundary s
mr_ERN_table_rdoc_raw <- as.data.frame(matrix(nrow=6, ncol=4))</pre>
mr_ERN_table_rdoc_raw[,1] <- c("Drift rate", "Drift rate (congruent)", "Drift rate (incongruent)", "Bounda
for (var in c("B11_avtz_D0_v_v_z","B11_avtz_D0_v_v_con_z","B11_avtz_D0_v_v_incon_z","B11_avtz_D0_v_a_z"
  eval(parse(text=paste0("
             model <- lmer(FCz_ERN_080_z ~ ",var," + flanker_score_rdoc_z + accuracy_incongruent_z + Ag</pre>
             mr_ERN_table_rdoc[i,2] <- paste0(round(tidy(model)$estimate[2],2), ifelse(tidy(model)$p.va
                                               make_CI(confint(model)[4,1], confint(model)[4,2]))
             mr_ERN_table_rdoc[i,3] <- paste0(round(tidy(model) $estimate[3],2), ifelse(tidy(model) $p.va</pre>
                                               make_CI(confint(model)[5,1], confint(model)[5,2]))
             mr ERN table rdoc[i,4] <- pasteO(round(tidy(model)$estimate[4],2), ifelse(tidy(model)$p.va
                                               make_CI(confint(model)[6,1], confint(model)[6,2]))
            mr_ERN_table_rdoc_raw[i,2] <- tidy(model)$estimate[2]</pre>
            mr_ERN_table_rdoc_raw[i,3] <- confint(model)[4,1]</pre>
            mr_ERN_table_rdoc_raw[i,4] <- confint(model)[4,2]</pre>
            mr_ERN_table_rdoc_raw[i,5] <- tidy(model)$estimate[3]</pre>
            mr_ERN_table_rdoc_raw[i,6] <- confint(model)[5,1]</pre>
            mr_ERN_table_rdoc_raw[i,7] <- confint(model)[5,2]</pre>
            mr_ERN_table_rdoc_raw[i,8] <- tidy(model)$estimate[4]</pre>
            mr_ERN_table_rdoc_raw[i,9] <- confint(model)[6,1]</pre>
            mr_ERN_table_rdoc_raw[i,10] <- confint(model)[6,2]</pre>
                                     ")))
  i=i+1
}
## Computing profile confidence intervals ...
## Computing profile confidence intervals ...
## Computing profile confidence intervals ...
```

```
## Computing profile confidence intervals ...
```

```
## Computing profile confidence intervals ...
## Computing profile confidence intervals \dots
## Computing profile confidence intervals ...
```

```
## Computing profile confidence intervals ...
i=1
mr_ERN_table_rdoc_congruent <- mr_ERN_table_rdoc</pre>
mr_ERN_table_rdoc_congruent_raw <- mr_ERN_table_rdoc_raw</pre>
for (var in c("B11_avtz_D0_v_v_z","B11_avtz_D0_v_v_con_z","B11_avtz_D0_v_v_incon_z","B11_avtz_D0_v_a_z"
  eval(parse(text=paste0("
              model <- lmer(FCz_ERN_080_z ~ ",var," + flanker_score_rdoc_z + accuracy_congruent_log_z +</pre>
              mr_ERN_table_rdoc_congruent[i,2] <- pasteO(round(tidy(model)$estimate[2],2), ifelse(tidy(model)$estimate[2],2)</pre>
                                                make_CI(confint(model)[4,1], confint(model)[4,2]))
              mr_ERN_table_rdoc_congruent[i,3] <- pasteO(round(tidy(model)$estimate[3],2), ifelse(tidy(model)$estimate[3],2)</pre>
                                                make_CI(confint(model)[5,1], confint(model)[5,2]))
              mr_ERN_table_rdoc_congruent[i,4] <- pasteO(round(tidy(model)$estimate[4],2), ifelse(tidy(model)$estimate[4],2)</pre>
                                                make_CI(confint(model)[6,1], confint(model)[6,2]))
             mr_ERN_table_rdoc_congruent_raw[i,2] <- tidy(model)$estimate[2]</pre>
             mr_ERN_table_rdoc_congruent_raw[i,3] <- confint(model)[4,1]</pre>
             mr_ERN_table_rdoc_congruent_raw[i,4] <- confint(model)[4,2]</pre>
             mr_ERN_table_rdoc_congruent_raw[i,5] <- tidy(model)$estimate[3]</pre>
             mr_ERN_table_rdoc_congruent_raw[i,6] <- confint(model)[5,1]</pre>
             mr_ERN_table_rdoc_congruent_raw[i,7] <- confint(model)[5,2]</pre>
             mr_ERN_table_rdoc_congruent_raw[i,8] <- tidy(model)$estimate[4]</pre>
             mr_ERN_table_rdoc_congruent_raw[i,9] <- confint(model)[6,1]</pre>
             mr_ERN_table_rdoc_congruent_raw[i,10] <- confint(model)[6,2]</pre>
                                       ")))
  i=i+1
}
## Computing profile confidence intervals ...
```

```
## Computing profile confidence intervals ...
```

```
## Computing profile confidence intervals ...
## Computing profile confidence intervals \dots
## Computing profile confidence intervals ...
```

```
colnames(mr_ERN_table_rdoc) <- c("Parameters", "Drift rate", "NIH Toolbox", "Raw accuracy")
colnames(mr_ERN_table_rdoc_congruent) <- c("Parameters", "Drift rate", "NIH Toolbox", "Raw accuracy (congruent)
write.csv(mr_ERN_table_rdoc, here("./work/tables/mr_ERN_table_rdoc.csv"))
write.csv(mr_ERN_table_rdoc_raw, here("./work/tables/mr_ERN_table_rdoc_raw.csv"))
write.csv(mr_ERN_table_rdoc_congruent, here("./work/tables/mr_ERN_table_rdoc_congruent.csv"))
write.csv(mr_ERN_table_rdoc_congruent_raw, here("./work/tables/mr_ERN_table_rdoc_congruent_raw.csv"))</pre>
```

Convergence with neuropsych measures

Inhibition & DDM

```
LDDM do3 rdoc no outliers$NL time rev <- -1*LDDM do3 rdoc no outliers$NL time
LDDM_do3_rdoc_no_outliers$Inhib_time_rev <- -1*LDDM_do3_rdoc_no_outliers$Inhib_time
LDDM_do3_rdoc_no_outliers$Inhib_time_rev_z <- scale(LDDM_do3_rdoc_no_outliers$Inhib_time_rev, center=T,
i=1
mr_INHIB_table_rdoc_nocov <- as.data.frame(matrix(nrow=6, ncol=4))</pre>
mr_INHIB_table_rdoc_nocov[,1] <- c("Drift rate", "Drift rate (congruent)", "Drift rate (incongruent)", "Bo
mr_INHIB_table_rdoc_cov <- as.data.frame(matrix(nrow=6, ncol=4))</pre>
mr_INHIB_table_rdoc_cov[,1] <- c("Drift rate","Drift rate (congruent)","Drift rate (incongruent)","Bound</pre>
mr_INHIB_table_rdoc_cov_raw <- as.data.frame(matrix(nrow=6, ncol=4))</pre>
mr_INHIB_table_rdoc_cov_raw[,1] <- c("Drift rate", "Drift rate (congruent)", "Drift rate (incongruent)", "]</pre>
for (var in c("B11_avtz_D0_v_v_z","B11_avtz_D0_v_v_con_z","B11_avtz_D0_v_v_incon_z","B11_avtz_D0_v_a_z"
  eval(parse(text=paste0("
             model <- lmer(Inhib_time_rev_z ~ ",var," + flanker_score_rdoc_z + accuracy_incongruent_z +</pre>
             mr INHIB table rdoc nocov[i,2] <- pasteO(round(tidy(model)$estimate[2],2), ifelse(tidy(mod
                                               make_CI(confint(model)[4,1], confint(model)[4,2]))
             mr_INHIB_table_rdoc_nocov[i,3] <- pasteO(round(tidy(model)$estimate[3],2), ifelse(tidy(mod</pre>
                                               make_CI(confint(model)[5,1], confint(model)[5,2]))
             mr_INHIB_table_rdoc_nocov[i,4] <- pasteO(round(tidy(model)$estimate[4],2), ifelse(tidy(mod</pre>
                                               make_CI(confint(model)[6,1], confint(model)[6,2]))
             model_cov <- lmer(Inhib_time_rev_z ~ ",var," + flanker_score_rdoc_z + accuracy_incongruent</pre>
             mr_INHIB_table_rdoc_cov[i,2] <- pasteO(round(tidy(model_cov)$estimate[2],2), ifelse(tidy(model_cov)$estimate[2],2)</pre>
                                               make_CI(confint(model_cov)[4,1], confint(model_cov)[4,2]))
             mr_INHIB_table_rdoc_cov[i,3] <- pasteO(round(tidy(model_cov)$estimate[3],2), ifelse(tidy(model_cov)$estimate[3],2)</pre>
                                               make_CI(confint(model_cov)[5,1], confint(model_cov)[5,2]))
             mr_INHIB_table_rdoc_cov[i,4] <- pasteO(round(tidy(model_cov)$estimate[4],2), ifelse(tidy(model_cov)$estimate[4],2)</pre>
                                               make_CI(confint(model_cov)[6,1], confint(model_cov)[6,2]))
             mr_INHIB_table_rdoc_cov_raw[i,2] <- tidy(model_cov)$estimate[2]</pre>
             mr INHIB table rdoc cov raw[i,3] <- confint(model cov)[4,1]
             mr_INHIB_table_rdoc_cov_raw[i,4] <- confint(model_cov)[4,2]</pre>
```

```
mr_INHIB_table_rdoc_cov_raw[i,5] <- tidy(model_cov)$estimate[3]
mr_INHIB_table_rdoc_cov_raw[i,6] <- confint(model_cov)[5,1]
mr_INHIB_table_rdoc_cov_raw[i,7] <- confint(model_cov)[5,2]

mr_INHIB_table_rdoc_cov_raw[i,8] <- tidy(model_cov)$estimate[4]
mr_INHIB_table_rdoc_cov_raw[i,9] <- confint(model_cov)[6,1]
mr_INHIB_table_rdoc_cov_raw[i,10] <- confint(model_cov)[6,2]

")))
i=i+1
}</pre>
```

```
## Computing profile confidence intervals ...
```

```
## Computing profile confidence intervals ...
## Computing profile confidence intervals \dots
## Computing profile confidence intervals ...
```

```
## Computing profile confidence intervals ...
## Computing profile confidence intervals \dots
## Computing profile confidence intervals ...
## Computing profile confidence intervals ...
## Computing profile confidence intervals \dots
## Computing profile confidence intervals ...
## Computing profile confidence intervals ...
## Computing profile confidence intervals ...
colnames(mr_INHIB_table_rdoc_nocov) <- c("Parameters", "Drift rate", "NIH Toolbox", "Raw accuracy (incon</pre>
colnames(mr_INHIB_table_rdoc_cov) <- c("Parameters", "Drift rate", "NIH Toolbox", "Raw accuracy (incongr
write.csv(mr_INHIB_table_rdoc_nocov, here("./work/tables/mr_INHIB_table_rdoc_nocov.csv"))
write.csv(mr_INHIB_table_rdoc_cov, here("./work/tables/mr_INHIB_table_rdoc_cov.csv"))
write.csv(mr_INHIB_table_rdoc_cov_raw, here("./work/tables/mr_INHIB_table_rdoc_cov_raw.csv"))
```

Set-shifting & DDM

```
# lapply(LDDM_do3_rdoc, attr, "label")
LDDM_do3_rdoc_no_outliers$NL_time_rev <- -1*LDDM_do3_rdoc_no_outliers$NL_time
LDDM_do3_rdoc_no_outliers$InSw_time_rev <- -1*LDDM_do3_rdoc_no_outliers$InSw_time
LDDM_do3_rdoc_no_outliers$exec_composite <- rowMeans(LDDM_do3_rdoc_no_outliers[,c("FF_cor","Switch_cor"
LDDM_do3_rdoc_no_outliers$exec_composite_z <- scale(LDDM_do3_rdoc_no_outliers$exec_composite, center=T,
i=1
mr_EXEC_table_rdoc_nocov <- as.data.frame(matrix(nrow=6, ncol=4))</pre>
mr_EXEC_table_rdoc_nocov[,1] <- c("Drift rate", "Drift rate (congruent)", "Drift rate (incongruent)", "Bou
mr_EXEC_table_rdoc_cov <- as.data.frame(matrix(nrow=6, ncol=4))</pre>
mr_EXEC_table_rdoc_cov[,1] <- c("Drift rate", "Drift rate (congruent)", "Drift rate (incongruent)", "Bound
mr_EXEC_table_rdoc_cov_raw <- as.data.frame(matrix(nrow=6, ncol=4))</pre>
mr_EXEC_table_rdoc_cov_raw[,1] <- c("Drift rate", "Drift rate (congruent)", "Drift rate (incongruent)", "B
for (var in c("B11_avtz_D0_v_v_z","B11_avtz_D0_v_v_con_z","B11_avtz_D0_v_v_incon_z","B11_avtz_D0_v_a_z"
  eval(parse(text=paste0("
             model <- lmer(exec_composite_z ~ ",var," + flanker_score_rdoc_z + accuracy_incongruent_z +</pre>
             mr_EXEC_table_rdoc_nocov[i,2] <- paste0(round(tidy(model)$estimate[2],2), ifelse(tidy(model)$)</pre>
                                              make_CI(confint(model)[4,1], confint(model)[4,2]))
             mr_EXEC_table_rdoc_nocov[i,3] <- paste0(round(tidy(model)$estimate[3],2), ifelse(tidy(model)$)</pre>
                                              make_CI(confint(model)[5,1], confint(model)[5,2]))
             mr_EXEC_table_rdoc_nocov[i,4] <- paste0(round(tidy(model)$estimate[4],2), ifelse(tidy(model)$)</pre>
                                              make_CI(confint(model)[6,1], confint(model)[6,2]))
             model_cov <- lmer(exec_composite_z ~ ",var," + flanker_score_rdoc_z + accuracy_z + Pred_FS</pre>
             mr_EXEC_table_rdoc_cov[i,2] <- paste0(round(tidy(model_cov)$estimate[2],2), ifelse(tidy(model_cov)$)</pre>
```

```
make_CI(confint(model_cov)[4,1], confint(model_cov)[4,2]))
             mr_EXEC_table_rdoc_cov[i,3] <- paste0(round(tidy(model_cov)$estimate[3],2), ifelse(tidy(model_cov)$)</pre>
                                              make_CI(confint(model_cov)[5,1], confint(model_cov)[5,2]))
             mr_EXEC_table_rdoc_cov[i,4] <- paste0(round(tidy(model_cov)$estimate[4],2), ifelse(tidy(model_cov)$)</pre>
                                              make_CI(confint(model_cov)[6,1], confint(model_cov)[6,2]))
            mr_EXEC_table_rdoc_cov_raw[i,2] <- tidy(model_cov)$estimate[2]</pre>
            mr EXEC table rdoc cov raw[i,3] <- confint(model cov)[4,1]
            mr_EXEC_table_rdoc_cov_raw[i,4] <- confint(model_cov)[4,2]</pre>
            mr_EXEC_table_rdoc_cov_raw[i,5] <- tidy(model_cov)$estimate[3]</pre>
            mr_EXEC_table_rdoc_cov_raw[i,6] <- confint(model_cov)[5,1]</pre>
            mr_EXEC_table_rdoc_cov_raw[i,7] <- confint(model_cov)[5,2]</pre>
            mr_EXEC_table_rdoc_cov_raw[i,8] <- tidy(model_cov)$estimate[4]</pre>
            mr_EXEC_table_rdoc_cov_raw[i,9] <- confint(model_cov)[6,1]</pre>
            mr_EXEC_table_rdoc_cov_raw[i,10] <- confint(model_cov)[6,2]</pre>
                                     ")))
  i=i+1
}
## Computing profile confidence intervals ...
## Computing profile confidence intervals \dots
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in confint.thpr(pp, level = level, zeta = zeta): bad spline fit for
## .sig01: falling back to linear interpolation
## Computing profile confidence intervals ...
```

```
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
```

```
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
## Warning in FUN(X[[i]], ...): non-monotonic profile for .sig01
## Warning in FUN(X[[i]], ...): bad spline fit for .sig01: falling back to linear
## interpolation
## Computing profile confidence intervals ...
```

```
## Computing profile confidence intervals ...
## Computing profile confidence intervals \dots
## Computing profile confidence intervals ...
## Computing profile confidence intervals ...
colnames(mr_EXEC_table_rdoc_nocov) <- c("Parameters", "Drift rate", "NIH Toolbox", "Raw accuracy (incong</pre>
colnames(mr_EXEC_table_rdoc_cov) <- c("Parameters", "Drift rate", "NIH Toolbox", "Raw accuracy (incongru
write.csv(mr EXEC table rdoc nocov, here("./work/tables/mr Exec table rdoc nocov.csv"))
write.csv(mr_EXEC_table_rdoc_cov, here("./work/tables/mr_Exec_table_rdoc_cov.csv"))
write.csv(mr_EXEC_table_rdoc_cov_raw, here("./work/tables/mr_EXEC_table_rdoc_cov_raw.csv"))
```

Heritability

```
LDDM_do3_rdoc_sib <- LDDM_do3_rdoc %>%
  pivot_wider(id_cols=F_ID, names_from=S_ID,
              values_from=c(B11_avtz_D0_v_v_z, B11_avtz_D0_v_v_con_z, B11_avtz_D0_v_v_incon_z,
                            B11_avtz_D0_v_a_z,B11_avtz_D0_v_t_z,B11_avtz_D0_v_z_z,flanker_score_rdoc_z,
LDDM_heritability_table <- data.frame(parameters= c("v", "v_congruent", "v_incongruent", "a", "t", "z",
                           ICC=c(icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_v_z_1","B11_avtz_D0_v_v_z_2")])
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_v_con_z_1","B11_avtz_D0_v_v_con_z_2")]
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_v_incon_z_1","B11_avtz_D0_v_v_incon_z
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_a_z_1","B11_avtz_D0_v_a_z_2")])$value
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_t_z_1","B11_avtz_D0_v_t_z_2")])$value
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_z_z_1","B11_avtz_D0_v_z_z_2")])$value
                           icc(LDDM_do3_rdoc_sib[c("flanker_score_rdoc_z_1","flanker_score_rdoc_z_2")])
                           icc(LDDM_do3_rdoc_sib[c("accuracy_incongruent_z_1","accuracy_incongruent_z_2
                           icc(LDDM_do3_rdoc_sib[c("accuracy_z_1", "accuracy_z_2")])$value),
                           p=c(icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_v_z_1","B11_avtz_D0_v_v_z_2")])$p
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_v_con_z_1","B11_avtz_D0_v_v_con_z_2")
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_v_incon_z_1","B11_avtz_D0_v_v_incon_z
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_a_z_1","B11_avtz_D0_v_a_z_2")])$p.val
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_t_z_1","B11_avtz_D0_v_t_z_2")])$p.val
                           icc(LDDM_do3_rdoc_sib[c("B11_avtz_D0_v_z_z_1","B11_avtz_D0_v_z_z_2")])$p.val
                           icc(LDDM_do3_rdoc_sib[c("flanker_score_rdoc_z_1","flanker_score_rdoc_z_2")])
                           icc(LDDM_do3_rdoc_sib[c("accuracy_incongruent_z_1","accuracy_incongruent_z_2
                           icc(LDDM_do3_rdoc_sib[c("accuracy_z_1", "accuracy_z_2")])$p.value))
write.csv(LDDM_heritability_table, here("./work/tables/rdoc_heritability_table.csv"))
```

Split half reliability (Spearman-Brown prophecy)

```
load(file=here("./work/data/LDDM_cleaning04_fullbeh_rdoc.RData"))

LDDM_cleaning04_rdoc_reliability <- LDDM_cleaning04_fullbeh_rdoc %>%
    group_by(subj_idx) %>%
    mutate(block = c(rep(1,30),rep(2,30),rep(3,30),rep(4,30),rep(5,30),rep(6,30),rep(7,30),rep(8,30),rep(group_by(subj_idx) %>%
    mutate(trial = 1:330)

for (s in seq(15,166,15)){
    eval(parse(text=paste0('
LDDM_first',s,'_1 <- LDDM_cleaning04_rdoc_reliability %>% filter(trial<=',s,')
LDDM_second',s,'_1 <- LDDM_cleaning04_rdoc_reliability %>% filter(trial< ',(s*2)+1,' & trial>',s,')

LDDM_first',s,'_2 <- LDDM_first',s,'_1 %>%
    group_by(subj_idx) %>% # per subject
    summarise(accuracy_score = sum(response==1)*(5/length(trial))) # accuracy score per NIH Toolbox manua
LDDM_second',s,'_2 <- LDDM_second',s,'_1 %>%
    group_by(subj_idx) %>%
```

```
summarise(accuracy_score = sum(response==1)*(5/length(trial))) # accuracy score per NIH Toolbox manua
LDDM_first',s,'_3 <- LDDM_first',s,'_1 %>%
  group_by(subj_idx) %>% # per subject
  filter(stim=="incongruent" & response==1) %>% # incongruent trials with correct response
  mutate(mean rt = mean(rt),
         sd_rt = sd(rt)) %>% # compute individual mean and sd RT for use below
  filter(rt>=0.1 & rt>(mean rt - 3*sd rt) & rt<(mean rt + 3*sd rt)) %>% # remove trials less than 100ms
  summarise(med_rt = median(rt)*1000) %>% # compute individual level median RT
  mutate(rt_score = 5-(5*((log(med_rt)-log(250))/(log(1000)-log(250)))))) # compute RT score to go into
LDDM_second',s,'_3 <- LDDM_second',s,'_1 %>%
  group_by(subj_idx) %>% # per subject
  filter(stim=="incongruent" & response==1) %>% # incongruent trials with correct response
  mutate(mean_rt = mean(rt),
         sd_rt = sd(rt)) %>% # compute individual mean and sd RT for use below
  filter(rt>=0.1 & rt>(mean_rt - 3*sd_rt) & rt<(mean_rt + 3*sd_rt)) %>% # remove trials less than 100ms
  summarise(med_rt = median(rt)*1000) %>% # compute individual level median RT
  mutate(rt_score = 5-(5*((log(med_rt)-log(250))/(log(1000)-log(250)))))) # compute RT score to go into
LDDM_first',s,' <- LDDM_first',s,'_1 %>%
  full_join(LDDM_first',s,'_2, by = "subj_idx") %>%
  full_join(LDDM_first',s,'_3, by="subj_idx") %>%
  group_by(subj_idx) %>% # per subject
  mutate(total_accuracy_perc = sum(response==1)/length(response)) %>% # compute total accuracy percenta
  summarise(flanker_score_list_rdoc = if_else(total_accuracy_perc>=0.8, accuracy_score+rt_score, accura
            accuracy = mean(total_accuracy_perc)) %>% # if accuracy is above 80% then add accuracy and
  transmute(ID=subj_idx, flanker_score_list_rdoc=flanker_score_list_rdoc, accuracy=accuracy) %>%
  group_by(ID) %>% # per subject
  summarise(flanker_score_rdoc = mean(flanker_score_list_rdoc),
           accuracy = mean(accuracy))
LDDM_second',s,' <- LDDM_second',s,'_1 %>%
  full_join(LDDM_second',s,'_2, by = "subj_idx") %>%
  full_join(LDDM_second',s,'_3, by="subj_idx") %>%
  group_by(subj_idx) %>% # per subject
  mutate(total_accuracy_perc = sum(response==1)/length(response)) %>% # compute total accuracy percenta
  summarise(flanker_score_list_rdoc = if_else(total_accuracy_perc>=0.8, accuracy_score+rt_score, accura
            accuracy = mean(total_accuracy_perc)) %>% # if accuracy is above 80% then add accuracy and
  transmute(ID=subj_idx, flanker_score_list_rdoc=flanker_score_list_rdoc, accuracy=accuracy) %>%
  group_by(ID) %>% # per subject
  summarise(flanker_score_rdoc = mean(flanker_score_list_rdoc),
           accuracy = mean(accuracy))
LDDM_first_accuracy',s,'<- LDDM_first',s,'_1 %>%
  full_join(LDDM_first',s,'_2, by = "subj_idx") %>%
  full_join(LDDM_first',s,'_3, by="subj_idx") %>%
  group_by(subj_idx, stim) %>% # per subject and condition
  mutate(accuracy_by_stim = sum(response==1)/length(response)) %>%
  summarise(accuracy_by_stim=mean(accuracy_by_stim)) %>%
  pivot_wider(names_from=stim, values_from=accuracy_by_stim, id_cols=subj_idx, names_prefix="accuracy_"
LDDM_second_accuracy',s,'<- LDDM_second',s,'_1 %>%
  full_join(LDDM_second',s,'_2, by = "subj_idx") %>%
  full_join(LDDM_second',s,'_3, by="subj_idx") %>%
  group_by(subj_idx, stim) %>% # per subject and condition
```

```
mutate(accuracy_by_stim = sum(response==1)/length(response)) %>%
  summarise(accuracy_by_stim=mean(accuracy_by_stim)) %>%
  pivot_wider(names_from=stim, values_from=accuracy_by_stim, id_cols=subj_idx, names_prefix="accuracy_"
r_half_',s,' <- cor(LDDM_first',s,'$flanker_score_rdoc, LDDM_second',s,'$flanker_score_rdoc, method="sp
r_half_ci_',s,' <- ci_cor(LDDM_first',s,'$flanker_score_rdoc, LDDM_second',s,'$flanker_score_rdoc, meth
r_sb_cilower_',s,' <- (2*r_half_ci_',s,'$interval[1])/(1+r_half_ci_',s,'$interval[1])
r sb ciupper ',s,' <- (2*r half ci ',s,'$interval[2])/(1+r half ci ',s,'$interval[2])
r_sb_',s,' <- (2*r_half_',s,')/(1+r_half_',s,')
racc_half_',s,' <- cor(LDDM_first_accuracy',s,'$accuracy_incongruent, LDDM_second_accuracy',s,'$accuracy
racc_half_ci_',s,' <- ci_cor(LDDM_first_accuracy',s,'$accuracy_incongruent, LDDM_second_accuracy',s,'$a</pre>
racc_sb_cilower_',s,' <- (2*racc_half_ci_',s,'$interval[1])/(1+racc_half_ci_',s,'$interval[1])</pre>
racc_sb_ciupper_',s,' <- (2*racc_half_ci_',s,'$interval[2])/(1+racc_half_ci_',s,'$interval[2])</pre>
racc_sb_',s,' <- (2*racc_half_',s,')/(1+racc_half_',s,')</pre>
')))
}
## Warning: Returning more (or less) than 1 row per 'summarise()' group was deprecated in dplyr 1.1.0.
## i Please use 'reframe()' instead.
## i When switching from 'summarise()' to 'reframe()', remember that 'reframe()' always returns an
   ungrouped data frame and adjust accordingly.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was generated.
## 'summarise()' has grouped output by 'subj_idx'. You can override using the
## '.groups' argument.
## Warning: Returning more (or less) than 1 row per 'summarise()' group was deprecated in dplyr 1.1.0.
## i Please use 'reframe()' instead.
## i When switching from 'summarise()' to 'reframe()', remember that 'reframe()' always returns an
   ungrouped data frame and adjust accordingly.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was generated.
## 'summarise()' has grouped output by 'subj_idx'. You can override using the '.groups' argument.
## 'summarise()' has grouped output by 'subj idx'. You can override using the '.groups' argument.
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## Warning: Returning more (or less) than 1 row per 'summarise()' group was deprecated in dplyr 1.1.0.
## i Please use 'reframe()' instead.
## i When switching from 'summarise()' to 'reframe()', remember that 'reframe()' always returns an
   ungrouped data frame and adjust accordingly.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was generated.
## 'summarise()' has grouped output by 'subj_idx'. You can override using the
## '.groups' argument.
## Warning: Returning more (or less) than 1 row per 'summarise()' group was deprecated in dplyr 1.1.0.
## i Please use 'reframe()' instead.
## i When switching from 'summarise()' to 'reframe()', remember that 'reframe()' always returns an
   ungrouped data frame and adjust accordingly.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was generated.
```

```
## 'summarise()' has grouped output by 'subj_idx'. You can override using the '.groups' argument.
## 'summarise()' has grouped output by 'subj_idx'. You can override using the '.groups' argument.
## 'summarise()' has grouped output by 'subj idx'. You can override using the '.groups' argument.
## Warning: Returning more (or less) than 1 row per 'summarise()' group was deprecated in dplyr 1.1.0.
## i Please use 'reframe()' instead.
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## ungrouped data frame and adjust accordingly.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was generated.
## 'summarise()' has grouped output by 'subj_idx'. You can override using the
## '.groups' argument.
## Warning: Returning more (or less) than 1 row per 'summarise()' group was deprecated in dplyr 1.1.0.
## i Please use 'reframe()' instead.
## i When switching from 'summarise()' to 'reframe()', remember that 'reframe()' always returns an
## ungrouped data frame and adjust accordingly.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was generated.
## 'summarise()' has grouped output by 'subj_idx'. You can override using the '.groups' argument.
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spearman_brown_rdoc <- data.frame(trials=seq(15,165,15),</pre>
                              rnih=rep(NA,length(seq(15,165,15))),
                              rnih_cilower=rep(NA,length(seq(15,165,15))),
                              rnih_ciupper=rep(NA,length(seq(15,165,15))),
                             racc=rep(NA,length(seq(15,165,15))),
                              racc_cilower=rep(NA,length(seq(15,165,15))),
                             racc_ciupper=rep(NA,length(seq(15,165,15))))
i=1
for (s in seq(15,165,15)){
  eval(parse(text=paste0('spearman_brown_rdoc[i,2] <- round(as.numeric(r_sb_',s,'),3)</pre>
                         spearman_brown_rdoc[i,3] <- round(as.numeric(r_sb_cilower_',s,'),3)</pre>
                          spearman_brown_rdoc[i,4] <- round(as.numeric(r_sb_ciupper_',s,'),3)</pre>
                          spearman_brown_rdoc[i,5] <- round(as.numeric(racc_sb_',s,'),3)</pre>
                          spearman_brown_rdoc[i,6] <- round(as.numeric(racc_sb_cilower_',s,'),3)</pre>
                          spearman_brown_rdoc[i,7] <- round(as.numeric(racc_sb_ciupper_',s,'),3)')))</pre>
 i=i+1
}
sttr_rdpc_ddm_splithalf <- read.csv(here("./Split_Half/RDoC_splithalf.csv"))</pre>
spearman_brown_rdoc$rdriftcon <- sttr_rdpc_ddm_splithalf$avtz_D0_v_vcon</pre>
spearman_brown_rdoc$rdriftincon <- sttr_rdpc_ddm_splithalf$avtz_D0_v_vinc</pre>
spearman_brown_rdoc$rboundary_separation <- sttr_rdpc_ddm_splithalf$avtz_D0_v_a</pre>
# qqplot(spearman_brown_rdoc, aes(x=trials, y=r, qroup=1)) +
  geom line() +
#
   geom_point() +
  scale y continuous(limits = c(0, 1)) +
  geom\_ribbon(aes(ymin = r\_ci\_lower, ymax = r\_ci\_upper), alpha = 0.2)
```

Correlations

ERN amplitude, Neuropsych, & DDM

```
all_measures_rdoc <- c("FCz_ERN_080_z","flanker_score_rdoc_z","accuracy_z","accuracy_incongruent_z","ac
                  "B11_avtz_D0_v_v_z", "B11_avtz_D0_v_v_con_z", "B11_avtz_D0_v_v_incon_z", "B11_avtz_D0_v_
rdoc_correlations <- correlation_matrix(LDDM_do3_rdoc_no_outliers[c(all_measures_rdoc)], type = c("spea
## Loading required package: Hmisc
## Registered S3 methods overwritten by 'Hmisc':
##
    method
##
     [.labelled
                            expss
##
    print.labelled
##
     as.data.frame.labelled expss
## Attaching package: 'Hmisc'
## The following objects are masked from 'package:DescTools':
##
##
       %nin%, Label, Mean, Quantile
## The following object is masked from 'package:psych':
##
##
       describe
## The following objects are masked from 'package:dplyr':
##
##
       src, summarize
## The following objects are masked from 'package:base':
##
##
       format.pval, units
write.csv(rdoc_correlations, file=here("./work/tables/rdoc_correlations.csv"))
# qqplot(filter(LDDM_do3_rdoc,FCz_ERN_080>-3 & FCz_ERN_080<3), aes(x=FCz_ERN_080_z, y=B11_avtz_D0_v_v_z
  geom_point() +
  stat_smooth(method="lm")
```

SOBP Stats

```
sum(!is.na(LDDM_do3_rdoc_no_outliers$B11_avtz_D0_v_v_z))
## [1] 381
```

```
summary(lmer(FCz_ERN_080_z ~ B11_avtz_D0_v_v_con_z + flanker_score_rdoc_z + accuracy_z + (1 | F_ID), LD
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: FCz_ERN_080_z ~ B11_avtz_D0_v_v_con_z + flanker_score_rdoc_z +
##
       accuracy_z + (1 | F_ID)
##
     Data: LDDM_do3_rdoc_no_outliers
##
## REML criterion at convergence: 1074.6
##
## Scaled residuals:
##
      Min 1Q Median
                               3Q
                                      Max
## -3.0111 -0.6375 0.0039 0.5331 2.6529
##
## Random effects:
## Groups Name
                        Variance Std.Dev.
## F_ID
            (Intercept) 0.03074 0.1753
## Residual
                        0.93613 0.9675
## Number of obs: 379, groups: F_ID, 230
##
## Fixed effects:
##
                         Estimate Std. Error
                                                    df t value Pr(>|t|)
## (Intercept)
                          0.01843
                                     0.05113 207.50333
                                                        0.360 0.718859
## B11_avtz_D0_v_v_con_z 0.19612
                                     0.05596 374.63952
                                                         3.505 0.000513 ***
                          0.02694
                                     0.06319 371.73516
## flanker_score_rdoc_z
                                                         0.426 0.670117
                          0.04420
                                     0.05854 374.92012
## accuracy_z
                                                         0.755 0.450746
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
             (Intr) B11__D fln___
## B11__D0___ -0.002
## flnkr_scr__ 0.004 -0.379
## accuracy_z -0.005 0.017 -0.473
summary(lmer(FCz_ERN_080_z ~ B11_avtz_D0_v_v_incon_z + flanker_score_rdoc_z + accuracy_z + (1 | F_ID),
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: FCz_ERN_080_z ~ B11_avtz_D0_v_v_incon_z + flanker_score_rdoc_z +
      accuracy_z + (1 | F_ID)
##
      Data: LDDM_do3_rdoc_no_outliers
##
## REML criterion at convergence: 1071
##
## Scaled residuals:
       \mathtt{Min}
                 1Q
                     Median
                                   3Q
                                           Max
## -2.95015 -0.64827 0.01161 0.59062 2.42517
##
## Random effects:
## Groups
                        Variance Std.Dev.
            Name
## F_ID
            (Intercept) 0.06428 0.2535
```

```
0.89720 0.9472
## Residual
## Number of obs: 379, groups: F_ID, 230
## Fixed effects:
                          Estimate Std. Error
                                                    df t value Pr(>|t|)
                          ## (Intercept)
## B11_avtz_D0_v_v_incon_z 0.352207
                                   0.090668 371.500485
                                                       3.885 0.000121 ***
                         ## flanker_score_rdoc_z
## accuracy z
                         -0.174432
                                   0.080129 373.465427 -2.177 0.030116 *
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Correlation of Fixed Effects:
             (Intr) B11__D fln___
## B11__D0____ 0.001
## flnkr_scr__ 0.003 -0.467
## accuracy_z -0.004 -0.686 -0.004
summary(lmer(FCz_ERN_080_z ~ B11_avtz_D0_v_v_z + (1 | F_ID), LDDM_do3_rdoc_no_outliers))
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: FCz ERN 080 z ~ B11 avtz D0 v v z + (1 | F ID)
##
     Data: LDDM_do3_rdoc_no_outliers
##
## REML criterion at convergence: 1063.1
##
## Scaled residuals:
       \mathtt{Min}
           1Q
                   Median
## -3.02131 -0.63300 0.00945 0.57350 2.58669
##
## Random effects:
## Groups Name
                      Variance Std.Dev.
## F_ID
           (Intercept) 0.03184 0.1784
                      0.92044 0.9594
## Residual
## Number of obs: 379, groups: F_ID, 230
## Fixed effects:
##
                    Estimate Std. Error
                                            df t value Pr(>|t|)
## (Intercept)
                    0.01844 0.05078 208.67382 0.363
                                                         0.717
## B11_avtz_D0_v_v_z 0.24422 0.05027 375.29562 4.858 1.74e-06 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Correlation of Fixed Effects:
             (Intr)
## B11_v_D0___ -0.001
```

Save datasets

```
save(LDDM_do3_rdoc_with_bad_data, file=here("./work/data/LDDM_do3_rdoc_with_bad_data.RData"))
save(LDDM_do3_rdoc, file=here("./work/data/LDDM_do3_rdoc.RData"))
save(LDDM_do3_rdoc_no_outliers, file=here("./work/data/LDDM_do3_rdoc_no_outliers.RData"))
save(spearman_brown_rdoc, file=here("./work/data/spearman_brown_rdoc.RData"))
```