Hippocampal subfield strophy in AD progression

Set up

```
packages <- c("here", "readxl", "dplyr", "ggplot2", "psych", "lmerTest",</pre>
           "lme4", "emmeans")
lapply(packages, library, character.only = TRUE)
## [[1]]
## [1] "here"
                                              "grDevices" "utils"
                    "stats"
                                 "graphics"
                                                                        "datasets"
## [7] "methods"
                    "base"
##
## [[2]]
## [1] "readxl"
                    "here"
                                 "stats"
                                              "graphics"
                                                           "grDevices" "utils"
                    "methods"
  [7] "datasets"
                                 "base"
##
##
  [[3]]
    [1] "dplyr"
                     "readxl"
                                  "here"
                                               "stats"
                                                            "graphics"
                                                                         "grDevices"
    [7] "utils"
                     "datasets"
                                  "methods"
                                               "base"
##
##
## [[4]]
##
    [1] "ggplot2"
                     "dplyr"
                                  "readxl"
                                               "here"
                                                            "stats"
                                                                         "graphics"
##
    [7] "grDevices"
                     "utils"
                                  "datasets"
                                               "methods"
                                                            "base"
##
   [[5]]
##
                                               "readxl"
##
    [1] "psych"
                     "ggplot2"
                                  "dplyr"
                                                            "here"
                                                                         "stats"
    [7] "graphics"
                     "grDevices" "utils"
                                               "datasets"
                                                            "methods"
                                                                         "base"
##
## [[6]]
##
    [1] "lmerTest"
                     "lme4"
                                  "Matrix"
                                               "psych"
                                                            "ggplot2"
                                                                         "dplyr"
                                  "stats"
    [7] "readxl"
                     "here"
                                               "graphics"
                                                            "grDevices" "utils"
   [13] "datasets"
                     "methods"
                                  "base"
##
## [[7]]
                     "lme4"
    [1] "lmerTest"
                                  "Matrix"
                                               "psych"
                                                            "ggplot2"
                                                                         "dplyr"
    [7] "readxl"
                     "here"
                                  "stats"
                                                            "grDevices" "utils"
                                               "graphics"
## [13] "datasets"
                     "methods"
                                  "base"
##
## [[8]]
                                  "lme4"
   [1] "emmeans"
                     "lmerTest"
                                               "Matrix"
                                                            "psych"
                                                                         "ggplot2"
   [7] "dplyr"
                     "readxl"
                                               "stats"
##
                                  "here"
                                                            "graphics"
                                                                         "grDevices"
## [13] "utils"
                     "datasets"
                                  "methods"
                                               "base"
```

Load data

```
# load all sample info
all_sample_info <- readxl::read_xlsx("datafile/Total.sample.final.xlsx", sheet = 1)
# load hippocampus subfields data
segHA_lh_AD <- read.table("datafile/hipposubfields_lh_AD.txt", header = T)</pre>
segHA rh AD <- read.table("datafile/hipposubfields rh AD.txt", header = T)</pre>
segHA_lh_CN <- read.table("datafile/hipposubfields.lh.CN.txt", header = T)</pre>
segHA_rh_CN <- read.table("datafile/hipposubfields.rh.CN.txt", header = T)</pre>
segHA_lh_MCI <- read.table("datafile/hipposubfields.lh.MCI.txt", header = T)</pre>
segHA_rh_MCI <- read.table("datafile/hipposubfields.rh.MCI.txt", header = T)</pre>
#View(seqHA_lh_AD)
# function
segHA_sub <- function(dat, gr,LorR) {</pre>
    segHA_tmp <- dat
    segHA_tmp_new <- as.data.frame(matrix(0, dim(segHA_tmp)[1], 1))</pre>
    segHA tmp new$subjid <- segHA tmp$Measure.volume</pre>
    segHA_tmp_new$CA1 <- segHA_tmp$CA1.head + segHA_tmp$CA1.body</pre>
    segHA_tmp_new$CA3 <- segHA_tmp$CA3.head + segHA_tmp$CA3.body</pre>
    segHA_tmp_new$CA4 <- segHA_tmp$CA4.head + segHA_tmp$CA4.body</pre>
    segHA_tmp_new$subiculum <- segHA_tmp$subiculum.head + segHA_tmp$subiculum.body
    segHA_tmp_new$presubiculum <- segHA_tmp$presubiculum.head + segHA_tmp$presubiculum.body
    segHA_tmp_new$GC.ML.DG <- segHA_tmp$GC.ML.DG.head + segHA_tmp$GC.ML.DG.head
    segHA tmp_new$molecular_layer_HP <- segHA_tmp$molecular_layer_HP.head +</pre>
    segHA_tmp$molecular_layer_HP.body
    segHA_tmp_new$parasubiculum <- segHA_tmp$parasubiculum</pre>
    segHA_tmp_new$Hippocampal_tail <- segHA_tmp$Hippocampal_tail</pre>
    segHA_tmp_new$HATA <- segHA_tmp$HATA
    segHA tmp new$fimbria <- segHA tmp$fimbria
    segHA tmp new$hippocampal.fissure <- segHA tmp$hippocampal.fissure
    segHA_tmp_new$Whole_hippocampus <- segHA_tmp$Whole_hippocampus</pre>
    #View(seqHA tmp new)
    colnames(segHA_tmp_new)
    segHA_tmp_new <- segHA_tmp_new[,-1]</pre>
    # merge
    all_data_subfileds <- merge(all_sample_info, segHA_tmp_new,
                        by.x = "Subject.ID", by.y = "subjid")
    # add estimated TIV
    eTIV <- read.table(paste0("datafile/eTIV ",gr,".txt"), header = T)</pre>
    all data subfileds <- merge(all data subfileds, eTIV, by.x = "Subject.ID",
                        by.y = "Subj")
```

```
all_data_subfileds$eTIV_ml <- round(all_data_subfileds$eTIV/1000,2)</pre>
    write.csv(all_data_subfileds, paste0("datafile/all_data_subfileds.",
                         gr,".",LorR,".csv"), row.names = F)
}
# run function
segHA sub(segHA lh AD, "AD", "lh")
segHA_sub(segHA_rh_AD, "AD", "rh")
segHA_sub(segHA_lh_CN,"CN","lh")
segHA_sub(segHA_rh_CN,"CN","rh")
segHA_sub(segHA_lh_MCI, "MCI", "lh")
segHA_sub(segHA_rh_MCI,"MCI","rh")
# organize data
gr <- "AD"
LorR <- "lh"
all_data_subfileds_AD.lh <- read.csv(paste0("datafile/all_data_subfileds.",
                      gr,".",LorR,".csv"))
gr <- "AD"
LorR <- "rh"
all_data_subfileds_AD.rh <- read.csv(paste0("datafile/all_data_subfileds.",</pre>
                      gr,".",LorR,".csv"))
gr <- "CN"
LorR <- "rh"
all_data_subfileds_CN.rh <- read.csv(paste0("datafile/all_data_subfileds.",
                      gr,".",LorR,".csv"))
gr <- "CN"
LorR <- "lh"
all_data_subfileds_CN.lh <- read.csv(paste0("datafile/all_data_subfileds.",
                      gr,".",LorR,".csv"))
gr <- "MCI"
LorR <- "rh"
all_data_subfileds_MCI.rh <- read.csv(paste0("datafile/all_data_subfileds.",
                      gr,".",LorR,".csv"))
gr <- "MCI"
LorR <- "lh"
all_data_subfileds_MCI.lh <- read.csv(paste0("datafile/all_data_subfileds.",
                      gr,".",LorR,".csv"))
#View(all_data_subfileds_AD.lh)
all_data_subfileds <- rbind.data.frame(cbind(all_data_subfileds_AD.lh, LorR = "Left"),
                       cbind(all_data_subfileds_CN.lh, LorR = "Left"),
                       cbind(all_data_subfileds_MCI.lh, LorR = "Left"),
                       cbind(all_data_subfileds_AD.rh, LorR = "Right"),
                       cbind(all_data_subfileds_CN.rh, LorR = "Right"),
                       cbind(all_data_subfileds_MCI.rh, LorR = "Right"))
dim(all_data_subfileds)
```

```
## [1] 758 30
all_data_subfileds$Research.Group <- factor(all_data_subfileds$Research.Group,</pre>
                           levels=c("CN","MCI","AD"))
all_data_subfileds$LorR <- factor(all_data_subfileds$LorR, levels = c("Left", "Right"))</pre>
dim(all_data_subfileds)
## [1] 758 30
writexl::write_xlsx(all_data_subfileds, "datafile/all_data_subfileds.xlsx", col_names = T)
CN converted & non-converted subjects
# all sample
all_data_subfileds_rm <- readxl::read_xlsx("datafile/all_data_subfileds.xlsx", sheet = 1)
dim(all_data_subfileds_rm)
## [1] 758 30
table(all data subfileds rm$Research.Group)
##
## AD CN MCI
## 290 288 180
# CN converted
converted_subj <- read.table("datafile/subj_CN_Converted.txt", header = T)</pre>
dim(converted_subj)
## [1] 28 2
all_data_subfileds_rm$group <- all_data_subfileds_rm$Research.Group
all_data_subfileds_rm$group <- as.character(all_data_subfileds_rm$group)</pre>
all_data_subfileds_rm$group[all_data_subfileds_rm$Research.Group == "CN" &
                    all data subfileds rm$Subject.ID %in%
                    converted_subj$subjid] <- "Converted_CN"</pre>
all_data_subfileds_rm$group[all_data_subfileds_rm$Research.Group == "CN" &
                    !all_data_subfileds_rm$Subject.ID %in%
                    converted_subj$subjid] <- "NonConverted_CN"</pre>
```

table(all_data_subfileds_rm\$group)

```
##
##
               AD
                      Converted CN
                                               MCI NonConverted_CN
               290
##
# period of conversion
summary(converted_subj$period_of_conversion)
##
      Min. 1st Qu. Median
                             Mean 3rd Qu.
                                              Max.
##
      6.00 12.00 24.00
                             36.18 51.00 105.00
sd(converted_subj$period_of_conversion)
## [1] 31.91662
# save file
write.csv(all_data_subfileds_rm, "datafile/all_data_subfileds_convertedCN.csv", row.names = F)
table(all_data_subfileds_rm$group)
##
                      Converted_CN
##
                                               MCI NonConverted CN
               AD
##
               290
                                               180
                                                               232
```

Demographic and clinical information

```
# summary of demographic and clinical info
all_data_subfileds_L <- all_data_subfileds_rm[all_data_subfileds_rm$LorR == "Left",]</pre>
# CN, MCI, and AD groups
mm <- table(all_data_subfileds_L$Research.Group, all_data_subfileds_L$Gender)
mm
##
##
          1 2
##
     AD 61 84
##
     CN 78 66
    MCI 42 48
##
mean(all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "CN"])
## [1] 74.77847
sd(all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "CN"])
## [1] 7.787763
```

```
mean(all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "MCI"])
## [1] 72.55889
sd(all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "MCI"])
## [1] 9.163299
mean(all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "AD"])
## [1] 74.94345
sd(all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "AD"])
## [1] 7.504441
# one way ANOVA
chisq.test(mm)
##
## Pearson's Chi-squared test
##
## data: mm
## X-squared = 4.2944, df = 2, p-value = 0.1168
ano_age <- anova(lm(Age ~ Research.Group, all_data_subfileds_L))</pre>
# CN-NC, CN-C, MCI, and AD
demo_info <- as.data.frame(matrix(0,7,9))</pre>
colnames(demo_info) <- c("CN_NC_mean", "CN_NC_range", "CN_C_mean", "CN_C_range",</pre>
             "MCI_mean", "MCI_range", "AD_mean", "AD_range", "p value")
rownames(demo_info) <- c("Gender", "Age", "Education", "CDR", "MMSE", "MoCA", "MIS")</pre>
# gender: chi-squared test
gender_gr <- table(all_data_subfileds_L$group, all_data_subfileds_L$Gender)</pre>
chisq.test(gender_gr)$p.value
## [1] 0.2309044
demo_info["Gender", 1] <- paste0(gender_gr["NonConverted_CN",2] ,"/",gender_gr["NonConverted_CN",1])</pre>
demo_info["Gender", 3] <- paste0(gender_gr["Converted_CN",2] ,"/",gender_gr["Converted_CN",1])</pre>
demo_info["Gender", 5] <- paste0(gender_gr["MCI",2] ,"/",gender_gr["MCI",1])</pre>
demo_info["Gender", 7] <- paste0(gender_gr["AD",2],"/",gender_gr["AD",1])</pre>
demo_info["Gender", 9] <- round(chisq.test(gender_gr)$p.value,3)</pre>
```

```
# Age, Education, CDR, MMSE, MoCA: mean, sd, range
all_data_subfileds_L$group <- factor(all_data_subfileds_L$group,</pre>
                     levels = c("NonConverted CN", "Converted CN", "MCI", "AD"))
mean_all <- describeBy(all_data_subfileds_L[, c("group","Age","Education",</pre>
                         "Global.CDR", "MMSE.Total.Score",
                         "MoCA.Total", "MoCA.MIS")],
                        group = "group",
                        digits = 3)
demo_info[c("Age","Education","CDR","MMSE","MoCA","MIS"), 1:8] <-</pre>
    cbind(paste0(round(mean_all[[1]]$mean[2:7],2),"±",
             round(mean_all[[1]]$sd[2:7],2)),
          paste0(round(mean_all[[1]]$min[2:7],2),"-",
                 round(mean_all[[1]]$max[2:7],2)),
          paste0(round(mean_all[[2]]$mean[2:7],2),"±",
                 round(mean_all[[2]]$sd[2:7],2)),
          paste0(round(mean_all[[2]]$min[2:7],2),"-",
                 round(mean_all[[2]]$max[2:7],2)),
          paste0(round(mean_all[[3]]$mean[2:7],2),"±",
                 round(mean_all[[3]]$sd[2:7],2)),
          paste0(round(mean_all[[3]]$min[2:7],2),"-",
                 round(mean_all[[3]]$max[2:7],2)),
           paste0(round(mean_all[[4]]$mean[2:7],2),"±",
                 round(mean_all[[4]]$sd[2:7],2)),
          paste0(round(mean_all[[4]]$min[2:7],2),"-",
                 round(mean_all[[4]]$max[2:7],2)))
# one way ANOVA
ano_age <- anova(lm(Age ~ group, all_data_subfileds_L))</pre>
ano_education <- anova(lm(Education ~ group, all_data_subfileds_L))</pre>
ano_MMSE <- anova(lm(MMSE.Total.Score ~ group, all_data_subfileds_L))</pre>
ano_MoCA <- anova(lm(MoCA.Total ~ group, all_data_subfileds_L))</pre>
ano_MIS <- anova(lm(MoCA.MIS ~ group, all_data_subfileds_L))</pre>
demo_info["Age", 9] <- round(ano_age$`Pr(>F)`[1],3)
demo_info["Education", 9] <- round(ano_education$`Pr(>F)`[1],3)
demo_info["MMSE", 9] <- round(ano_MMSE$`Pr(>F)`[1],4)
demo_info["MoCA", 9] <- round(ano_MoCA$`Pr(>F)`[1],4)
demo_info["MIS", 9] <- round(ano_MIS$`Pr(>F)`[1],4)
# Age
t.test(all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "AD"],
       all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "CN"])
##
  Welch Two Sample t-test
## data: all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "AD"] and all_data_subfileds_
## t = 0.18336, df = 286.45, p-value = 0.8546
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -1.606002 1.935954
```

```
## sample estimates:
## mean of x mean of y
## 74.94345 74.77847
t.test(all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "CN"],
       all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "MCI"])
##
## Welch Two Sample t-test
##
## data: all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "CN"] and all_data_subfileds_
## t = 1.9074, df = 166.39, p-value = 0.05819
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.07788104 4.51704771
## sample estimates:
## mean of x mean of y
## 74.77847 72.55889
t.test(all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "AD"],
       all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "MCI"])
##
## Welch Two Sample t-test
## data: all_data_subfileds_L$Age[all_data_subfileds_L$Research.Group == "AD"] and all_data_subfileds_
## t = 2.0744, df = 161.25, p-value = 0.03963
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 0.1145476 4.6545712
## sample estimates:
## mean of x mean of y
## 74.94345 72.55889
# CDR
kruskal.test(Global.CDR ~ Research.Group, data = all_data_subfileds_L)
##
## Kruskal-Wallis rank sum test
##
## data: Global.CDR by Research.Group
## Kruskal-Wallis chi-squared = 325.76, df = 2, p-value < 2.2e-16
demo_info["CDR", 9] <- "< 0.001"
knitr::kable(demo_info[, -9])
```

	CN_NC_n	ne@aN_NC_ra	ngeN_C_me	a6N_C_ran	ngMCI_mean	MCI_rang	eAD_mean	AD_range
Gender	53/63	0	13/15	0	48/42	0	84/61	0
Age	73.98 ± 7.82	58.4 - 91.4	$78.1 {\pm} 6.8$	66.7 - 93.2	72.56 ± 9.16	55.2 - 97.4	74.94 ± 7.5	56-91

```
CN_NC_meaN_NC_rangeN_C_meaCN_C_rangeMCI_mean_MCI_rangeAD_mean_AD_range
Education 17.02\pm2.31 12-20
                                      16 \pm 2.09
                                                  12-20
                                                               16.08\pm2.67 8-20
                                                                                        15.52\pm2.66 8-20
CDR
                        0 - 0
                                      0\pm0
                                                  0 - 0
                                                               0.52\pm0.09 0.5-1
                                                                                       0.81 \pm 0.35
                                                                                                    0.5 - 2
           0\pm0
MMSE
           29.03 \pm 1.08 26-30
                                      29 \pm 1.19
                                                  25-30
                                                               27.27 \pm 2.43 16-30
                                                                                       22.72\pm3.04 5-29
MoCA
           24.36\pm1.77 18-28
                                     24.18\pm1.89 20-27
                                                               22.63\pm3.47 10-29
                                                                                       18.14 \pm 4.75 \ 9-27
MIS
           7.55\pm2.57 0-13
                                     7.46 \pm 3.25 0-12
                                                               7.68 \pm 3.58 \quad 0-14
                                                                                       6.88 \pm 3.72
                                                                                                    0 - 15
```

```
writexl::write_xlsx(demo_info, "results/demo_info.xlsx")
# subjects with education
length(which(!is.na(all_data_subfileds_L$Education[all_data_subfileds_L$Research.Group == "CN"])))
## [1] 138
length(which(!is.na(all_data_subfileds_L$Education[all_data_subfileds_L$Research.Group == "MCI"])))
## [1] 89
length(which(!is.na(all_data_subfileds_L$Education[all_data_subfileds_L$Research.Group == "AD"])))
## [1] 138
# subjects with MMSE
length(which(!is.na(all data subfileds L$MMSE.Total.Score[all data subfileds L$Research.Group == "CN"])
## [1] 144
length(which(!is.na(all_data_subfileds_L$MMSE.Total.Score[all_data_subfileds_L$Research.Group == "MCI"]
## [1] 90
length(which(!is.na(all_data_subfileds_L$MMSE.Total.Score[all_data_subfileds_L$Research.Group == "AD"])
## [1] 145
# subjects with MoCA
length(which(!is.na(all_data_subfileds_L$MoCA.Total[all_data_subfileds_L$Research.Group == "CN"])))
## [1] 134
length(which(!is.na(all data subfileds L$MoCA.Total[all data subfileds L$Research.Group == "MCI"])))
```

[1] 79

```
length(which(!is.na(all_data_subfileds_L$MoCA.Total[all_data_subfileds_L$Research.Group == "AD"])))
## [1] 84
```

Hippocampus and hippocampal subfields in CN-NC, CN-C, MCI, and AD groups

```
subfiled_sum <- as.data.frame(matrix(0,length(c(15:27,29)),8))</pre>
for (LorR in c( "Left", "Right")) {
    tmp <- all_data_subfileds_rm[all_data_subfileds_rm$LorR == LorR,]</pre>
    mean_all \leftarrow describeBy(tmp[, c(15:27,29,31)], group = "group", digits = 3)
    subfiled_sum[1:length(c(15:27,29)), 1:8] <-
        cbind(paste0(round(mean_all[[1]]$mean[1:14],2),"±",
                 round(mean_all[[1]]$sd[1:14],2)),
             paste0(round(mean_all[[1]]$min[1:14],2),"-",
                    round(mean all[[1]]$max[1:14],2)),
             paste0(round(mean_all[[2]]$mean[1:14],2),"±",
                    round(mean_all[[2]]$sd[1:14],2)),
             paste0(round(mean_all[[2]]$min[1:14],2),"-",
                    round(mean_all[[2]]$max[1:14],2)),
             paste0(round(mean_all[[3]]$mean[1:14],2),"±",
                    round(mean_all[[3]]$sd[1:14],2)),
             paste0(round(mean_all[[3]]$min[1:14],2),"-",
                    round(mean_all[[3]]$max[1:14],2)),
             paste0(round(mean_all[[4]]$mean[1:14],2),"±",
                    round(mean_all[[4]]$sd[1:14],2)),
             paste0(round(mean_all[[4]]$min[1:14],2),"-",
                    round(mean all[[4]]$max[1:14],2)))
#View(subfiled sum)
row.names(subfiled_sum) <- colnames(tmp[, c(15:27,29)])</pre>
writexl::write xlsx(subfiled sum,paste0("results/subfiled sum ",LorR,".xlsx"))
```

Violin plots for each group

```
all_data_subfileds_rm$Research.Group <- factor(all_data_subfileds_rm$Research.Group,
                           levels = c("CN", "MCI", "AD"))
for (sub in colnames(all_data_subfileds_rm)[15:27]) {
    ggplot(all_data_subfileds_rm, aes_string(x = "LorR", y =sub, color = "group")) +
        geom_violin( position = position_dodge(width=0.9))+
        scale color brewer(palette = "Set1")+
        #geom_point(position = position_dodge(width=0.9)) +
        stat_summary(fun=mean, size = 3, position = position_dodge(width=0.9),
             geom = "point", shape=4)+ # shape=1
    guides(colour = "none") +
        theme(axis.text = element_text(size = 12, face = "bold"),
              axis.title = element_blank())+
        theme(panel.border = element_blank(),
              panel.background = element_blank(),
              panel.grid = element_blank(),
              axis.line = element_line(colour = "black"))
        ggsave(paste0("results/","boxplot_point.",sub,".convertedCN.png"),
           width = 4, height = 3, units = c("in"), dpi = 200)
```

Group differences in subfields

```
all_data_subfileds_rm <- read.csv("datafile/all_data_subfileds_convertedCN.csv")</pre>
all_data_subfileds_rm$group <- factor(all_data_subfileds_rm$group, levels =</pre>
                         c("NonConverted_CN", "Converted_CN",
                           "MCI", "AD"))
table(all_data_subfileds_rm$group)
##
## NonConverted_CN
                       Converted_CN
                                                                    AD
##
                232
                                                                   290
                                                  180
gr_diff <- as.data.frame(matrix(0, length(15:27), 19))</pre>
all_data_subfileds_rm$LorR <- as.character(all_data_subfileds_rm$LorR)</pre>
all_data_subfileds_rm$Research.Group <-
    as.factor(all_data_subfileds_rm$Research.Group)
for (subfields in colnames(all_data_subfileds_rm)[15:27]) {
    for (LorR in c("Left", "Right")) {
    i <- i+1
    fl <- formula(pasteO(subfields, " ~ group + Age + Gender +</pre>
                  eTIV ml + Education"))
```

```
diff <- anova(lm(f1, all_data_subfileds_rm[all_data_subfileds_rm$LorR == LorR,]))</pre>
    gr_diff[i,1:2] <- c(subfields,LorR)</pre>
    gr_diff[i, 3:7] <- round(diff$`Pr(>F)`[1:5],4)
    # post-hoc analysis
    tt <- pairwise.t.test(all_data_subfileds_rm[all_data_subfileds_rm$LorR == LorR,
                              subfields],
                   all_data_subfileds_rm[all_data_subfileds_rm$LorR == LorR,
                                "group"],
             p.adjust="fdr", pool.sd = T)
    gr_diff[i,8] <- "NonConverted_CN vs. Converted_CN"</pre>
    gr_diff[i,9] <- round(tt$p.value[1,1],4)</pre>
    gr_diff[i,10] <- "NonConverted_CN vs. MCI"</pre>
    gr_diff[i,11] <- round(tt$p.value[2,1],4)</pre>
    gr_diff[i,12] <- "NonConverted_CN vs. AD"</pre>
    gr_diff[i,13] <- round(tt$p.value[3,1],4)</pre>
    gr_diff[i,14] <- "Converted_CN vs. MCI"</pre>
    gr_diff[i,15] <- round(tt$p.value[2,2],4)</pre>
    gr_diff[i,16] <- "Converted_CN vs. AD"</pre>
    gr_diff[i,17] <- round(tt$p.value[3,2],4)</pre>
    gr_diff[i,18] <- "MCI vs. AD"</pre>
    gr_diff[i,19] <- round(tt$p.value[3,3],4)</pre>
    }
}
colnames(gr_diff) <- c("subfields", "Hemisphere", "Group", "Age", "Gender", "eTIV",</pre>
                "Education", "contrast1", "p-value1", "contrast2", "p-value2",
                "contrast3", "p-value3", "contrast4", "p-value4", "contrast5",
                "p-value5", "contrast6", "p-value6")
#View(qr_diff)
writexl::write_xlsx(gr_diff, "results/group_diff_all.xlsx", col_names = T)
```

Correlations between subfields and MoCA MIS

[1] 758 31

```
# regression analysis
cor_sum <- as.data.frame(matrix(0, 2*4*length(15:27), 0))
dim(all_data_subfileds_rm)</pre>
```

```
i <- 0
for (test in c( "MoCA.MIS")) {
for (gr in c("NonConverted CN", "Converted CN", "MCI", "AD")) {
    for (subfields in colnames(all_data_subfileds_rm)[15:27]) {
        for (LorR in c("Left", "Right")) {
        i <- i+1
        tmp <- all_data_subfileds_rm[all_data_subfileds_rm$group == gr</pre>
                           & all_data_subfileds_rm$LorR == LorR,]
            mycor <- cor.test(tmp[,subfields], tmp[, test])</pre>
    ff <- formula(pasteO(test, " ~ ",subfields, "+ Age+ Gender + Education + eTIV_ml"))</pre>
    11 <- anova(lm(ff, tmp))</pre>
    cor_sum$test[i] <- test</pre>
    cor_sum$group[i] <- gr</pre>
    cor_sum$subfields[i] <- subfields</pre>
    cor_sum$LorR[i] <- LorR</pre>
    cor_sum$`r-value`[i] <- round(mycor$estimate,3)</pre>
    cor_sum$`p-value`[i] <- round(mycor$p.value,3)</pre>
    cor_sum$Pr[i] <- 11$`Pr(>F)`[1]
}
    }
}
}
writexl::write_xlsx(cor_sum, paste0("results/cor_sum_MoCA.xlsx"), col_names = T)
#View(cor_sum)
# check significant results
cor_sum_total <- readxl::read_xlsx(paste0("results/cor_sum_MoCA.xlsx"), sheet = 1)</pre>
#View(cor_sum_total)
#knitr::kable(cor_sum_total)
test <- "MoCA.MIS" # "MoCA.MIS"
which(p.adjust(cor_sum_total$Pr[cor_sum_total$test == test &
                     cor_sum_total$group == "AD" & cor_sum_total$LorR == "Left"]) < 0.05)</pre>
## integer(0)
which(p.adjust(cor_sum_total$Pr[cor_sum_total$test == test &
                     cor_sum_total$group == "AD" & cor_sum_total$LorR == "Right"]) < 0.05)</pre>
## [1] 5
which(p.adjust(cor_sum_total$Pr[cor_sum_total$test == test &
                     cor_sum_total$group == "MCI" &
                     cor_sum_total$LorR == "Left"]) < 0.05)</pre>
```

integer(0)

Scatter plots for significant correlation results in CN-NC and AD

```
# scatter plots
tmp <- all_data_subfileds_rm[all_data_subfileds_rm$Research.Group == "AD" &</pre>
                   all_data_subfileds_rm$LorR == "Right",]
plm <- lm(formula = MoCA.MIS ~ presubiculum + Gender + Age +eTIV_ml + Education,
      data=tmp)
anova(plm)
## Analysis of Variance Table
##
## Response: MoCA.MIS
               Df Sum Sq Mean Sq F value
## presubiculum 1 117.77 117.774 9.1063 0.003492 **
                   0.22 0.224 0.0173 0.895608
## Gender
                1
## Age
               1 1.30
                         1.301 0.1006 0.752026
## eTIV_ml
              1 0.00 0.000 0.0000 0.997519
## Education 1 0.12 0.120 0.0093 0.923406
## Residuals 74 957.07 12.933
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
tmp$group <- droplevels(tmp$group)</pre>
cor.test(tmp$MoCA.MIS[tmp$group == "AD"],
     tmp$presubiculum[tmp$group == "AD"])
```

```
##
## Pearson's product-moment correlation
##
## data: tmp$MoCA.MIS[tmp$group == "AD"] and tmp$presubiculum[tmp$group == "AD"]
## t = 2.6595, df = 83, p-value = 0.009389
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.07136404 0.46554635
## sample estimates:
##
         cor
## 0.2802261
# plot with lm for outcome ~ pred + factor
p <- ggplot(data=tmp, aes(x=presubiculum, y=MoCA.MIS, color = group)) +</pre>
    geom_point(size = 5, col = '#984EA3', alpha = 0.8) + #col = 'blue',
    geom_smooth(method = "lm", mapping=aes(y=predict(plm,tmp)),
           col = 'black',linewidth = 2, se = FALSE) +
    guides(color = "none") +
    labs(x = "", y = "") +
    theme(legend.title = element_text(colour="black", size=14, face="bold"),
          legend.text = element_text(colour="black", size=14, face="bold")) +
   theme(plot.title = element_text(hjust = 0.5))+
   theme(axis.title = element_blank(),
          axis.text = element_text(size = 16, face = "bold")) +
   theme(panel.background = element_blank(),
          panel.border = element_blank(),
          panel.grid = element blank(),
          axis.line = element line(colour = "black")) +
    coord_cartesian(ylim=c(0,15), xlim = c(150, 400)) +
    scale_y_continuous(breaks = seq(0,15,5))
   print(p)
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

```
ggsave(paste0("results/","MoCA.MIS_right.presubiculum.AD.png"), width = 6,
height = 4, units = c("in"), dpi = 200)
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

##

cor

0.3510998

```
##
## Pearson's product-moment correlation
##
## data: tmp$MoCA.MIS[tmp$group == "NonConverted_CN"] and tmp$presubiculum[tmp$group == "NonConverted_"
## t = 3.824, df = 104, p-value = 0.0002241
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.1718542 0.5078429
## sample estimates:
```

```
plm <- lm(formula = MoCA.MIS ~ presubiculum + Gender + Age +eTIV_ml + Education,</pre>
     data=tmp)
anova(plm)
## Analysis of Variance Table
## Response: MoCA.MIS
               Df Sum Sq Mean Sq F value
                                            Pr(>F)
## presubiculum 1 87.01 87.008 14.6707 0.0002259 ***
                          0.005 0.0008 0.9780065
## Gender
                1
                   0.00
                1 13.69 13.695 2.3091 0.1318382
## Age
               1 0.41
                          0.411 0.0693 0.7929319
## eTIV_ml
               1 0.89
                          0.886 0.1494 0.6999413
## Education
## Residuals 98 581.22
                          5.931
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
tmp$group <- droplevels(tmp$group)</pre>
# plot with lm for outcome ~ pred + factor
p <- ggplot(data=tmp, aes(x=presubiculum, y=MoCA.MIS, color = group)) +
   geom_point(size = 5, col = '#E41A1C', alpha = 0.8) + #col = 'blue',
   geom_smooth(method = "lm", mapping=aes(y=predict(plm,tmp)),
           col = 'black',linewidth = 2, se = FALSE) +
   guides(color = "none") +
   labs(x = "", y = "") +
   theme(legend.title = element_text(colour="black", size=14, face="bold"),
         legend.text = element_text(colour="black", size=14, face="bold")) +
   theme(plot.title = element text(hjust = 0.5))+
   theme(axis.title = element_blank(),
         axis.text = element_text(size = 16, face = "bold")) +
   theme(panel.background = element_blank(),
         panel.border = element_blank(),
         panel.grid = element_blank(),
         axis.line = element_line(colour = "black")) +
    coord_cartesian(ylim=c(0,15), xlim = c(150, 450)) +
    scale_y_continuous(breaks = seq(0,15,5))
return(anova(plm))
## Analysis of Variance Table
##
## Response: MoCA.MIS
               Df Sum Sq Mean Sq F value
                                            Pr(>F)
## presubiculum 1 87.01 87.008 14.6707 0.0002259 ***
## Gender
                1 0.00
                          0.005 0.0008 0.9780065
                1 13.69 13.695 2.3091 0.1318382
## Age
## eTIV_ml
                          0.411 0.0693 0.7929319
               1
                   0.41
## Education
               1 0.89
                          0.886 0.1494 0.6999413
## Residuals 98 581.22
                          5.931
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

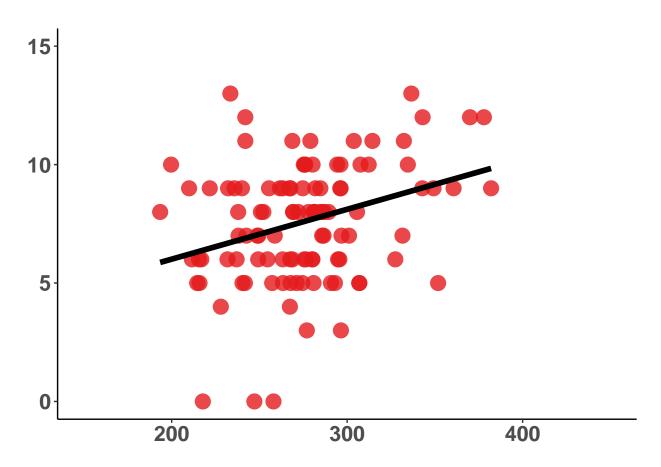
```
## 'geom_smooth()' using formula = 'y ~ x'
 15
 10
  5
  0
                   200
                                             300
                                                                      400
ggsave(paste0("results/","MoCA.MIS.left.presubiculum.CN.png"), width = 6,
           height = 4, units = c("in"), dpi = 200)
## 'geom_smooth()' using formula = 'y ~ x'
# CN-NC, right presubiculum
    tmp <- all_data_subfileds_rm[all_data_subfileds_rm$group == "NonConverted_CN" &</pre>
                        all_data_subfileds_rm$LorR == "Right",]
    cor.test(tmp$MoCA.MIS[tmp$group == "NonConverted_CN"],
         tmp$presubiculum[tmp$group == "NonConverted_CN"])
   Pearson's product-moment correlation
##
## data: tmp$MoCA.MIS[tmp$group == "NonConverted_CN"] and tmp$presubiculum[tmp$group == "NonConverted_CN"]
## t = 3.2388, df = 104, p-value = 0.001612
\#\# alternative hypothesis: true correlation is not equal to 0
```

print(p)

```
## 95 percent confidence interval:
## 0.1187933 0.4665089
## sample estimates:
##
        cor
## 0.3026896
plm <- lm(formula = MoCA.MIS ~ presubiculum + Gender + Age +eTIV_ml + Education,</pre>
     data=tmp)
anova(plm)
## Analysis of Variance Table
## Response: MoCA.MIS
               Df Sum Sq Mean Sq F value Pr(>F)
## presubiculum 1 65.28 65.280 10.7136 0.00147 **
                1 0.00 0.004 0.0006 0.98012
## Gender
## Age
                1 19.15 19.152 3.1432 0.07935 .
               1 1.21
                          1.206 0.1979 0.65741
## eTIV_ml
## Education
               1 0.44
                           0.445 0.0730 0.78765
## Residuals 98 597.14
                           6.093
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
tmp$group <- droplevels(tmp$group)</pre>
# plot with lm for outcome ~ pred + factor
p <- ggplot(data=tmp, aes(x=presubiculum, y=MoCA.MIS, color = group)) +
   geom_point(size = 5, col = '#E41A1C', alpha = 0.8) + #col = 'blue',
   geom_smooth(method = "lm", mapping=aes(y=predict(plm,tmp)),
           col = 'black',linewidth = 2, se = FALSE) +
   guides(color = "none") +
   labs(x = "", y = "") +
   theme(legend.title = element_text(colour="black", size=14, face="bold"),
         legend.text = element_text(colour="black", size=14, face="bold")) +
   theme(plot.title = element_text(hjust = 0.5))+
   theme(axis.title = element_blank(),
         axis.text = element text(size = 16, face = "bold")) +
   theme(panel.background = element_blank(),
         panel.border = element_blank(),
         panel.grid = element_blank(),
         axis.line = element_line(colour = "black")) +
    coord_cartesian(ylim=c(0,15), xlim = c(150, 450)) +
   scale_y_continuous(breaks = seq(0,15,5))
return(anova(plm))
## Analysis of Variance Table
##
## Response: MoCA.MIS
               Df Sum Sq Mean Sq F value Pr(>F)
## presubiculum 1 65.28 65.280 10.7136 0.00147 **
## Gender
                1 0.00 0.004 0.0006 0.98012
```

```
## Age    1 19.15 19.152 3.1432 0.07935 .
## eTIV_ml    1 1.21 1.206 0.1979 0.65741
## Education    1 0.44 0.445 0.0730 0.78765
## Residuals    98 597.14 6.093
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
print(p)
```

'geom_smooth()' using formula = 'y \sim x'



```
ggsave(paste0("results/","MoCA.MIS.right.presubiculum.CN.png"), width = 6,
height = 4, units = c("in"), dpi = 200)
```

'geom_smooth()' using formula = 'y ~ x'