STOPPD mean diffusivity analysis (left entorhinal cortex)

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This script analyses mean diffusivity in the left entorhinal cortex.

```
#load libraries
library(tidyverse)
library(lme4)
library(growthmodels)

#bring in subject info (generated by 03_STOPPD_masterDF.Rmd)

# then take only the subjects who completed (n= 74)

df <- read_csv('../generated_csvs/STOPPD_masterDF_2018-09-06.csv') %>%
    mutate(STUDYID = as.character(STUDYID)) %>%
    filter(second_complete == "Yes")

#rename timepoint variable for clarity
colnames(df)[colnames(df)=="second_timepoint"] <- "category"

#make a datediff column for time between scans
df$dateDiff <- as.numeric(round(difftime(df$second_date, df$first_date, units = "days"), 0))</pre>
```

Known exclusion reasons

subject 210030 (MAS): serious outlier.. it appears that this participants baseline is acutally a mislabled repeat of the subject 210022's baseline data. Calculating the difference in cortical thinkness between two different people was a relatively large value..

subject 320032 (PMC): incidental findings more atrophy, should exclude

known DWI issues

 ${f subject\ 410012\ timepoint\ 02\ ->\ scan\ was\ blacklisted\ "aborted"}$ for system failure..no DWI for this participant

subject 220009_timepoint 01 -> scan was also incomplete (this participant was only able complete the T1w)

```
So we will filter the data table to exclude these 4 participants (final n=70)
```

```
df <- filter(df, !(STUDYID %in% c("210030", "320032", "410012", "220009")))
```

mangling the Mean Diffusivity cata data

Erin reran the enigma DTI pipeline for only PMC using a different skull stripping parameter (-fa 0.7 to BET). We will use these numbers instead of the others in the archive here..

```
#bring in MD data (from the filesystem)
MD_most <- read_csv('.../data/enigma-DTI_archive_201809/enigmaDTI-MD-results.csv')
MD_PMC <- read_csv('../data/enigma-DTI_PMCredo_201809/enigmaDTI-MD-results.csv')
# separate id into it's parts and then drop old PMC data
MD most <- MD most %>%
  separate(id, into = c("study", "site", "STUDYID", "timepoint")) %>%
  filter(site != "PMC")
# separate the PMC subject id into it's parts and then bind to the data from the other sites
MD <- MD_PMC %>%
  separate(id, into = c("study", "site", "STUDYID", "timepoint")) %>%
  bind_rows(MD_most)
# drop acute ("00") and other ("03") timepoints from the analysis
MD <- MD %>%
  filter(!(timepoint %in% c("00", "03"))) %>%
  gather(tract, MD, ends_with("MD")) %>%
  unite(tract_timepoint, tract, timepoint) %>%
  spread(tract_timepoint, MD)
```

check for missing MD data

```
# filter the master spreadsheet for the list of completers (no output means we are ok)
df %>%
  anti_join(MD, by = "STUDYID") %>%
  summarise(`Number of missing MD values` = n()) %>%
  knitr::kable()
```

 $\frac{\text{Number of missing MD values}}{0}$

merge (i.e. join) the MD data with the clinical scores

```
all_MD <- df %>%
  select(STUDYID, sex, age, randomization, category, dateDiff) %>%
  left_join(MD, by = "STUDYID")

all_MD %>%
  filter(is.na(AverageFA_MD_01)) %>%
  summarise(`Number of missing timepoint 1 MD values` = n()) %>%
```

```
Number of missing timepoint 1 MD values

0

all_MD %>%
filter(is.na(AverageFA_MD_02)) %>%
summarise(`Number of missing timepoint 2 MD values` = n()) %>%
knitr::kable()

Number of missing timepoint 2 MD values

0

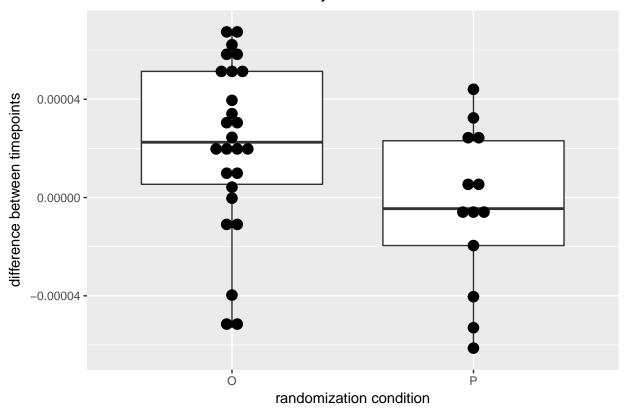
#write out clean MD speadsheet (required for subsequent MD analyses)
write.csv(all_MD, '../generated_csvs/STOPPD_MDclean.csv', row.names = FALSE)
```

RCT only

```
#boxplot of difference in MD in whole skeleton (y axis) by randomization group (x axis)
ggplot(RCT_MD, aes(x= randomization, y = diffAverageSkel_MD)) +
   geom_boxplot(outlier.shape = NA) +
   geom_dotplot(binaxis = 'y', stackdir = 'center') +
   ggtitle("Whole skeleton mean diffusivity") +
   xlab("randomization condition") +
   ylab("difference between timepoints")
```

`stat_bindot()` using `bins = 30`. Pick better value with `binwidth`.

Whole skeleton mean diffusivity



 ${\it\#Note: the outlying participant is again 210030; this participant has {\it NOT been excluded from subsequent}}$

```
fit_rct <- lm(diffAverageSkel_MD ~ randomization, data= RCT_MD)</pre>
print(fit_rct)
##
## Call:
## lm(formula = diffAverageSkel_MD ~ randomization, data = RCT_MD)
## Coefficients:
##
      (Intercept)
                   randomizationP
       0.00002180
                      -0.00002606
summary(fit_rct)
##
## Call:
## lm(formula = diffAverageSkel_MD ~ randomization, data = RCT_MD)
## Residuals:
                                   Median
##
                          1Q
## -0.000074955 -0.000016430 -0.000000265 0.000028775 0.000048305
##
## Coefficients:
##
                      Estimate
                                 Std. Error t value Pr(>|t|)
                   0.000021795 0.000006622
                                             3.291 0.0022 **
## (Intercept)
## randomizationP -0.000026061 0.000011470 -2.272
## ---
```

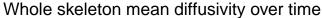
```
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.00003377 on 37 degrees of freedom
## Multiple R-squared: 0.1224, Adjusted R-squared: 0.09872
## F-statistic: 5.162 on 1 and 37 DF, p-value: 0.02899
#run linear model with covariates of sex and age
fit_rct <- lm(diffAverageSkel_MD ~ randomization + sex + age, data= RCT_MD)
print(fit_rct)
##
## Call:
## lm(formula = diffAverageSkel_MD ~ randomization + sex + age,
       data = RCT_MD)
##
## Coefficients:
##
      (Intercept) randomizationP
                                             sexM
                                                              age
## 0.00001578270 -0.00002642399
                                   0.00000240912
                                                   0.0000009383
summary(fit_rct)
##
## Call:
## lm(formula = diffAverageSkel_MD ~ randomization + sex + age,
##
       data = RCT_MD)
##
## Residuals:
           Min
                         10
                                  Median
                                                   30
                                                                Max
## -0.000078858 -0.000015710 0.000001514 0.000027177 0.000051303
##
## Coefficients:
##
                                     Std. Error t value Pr(>|t|)
                       Estimate
## (Intercept)
                  0.00001578270 0.00002188374
                                                0.721
                                                         0.4756
## randomizationP -0.00002642399 0.00001208614 -2.186
                                                          0.0356 *
## sexM
                  0.00000240912 0.00001153089
                                                 0.209
                                                          0.8357
                  0.0000009383 0.0000040916
                                                 0.229
## age
                                                          0.8200
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.00003466 on 35 degrees of freedom
## Multiple R-squared: 0.1256, Adjusted R-squared: 0.0506
## F-statistic: 1.675 on 3 and 35 DF, p-value: 0.1901
#run linear model with covariates of sex, age and site
fit_rct <- lm(diffAverageSkel_MD ~ randomization + sex + age + site, data= RCT_MD)
print(fit_rct)
##
## Call:
## lm(formula = diffAverageSkel_MD ~ randomization + sex + age +
       site, data = RCT_MD)
##
##
## Coefficients:
##
      (Intercept) randomizationP
                                             sexM
                                                              age
##
     0.0000015618
                  -0.0000283856
                                     0.0000026839
                                                    0.000003249
##
          siteMAS
                          siteNKI
                                          sitePMC
```

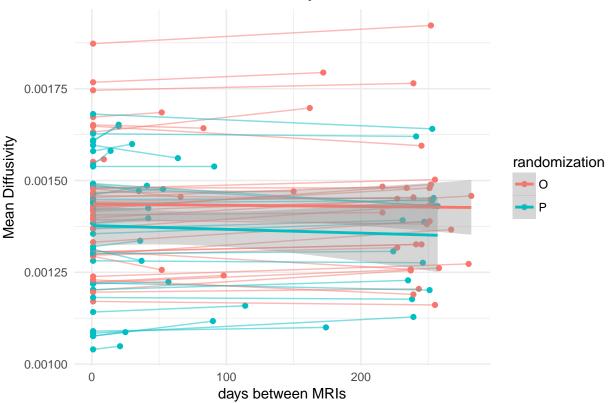
```
0.0000082173
                    0.0000118892
                                   -0.0000128727
summary(fit_rct)
##
## Call:
## lm(formula = diffAverageSkel_MD ~ randomization + sex + age +
##
      site, data = RCT_MD)
##
## Residuals:
##
                         1Q
                                  Median
## -0.000072052 -0.000012354 0.000002321 0.000020826 0.000049127
## Coefficients:
                                  Std. Error t value Pr(>|t|)
##
                      Estimate
## (Intercept)
                  0.0000015618 0.0000248965
                                              0.063
                                                      0.9504
## randomizationP -0.0000283856 0.0000124375 -2.282
                                                       0.0293 *
## sexM
                  0.0000026839 0.0000119040
                                              0.225
                                                      0.8231
                  0.0000003249
                                0.0000004581
                                               0.709
                                                      0.4834
## age
                                                      0.6107
## siteMAS
                 0.0000082173 0.0000159846
                                              0.514
                  0.0000118892 0.0000148117
## siteNKI
                                               0.803
                                                       0.4281
## sitePMC
                 -0.0000128727 0.0000181798 -0.708
                                                      0.4840
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.00003532 on 32 degrees of freedom
## Multiple R-squared: 0.1698, Adjusted R-squared: 0.01415
## F-statistic: 1.091 on 6 and 32 DF, p-value: 0.3889
```

RCT & Relapse (with time as factor)

```
RCTRelapse_wholeskelMD <- RCTRelapse_MD %>%
    filter(Tract == "AverageFA")
#plot

RCTRelapse_wholeskelMD %>%
    ggplot(aes(x=model_days, y=MD, colour=randomization)) +
    geom_point() +
    geom_line(aes(group=STUDYID), alpha = 0.5) +
    geom_smooth(method="lm", formula=y~poly(x,1)) +
    ggtitle("Whole skeleton mean diffusivity over time") +
    xlab("days between MRIs") +
    ylab("Mean Diffusivity") +
    theme_minimal()
```





```
#run mixed linear model, with covariates
fit_all <- lmer(MD ~ randomization*model_days + sex + age + (1|STUDYID), data= RCTRelapse_wholeskelMD)</pre>
summary(fit_all)
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: MD ~ randomization * model_days + sex + age + (1 | STUDYID)
##
     Data: RCTRelapse_wholeskelMD
##
## REML criterion at convergence: -2172.6
## Scaled residuals:
                     Median
                 1Q
## -1.79473 -0.45375 -0.02211 0.38740
##
## Random effects:
  Groups
           Name
                       Variance
                                       Std.Dev.
## STUDYID (Intercept) 0.000000162720 0.00012756
                       0.000000004398 0.00002097
   Residual
## Number of obs: 140, groups: STUDYID, 70
##
## Fixed effects:
##
                                              Std. Error
                                 Estimate
## (Intercept)
                            0.00097123189 0.00005922449 66.23783533131
## randomizationP
```

0.00000008709 0.00000002214 68.19357217598 0.00003295832 0.00003101516 65.99532783955

model_days

sexM

```
0.00000798688 0.00000100086 65.99338155713
## age
## randomizationP:model_days -0.00000009381 0.00000003809 68.49572304325
                            t value
                                                Pr(>|t|)
                             16.399 < 0.000000000000000000000 ***
## (Intercept)
## randomizationP
                             -2.152
                                                0.034962 *
## model days
                                                0.000198 ***
                              3.934
## sexM
                              1.063
                                                0.291813
## age
                              7.980
                                         0.000000000288 ***
## randomizationP:model_days -2.463
                                                0.016298 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##
              (Intr) rndmzP mdl_dy sexM
## randomiztnP -0.197
## model_days -0.043 0.075
              -0.166 0.068 0.002
## sexM
## age
              -0.898 -0.072 0.004 -0.096
## rndmztnP:m_ 0.024 -0.098 -0.581 0.003 -0.002
rm('df', 'fit_all', 'fit_rct', 'MD', 'plot', 'RCT_FA', 'RCTRelapse_FA')
```

just wanna through in one site model..

```
#run mixed linear model, with covariates
fit_all <- lmer(MD ~ randomization*model_days + sex + age + site + (1|STUDYID), data= RCTRelapse_wholes
summary(fit all)
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula:
## MD ~ randomization * model_days + sex + age + site + (1 | STUDYID)
##
     Data: RCTRelapse_wholeskelMD
##
## REML criterion at convergence: -2165.5
##
## Scaled residuals:
##
       Min
             1Q
                   Median
                                30
## -1.75027 -0.45651 -0.00872 0.40133 1.90149
##
## Random effects:
## Groups Name
                      Variance
                                    Std.Dev.
## STUDYID (Intercept) 0.000000078425 0.00008856
## Residual
                      0.000000004399 0.00002097
## Number of obs: 140, groups: STUDYID, 70
##
## Fixed effects:
##
                                           Std. Error
                               Estimate
                                                                df
## (Intercept)
                          0.00107620814 0.00004345570 63.38845538623
## randomizationP
                          ## model_days
                           0.00006003452 0.00002200440 62.95205381331
## sexM
```

```
0.00000760310 0.00000070508 62.94985076260
## age
                         ## siteMAS
## siteNKI
                         ## sitePMC
                         -0.00018734952  0.00003128515  62.94544058940
## randomizationP:model_days -0.00000009601 0.00000003802 68.96745578641
                                          Pr(>|t|)
                        t value
## (Intercept)
                         24.766 < 0.0000000000000000 ***
## randomizationP
                         -1.583
                                          0.118134
                         3.898
## model_days
                                          0.000224 ***
## sexM
                          2.728
                                          0.008241 **
## age
                         10.783 0.00000000000000614 ***
                         -6.836 0.00000003867958716 ***
## siteMAS
                         -6.064 0.000000082815682000 ***
## siteNKI
                         -5.988 0.000000111344241240 ***
## sitePMC
## randomizationP:model_days -2.525
                                          0.013880 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Correlation of Fixed Effects:
    (Intr) rndmzP mdl_dy sexM age sitMAS sitNKI sitPMC
## randomiztnP -0.136
## model_days -0.059 0.105
         -0.124 0.087 0.003
## sexM
            -0.878 -0.083 0.006 -0.094
## age
## siteMAS -0.286 -0.152 0.004 -0.066 0.113
## siteNKI -0.175 -0.124 -0.008 -0.125 0.015 0.346
## sitePMC -0.153 -0.094 -0.001 -0.149 -0.005 0.332 0.320
## rndmztnP:m_ 0.031 -0.138 -0.582 0.004 -0.003 0.004 0.007 0.002
```