FACEBOOK use analysis

Rui Pei

2020-06-29

Contents

1. Participants **5**

2. Data description **5**
Habitual FB use. . . . 5
Emotional FB use. . . . 6
ROIS 7
IRI 14
Need threat scale 15

3. Linking FB use with self-report **17**
IRI 17
NTS 17

4. Linking FB use with neural data **18**
Functional ROIS 18
Dufour mentalzing meta-analytic ROIS 19
Neurosynth mentalzing ROIS 21
Vijayakumar social exclusion meta-analytic ROIs 23
dACC ROIS 24

5. Link between neural activation in mentalizing regions and IRI **25**
Dufour et al. meta-analytic regions 25
Neurosynth regions 26

6. Link between neural activation in cyberball and NTS **27**
Functional ROIS 29
Dufour mentalzing meta-analytic ROIS 29
Vijayakumar social exclusion meta-analytic ROIs 30
dACC ROIS 31

```
rm(list = ls())
suppressPackageStartupMessages(library(ggplot2))
suppressPackageStartupMessages(library(tidyverse))
suppressPackageStartupMessages(library(plyr))
suppressPackageStartupMessages(library(dplyr))
suppressPackageStartupMessages(library(knitr))
suppressPackageStartupMessages(library(kableExtra))
suppressPackageStartupMessages(library(ggpubr))
suppressPackageStartupMessages(library(pander))
suppressPackageStartupMessages(library(jtools))
# suppressPackageStartupMessages(library(stringr))
```

```r
# suppressPackageStartupMessages(library(broom.mixed))
# suppressPackageStartupMessages(library(sjPlot))
# suppressPackageStartupMessages(library(huxtable))
# suppressPackageStartupMessages(library(ggstance))
```

```r
pd <- position_dodge(0.3) # move them .05 to the left and right

# Multiple plot function
#
# ggplot objects can be passed in ..., or to plotlist (as a list of ggplot objects)
# - cols:   Number of columns in layout
# - layout: A matrix specifying the layout. If present, 'cols' is ignored.
#
# If the layout is something like matrix(c(1,2,3,3), nrow=2, byrow=TRUE),
# then plot 1 will go in the upper left, 2 will go in the upper right, and
# 3 will go all the way across the bottom.
#
multiplot <- function(..., plotlist=NULL, file, cols=1, layout=NULL) {
  library(grid)

  # Make a list from the ... arguments and plotlist
  plots <- c(list(...), plotlist)

  numPlots = length(plots)

  # If layout is NULL, then use 'cols' to determine layout
  if (is.null(layout)) {
    # Make the panel
    # ncol: Number of columns of plots
    # nrow: Number of rows needed, calculated from # of cols
    layout <- matrix(seq(1, cols * ceiling(numPlots/cols)),
                     ncol = cols, nrow = ceiling(numPlots/cols))
  }

 if (numPlots==1) {
    print(plots[[1]])

  } else {
    # Set up the page
    grid.newpage()
    pushViewport(viewport(layout = grid.layout(nrow(layout), ncol(layout))))

    # Make each plot, in the correct location
    for (i in 1:numPlots) {
      # Get the i,j matrix positions of the regions that contain this subplot
      matchidx <- as.data.frame(which(layout == i, arr.ind = TRUE))

      print(plots[[i]], vp = viewport(layout.pos.row = matchidx$row,
                                      layout.pos.col = matchidx$col))
    }
  }
}
numericcharacters <- function(x) {
  !any(is.na(suppressWarnings(as.numeric(x)))) & is.character(x)
```

```
}

scale1 <- function(x) scale(x)[,1]

construct_coef_df = function(dvs,df){
  fhab_mds = lapply(dvs, function(x) {
    lm(substitute(i ~ fhab + study_t + scanner_t, list(i = as.name(x))), data = df)
})
  fhab_sums = lapply(fhab_mds, summary)

  femo_mds = lapply(dvs, function(x) {
      lm(substitute(i ~ femo + study_t + scanner_t, list(i = as.name(x))), data = df)
  })
  femo_sums = lapply(femo_mds, summary)

  fhab_betas = c()
  femo_betas = c()
  fhab_se = c()
  femo_se = c()
  for (i in c(1:length(dvs))){
    fhab_betas = c(fhab_betas,fhab_sums[[i]]$coefficients[2,1])
    femo_betas = c(femo_betas,femo_sums[[i]]$coefficients[2,1] )
    fhab_se = c(fhab_se, fhab_sums[[i]]$coefficients[2,2])
    femo_se = c(femo_se, femo_sums[[i]]$coefficients[2,2])
  }

  coef_df = data.frame("IV" = c(rep("Habitual FB use", length(dvs)), rep("Emotional FB use", length(dvs)
                       "DV" = c(dvs,dvs),
                       "beta" = c(fhab_betas, femo_betas),
                       "se" = c(fhab_se, femo_se))
  coef_df = coef_df %>%
  mutate(IV = factor(coef_df$IV, levels=c("Habitual FB use", "Emotional FB use")))
  return(coef_df)
}

plot_coef_df = function(coef_dv){
  g1 = ggplot(coef_df, aes(x=DV, y=beta, colour=IV)) +
    geom_errorbar(aes(ymin= beta-1.96*se, ymax= beta+1.96*se), width=.1, position=pd) +
    geom_point(position=pd) + theme_pubr() +
  scale_color_manual(values = c("Habitual FB use" = "dodgerblue", "Emotional FB use" = "forestgreen")) +
   geom_hline(yintercept=0, linetype="dashed",
                color = "grey", size=1) +
  rotate_x_text(45)
  return(g1)
}

construct_coef_df_iri = function(dvs,df){
  mds1 = lapply(dvs, function(x) {
    lm(substitute(i ~ IRI_Perspective_Taking + study_t + scanner_t, list(i = as.name(x))), data = df)
})
  sums1 = lapply(mds1, summary)

  mds2 = lapply(dvs, function(x) {
```

```
    lm(substitute(i ~ IRI_Fantasy + study_t + scanner_t, list(i = as.name(x))), data = df)
})
  sums2 = lapply(mds2, summary)

  mds3 = lapply(dvs, function(x) {
    lm(substitute(i ~ IRI_Empathic_Concern + study_t + scanner_t, list(i = as.name(x))), data = df)
})
  sums3 = lapply(mds3, summary)

  mds4 = lapply(dvs, function(x) {
    lm(substitute(i ~ IRI_Personal_Distress + study_t + scanner_t, list(i = as.name(x))), data = df)
})
  sums4 = lapply(mds4, summary)

  betas1 = c()
  betas2 = c()
  betas3 = c()
  betas4 = c()
  se1 = c()
  se2 = c()
  se3 = c()
  se4 = c()
  for (i in c(1:length(dvs))){
    betas1 = c(betas1,sums1[[i]]$coefficients[2,1])
    betas2 = c(betas2,sums2[[i]]$coefficients[2,1])
    betas3 = c(betas3,sums3[[i]]$coefficients[2,1])
    betas4 = c(betas4,sums4[[i]]$coefficients[2,1])
    se1 = c(se1, sums1[[i]]$coefficients[2,2])
    se2 = c(se2, sums2[[i]]$coefficients[2,2])
    se3 = c(se3, sums3[[i]]$coefficients[2,2])
    se4 = c(se4, sums4[[i]]$coefficients[2,2])
  }

  coef_df = data.frame("IV" = c(rep("Perspective taking", length(dvs)),
                                rep("Fantasy", length(dvs)),
                                rep("Empathic concern", length(dvs)),
                                rep("Personal distress", length(dvs))),
                       "DV" = c(dvs,dvs,dvs,dvs),
                       "beta" = c(betas1, betas2,betas3,betas4),
                       "se" = c(se1, se2,se3,se4))
  coef_df = coef_df
  return(coef_df)
}

plot_coef_iri = function(coef_dv){
  g1 = ggplot(coef_df, aes(x=DV, y=beta, colour=IV)) +
    geom_errorbar(aes(ymin= beta-1.96*se, ymax= beta+1.96*se), width=.1, position=pd) +
    geom_point(position=pd) + theme_pubr()  +
   geom_hline(yintercept=0, linetype="dashed",
                color = "grey", size=1) +
  rotate_x_text(45)
  return(g1)
}
```

```
path =  "/Users/Rui/Documents/GitHub/facebookUse/"
setwd(path)
## socialMedia data
socialMedia = read.csv(paste0(path, '0_facebook/tps2_facebook.csv'),na.strings=c("","NA"), stringsAsFac

socialMedia = socialMedia %>%
  mutate(TPS_ID = as.factor(TPS_ID)) %>%
  mutate_if(is.character,as.numeric) %>%
  mutate(fhab = rowMeans(.[,10:19], na.rm = FALSE),
         femo = rowMeans(.[c(8,9,20:39)], na.rm = FALSE), ## include the two "feel connected" questions
         TPS_ID = as.character(TPS_ID)) %>%
  select(TPS_ID, fhab, femo)

cyber = read.csv(paste0(path, '0_cyberball/cyberball_roi.csv'), stringsAsFactors = FALSE)

age = read.csv(paste0(path, '0_surveys/tps12_scanner_study_age.csv'), stringsAsFactors = FALSE) %>%
  select(TPS_ID, age, study_t, scanner_t)



iri = read.csv(paste0(path, '0_surveys/iri_tps12.csv'), stringsAsFactors = FALSE)
iri = iri %>%
  mutate(IRI_mean = rowMeans(select(iri, starts_with("IRI_")), na.rm = FALSE))


nts = read.csv(paste0(path, '0_surveys/nts.csv'), stringsAsFactors = FALSE)

df = cyber %>%
  left_join(socialMedia, by = "TPS_ID") %>%
  left_join(age, by = "TPS_ID") %>%
  left_join(iri, by = "TPS_ID") %>%
  left_join(nts, by = "TPS_ID")


# Remove NAs
df = df %>% drop_na(age,fhab,femo, NTS_Belongingness,
                    NTS_SelfEsteem, NTS_Control,
                    NTS_MeaningfulExistence, study_t, scanner_t,
                    IRI_Perspective_Taking, IRI_Fantasy,
                    IRI_Empathic_Concern, IRI_Personal_Distress)
```

1. Participants

We include 59 participants in this report.

2. Data description

Habitual FB use.

1 - Strongly disagree 7 - Strongly agree

1. Using Facebook is something I do automatically.

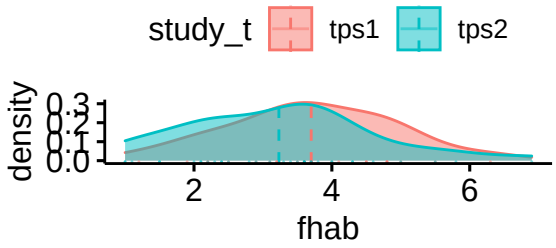
2. Using Facebook is something I do without meaning to do it.

3. Using Facebook is something I do without thinking.

4. Using Facebook is something I start doing before I realize I'm doing it.
5. Using Facebook is something that would require effort not to do it.
6. Using Facebook is something I do without having to consciously remember.
7. Using Facebook is something that belongs to my daily routine.

8. Using Facebook is something I would find hard not to do.
9. Using Facebook is something I have no need to think about doing.
10. Using Facebook is something that's typically "me".

```
ggdensity(df, x = "fhab",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```

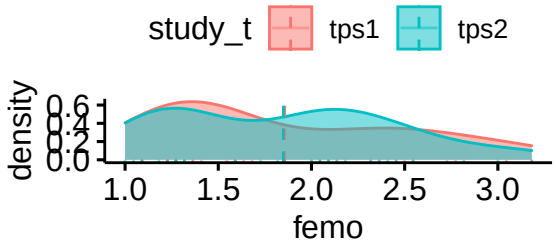


Emotional FB use.

1. I feel connected to my friends when I use Facebook.
2. I feel connected to my family members when I use Facebook.

I get an urge to post on Facebook from my computer or phone the moment when ... 3. something makes me feel amused
4. something makes me feel surprised 5. something makes me feel awed 6. something makes me feel loved 7. something makes me feel proud
8. something makes me feel excited
9. something makes me feel grateful 10. something makes me feel happy
11. something makes me feel inspired
12. something makes me feel confident
13. something makes me feel angry
14. something makes me feel nervous 15. something makes me feel awkward 16. something makes me feel stressed
17. something makes me feel jealous 18. something makes me feel lonely
19. something makes me feel scared
20. something makes me feel upset
21. something makes me feel ashamed 22. something makes me feel guilty

```
ggdensity(df, x = "femo",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```



ROIS

Scrit to obtain the ROI values at jupyter hub.

```
roi_names = colnames(df)[c(3:20,23)]
source = c("functional ROI",
           "functional ROI",
           "functional ROI",
           "functional ROI",
           "Dufour et al., 2013","Dufour et al., 2013","Dufour et al., 2013",
           "Dufour et al., 2013","Dufour et al., 2013","Dufour et al., 2013",
           "Dufour et al., 2013",
           "Vijayakumar et al., 2017","Vijayakumar et al., 2017",
           "Vijayakumar et al., 2017","Vijayakumar et al., 2017",
           "Vijayakumar et al., 2017","Vijayakumar et al., 2017",
           "Vijayakumar et al., 2017","Vijayakumar et al., 2017")
img_folder = '/Users/Rui/Box Sync/CurrentProjects_Penn/TPS12_BART/99_fomo/data/cyberball/ROIs/png/'
imgs = paste0(img_folder, roi_names, '.png')

rois = data.frame("name" = roi_names, "source" = source, "image" = imgs)


rois %>%
  mutate(
    image %>% pander::pandoc.image.return()
    ) %>%
  select(-image) %>%
  pander()
```

Table 1: Table continues below

name	source
subacc_roi_mask	functional ROI
rInsula_roi_mask	functional ROI
lInsula_roi_mask	functional ROI
functional_combined_roi_mask	functional ROI
dmpfc_roi_mask	Dufour et al., 2013
mmpfc_roi_mask	Dufour et al., 2013
vmpfc_roi_mask	Dufour et al., 2013
mpfc_roi_mask	Dufour et al., 2013
precuneus_roi_mask	Dufour et al., 2013
lTpj_roi_mask	Dufour et al., 2013
rTpj_roi_mask	Dufour et al., 2013
rSts_roi_mask	Vijayakumar et al., 2017

name	source
saxe_combined_mask	Vijayakumar et al., 2017
ns_ment_mask	Vijayakumar et al., 2017
ns_ment_clust1_mask	Vijayakumar et al., 2017
ns_ment_clust2_mask	Vijayakumar et al., 2017
ns_ment_clust3_mask	Vijayakumar et al., 2017
ns_ment_clust4_mask	Vijayakumar et al., 2017
ns_ment_clust7_mask	Vijayakumar et al., 2017

image %>% pander::pandoc.image.return()

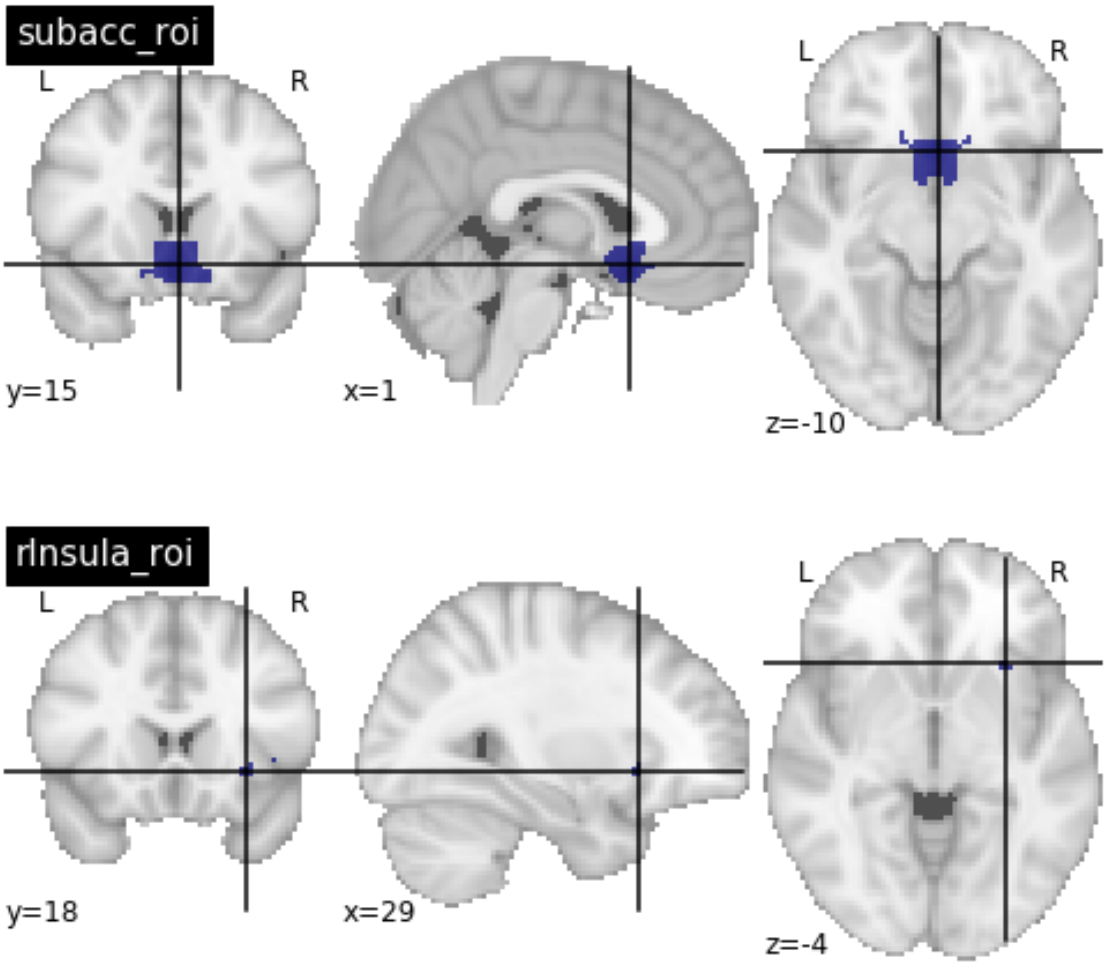


image %>% pander::pandoc.image.return()

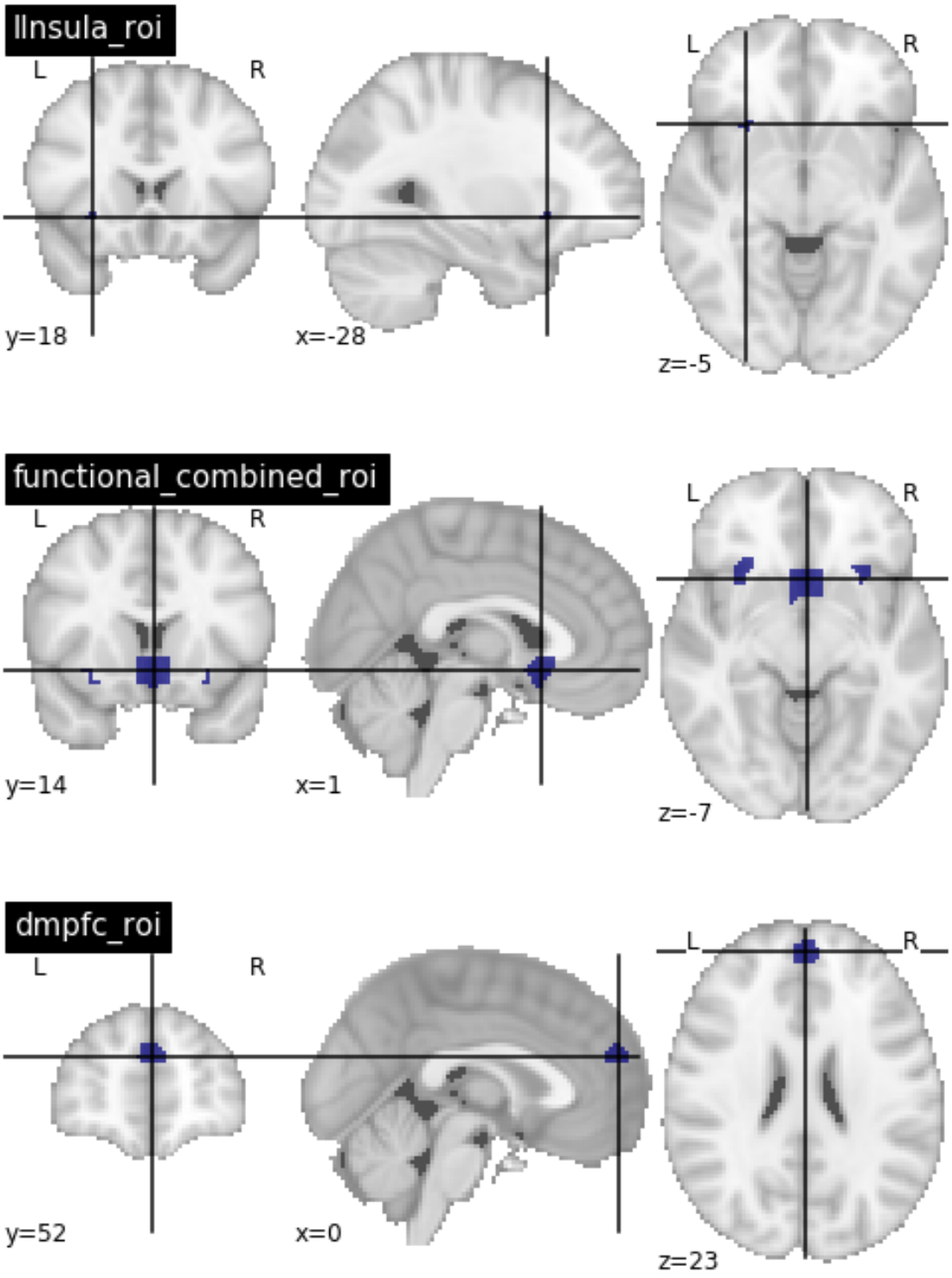


image %>% pander::pandoc.image.return()

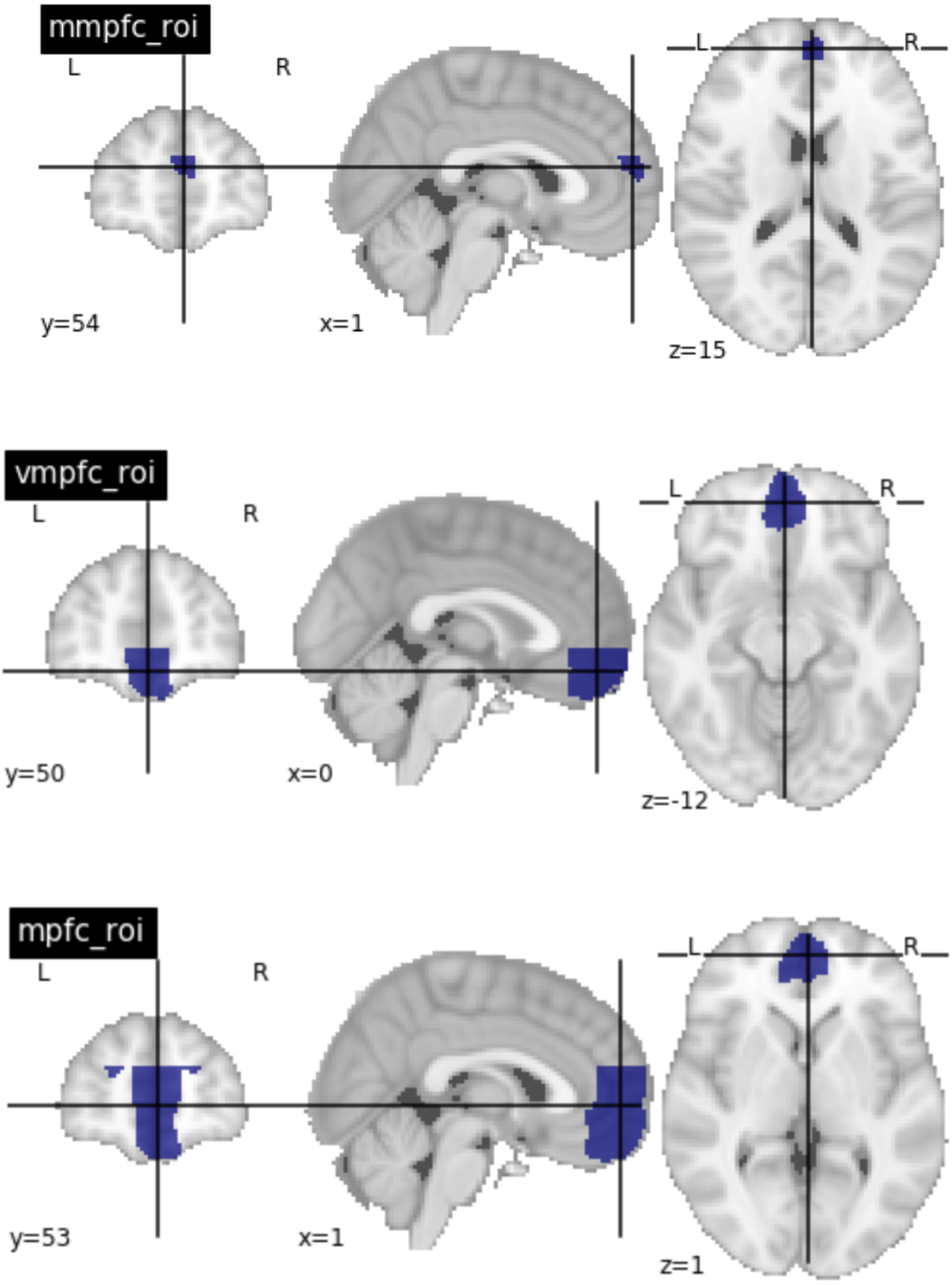


image %>% pander::pandoc.image.return()

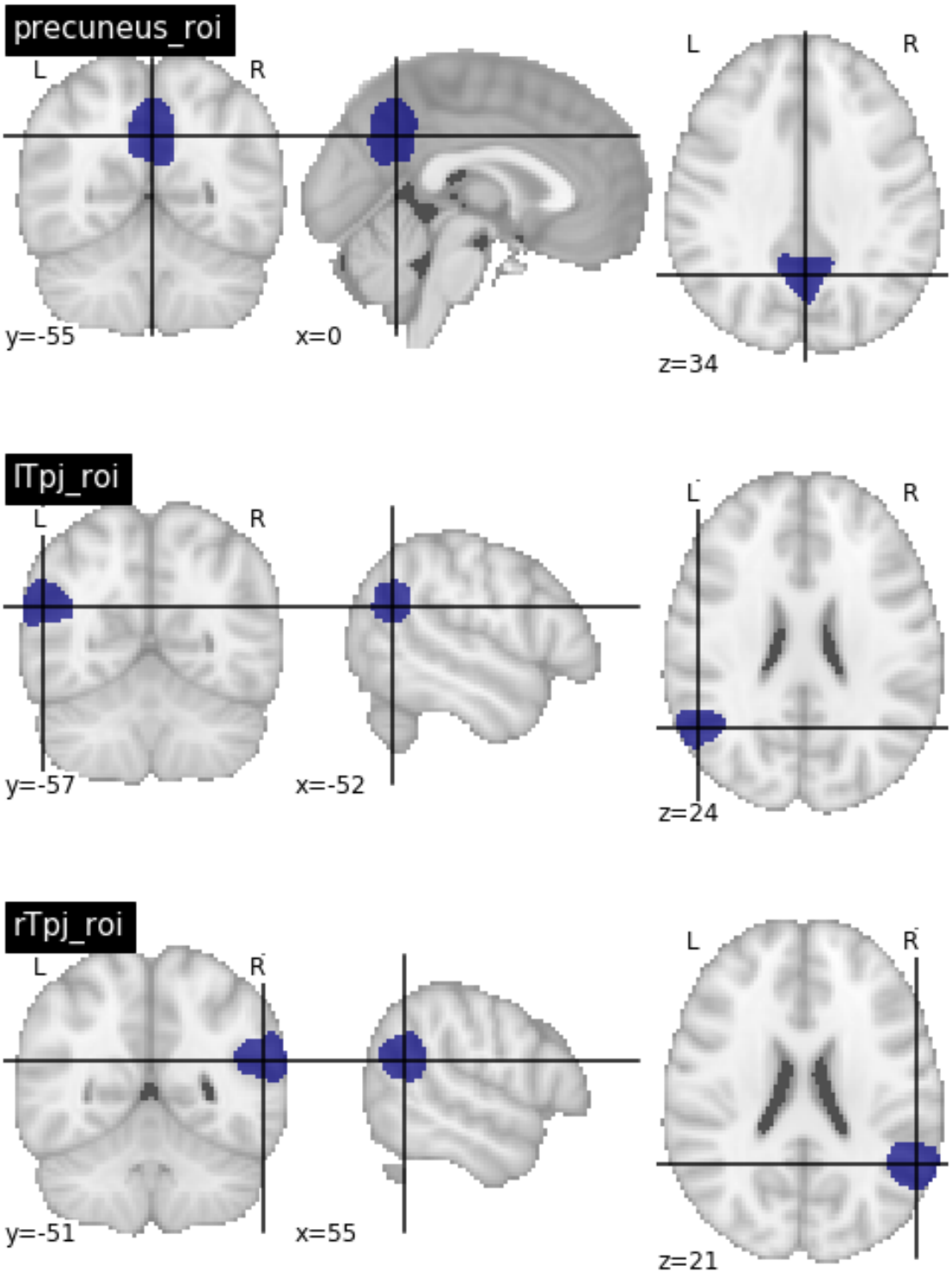


image %>% pander::pandoc.image.return()

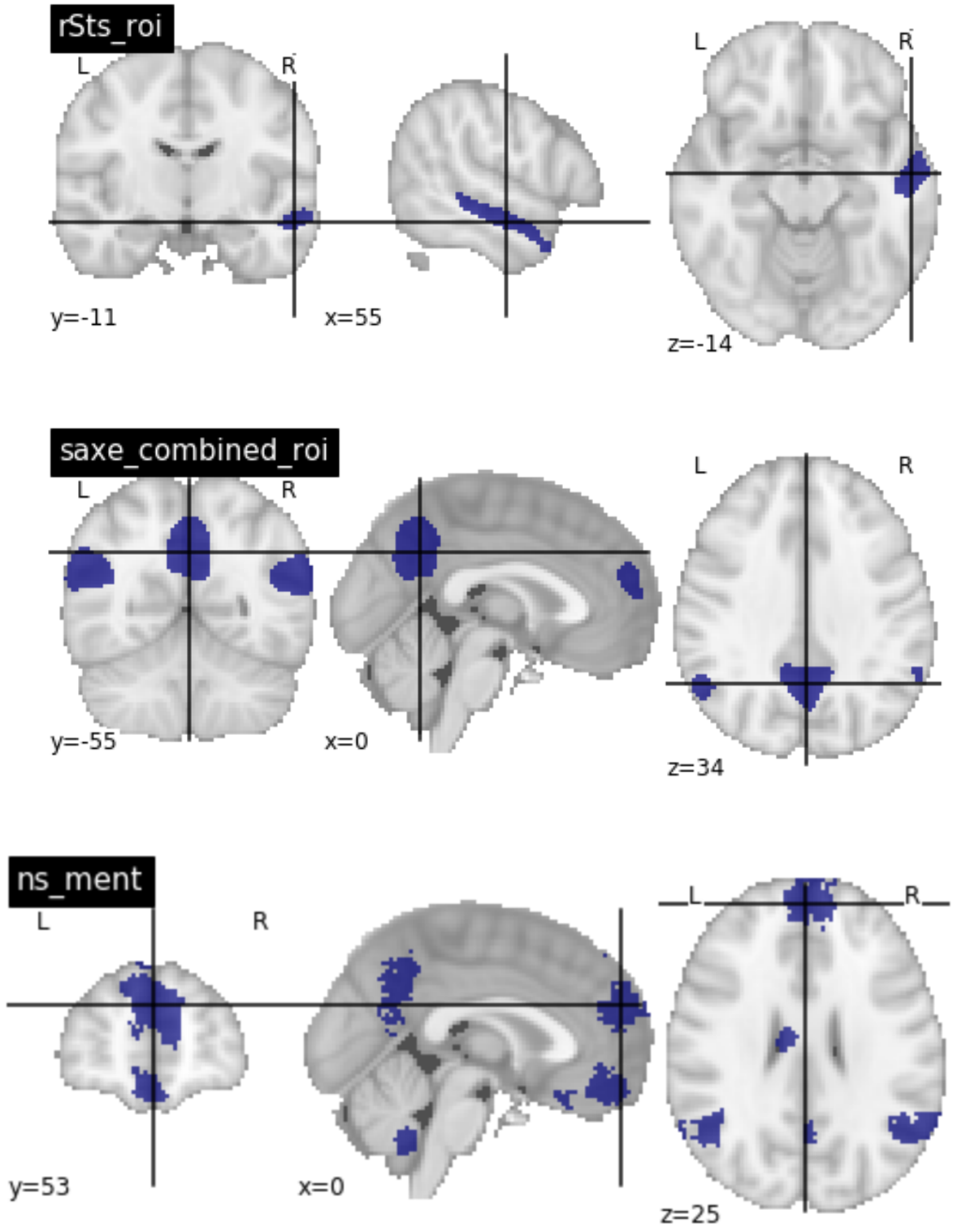


image %>% pander::pandoc.image.return()

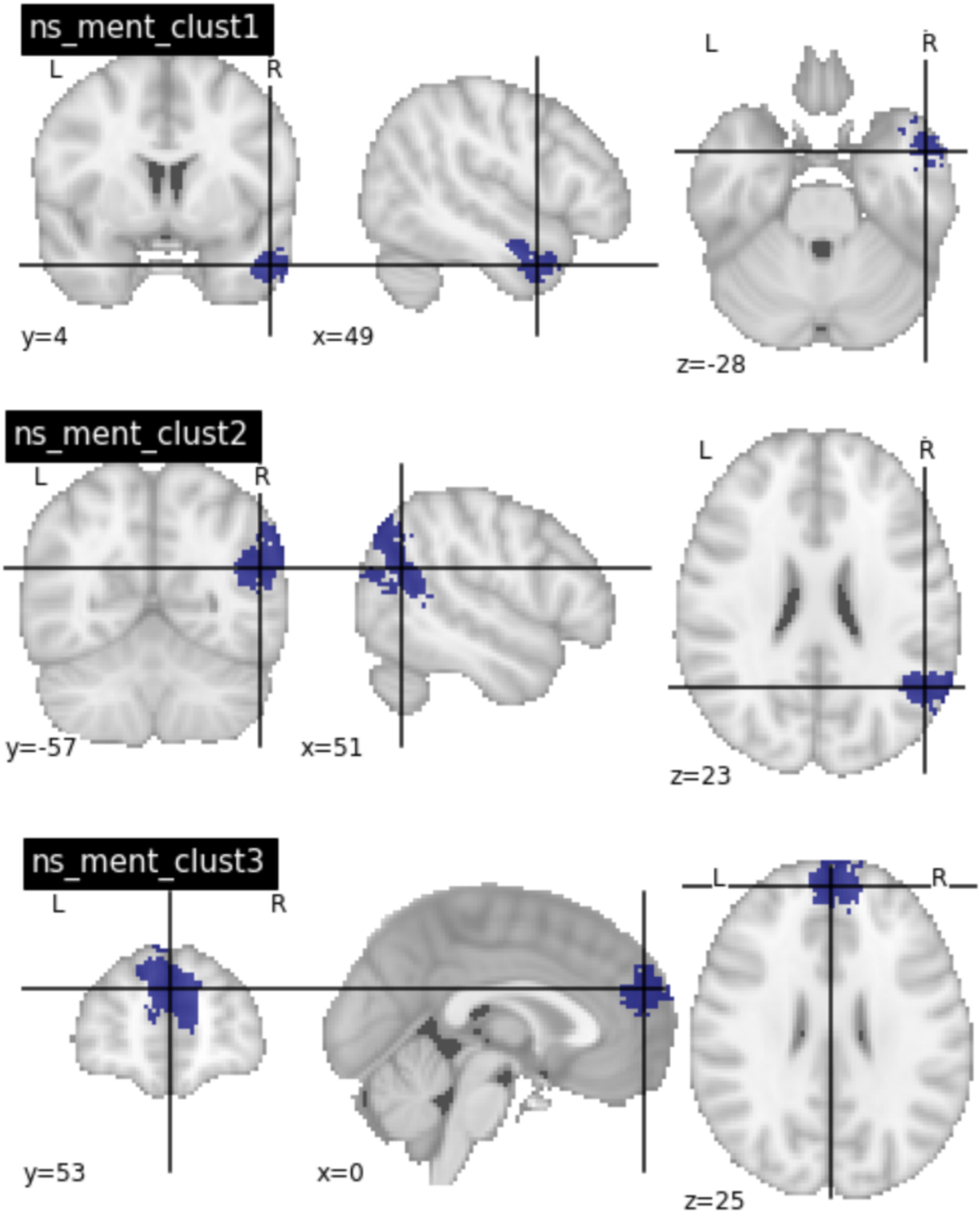
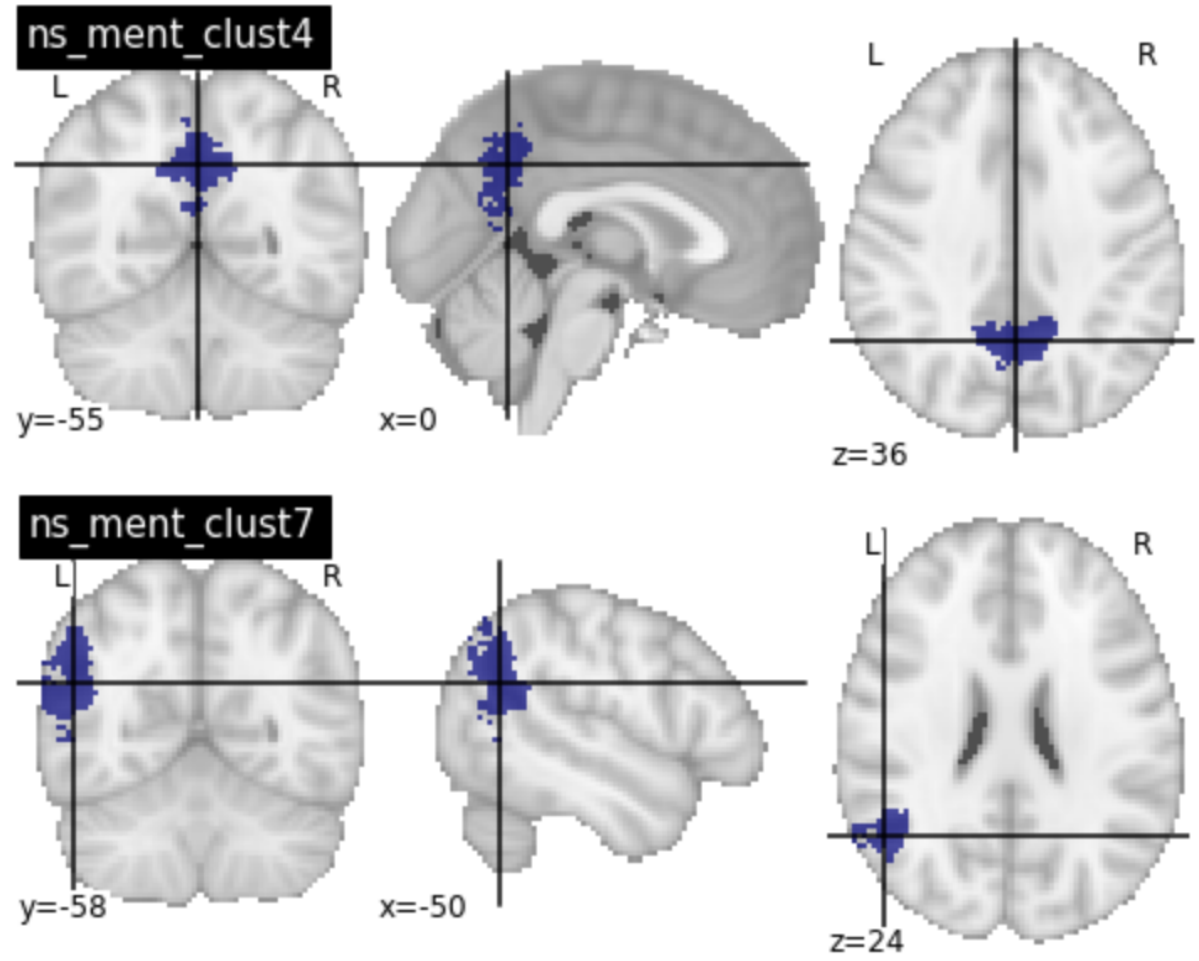


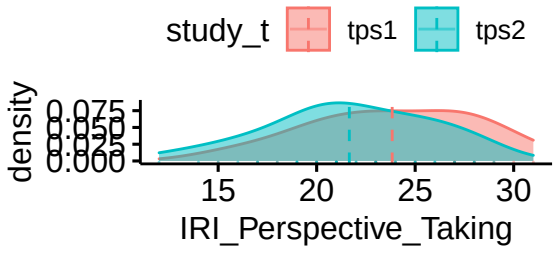
image %>% pander::pandoc.image.return()



IRI

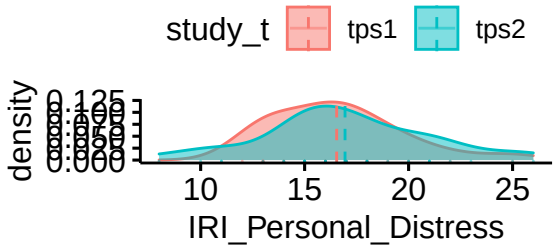
Perspective taking

```
ggdensity(df, x = "IRI_Perspective_Taking",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```



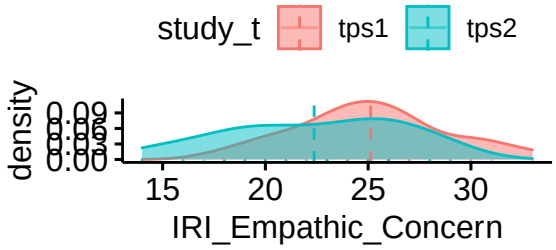
Perspective taking

```
ggdensity(df, x = "IRI_Personal_Distress",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```



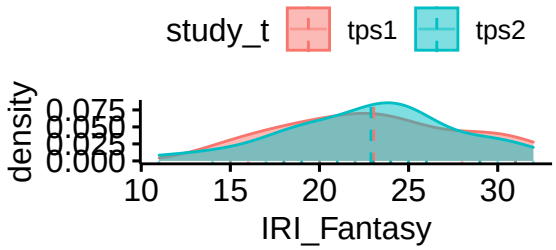
Empathic concern

```
ggdensity(df, x = "IRI_Empathic_Concern",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```



Fantasy

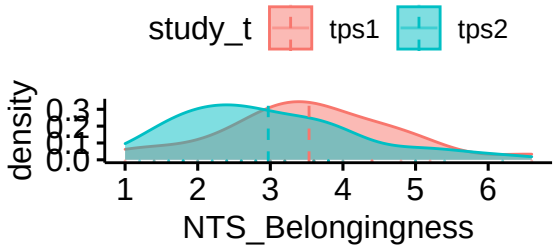
```
ggdensity(df, x = "IRI_Fantasy",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```



Need threat scale

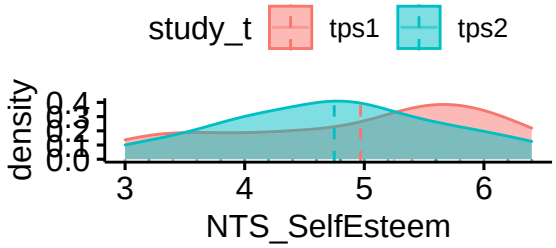
Belongingness

```
ggdensity(df, x = "NTS_Belongingness",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```



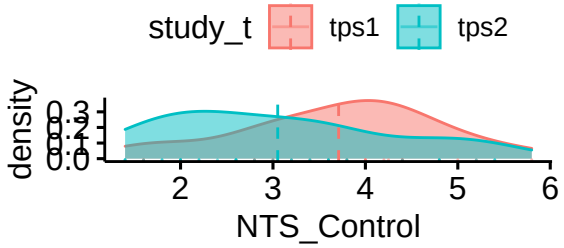
Self esteem

```
ggdensity(df, x = "NTS_SelfEsteem",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```



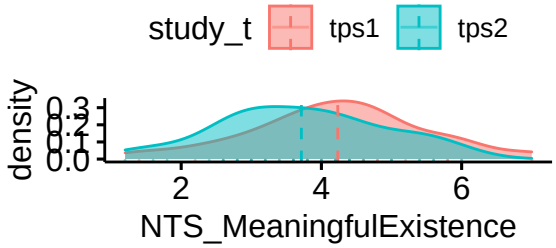
Control

```
ggdensity(df, x = "NTS_Control",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```



MeaningfulExistence

```
ggdensity(df, x = "NTS_MeaningfulExistence",
   add = "mean", rug = TRUE,
   color = "study_t", fill = "study_t")
```

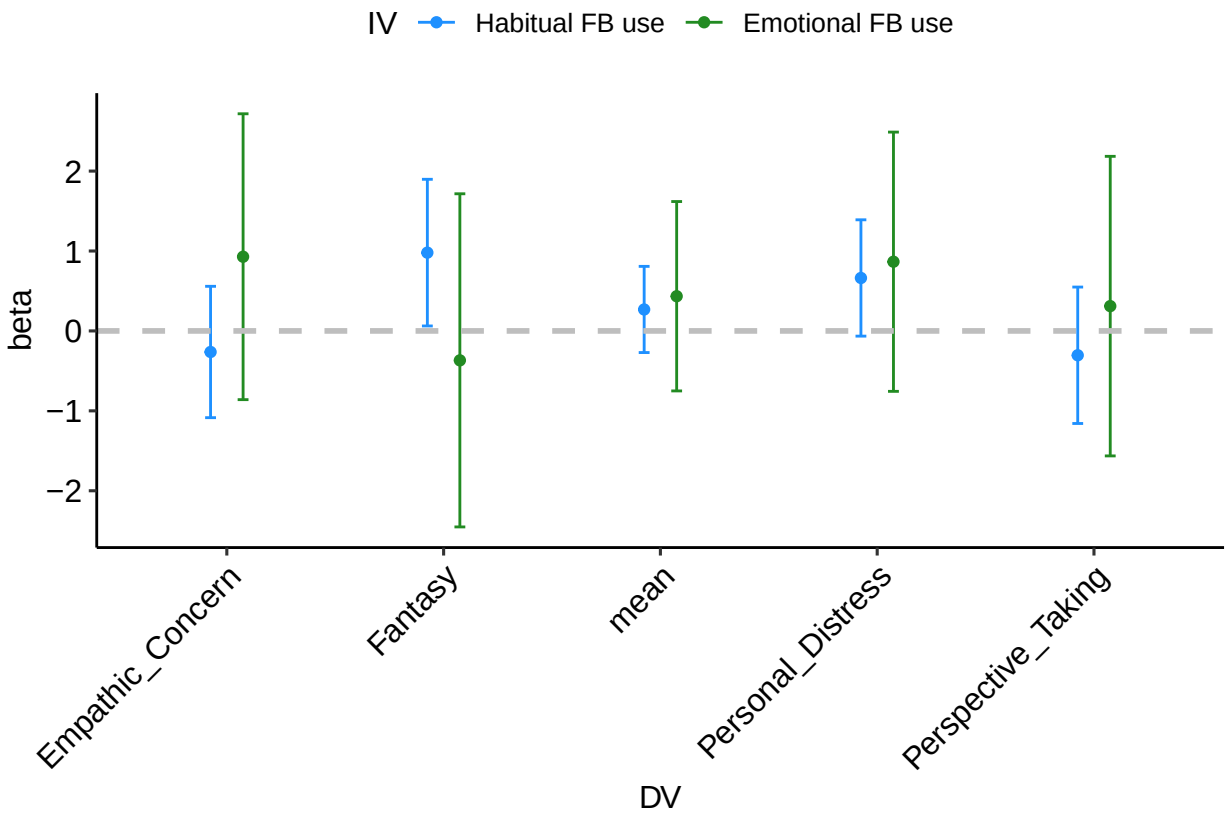


3. Linking FB use with self-report

```
# ## scale the IVs and DVs
# df = df %>%
#   mutate_at(c(3:26,29:37), scale1)
```

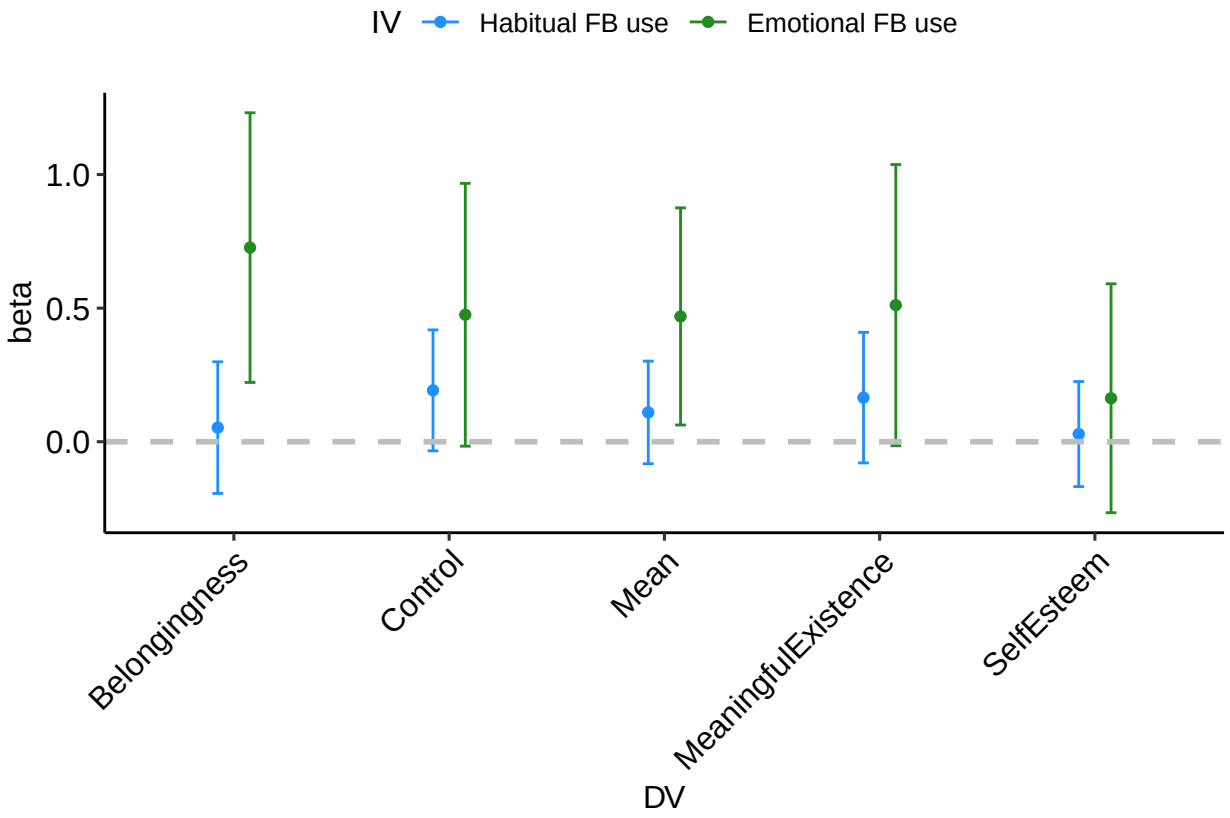
IRI

```
dvs = c("IRI_Perspective_Taking","IRI_Fantasy",
                  "IRI_Empathic_Concern","IRI_Personal_Distress", "IRI_mean")

coef_df = construct_coef_df(dvs, df) %>%
  mutate(DV = gsub("IRI_", "", DV))

plot_coef_df(coef_dv)
```



NTS

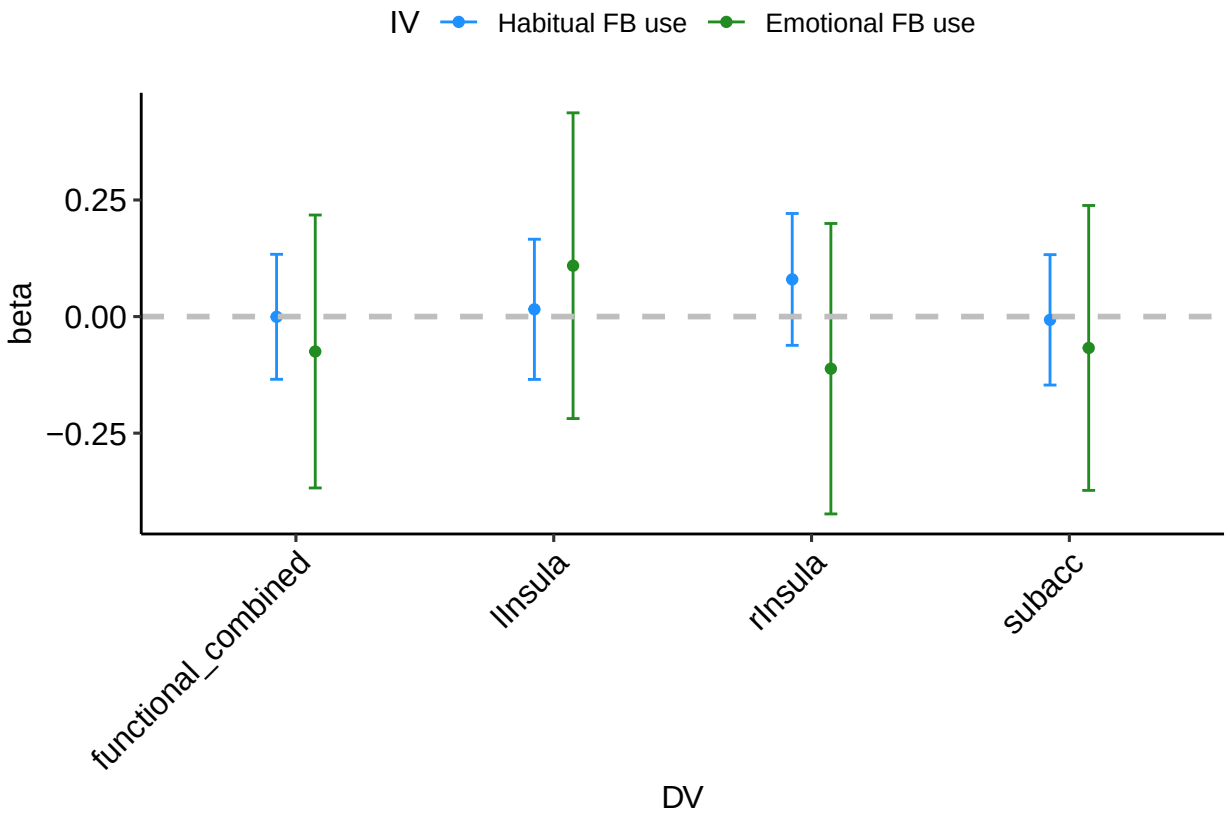
```
dvs = c("NTS_Belongingness","NTS_SelfEsteem",
        "NTS_Control", "NTS_MeaningfulExistence","NTS_Mean")

coef_df = construct_coef_df(dvs, df) %>%
  mutate(DV = gsub("NTS_", "", DV))
plot_coef_df(coef_dv)
```



4. Linking FB use with neural data

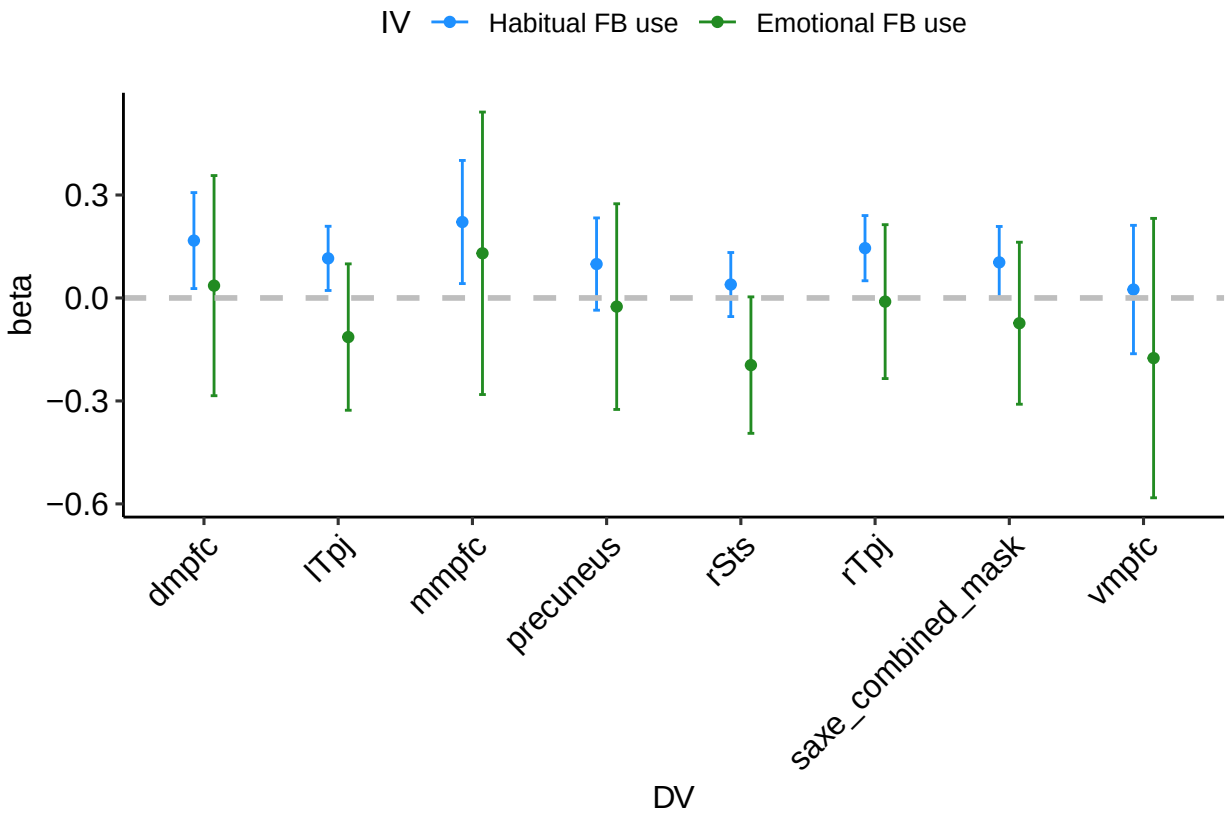
Functional ROIS

```
dvs = c("subacc_roi_mask","rInsula_roi_mask",
        "lInsula_roi_mask", "functional_combined_roi_mask")

coef_df = construct_coef_df(dvs, df) %>%
  mutate(DV = gsub("_roi_mask", "", DV))
plot_coef_df(coef_dv)
```



Dufour mentalzing meta-analytic ROIS

```
dvs = c("vmpfc_roi_mask","dmpfc_roi_mask","mmpfc_roi_mask", "precuneus_roi_mask",          "lTpj_roi_
        "saxe_combined_mask")

coef_df = construct_coef_df(dvs, df) %>%
  mutate(DV = gsub("_roi_mask", "", DV))
plot_coef_df(coef_dv)
```




```
md = lm(saxe_combined_mask ~ fhab +age + study_t + scanner_t, data = df)
summ(md, digits = 3)
```

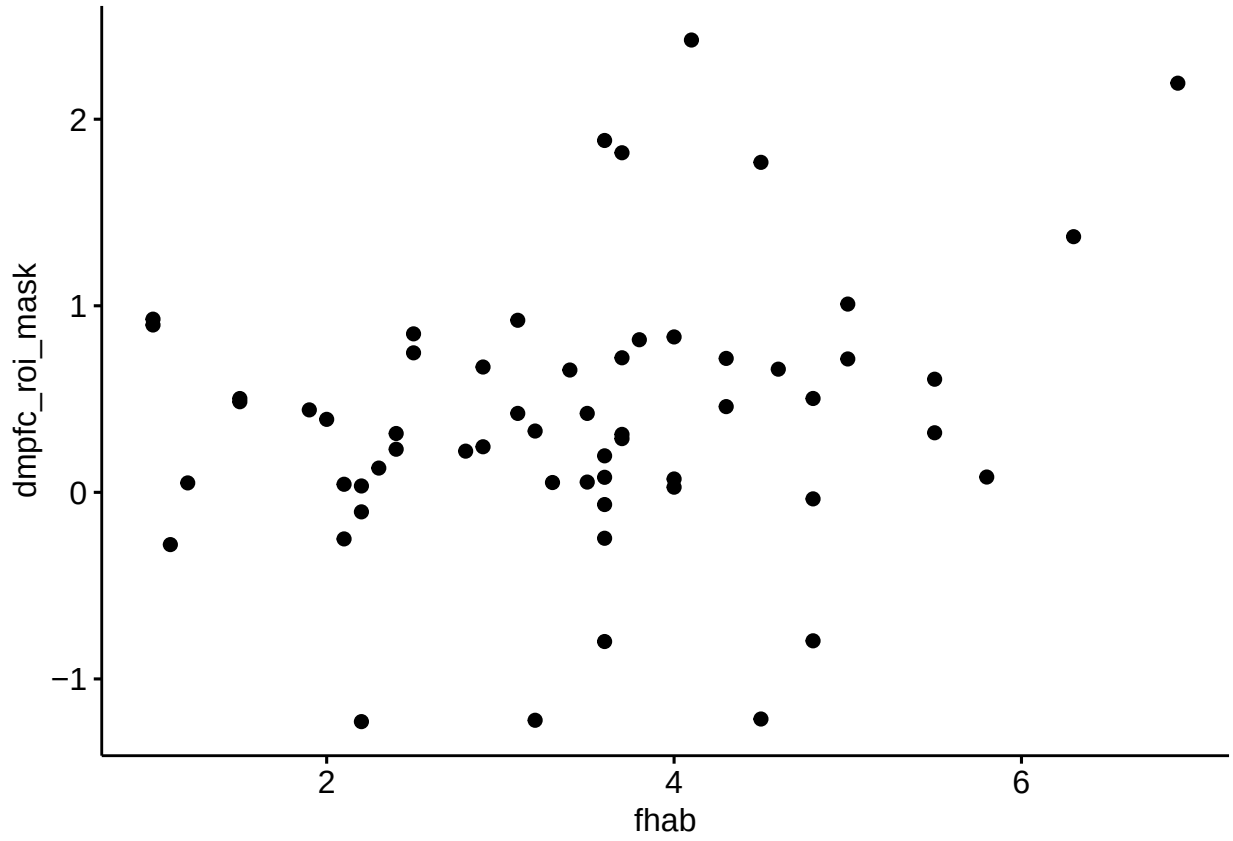
Observations	59
Dependent variable	saxe_combined_mask
Type	OLS linear regression

F(4,54)	2.216
R^2	0.141
Adj. R^2	0.077

	Est.	S.E.	t val.	p
(Intercept)	0.175	3.271	0.054	0.957
fhab	0.101	0.057	1.778	0.081
age	-0.035	0.190	-0.184	0.854
study_ttps2	0.528	0.204	2.593	0.012
scanner_tscanner2	-0.323	0.200	-1.615	0.112

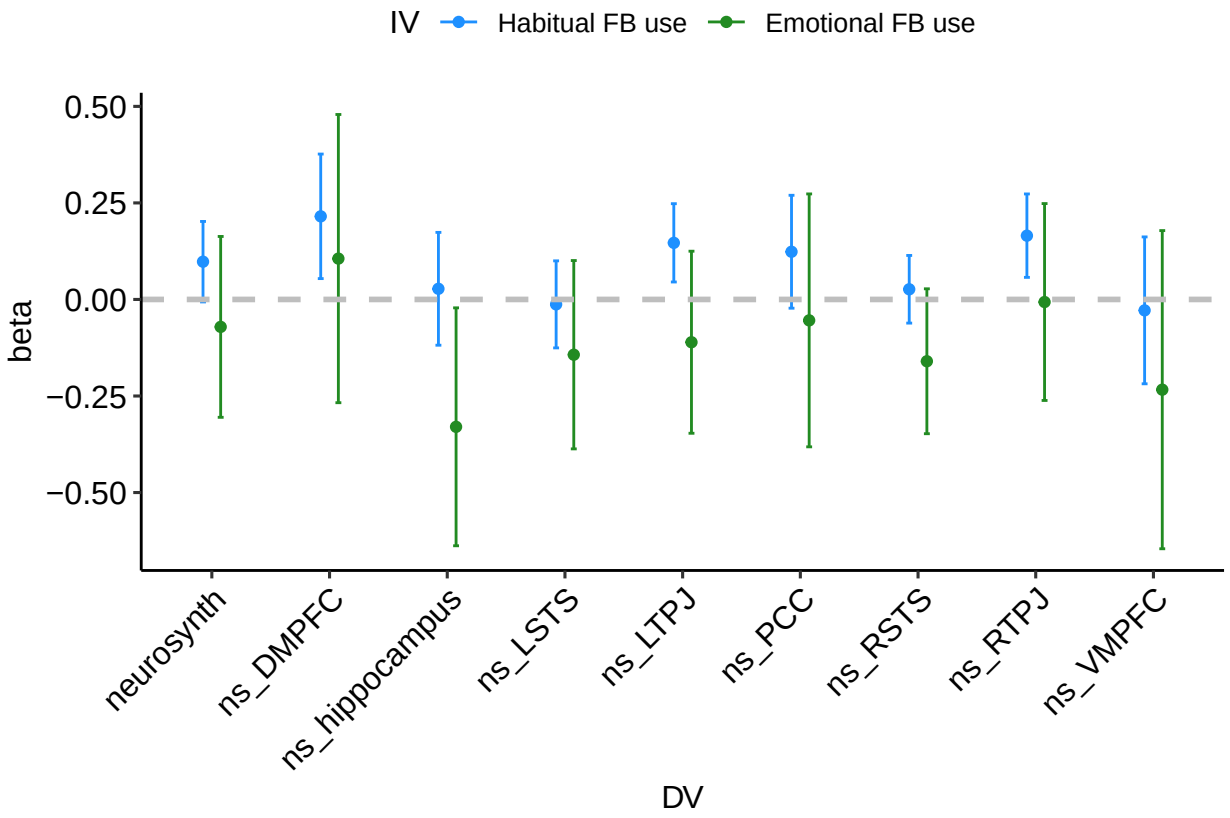
Standard errors: OLS

```
ggscatter(df, "fhab", "dmpfc_roi_mask")
```



Neurosynth mentalzing ROIS

```
df = df %>%
  dplyr::rename(.,neurosynth_mask = ns_ment_mask,
                ns_RSTS = ns_ment_clust1_mask,
                ns_RTPJ = ns_ment_clust2_mask,
                ns_DMPFC = ns_ment_clust3_mask,
                ns_PCC = ns_ment_clust4_mask,
                ns_VMPFC = ns_ment_clust5_mask,
                ns_hippocampus = ns_ment_clust6_mask,
                ns_LTPJ = ns_ment_clust7_mask,
                ns_LSTS = ns_ment_clust8_mask)
dvs = c("neurosynth_mask","ns_RSTS", "ns_RTPJ",
        "ns_DMPFC","ns_PCC" ,"ns_VMPFC",
        "ns_hippocampus","ns_LTPJ","ns_LSTS")

coef_df = construct_coef_df(dvs, df) %>%
  mutate(DV = gsub("_mask", "", DV))
plot_coef_df(coef_dv)
```




```
md = lm(neurosynth_mask ~ fhab + age + study_t + scanner_t, data = df)
summ(md, digits = 3)
```

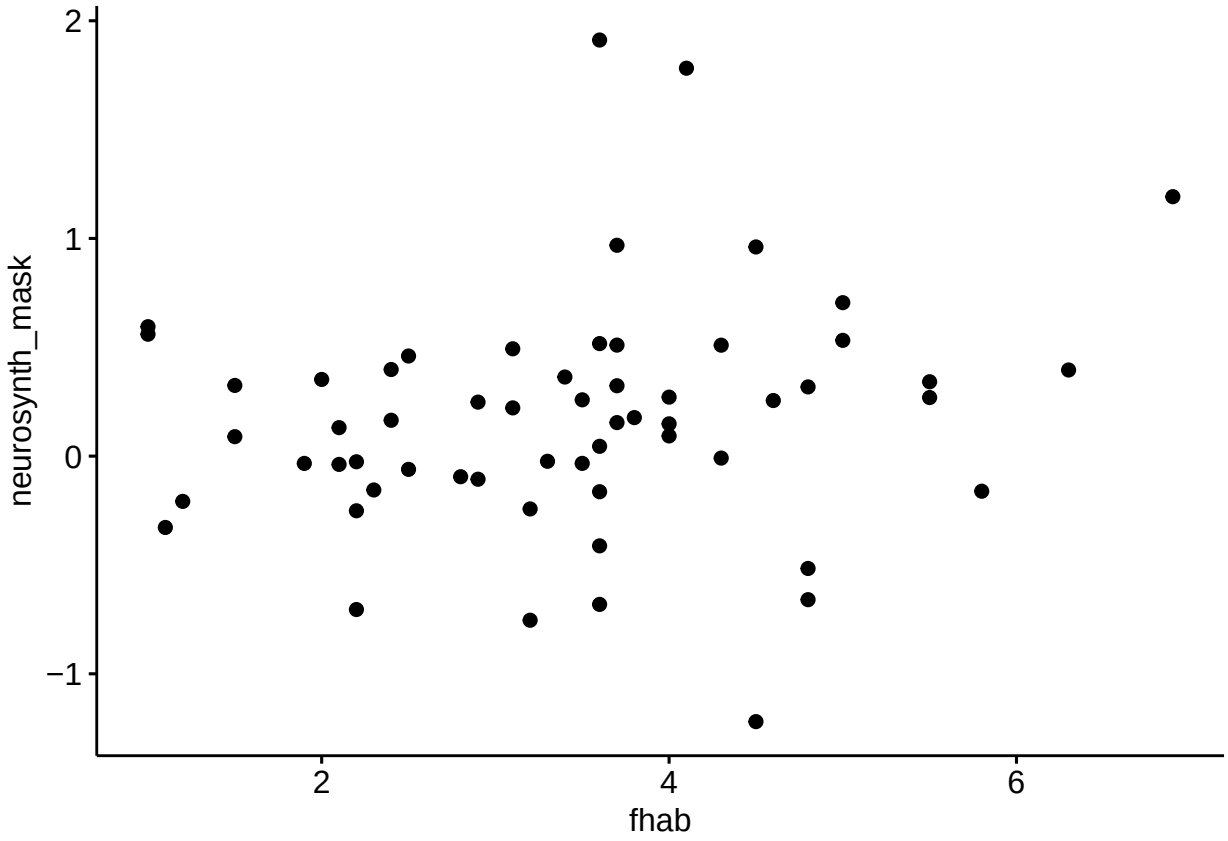
Observations	59
Dependent variable	neurosynth_mask
Type	OLS linear regression

F(4,54)	1.761
R^2	0.115
Adj. R^2	0.050

	Est.	S.E.	t val.	p
(Intercept)	0.603	3.255	0.185	0.854
fhab	0.093	0.056	1.653	0.104
age	-0.053	0.189	-0.282	0.779
study_ttps2	0.451	0.203	2.222	0.030
scanner_tscanner2	-0.277	0.199	-1.391	0.170

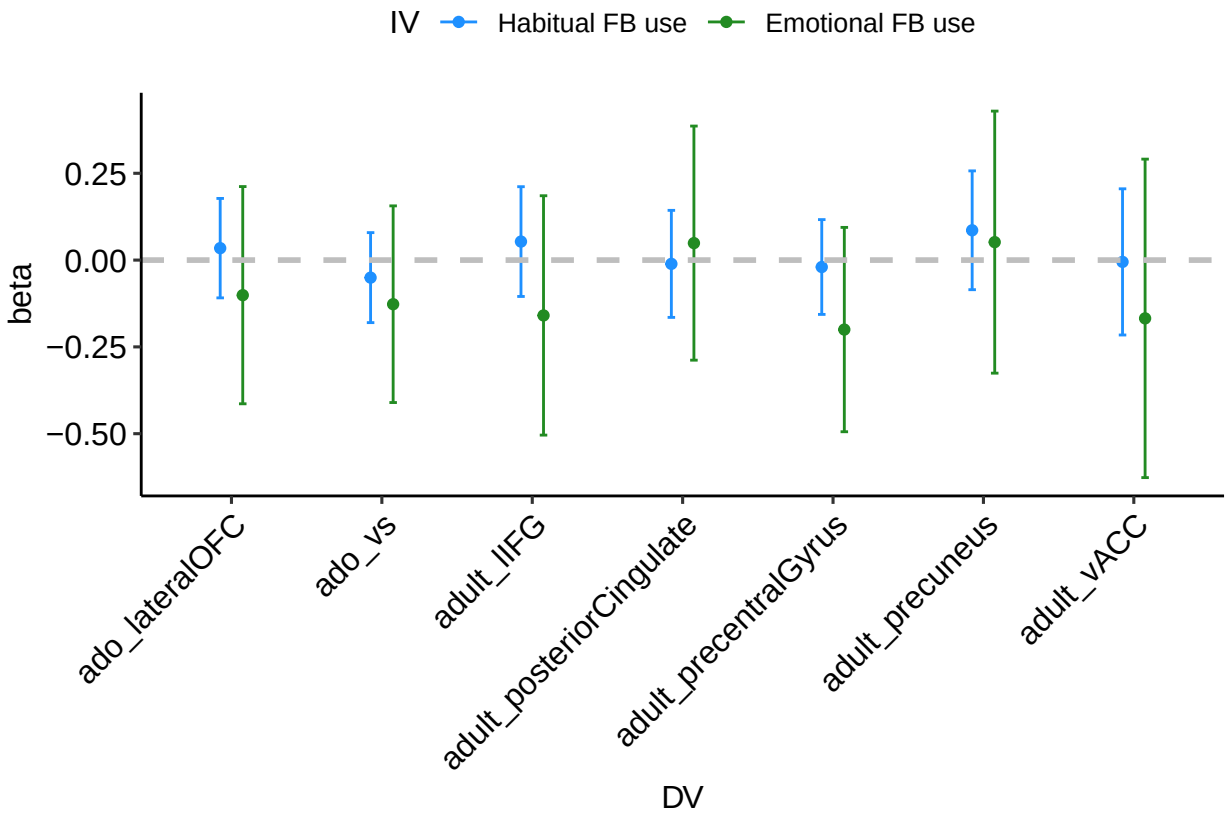
Standard errors: OLS

```
ggscatter(df, "fhab", "neurosynth_mask")
```



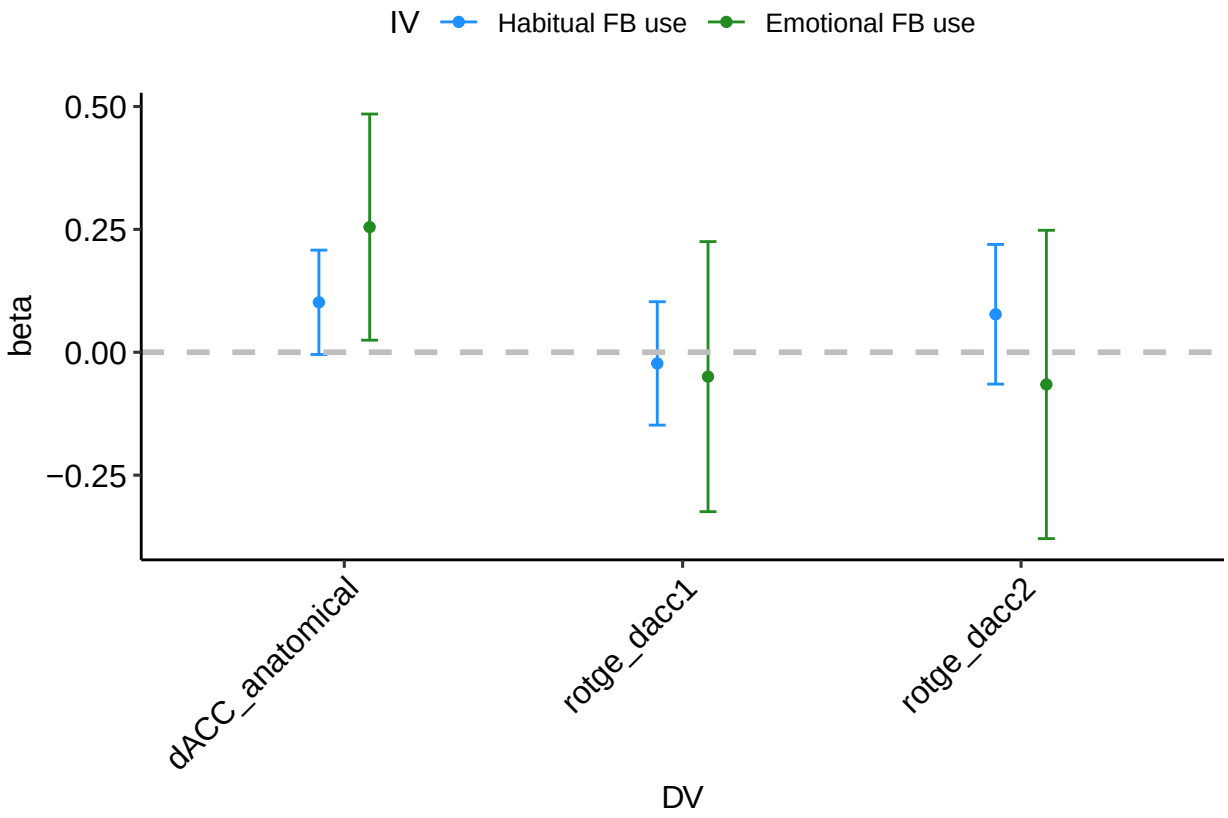
Vijayakumar social exclusion meta-analytic ROIs

```
dvs = c("adult_lIFG_mask","adult_vACC_mask",
        "adult_posteriorCingulate_mask","adult_precuneus_mask",
        "adult_precentralGyrus_mask" ,
        "ado_lateralOFC_mask", "ado_vs_mask")

coef_df = construct_coef_df(dvs, df) %>%
  mutate(DV = gsub("_mask", "", DV))
plot_coef_df(coef_dv)
```



dACC ROIS

```
dvs = c("rotge_dacc1_mask","rotge_dacc2_mask",
        "dACC_anatomical")

coef_df = construct_coef_df(dvs, df) %>%
  mutate(DV = gsub("_mask", "", DV))
plot_coef_df(coef_dv)
```

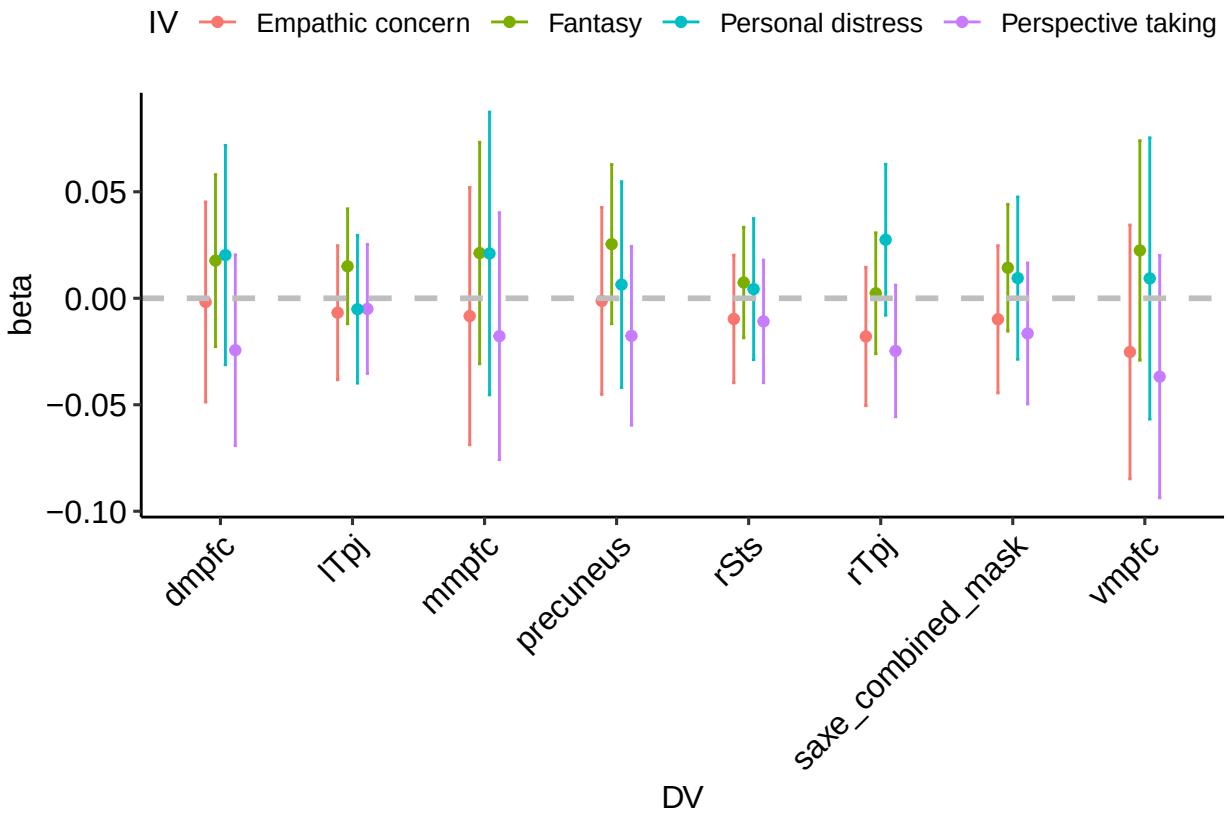



```
# md = lm(dACC_anatomical_mask ~ femo + age + study_t + scanner_t, data = df)
# summary(md)
```

5. Link between neural activation in mentalizing regions and IRI

Dufour et al. meta-analytic regions

```
dvs = c("vmpfc_roi_mask","dmpfc_roi_mask","mmpfc_roi_mask", "precuneus_roi_mask",          "lTpj_roi_
        "saxe_combined_mask")

coef_df = construct_coef_df_iri(dvs, df) %>%
  mutate(DV = gsub("_roi_mask", "", DV))
plot_coef_iri(coef_dv)
```




```
md = lm(saxe_combined_mask ~ IRI_Fantasy + age + study_t + scanner_t, data = df)
summ(md)
```

Observations	59
Dependent variable	saxe_combined_mask
Type	OLS linear regression

F(4,54)	1.62
R^2	0.11
Adj. R^2	0.04

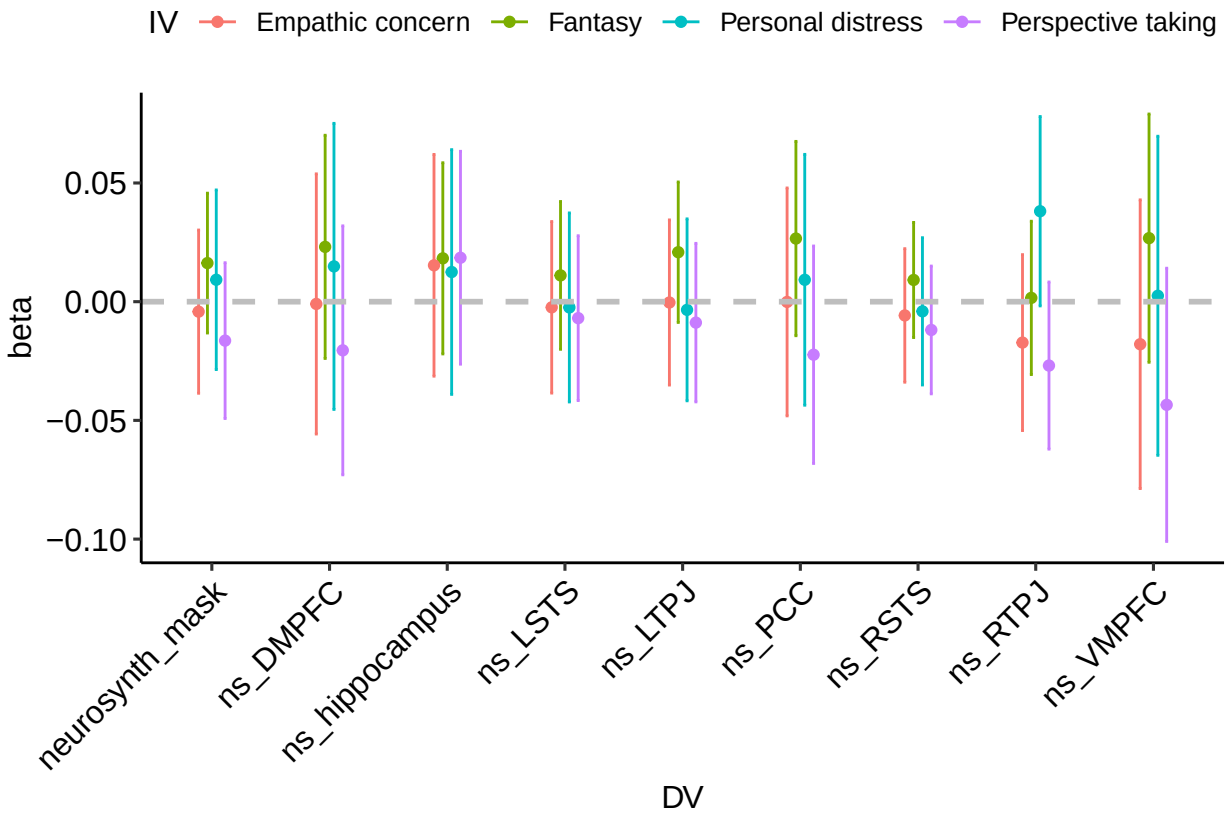
	Est.	S.E.	t val.	p
(Intercept)	2.14	3.11	0.69	0.50
IRI_Fantasy	0.02	0.02	0.99	0.32
age	-0.15	0.19	-0.81	0.42
study_ttps2	0.41	0.21	1.96	0.05
scanner_tscanner2	-0.21	0.21	-0.98	0.33

Standard errors: OLS

Neurosynth regions

```
dvs = c("neurosynth_mask","ns_RSTS", "ns_RTPJ",
        "ns_DMPFC","ns_PCC" ,"ns_VMPFC",
```

```
        "ns_hippocampus","ns_LTPJ","ns_LSTS")

coef_df = construct_coef_df_iri(dvs, df) %>%
  mutate(DV = gsub("_roi_mask", "", DV))
plot_coef_iri(coef_dv)
```




```
md = lm(neurosynth_mask ~ IRI_Fantasy + age + study_t + scanner_t, data = df)
summ(md)
```

Observations	59
Dependent variable	neurosynth_mask
Type	OLS linear regression

F(4,54)	1.37
R^2	0.09
Adj. R^2	0.03

6. Link between neural activation in cyberball and NTS

```
## function
construct_coef_df_nts = function(dvs,df){
  mds1 = lapply(dvs, function(x) {
    lm(substitute(i ~ NTS_Belongingness + study_t + scanner_t, list(i = as.name(x))), data = df)
```

	Est.	S.E.	t val.	p
(Intercept)	2.39	3.08	0.78	0.44
IRI_Fantasy	0.02	0.02	1.14	0.26
age	-0.16	0.18	-0.89	0.38
study_ttps2	0.33	0.20	1.61	0.11
scanner_tscanner2	-0.16	0.21	-0.75	0.46

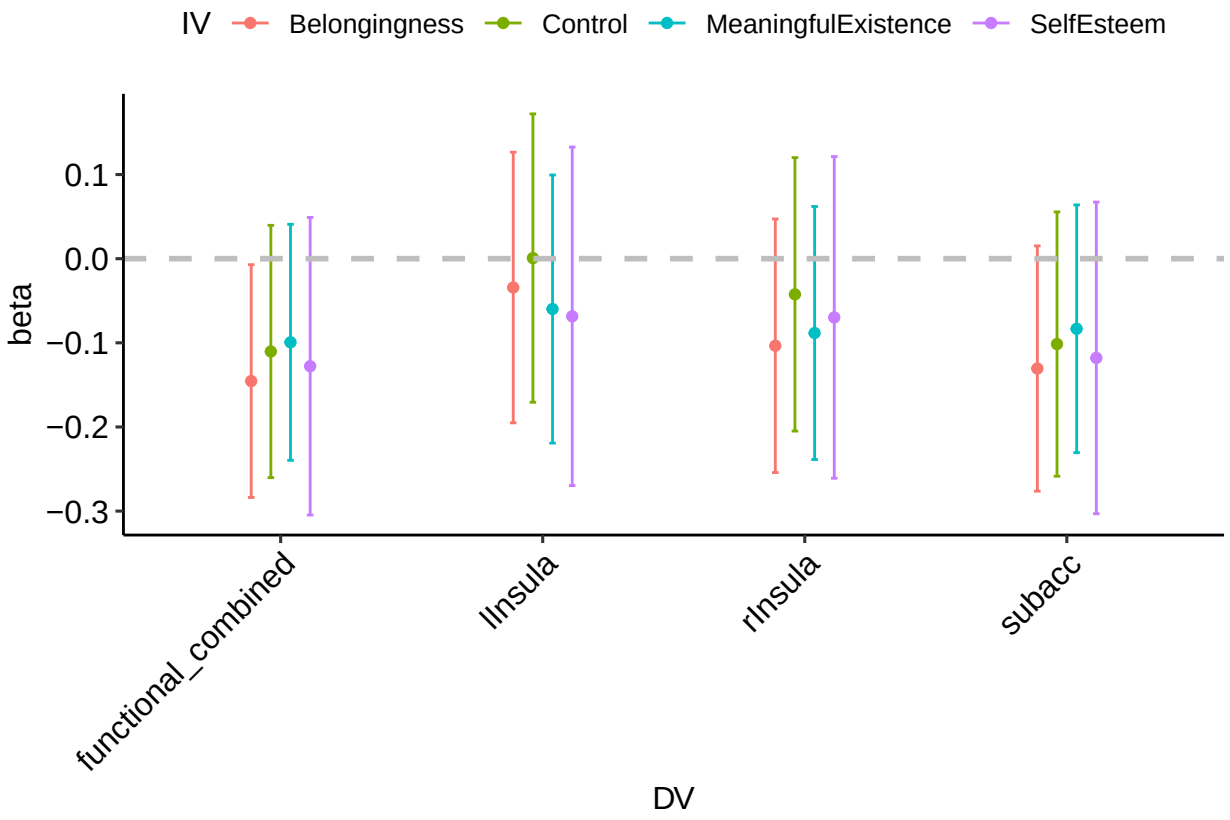
Standard errors: OLS

```
})
  sums1 = lapply(mds1, summary)

  mds2 = lapply(dvs, function(x) {
    lm(substitute(i ~ NTS_SelfEsteem + study_t + scanner_t, list(i = as.name(x))), data = df)
})
  sums2 = lapply(mds2, summary)

  mds3 = lapply(dvs, function(x) {
    lm(substitute(i ~ NTS_Control + study_t + scanner_t, list(i = as.name(x))), data = df)
})
  sums3 = lapply(mds3, summary)

  mds4 = lapply(dvs, function(x) {
    lm(substitute(i ~ NTS_MeaningfulExistence + study_t + scanner_t, list(i = as.name(x))), data = df)
})
  sums4 = lapply(mds4, summary)

  betas1 = c()
  betas2 = c()
  betas3 = c()
  betas4 = c()
  se1 = c()
  se2 = c()
  se3 = c()
  se4 = c()
  for (i in c(1:length(dvs))){
    betas1 = c(betas1,sums1[[i]]$coefficients[2,1])
    betas2 = c(betas2,sums2[[i]]$coefficients[2,1])
    betas3 = c(betas3,sums3[[i]]$coefficients[2,1])
    betas4 = c(betas4,sums4[[i]]$coefficients[2,1])
    se1 = c(se1, sums1[[i]]$coefficients[2,2])
    se2 = c(se2, sums2[[i]]$coefficients[2,2])
    se3 = c(se3, sums3[[i]]$coefficients[2,2])
    se4 = c(se4, sums4[[i]]$coefficients[2,2])
  }

  coef_df = data.frame("IV" = c(rep("Belongingness", length(dvs)),
                                rep("SelfEsteem", length(dvs)),
                                rep("Control", length(dvs)),
                                rep("MeaningfulExistence", length(dvs))),
                       "DV" = c(dvs,dvs,dvs,dvs),
                       "beta" = c(betas1, betas2,betas3,betas4),
                       "se" = c(se1, se2,se3,se4))
```

```
  coef_df = coef_df
  return(coef_df)
}
```

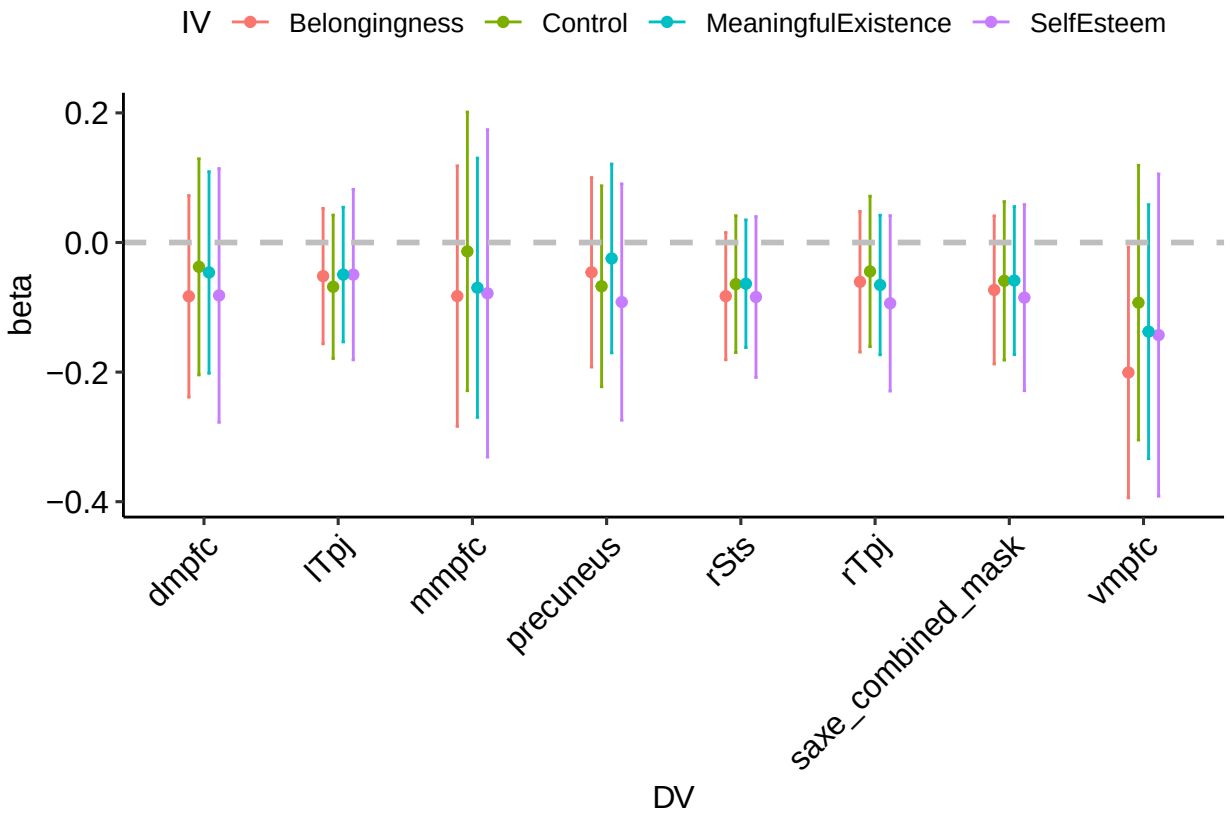
Functional ROIS

```
dvs = c("subacc_roi_mask","rInsula_roi_mask",
        "lInsula_roi_mask", "functional_combined_roi_mask")

coef_df = construct_coef_df_nts(dvs, df) %>%
  mutate(DV = gsub("_roi_mask", "", DV))
plot_coef_iri(coef_dv)
```



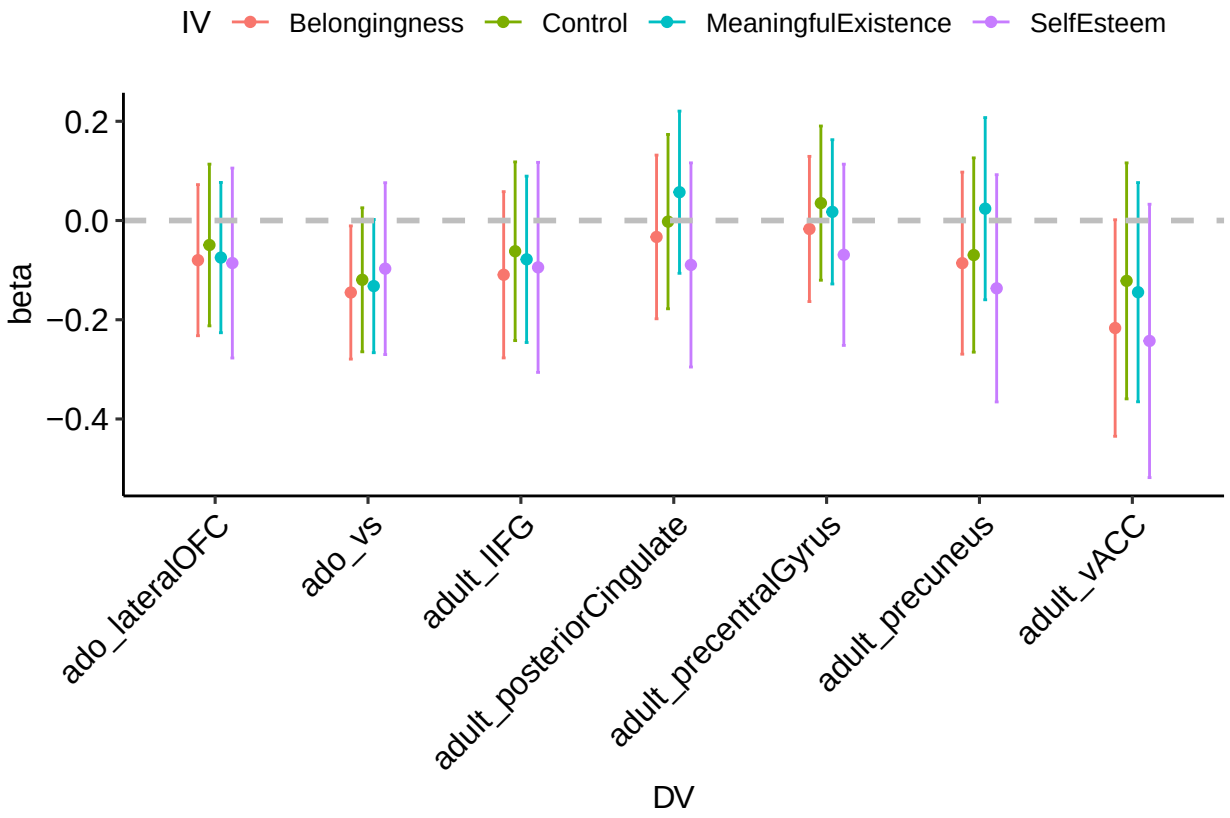
Dufour mentalzing meta-analytic ROIS

```
dvs = c("vmpfc_roi_mask","dmpfc_roi_mask","mmpfc_roi_mask", "precuneus_roi_mask",          "lTpj_roi_
        "saxe_combined_mask")
coef_df = construct_coef_df_nts(dvs, df) %>%
  mutate(DV = gsub("_roi_mask", "", DV))
plot_coef_iri(coef_dv)
```



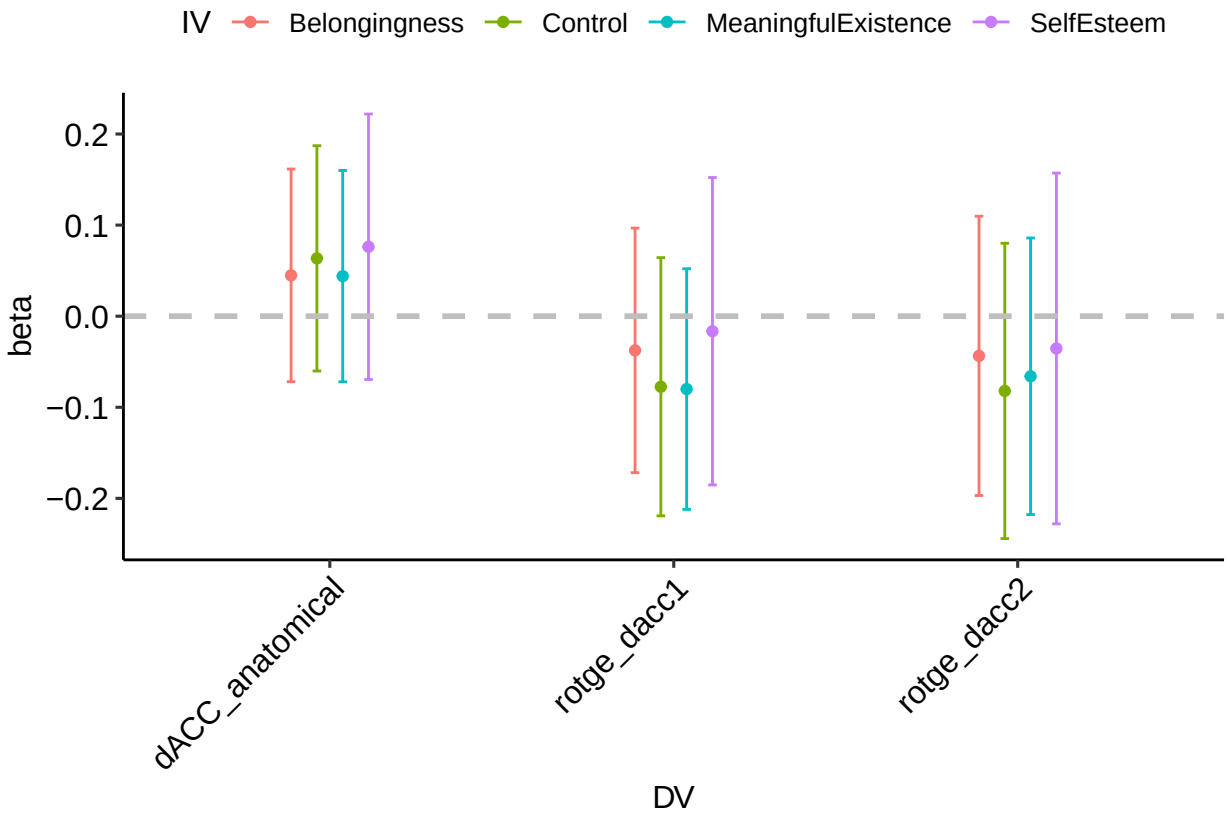
Vijayakumar social exclusion meta-analytic ROIs

```
dvs = c("adult_lIFG_mask","adult_vACC_mask",
        "adult_posteriorCingulate_mask","adult_precuneus_mask",
        "adult_precentralGyrus_mask" ,
        "ado_lateralOFC_mask", "ado_vs_mask")

coef_df = construct_coef_df_nts(dvs, df) %>%
  mutate(DV = gsub("_mask", "", DV))
plot_coef_iri(coef_dv)
```



dACC ROIS

```
dvs = c("rotge_dacc1_mask","rotge_dacc2_mask",
        "dACC_anatomical")

coef_df = construct_coef_df_nts(dvs, df) %>%
  mutate(DV = gsub("_mask", "", DV))
plot_coef_iri(coef_dv)
```


IV
Belongingness
Control
MeaningfulExistence
SelfEsteem
beta
0.2
0.1
0.0
−0.1
−0.2
dACC_anatomical
rotge_dacc1
rotge_dacc2
DV