# Analyses for event segmentation thesis

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These are the final analyses reported in the thesis.

# Main analyses

The analyses for the main research questions: responses for bounds versus perms, modulation by salience, and plotting these with the FIR models.

## Calculate significance of bounds vs perms and plot results

Define which data to use for all results:

```
re_time <- '3s'
if (re_time == '3s') {is_3s <- TRUE} else {is_3s <- FALSE}</pre>
```

### Statistical analyses

```
datapath <- paste(indir,'AVG_boundperms_',re_time,'_all.csv', sep="")
data <- read_csv(datapath, col_types = cols())</pre>
```

Warning: Missing column names filled in: 'X1' [1]

```
results <- data.frame(area=areas, beta=0, p=1, perm_mean=0, perm_sd=0)
for (a in areas) {
   perm_betas <- data$beta[data$perm != 0 & data$area == a]
   bound_beta <- data$beta[data$perm_betas > bound_beta]/length(perm_betas) # in fact we just want 1-tailed
   results$beta[results$area == a] <- bound_beta
   results$p[results$area == a] <- p
   results$perm_mean[results$area == a] <- mean(perm_betas)
   results$perm_sd[results$area == a] <- sd(perm_betas)
}
results$p_adj <- p.adjust(results$p, method='holm')
results[results$area == areas[49],]</pre>
```

```
49 Hippocampus 13.26718 0.002 -0.008989224 4.532317 0.076
topareas <- results[results$p_adj < 0.05,]</pre>
topareas <- topareas[order(topareas$beta, decreasing = TRUE),]</pre>
topareas
                                        area
                                                  beta p
                                                           perm_mean perm_sd
30
         Cingulate Gyrus, posterior division 33.16503 0 -0.49977993 5.279699
47
                       Supracalcarine Cortex 31.98627 0 -0.40716621 4.712508
24
                       Intracalcarine Cortex 29.30523 0 -0.40588941 4.398255
                               Cuneal Cortex 27.72143 0 -0.36579824 5.289758
32
29
          Cingulate Gyrus, anterior division 26.18845 0 -0.49351962 4.547156
31
                           Precuneous Cortex 24.28142 0 -0.08424546 5.692329
36
                               Lingual Gyrus 24.26136 0 -0.32052527 4.311826
2
                              Insular Cortex 21.53005 0 0.18433728 4.460171
41
                    Frontal Operculum Cortex 19.22037 0 -0.80496618 4.639239
35 Parahippocampal Gyrus, posterior division 17.98500 0 -0.03943910 4.632752
                       Frontal Medial Cortex 17.93846 0 0.02131449 4.398071
   p_adj
30
       0
47
       0
24
       0
32
       0
29
       0
31
       0
36
       0
2
       0
41
       0
35
       0
25
       0
for (i in 1:nrow(topareas)) {
  writeLines(paste(topareas$area[i], round(topareas$beta[i],3), round(topareas$p_adj[i],6), sep = '\t')
}
Cingulate Gyrus, posterior division 33.165 0
Supracalcarine Cortex
                        31.986 0
Intracalcarine Cortex
                        29.305 0
Cuneal Cortex
              27.721 0
Cingulate Gyrus, anterior division 26.188 0
Precuneous Cortex
                    24.281 0
Lingual Gyrus
                24.261 0
Insular Cortex 21.53
Frontal Operculum Cortex
                            19.22
Parahippocampal Gyrus, posterior division
                                            17.985 0
Frontal Medial Cortex
                        17.938 0
write_csv(results, paste(outdir, 'boundperm_results_',re_time,'.csv', sep=""))
sigs <- as.numeric(rownames(topareas))</pre>
```

perm\_mean perm\_sd p\_adj

area

sigs <- append(sigs, 49)

beta

р

#### Plot bounds vs perms

We'll only plot this for HC, the rest will make do with FIRs and stats

```
# Hippocampus
a = c(49)
asel <- areas[a]</pre>
bounds <- subset(data, perm == 0 & area %in% asel)</pre>
perms <- subset(data, perm != 0 & area %in% asel)</pre>
png(paste(figdir, 'bounds_vs_perms_',re_time,'_HC.png',sep=""), width=550, height=400)
ggplot(data=perms, aes(group=area)) +
  geom histogram(aes(x=beta), binwidth = 2) +
  geom_vline(mapping=aes(xintercept=beta), data=bounds, color='red', size=2) +
  facet wrap( \sim area, ncol=2) +
  labs(x='beta values', y='count (total 1000)', title=paste('Latency -', re_time, sep="")) +
  scale_x_continuous(limits=c(-17,17), breaks=seq(-15,15,5)) +
  theme_grey(base_size = 25)
dev.off()
pdf
  2
```

#### Plot FIRs

We'll only use the ones from 3s, with an added line for the 1s bound. In addition to HC we'll plot all areas that are significant in either condition, because the reader should have all that info available – it sucks when you don't have it. For this first part we plot the grand average over all saliences as that's more clearly interpretable in conjunction with the cortical plots.

```
datapath <- paste(indir,'AVG_fir_',re_time,'_all.csv', sep="")
data <- read_csv(datapath, col_types = cols())</pre>
```

Warning: Missing column names filled in: 'X1' [1]

```
data$cond <- as.factor(data$cond)
data$cond <- ordered(data$cond, levels=c('high','mid','low','all'))

data <- filter(data, grepl('delay', regressor))
data$delay <- as.numeric(gsub(data$regressor, pattern="[^0-9]", replacement=""))
data$delay <- data$delay - 8  # adjust latencies so that behav boundary is 0
areas <- unique(data$area)

# Plot HC large
conds <- c('high','mid','low')
asel <- areas[49]
plotdata <- subset(data, area %in% asel & cond %in% conds)
png(paste(figdir,'bounds_FIRs_',re_time,'_HC.png', sep=""), width=1100, height=700)
pd <- position_dodge(0.2)
ggplot(plotdata, aes(x=delay, y=beta, color=cond)) +
    geom_hline(yintercept = 0, color='grey50', size=1) +
    geom_vline(xintercept = -3, color='blue3', size=2) + # latency = -3</pre>
```

```
geom_vline(xintercept = -1, color='green3', size=2) + # latency = -1
  geom_point(position=pd, size=8) +
  geom_line(position=pd, size=3) +
  geom_errorbar(aes(ymin=beta-se, ymax=beta+se), width=0.5, size=1, position=pd) +
  scale_color_grey(name='salience') +
  scale_x_continuous(breaks=seq(-8,7)) +
  labs(title=paste('Modulation by salience')) +
  xlab('delay (sec)') +
  facet_wrap(~ area, ncol=1) +
  theme bw(base size = 24) +
  theme(panel.grid=element_blank())
dev.off()
pdf
  2
# Plot others small, grand average
conds <- c('all')</pre>
asel <- areas[sigs]</pre>
plotdata <- subset(data, area %in% asel & cond %in% conds)</pre>
pd <- position_dodge(0.2)</pre>
png(paste(figdir, 'bounds_FIRs_',re_time, '_avg_cortical.png', sep=""), width=1400, height=700)
ggplot(plotdata, aes(x=delay, y=beta)) +
  geom hline(yintercept = 0, color='grey50', size=1) +
  geom_vline(xintercept = -3, color='blue3', size=2) + # latency = -3
  geom_vline(xintercept = -1, color='green3', size=2) + # latency = -1
  geom_point(position=pd, size=3) +
  geom_line(position=pd, size=1) +
  geom_errorbar(aes(ymin=beta-se, ymax=beta+se), width=0.3, size=0.7, position=pd) +
  #scale_color_grey(name='salience') +
  scale_x_continuous(breaks=seq(-8,7)) +
  scale_y_continuous(limits=c(-7,12), breaks=seq(-5,10,5)) +
  labs(title=paste('FIR responses')) +
  xlab('delay (sec)') +
  facet_wrap(~ area, ncol=4, scales='free') +
  theme_bw(base_size = 20) +
  theme(panel.grid=element_blank())
dev.off()
pdf
  2
```

## Modulation by salience

```
suppressMessages(library(lme4))
suppressMessages(library(lmerTest))
```

Warning: package 'lmerTest' was built under R version 4.0.2

```
datapath <- paste(indir, 'betas_series_long_',re_time,'.csv', sep="")</pre>
data <- read_csv(datapath, col_types = cols())</pre>
Warning: Missing column names filled in: 'X1' [1]
boundpath <- paste('/Users/jenska/code/python/eventcode/1_create_boundaries/out/boundaries_f',re_time,'
bounds <- read_csv(boundpath, col_types = cols())</pre>
Warning: Missing column names filled in: 'X1' [1]
areas <- unique(data$area)</pre>
# Salience for bounds (bins: 5-6, 7-9, 10-17)
bounds$salience <- 0</pre>
bounds$salience[bounds$nobs >= 7] <- 1</pre>
bounds$salience[bounds$nobs >= 10] <- 2</pre>
# Clean some unnecessary columns
bounds <- subset(bounds, select=c(id,nobs,salience,meanvol,voldiff))</pre>
data <- subset(data, select=-c(X1))</pre>
data <- dplyr::rename(data, id=bound)</pre>
# Also remove the post-hoc bounds from data
data <- data[data$id != 999,]</pre>
# Join the data frames
data <- inner_join(data,bounds, by="id")</pre>
# Check correlations
cor.test(bounds$salience, bounds$meanvol)
    Pearson's product-moment correlation
data: bounds$salience and bounds$meanvol
t = -2.4113, df = 77, p-value = 0.01828
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
-0.45917825 -0.04658916
sample estimates:
     cor
-0.26497
cor.test(bounds$salience, bounds$voldiff)
    Pearson's product-moment correlation
data: bounds$salience and bounds$voldiff
t = 0.86378, df = 77, p-value = 0.3904
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
-0.1258731 0.3123093
```

```
sample estimates:
     cor
0.0979638
cor.test(bounds$nobs, bounds$meanvol)
   Pearson's product-moment correlation
data: bounds$nobs and bounds$meanvol
t = -2.4307, df = 77, p-value = 0.0174
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
-0.46085634 -0.04871285
sample estimates:
     cor
-0.266948
cor.test(bounds$nobs, bounds$voldiff)
   Pearson's product-moment correlation
data: bounds$nobs and bounds$voldiff
t = 1.2793, df = 77, p-value = 0.2046
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
-0.07937505 0.35408284
sample estimates:
     cor
0.1442673
# Run qlm on HC, salience
datasel <- data[data$area == areas[49],]</pre>
m1 <- lmer(beta ~ salience + meanvol + voldiff + (1 | subj) + (1 | id), datasel)
anova(m1)
Type III Analysis of Variance Table with Satterthwaite's method
        Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
salience 189669 189669 1
                               62 4.9862 0.02917 *
meanvol 7133 7133
                                62 0.1875 0.66650
                         1
voldiff 2048
                        1
                                62 0.0538 0.81726
                  2048
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
m11 <- lmer(beta ~ salience + (1 | subj) + (1 | id), datasel)
anova(m11)
Type III Analysis of Variance Table with Satterthwaite's method
        Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
salience 231126 231126 1
                              64 6.0761 0.0164 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Run qlm on HC, nObs
datasel <- data[data$area == areas[49],]</pre>
m2 <- lmer(beta ~ nobs + meanvol + voldiff + (1 | subj) + (1 | id), datasel)
anova(m2)
Type III Analysis of Variance Table with Satterthwaite's method
        Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
        204721 204721
                                62 5.3819 0.02365 *
                           1
nobs
                                62 0.0803 0.77777
meanvol
          3056
                  3056
                           1
                  1018
voldiff 1018
                           1
                                62 0.0268 0.87056
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
m21 <- lmer(beta ~ nobs + (1 | subj) + (1 | id), datasel)</pre>
anova(m21)
Type III Analysis of Variance Table with Satterthwaite's method
     Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
nobs 252186 252186
                        1 63.999 6.6297 0.01235 *
Signif. codes: 0 '***, 0.001 '**, 0.01 '*, 0.05 '.', 0.1 ', 1
# Run for all ROIs
results <- data.frame(area=areas, f_sal=0, p_sal=1, f_obs=0, p_obs=1)
for (i in c(1:49)) {
  datasel <- data[data$area == areas[i],]</pre>
  anv1 <- anova(lmer(beta ~ salience + meanvol + voldiff + (1 | subj) + (1 | id), datasel))
  anv2 <- anova(lmer(beta ~ nobs + meanvol + voldiff + (1 | subj) + (1 | id), datasel))
 results$f_sal[i] <- anv1$`F value`[1]</pre>
 results$p sal[i] <- anv1$`Pr(>F)`[1]
 results$f_obs[i] <- anv2$`F value`[1]</pre>
  results$p_obs[i] <- anv2$`Pr(>F)`[1]
boundary (singular) fit: see ?isSingular
```

```
boundary (singular) fit: see ?isSingular
results$p_sal_adj <- p.adjust(results$p_sal, method='holm')</pre>
results$p_obs_adj <- p.adjust(results$p_obs, method='holm')</pre>
# Select results
topareas <- results[results$p_sal_adj < 0.05 | results$p_obs_adj < 0.05,]
topareas <- topareas[order(topareas$f_obs, decreasing = T),]</pre>
topareas
                                          f_sal
                                                        p_sal
                                                                  f_obs
                                  area
30 Cingulate Gyrus, posterior division 16.16975 0.0001596161 25.211842
29 Cingulate Gyrus, anterior division 16.56074 0.0001356087 21.466149
36
                         Lingual Gyrus 15.28863 0.0002313237 19.312237
24
                 Intracalcarine Cortex 16.20530 0.0001572619 17.745262
47
                 Supracalcarine Cortex 15.42152 0.0002186605 16.438793
41
              Frontal Operculum Cortex 12.39913 0.0008119068 13.314467
                    Subcallosal Cortex 12.26186 0.0004684480 6.055388
                  p_sal_adj
                               p obs adj
          p_obs
30 4.617616e-06 0.007548569 0.0002262632
29 1.898349e-05 0.006644829 0.0009112077
36 4.424307e-05 0.010409565 0.0020794244
24 8.325590e-05 0.007548569 0.0038297716
47 1.426661e-04 0.010058383 0.0064199744
41 5.417898e-04 0.034911991 0.0238387500
27 1.666250e-02 0.020611711 0.4998750318
results[results$area == areas[49],]
                                                 p_obs p_sal_adj p_obs_adj
                  f sal
                             p_sal
                                      f_obs
49 Hippocampus 4.986202 0.02916968 5.381894 0.02365382 0.9625996 0.6445788
for (i in 1:nrow(topareas)) {
  writeLines(paste(topareas$area[i], round(topareas$f_sal[i],3), round(topareas$p_sal[i],3), round(topareas$p_sal[i],3)
Cingulate Gyrus, posterior division 16.17
                                                0.008
                                                         25.212 0
Cingulate Gyrus, anterior division 16.561 0
                                                0.007
                                                        21.466 0
                                                                     0.001
Lingual Gyrus 15.289 0
                            0.01
                                    19.312 0
                                                0.002
```

17.745 0 0.004

0.008

Intracalcarine Cortex 16.205 0

```
Supracalcarine Cortex 15.422 0 0.01 16.439 0 0.006
Frontal Operculum Cortex 12.399 0.001 0.035 13.314 0.001 0.024
Subcallosal Cortex 12.262 0 0.021 6.055 0.017 0.5
```

```
sigs <- as.numeric(rownames(topareas))
sigs <- append(sigs, 49)</pre>
```

Without confounds

```
# Run for all ROIs
results_nocon <- data.frame(area=areas, f_sal=0, p_sal=1, f_obs=0, p_obs=1)

for (i in c(1:49)) {
   datasel <- data[data$area == areas[i],]
   anv1 <- anova(lmer(beta ~ salience + (1 | subj) + (1 | id), datasel))
   anv2 <- anova(lmer(beta ~ nobs + (1 | subj) + (1 | id), datasel))
   results_nocon$f_sal[i] <- anv1$^F value^[1]
   results_nocon$p_sal[i] <- anv1$^Pr(>F)^[1]
   results_nocon$f_obs[i] <- anv2$^F value^[1]
   results_nocon$p_obs[i] <- anv2$^Pr(>F)^[1]
}
```

```
boundary (singular) fit: see ?isSingular
```

```
results_nocon$p_sal_adj <- p.adjust(results_nocon$p_sal, method='holm')</pre>
results_nocon$p_obs_adj <- p.adjust(results_nocon$p_obs, method='holm')</pre>
# Select results
topareas_nocon <- results_nocon[results_nocon$p_sal_adj < 0.05 | results_nocon$p_obs_adj < 0.05,]
topareas_nocon <- topareas_nocon[order(topareas_nocon$f_obs, decreasing = T),]</pre>
topareas_nocon
                                  area
                                            f_sal
                                                         p_sal
                                                                   f_obs
   Cingulate Gyrus, anterior division 20.251710 2.931893e-05 25.857183
30 Cingulate Gyrus, posterior division 14.236319 3.548487e-04 21.134587
                         Lingual Gyrus 16.518809 1.341377e-04 20.238947
                 Intracalcarine Cortex 16.956117 1.117642e-04 18.092706
24
41
              Frontal Operculum Cortex 15.931704 1.717041e-04 17.332846
47
                 Supracalcarine Cortex 15.776815 1.833296e-04 16.378016
28
                   Paracingulate Gyrus 8.235508 5.560652e-03 12.671881
                    Subcallosal Cortex 12.302040 4.585032e-04 6.225487
27
                  p_sal_adj
                               p_obs_adj
          p_obs
29 3.441027e-06 0.001436627 0.0001686103
30 2.070082e-05 0.015613342 0.0009936394
36 2.946771e-05 0.006304473 0.0013849824
24 6.994200e-05 0.005364680 0.0032173318
41 9.559825e-05 0.007898389 0.0043019214
47 1.422910e-04 0.008249830 0.0062608051
28 7.061029e-04 0.205744115 0.0303624239
27 1.264096e-02 0.019715637 0.3829092245
results nocon[results nocon$area == areas[49],]
          area
                  f_sal
                            p_sal
                                     f_obs
                                                 p_obs p_sal_adj p_obs_adj
49 Hippocampus 6.076052 0.0163965 6.629705 0.01235191 0.5738776 0.3829092
for (i in 1:nrow(topareas_nocon)) {
  writeLines(paste(topareas_nocon$area[i], round(topareas_nocon$f_sal[i],3), round(topareas_nocon$p_sal
}
Cingulate Gyrus, anterior division 20.252 0
                                                 0.001
                                                         25.857 0
                                                                     0
Cingulate Gyrus, posterior division 14.236 0
                                                0.016
                                                         21.135 0
                                                                     0.001
Lingual Gyrus
               16.519 0
                            0.006
                                    20.239 0
                                                 0.001
Intracalcarine Cortex
                                    0.005
                        16.956 0
                                             18.093 0
                                                         0.003
Frontal Operculum Cortex
                            15.932
                                    0
                                        0.008
                                                 17.333
                                                         0
                                                             0.004
Supracalcarine Cortex
                       15.777 0
                                    0.008
                                             16.378 0
                                                         0.006
Paracingulate Gyrus 8.236
                            0.006
                                    0.206
                                             12.672 0.001
                                                             0.03
Subcallosal Cortex 12.302 0
                                0.02
                                        6.225
                                                0.013
                                                         0.383
sigs_nocon <- as.numeric(rownames(topareas_nocon))</pre>
sigs_nocon <- append(sigs, 49)</pre>
```

### Plot FIRs

With salience

```
datapath <- paste(indir,'AVG_fir_',re_time,'_all.csv', sep="")</pre>
data <- read_csv(datapath, col_types = cols())</pre>
Warning: Missing column names filled in: 'X1' [1]
data$cond <- as.factor(data$cond)</pre>
data$cond <- ordered(data$cond, levels=c('high','mid','low','all'))</pre>
data <- filter(data, grepl('delay', regressor))</pre>
data$delay <- as.numeric(gsub(data$regressor, pattern="[^0-9]", replacement=""))
data$delay <- data$delay - 8 # adjust latencies so that behav boundary is 0
areas <- unique(data$area)</pre>
conds <- c('high', 'mid', 'low')</pre>
# Plot all with salience
asel <- areas[sigs]</pre>
plotdata <- subset(data, area %in% asel & cond %in% conds)</pre>
pd <- position_dodge(0.2)</pre>
png(paste(figdir, 'bounds_FIRs_', re_time, '_sal_cortical.png', sep=""), width=1400, height=500)
ggplot(plotdata, aes(x=delay, y=beta, color=cond)) +
  geom_hline(yintercept = 0, color='grey50', size=1) +
  geom_vline(xintercept = -3, color='blue3', size=2) + # latency = -3
  #geom_vline(xintercept = -1, color='green3', size=2) + # latency = -1
  geom_point(position=pd, size=3) +
  geom line(position=pd, size=1) +
  geom errorbar(aes(ymin=beta-se, ymax=beta+se), width=0.3, size=0.7, position=pd) +
  scale_color_grey(name='salience') +
  scale_x_continuous(breaks=seq(-8,7)) +
  scale_y_continuous(limits=c(-7,18), breaks=seq(-5,15,5)) +
  labs(title=paste('FIR responses')) +
  xlab('delay (sec)') +
  facet_wrap(~ area, ncol=4, scales='free') +
  theme_bw(base_size = 20) +
  theme(panel.grid=element_blank())
dev.off()
pdf
```

# Secondary analyses: familiarity effects

Nothing is even close to significant at 3s, even less at 1s – does it make sense to even plot any of these? Well maybe HC has theoretical interest, so we should mention it.

```
library(effsize)
```

Warning: package 'effsize' was built under R version 4.0.2

```
datapath <- paste(indir,'betas_boundperms_long_',re_time,'.csv', sep="")</pre>
data <- read_csv(datapath, col_types = cols())</pre>
Warning: Missing column names filled in: 'X1' [1]
areas <- unique(data$area)</pre>
# Get subj info and merge
subjdata <- read_csv('code/python/eventcode/subj_info.csv', col_types = cols())</pre>
data <- merge(data, subjdata)</pre>
results <- data.frame(area = areas, mean_lis1=0, mean_lis2=0, t=0, p=1, d=0)
for (a in areas) {
 lis1_betas <- data$beta[data$listening == 'first' & data$area == a]</pre>
  lis2 betas <- data$beta[data$listening == 'second' & data$area == a]
  stat <- t.test(lis1 betas,lis2 betas)</pre>
  eff <- cohen.d(lis1_betas, lis2_betas)</pre>
  results$mean_lis1[results$area == a] <- stat$estimate[1]</pre>
  results$mean_lis2[results$area == a] <- stat$estimate[2]</pre>
  results$t[results$area == a] <- stat$statistic</pre>
  results$p[results$area == a] <- stat$p.value
  results$d[results$area == a] <- eff$estimate</pre>
results$p_adj <- p.adjust(results$p, method='holm')</pre>
rm(lis1_betas,lis2_betas,stat,eff)
results[results$area == areas[49],]
          area mean_lis1 mean_lis2
                                             t
49 Hippocampus 14.36218 12.08093 0.2906752 0.7725962 0.08264152
topareas <- results[results$d > 0.3,]
topareas <- topareas[order(topareas$p, decreasing = FALSE),]</pre>
topareas
                                            area mean_lis1 mean_lis2
11
      Middle Temporal Gyrus, anterior division -0.8446181 -15.926837 1.827171
43
                     Parietal Operculum Cortex 9.4054551 0.949726 1.350540
15 Inferior Temporal Gyrus, posterior division 2.4565525 -4.607025 1.217448
42
                       Central Opercular Cortex 6.5383614 -2.131595 1.191225
29
            Cingulate Gyrus, anterior division 29.6676144 22.419352 1.141616
32
                                  Cuneal Cortex 31.4039873 23.731998 1.127764
30
           Cingulate Gyrus, posterior division 37.3968033 28.580617 1.067881
            Temporal Occipital Fusiform Cortex 12.9878692 4.736630 1.065569
39
                       d p_adj
11 0.07389533 0.5156857
43 0.18321916 0.3823119
15 0.22969264 0.3466994
42 0.23947423 0.3373737
29 0.25935049 0.3235867
```

```
32 0.26502910 0.3185144
30 0.29097949 0.3026303
39 0.29195438 0.3002527
for (i in 1:nrow(topareas)) {
 writeLines(paste(topareas$area[i], round(topareas$mean_lis1[i],3), round(topareas$mean_lis2[i],3), r
}
Middle Temporal Gyrus, anterior division
                                           -0.845 -15.927 0.516
                                                                  0.074
                                                                          1
Parietal Operculum Cortex
                          9.405
                                   0.95
                                           0.382
                                                  0.183
                                                          1
Inferior Temporal Gyrus, posterior division 2.457
                                                                  0.23
                                                  -4.607 0.347
                                                                          1
                                   -2.132 0.337
Central Opercular Cortex
                                                  0.239
                           6.538
                                                          1
Cingulate Gyrus, anterior division 29.668 22.419 0.324
                                                          0.259
              31.404 23.732 0.319
Cuneal Cortex
                                      0.265
                                                          0.291
Cingulate Gyrus, posterior division 37.397 28.581 0.303
Temporal Occipital Fusiform Cortex 12.988 4.737
                                                  0.3 0.292
```

### Plot swarms

I.e. the values that went into the t-test

```
library(ggbeeswarm)
```

Warning: package 'ggbeeswarm' was built under R version 4.0.2

```
asel <- areas[49]
plotdata <- subset(data, area %in% asel)
png(paste(figdir,'fam_swarm_',re_time, '_HC.png', sep=""), width=600, height=500)
ggplot(plotdata, aes(x=listening, y=beta)) +
   geom_violin(color='grey40', fill='bisque2', size=1, draw_quantiles=0.5) +
   geom_quasirandom(dodge.width=1, size=4) +
   labs(title=paste('Effect of familiarity on HC response')) +
   xlab('Listening') +
   theme_bw(base_size = 24) +
   theme(panel.grid=element_blank())
dev.off()</pre>
```

pdf 2

## Plot FIRs by grp

For HC

```
datapath1 <- paste(indir,'AVG_fir_',re_time,'_1st.csv', sep="")
data1 <- read_csv(datapath1, col_types = cols())</pre>
```

Warning: Missing column names filled in: 'X1' [1]

```
datapath2 <- paste(indir,'AVG_fir_',re_time,'_2nd.csv', sep="")</pre>
data2 <- read_csv(datapath2, col_types = cols())</pre>
Warning: Missing column names filled in: 'X1' [1]
data1 <- subset(data1, cond=='all')</pre>
data2 <- subset(data2, cond=='all')</pre>
data1$lis <- '1st'
data2$lis <- '2nd'
data <- rbind(data1, data2)</pre>
rm(datapath1,datapath2,data1,data2)
data <- filter(data, grepl('delay', regressor))</pre>
data$delay <- as.numeric(gsub(data$regressor, pattern="[^0-9]", replacement=""))
data$delay <- data$delay - 5
areas <- unique(data$area)
# Plot HC
asel <- areas[49]
plotdata <- subset(data, area %in% asel)</pre>
png(paste(figdir, 'fam_FIRs_', re_time, '_HC.png', sep=""), width=1000, height=600)
pd <- position_dodge(0.2)</pre>
ggplot(plotdata, aes(x=delay, y=beta, color=lis)) +
  geom_hline(yintercept = 0, color='grey50', size=1) +
  geom vline(xintercept = 0, color='blue3', size=2) + # delay = 0
  geom_point(position=pd, size=8) +
  geom_line(position=pd, size=3) +
  geom_errorbar(aes(ymin=beta-se, ymax=beta+se), width=0.5, size=1, position=pd) +
  scale_color_grey(name='Listening') +
  scale_x_continuous(breaks=seq(-5,10)) +
  labs(title=paste('Modulation by familiarity')) +
  xlab('delay (sec)') +
  #facet_wrap(~ area, ncol=1) +
  theme_bw(base_size = 24) +
  theme(panel.grid=element_blank())
dev.off()
pdf
```

# Test bounds against non-bounds (audio gaps)

```
datapath <- paste(indir,'AVG_audioperms_',re_time,'_all.csv', sep="")
data <- read_csv(datapath, col_types = cols())</pre>
```

Warning: Missing column names filled in: 'X1' [1]

2

```
areas <- unique(data$area)</pre>
results <- data.frame(area=areas, beta=0, p=1, perm_mean=0, perm_sd=0)
for (a in areas) {
  perm_betas <- data$beta[data$perm != 0 & data$area == a]</pre>
  bound_beta <- data$beta[data$perm == 0 & data$area == a]</pre>
 p <- length(perm_betas[perm_betas > bound_beta])/length(perm_betas) # in fact we just want 1-tailed
 results$beta[results$area == a] <- bound beta
 results$p[results$area == a] <- p</pre>
  results$perm_mean[results$area == a] <- mean(perm_betas)</pre>
  results$perm_sd[results$area == a] <- sd(perm_betas)</pre>
results$p_adj <- p.adjust(results$p, method='holm')</pre>
results[results$area == areas[49],]
                   beta
                            p perm_mean perm_sd p_adj
          area
49 Hippocampus 10.69904 0.038 1.866707 5.190742 0.988
topareas <- results[results$p_adj < 0.05,]</pre>
topareas <- topareas[order(topareas$beta, decreasing = TRUE),]</pre>
topareas
                                            beta p perm_mean perm_sd p_adj
                                  area
  Cingulate Gyrus, anterior division 35.86869 0 4.766185 4.902877
2
                        Insular Cortex 32.23890 0 2.464093 4.757457
                                                                          0
32
                         Cuneal Cortex 27.61241 0 2.679093 5.498810
                                                                          0
47
                 Supracalcarine Cortex 27.29055 0 4.877853 4.849920
                                                                          0
                 Intracalcarine Cortex 25.83786 0 3.868627 4.599572
24
41
              Frontal Operculum Cortex 23.91681 0 6.535029 4.799281
                                                                          0
36
                         Lingual Gyrus 23.42647 0 1.656477 4.634595
                                                                          0
40
              Occipital Fusiform Gyrus 22.82935 0 4.721020 5.492654
                                                                          0
30 Cingulate Gyrus, posterior division 22.81602 0 4.309939 5.734724
                        Occipital Pole 17.49597 0 1.498964 4.420203
48
                                                                          0
for (i in 1:nrow(topareas)) {
  writeLines(paste(topareas$area[i], round(topareas$beta[i],3), round(topareas$p_adj[i],6), sep = '\t')
}
Cingulate Gyrus, anterior division 35.869 0
Insular Cortex 32.239 0
Cuneal Cortex 27.612 0
Supracalcarine Cortex 27.291 0
Intracalcarine Cortex
                        25.838 0
Frontal Operculum Cortex
                            23.917 0
Lingual Gyrus
               23.426 0
Occipital Fusiform Gyrus
                            22.829 0
Cingulate Gyrus, posterior division 22.816 0
Occipital Pole 17.496 0
write_csv(results, paste(outdir, 'audioperm_results_',re_time, '.csv', sep=""))
sigs <- as.numeric(rownames(topareas))</pre>
```

### Plot

Not sure if it makes sense to plot these... well maybe yes, there's room in the fig

```
# Plot HC
a = c(49)
asel <- areas[a]</pre>
bounds <- subset(data, perm == 0 & area %in% asel)</pre>
perms <- subset(data, perm != 0 & area %in% asel)</pre>
png(paste(figdir,'bounds_vs_audio_',re_time,'_HC.png',sep=""), width=500, height=400)
ggplot(data=perms, aes(group=area)) +
 geom_histogram(aes(x=beta), binwidth = 2) +
  geom vline(mapping=aes(xintercept=beta), data=bounds, color='red', size=2) +
 facet wrap( ~ area,ncol=2) +
 labs(x='beta values', y='count (total 1000)', title='Boundaries vs. audiogaps') +
  theme_grey(base_size = 25)
dev.off()
pdf
 2
# Plot others
asel <- areas[sigs]</pre>
bounds <- subset(data, perm == 0 & area %in% asel)</pre>
perms <- subset(data, perm != 0 & area %in% asel)</pre>
png(paste(figdir, 'bounds_vs_audio_',re_time, '_cortical.png', sep=""), width=1300, height=400)
ggplot(data=perms, aes(group=area)) +
 geom_histogram(aes(x=beta), binwidth = 2) +
  geom_vline(mapping=aes(xintercept=beta), data=bounds, color='red', size=1.5) +
  facet_wrap( ~ area,nrow=2) +
  labs(x='beta values', y='count (total 1000)') +
 theme_grey(base_size = 19)
dev.off()
pdf
  2
rm(data,bounds,perms,results,topareas,areas,a,asel,bound_beta,perm_betas,i,datapath,plotlist,p,sigs)
Warning in rm(data, bounds, perms, results, topareas, areas, a, asel,
bound_beta, : object 'plotlist' not found
```