

Analyses for event segmentation thesis

Jenni Saaristo

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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 <- '1s'
if (re_time == '3s') {is_3s <- TRUE} else {is_3s <- FALSE}
```

Statistical analyses

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

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

```
areas <- unique(data$area)

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$perm == 0 & data$area == a]
  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
  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],]
```

```

      area      beta      p perm_mean perm_sd p_adj
49 Hippocampus 5.500318 0.115 0.03081762 4.384556      1

```

```

topareas <- results[results$p_adj < 0.05,]
topareas <- topareas[order(topareas$beta, decreasing = TRUE),]
topareas

```

```

      area      beta p perm_mean perm_sd
30 Cingulate Gyrus, posterior division 32.05827 0 -0.59157923 5.201382
31 Precuneous Cortex 25.72171 0 -0.34749195 5.673152
47 Supracalcarine Cortex 25.57428 0 -0.61049747 4.718722
24 Intracalcarine Cortex 21.72983 0 -0.59643287 4.344622
32 Cuneal Cortex 21.21921 0 -0.72466522 5.255012
35 Parahippocampal Gyrus, posterior division 19.07078 0 -0.09085367 4.567164
36 Lingual Gyrus 16.97368 0 -0.53987416 4.273520
25 Frontal Medial Cortex 16.08432 0 0.25332684 4.395494
      p_adj
30      0
31      0
47      0
24      0
32      0
35      0
36      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 32.058 0
Precuneous Cortex 25.722 0
Supracalcarine Cortex 25.574 0
Intracalcarine Cortex 21.73 0
Cuneal Cortex 21.219 0
Parahippocampal Gyrus, posterior division 19.071 0
Lingual Gyrus 16.974 0
Frontal Medial Cortex 16.084 0

```

```

write_csv(results, paste(outdir, 'boundperm_results_', re_time, '.csv', sep = ""))
sigs <- as.numeric(rownames(topareas))

```

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]
bounds <- subset(data, perm == 0 & area %in% asel)
perms <- subset(data, perm != 0 & area %in% asel)

```

```
png(paste(figdir, 'bounds_vs_perms_', re_time, '_HC.png', sep=""), width=1000, height=600)
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. permutations') +
  theme_grey(base_size = 25)
dev.off()
```

pdf
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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="")
data <- read_csv(datapath, col_types = cols())
```

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

```
boundpath <- paste('/Users/jenska/code/python/eventcode/1_create_boundaries/out/boundaries_f', re_time, '.csv', sep="")
bounds <- read_csv(boundpath, col_types = cols())
```

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

```
areas <- unique(data$area)

# Salience for bounds (bins: 5-6, 7-9, 10-17)
bounds$salience <- 0
bounds$salience[bounds$nobs >= 7] <- 1
bounds$salience[bounds$nobs >= 10] <- 2

# Clean some unnecessary columns
bounds <- subset(bounds, select=c(id,nobs,salience,meanvol,voldiff))
data <- subset(data, select=-c(X1))
data <- dplyr::rename(data, id=bound)

# Also remove the post-hoc bounds from data
data <- data[data$id != 999,]

# Join the data frames
data <- inner_join(data, bounds, by="id")

# Run glm on HC, salience
```

```

dataset1 <- data[data$area == areas[49],]
m1 <- lmer(beta ~ salience + meanvol + voldiff + (1 | subj) + (1 | id), dataset1)
anova(m1)

```

Type III Analysis of Variance Table with Satterthwaite's method

	Sum Sq	Mean Sq	NumDF	DenDF	F value	Pr(>F)
salience	78069	78069	1	63.999	2.0636	0.1557
meanvol	12356	12356	1	63.999	0.3266	0.5697
voldiff	18804	18804	1	63.999	0.4970	0.4834

Run glm on HC, nObs

```

dataset1 <- data[data$area == areas[49],]
m2 <- lmer(beta ~ nobis + meanvol + voldiff + (1 | subj) + (1 | id), dataset1)
anova(m2)

```

Type III Analysis of Variance Table with Satterthwaite's method

	Sum Sq	Mean Sq	NumDF	DenDF	F value	Pr(>F)
nobis	91538	91538	1	64	2.4196	0.1248
meanvol	2541	2541	1	64	0.0672	0.7964
voldiff	18754	18754	1	64	0.4957	0.4839

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)) {
  dataset1 <- data[data$area == areas[i],]
  anv1 <- anova(lmer(beta ~ salience + meanvol + voldiff + (1 | subj) + (1 | id), dataset1))
  anv2 <- anova(lmer(beta ~ nobis + meanvol + voldiff + (1 | subj) + (1 | id), dataset1))
  results$f_sal[i] <- anv1$`F value`[1]
  results$p_sal[i] <- anv1$`Pr(>F)`[1]
  results$f_obs[i] <- anv2$`F value`[1]
  results$p_obs[i] <- anv2$`Pr(>F)`[1]
}

```

boundary (singular) fit: see ?isSingular

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```

results$p_sal_adj <- p.adjust(results$p_sal, method='holm')
results$p_obs_adj <- p.adjust(results$p_obs, method='holm')

```

Select results

```

topareas <- results[results$p_sal_adj < 0.05 | results$p_obs_adj < 0.05,]
topareas <- topareas[order(topareas$f_sal, decreasing = T),]
topareas

```

	area	f_sal	p_sal	f_obs	p_obs	p_sal_adj
36 Lingual Gyru	9.328812	0.003285878	12.99564	0.0006114604	0.161008	
		p_obs_adj				
36	0.02996156					

```
results[results$area == areas[49],]
```

	area	f_sal	p_sal	f_obs	p_obs	p_sal_adj	p_obs_adj
49 Hippocampus	2.063564	0.1557276	2.419582	0.1247585		1	1

```
for (i in 1:nrow(topareas)) {
  writeLines(paste(topareas$area[i], round(topareas$f_sal[i],3), round(topareas$p_sal[i],3), round(topa
})
```

Lingual Gyrus	9.329	0.003	0.161	12.996	0.001	0.03
---------------	-------	-------	-------	--------	-------	------

Plotting FIRs

We'll actually 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.

```
datapath <- paste(indir,'AVG_fir_',re_time,'_all.csv', sep="")
data <- read_csv(datapath, col_types = cols())
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 - 5

areas <- unique(data$area)
conds <- c('high','mid','low')
#conds <- c('all')

# Plot HC
asel <- areas[49]
plotdata <- subset(data, area %in% asel & cond %in% conds)
png(paste(figdir,'bounds_FIRs_',re_time,'_HC.png', sep=""), width=1000, height=600)
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 = 0, color='blue3', size=2) + # delay = 0
  geom_vline(xintercept = 2, color='green3', size=2) + # delay = 2
  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(-5,10)) +
  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()
```

```

# Plot others
plotlist = list()
anums = sigs
for (i in 1:length(anums)) {
  asel <- areas[anums[i]]
  plotdata <- subset(data, area %in% asel & cond %in% conds)
  pd <- position_dodge(0.2)
  p <- ggplot(plotdata, aes(x=delay, y=beta, color=cond)) +
    geom_hline(yintercept = 0, color='grey50', size=1) +
    geom_vline(xintercept = 0, color='blue3', size=2) + # delay = 0
    geom_vline(xintercept = 2, color='green3', size=2) + # delay = 2
    geom_point(position=pd, size=6) +
    geom_line(position=pd, size=2) +
    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(-5,10)) +
    scale_y_continuous(limits=c(-7,18), breaks=seq(-5,15,5)) +
    labs(title=paste('Modulation by salience')) +
    xlab('delay (sec)') +
    facet_wrap(~ area, ncol=1) +
    theme_bw(base_size = 24) +
    theme(panel.grid=element_blank())
  plotlist[[i]] <- p
}

for (i in 1:length(anums)) {
  png(paste(figdir, 'bounds_FIRs_', re_time, '_', anums[[i]], '.png', sep=""), width=800, height=500)
  print(plotlist[[i]])
  dev.off()
}

```

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="")
data <- read_csv(datapath, col_types = cols())
```

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

```
areas <- unique(data$area)

# Get subj info and merge
subjdata <- read_csv('code/python/eventcode/subj_info.csv', col_types = cols())
data <- merge(data, subjdata)
```

```

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$area == a]
  lis2_betas <- data$beta[data$listening == 'second' & data$area == a]
  stat <- t.test(lis1_betas, lis2_betas)
  eff <- cohen.d(lis1_betas, lis2_betas)
  results$mean_lis1[results$area == a] <- stat$estimate[1]
  results$mean_lis2[results$area == a] <- stat$estimate[2]
  results$t[results$area == a] <- stat$statistic
  results$p[results$area == a] <- stat$p.value
  results$d[results$area == a] <- eff$estimate
}
results$p_adj <- p.adjust(results$p, method='holm')

rm(lis1_betas, lis2_betas, stat, eff)

results[results$area == areas[49],]

```

	area	mean_lis1	mean_lis2	t	p	d	p_adj
49	Hippocampus	4.518362	6.564103	-0.3168905	0.7528393	-0.09058043	1

```

topareas <- results[results$d > 0.3,]
topareas <- topareas[order(topareas$p, decreasing = FALSE),]
topareas

```

	area	mean_lis1	mean_lis2	t	p	d	p_adj
31	Precuneous Cortex	33.55874	17.23161	2.213658	0.03164716	0.6256473	1
32	Cuneal Cortex	25.85798	16.19388	1.166743	0.24925551	0.3315474	1

```

for (i in 1:nrow(topareas)) {
  writeLines(paste(topareas$area[i], round(topareas$mean_lis1[i],3), round(topareas$mean_lis2[i],3), r
}

```

Precuneous Cortex	33.559	17.232	0.626	0.032	1
Cuneal Cortex	25.858	16.194	0.332	0.249	1

Plot FIRs by grp?

Yes this should be done, though only for HC

```

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

```

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

```

datapath2 <- paste(indir, 'AVG_fir_', re_time, '_2nd.csv', sep='')
data2 <- read_csv(datapath2, col_types = cols())

```

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

```
data1 <- subset(data1, cond=='all')
data2 <- subset(data2, cond=='all')
data1$lis <- '1st'
data2$lis <- '2nd'
data <- rbind(data1, data2)
rm(datapath1,datapath2,data1,data2)

data <- filter(data, grepl('delay', regressor))
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)
png(paste(figdir, 'fam_FIRs_', re_time, '_HC.png', sep=""), width=1000, height=600)
pd <- position_dodge(0.2)
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_vline(xintercept = 2, color='green3', size=2) + # delay = 2
  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
2

Test bounds against non-bounds (audio gaps)

```
datapath <- paste(indir, 'AVG_audioperms_', re_time, '_all.csv', sep="")
data <- read_csv(datapath, col_types = cols())
areas <- unique(data$area)

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$perm == 0 & data$area == a]
  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
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],]

topareas <- results[results$p_adj < 0.05,]
topareas <- topareas[order(topareas$beta, decreasing = TRUE),]
topareas

for (i in 1:nrow(topareas)) {
  writeLines(paste(topareas$area[i], round(topareas$beta[i],3), round(topareas$p_adj[i],6), sep = '\t'))
}

write_csv(results, paste(outdir,'audioperm_results_',re_time,'.csv', sep=""))
sigs <- as.numeric(rownames(topareas))

```

Plot

Not sure if it makes sense to plot these... for HC maybe? It would be nice to be able to plot FIRs contrasted with a non-bound, but I guess it doesn't really make sense...

```

# Plot HC
a = c(49)
asel <- areas[a]
bounds <- subset(data, perm == 0 & area %in% asel)
perms <- subset(data, perm != 0 & area %in% asel)
png(paste(figdir,'bounds_vs_audio_',re_time,'_HC.png',sep=""), width=1000, height=600)
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()

# Plot others
plotlist = list()
for (i in 1:length(sigs)) {
  asel <- areas[sigs[i]]
  bounds <- subset(data, perm == 0 & area %in% asel)
  perms <- subset(data, perm != 0 & area %in% asel)
  p <- 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)
  plotlist[[i]] <- p
}
for (i in 1:length(sigs)) {
  png(paste(figdir,'bounds_vs_audio_',re_time, '_', sigs[[i]], '.png', sep=""), width=800, height=500)

```

```
print(plotlist[[i]])  
dev.off()  
}
```