F03 mci emotion mixed models.R

2021-02-25

```
## MCI EMO MIXED MODELS SCRIPT ##
# Computes linear mixed-effects regression models with simple contrast coding for the fixed effects of semantics and
# emotional context. Thus, in each model, the estimate of the intercept is the grand mean, while the estimates of the
# slopes contrast "treatment" levels to their respective reference levels (semantics: violation - intuitive, mci -
# intuitive; emotional context (negative - neutral). The maximal random effects structure is used with all by-
# participant and by-item random slopes and random intercepts. Correlations between random effects are removed if the
# model fails two converge with two different numerical optimizers. Planned follow-up contrasts are computed for the
# main effects and the effects of semantics separately within each type of emotional context.
# Load packages
                      # version 7.3-51.6
library(MASS)
library(lme4)
                      # version 1.1-23
library(lmerTest)
                      # version 3.1-2
                      # version 0.27-2
library(afex)
                      # version 1.4.8
library(emmeans)
library(tidyverse)
                      # Version 1.3.0
library(magrittr)
                      # Version 1.5
# Load preprocessed data
a1 <- readRDS("EEG/export/a1.RDS")</pre>
# Remove trials with errors or invalid RTs/ERPs
a1 %<>% filter(!error) %>% na.omit()
# Define simple contrast coding for context emotionality (negative - neutral)
      HO(Intercept): (mu1+mu2)/2 = 0 <-> mu1+mu2 = 0
     HO(Slope): -mu1 + mu2 = 0
```

```
with mu1 = mean of the neutral contexts and mu2 = mean of the neg contexts
t(contrasts.context \leftarrow t(cbind(c("neu" = -1, "neg" = 1))))
       [.1]
##
## neu -1
## neg
contrasts(a1$context) <- ginv(contrasts.context)</pre>
# Define simple contrast coding for semantics (violation - intuitive, mci - intuitive)
      HO(Intercept): (mu1+mu2+mu3)/3 = 0 <-> mu1+mu2+mu3 = 0
     HO(Slope1): -1*mu1 + 1*mu2 + 0*mu3 = 0
     HO(Slope2): -1*mu1 + 0*mu2 + 1*mu3 = 0
      with mu1 = mean of intuitive concepts, mu2 = mean of violations, mu3 = mean of MCIs
t(contrasts.semantics <- t(cbind(c("int" = -1, "vio" = 1, "mci" = 0),
                                 c("int" = -1, "vio" = 0, "mci" = 1))))
       [,1] [,2]
## int -1 -1
## vio 1
## mci
contrasts(a1$semantics) <- ginv(contrasts.semantics)</pre>
## LINEAR MIXED-EFFECTS MODELS ## ----
# LMM for valence ratings (converged on first attempt)
mod valence <- lmer(ValenzResp ~ context + (context|participant) + (context|item),</pre>
                    data = a1, control = lmerControl(calc.derivs = FALSE))
# LMM for arousal ratings (converged on first attempt)
mod aroursal <- lmer(ArousalResp ~ context + (context|participant) + (context|item),</pre>
                     data = a1, control = lmerControl(calc.derivs = FALSE))
# LMM for verb-related N400 (converged after changing the optimizer and removing correlations between REs)
mod_N400_verb <- lmer_alt(N400_verb ~ semantics*context + (semantics*context||participant) + (semantics*context||item),</pre>
                          data = a1, control = lmerControl(calc.derivs = FALSE,
                                                            optimizer = "bobyqa",
                                                            optCtrl = list(maxfun = 2e5)))
```

```
# LMM for picture-related N400 (converged after changing the optimizer)
mod_N400_pict <- lmer(N400_pict ~ semantics*context + (semantics*context|participant) + (semantics*context|item),</pre>
                      data = a1, control = lmerControl(calc.derivs = FALSE,
                                                      optimizer = "bobyga",
                                                       optCtrl = list(maxfun = 2e5)))
# LMM for verb-related P600 (converged after changing the optimizer and removing correlations between REs)
mod P600 verb <- lmer alt(P600 verb ~ semantics*context + (semantics*context||participant) + (semantics*context||item),
                         data = a1, control = lmerControl(calc.derivs = FALSE,
                                                           optimizer = "bobyga",
                                                           optCtrl = list(maxfun = 2e5)))
# Create a list of all models
models <- list("VALENCE" = mod_valence, "AROUSAL" = mod_aroursal, "N400_VERB" = mod_N400_verb,
               "N400_PICT" = mod_N400_pict, "P600_VERB" = mod_P600_verb)
# F-tests (type III tests)
(tests <- map(models, anova))</pre>
## $VALENCE
## Type III Analysis of Variance Table with Satterthwaite's method
           Sum Sq Mean Sq NumDF DenDF F value
                                                 Pr(>F)
## context 87.314 87.314
                             1 37.808 164.2 2.482e-15 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## $AROUSAL
## Type III Analysis of Variance Table with Satterthwaite's method
           Sum Sq Mean Sq NumDF DenDF F value
## context 51.534 51.534
                             1 37.672 83.284 4.413e-11 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## $N400_VERB
## Type III Analysis of Variance Table with Satterthwaite's method
                     Sum Sq Mean Sq NumDF DenDF F value
                                                             Pr(>F)
## semantics
                     290.572 145.286
                                        2 100.801 8.2640 0.0004748 ***
## context
                       0.355 0.355
                                      1 24.294 0.0202 0.8881094
## semantics:context 42.207 21.104
                                      2 71.656 1.2004 0.3070538
```

```
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## $N400 PICT
## Type III Analysis of Variance Table with Satterthwaite's method
                    Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
                    23.533 11.766
                                     2 37.017 0.7277 0.48981
## semantics
## context
                    0.085
                          0.085
                                  1 44.147 0.0053 0.94243
## semantics:context 125.895 62.948
                                  2 52.128 3.8929 0.02656 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## $P600 VERB
## Type III Analysis of Variance Table with Satterthwaite's method
                   Sum Sq Mean Sq NumDF     DenDF F value Pr(>F)
## semantics
                  47.810 23.9051
                                    2 102.057 1.2009 0.3051
## context
                   0.207 0.2066
                                 1 29.669 0.0104 0.9195
## semantics:context 31.597 15.7984
                                    2 192.921 0.7936 0.4537
# Allow emmeans to use Satterthwaites p-values
emm options(lmer.df = "Satterthwaite", lmerTest.limit = Inf)
# Follow-up contrasts for the main effect of semantics
(means semantics <- map(models[c("N400 VERB", "N400 PICT", "P600 VERB")],function(x){
 emmeans(x, trt.vs.ctrl ~ semantics, infer = TRUE, adjust = "bonferroni")$contrasts
}))
## NOTE: Results may be misleading due to involvement in interactions
## NOTE: Results may be misleading due to involvement in interactions
## NOTE: Results may be misleading due to involvement in interactions
## $N400_VERB
## contrast estimate
                       SE df lower.CL upper.CL t.ratio p.value
## vio - int -0.0288 0.118 90.5 -0.299
                                       0.241 -0.243 1.0000
## mci - int -0.3847 0.102 90.7 -0.617 -0.152 -3.771 0.0006
## Results are averaged over the levels of: context
## Degrees-of-freedom method: satterthwaite
```

```
## Confidence level used: 0.95
## Conf-level adjustment: bonferroni method for 2 estimates
## P value adjustment: bonferroni method for 2 tests
## $N400 PICT
## contrast estimate
                         SE
                            df lower.CL upper.CL t.ratio p.value
   vio - int
              0.073 0.129 35.5
                                   -0.228
                                             0.374 0.567 1.0000
## mci - int -0.121 0.138 37.7 -0.444
                                             0.202 -0.872 0.7774
##
## Results are averaged over the levels of: context
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## Conf-level adjustment: bonferroni method for 2 estimates
## P value adjustment: bonferroni method for 2 tests
## $P600_VERB
## contrast estimate
                         SE df lower.CL upper.CL t.ratio p.value
## vio - int 0.1149 0.109 85.1
                                   -0.133
                                             0.363 1.056 0.5877
## mci - int -0.0794 0.132 83.2
                                  -0.380
                                             0.221 -0.603 1.0000
## Results are averaged over the levels of: context
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## Conf-level adjustment: bonferroni method for 2 estimates
## P value adjustment: bonferroni method for 2 tests
# Follow-up contrasts for the main effect of context
(means_context <- map(models, function(x){</pre>
 emmeans(x, trt.vs.ctrl ~ context, infer = TRUE, adjust = "bonferroni")$contrasts
}))
## NOTE: Results may be misleading due to involvement in interactions
## NOTE: Results may be misleading due to involvement in interactions
## NOTE: Results may be misleading due to involvement in interactions
## $VALENCE
## contrast estimate
                        SE df lower.CL upper.CL t.ratio p.value
                                   -1.63 -1.19 -12.814 <.0001
## neg - neu
                -1.41 0.11 37.8
## Degrees-of-freedom method: satterthwaite
```

```
## Confidence level used: 0.95
## $AROUSAL
## contrast estimate
                         SE df lower.CL upper.CL t.ratio p.value
   neg - neu
                 1.04 0.114 37.7
                                    0.809
                                              1.27 9.126 <.0001
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## $N400_VERB
                          SE df lower.CL upper.CL t.ratio p.value
## contrast estimate
   neg - neu 0.0135 0.0946 24.3 -0.182
                                              0.209 0.142 0.8881
## Results are averaged over the levels of: semantics
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## $N400_PICT
                          SE df lower.CL upper.CL t.ratio p.value
## contrast estimate
## neg - neu -0.00706 0.0972 44.1 -0.203
                                            0.189 -0.073 0.9424
## Results are averaged over the levels of: semantics
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## $P600_VERB
                         SE
                            df lower.CL upper.CL t.ratio p.value
## contrast estimate
                                   -0.229
## neg - neu
                0.012 0.118 29.7
                                             0.253 0.102 0.9195
## Results are averaged over the levels of: semantics
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
# Follow-up contrasts for semantics within each contexts
(means_nested <- map(models[c("N400_VERB", "N400_PICT", "P600_VERB")], function(x){</pre>
  emmeans(x, trt.vs.ctrl ~ semantics|context, infer = TRUE, adjust = "bonferroni")$contrasts
}))
## $N400 VERB
```

context = neu:

```
contrast estimate
                        SE df lower.CL upper.CL t.ratio p.value
## vio - int
              -0.174 0.163 182
                                 -0.543
                                         0.196 -1.062 0.5791
## mci - int
             -0.531 0.151 117
                                 -0.874 -0.188 -3.517 0.0012
## context = neg:
   contrast estimate
                        SE df lower.CL upper.CL t.ratio p.value
                                 -0.253
   vio - int
             0.116 0.163 182
                                          0.485 0.711 0.9560
   mci - int -0.238 0.151 116
                                 -0.581
                                          0.104 - 1.579 0.2342
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## Conf-level adjustment: bonferroni method for 2 estimates
## P value adjustment: bonferroni method for 2 tests
##
## $N400_PICT
## context = neu:
## contrast estimate
                        SE df lower.CL upper.CL t.ratio p.value
## vio - int -0.0363 0.163 34.4 -0.419 0.346616 -0.222 1.0000
## mci - int -0.4060 0.175 47.1 -0.811 -0.000838 -2.320 0.0494
## context = neg:
   contrast estimate
                        SE
                            df lower.CL upper.CL t.ratio p.value
## vio - int 0.1823 0.178 43.2
                                -0.230 0.594863 1.026 0.6209
## mci - int 0.1645 0.170 35.8 -0.233 0.561694 0.969 0.6778
##
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## Conf-level adjustment: bonferroni method for 2 estimates
## P value adjustment: bonferroni method for 2 tests
##
## $P600 VERB
## context = neu:
                        SE df lower.CL upper.CL t.ratio p.value
## contrast estimate
## vio - int -0.0214 0.156 184
                                 -0.374
                                         0.331 -0.137 1.0000
   mci - int -0.1145 0.171 233
                                -0.500
                                         0.271 -0.671 1.0000
## context = neg:
## contrast estimate
                        SE df lower.CL upper.CL t.ratio p.value
## vio - int 0.2511 0.156 183 -0.101
                                         0.603 1.613 0.2168
```

```
## mci - int -0.0443 0.171 232 -0.429 0.341 -0.260 1.0000
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## Conf-level adjustment: bonferroni method for 2 estimates
## P value adjustment: bonferroni method for 2 tests
# Follow-up contrasts for contexts within each semantic condition
(means_nested_rev <- map(models[c("N400_VERB", "N400_PICT", "P600_VERB")], function(x){</pre>
 emmeans(x, trt.vs.ctrl ~ context|semantics, infer = TRUE, adjust = "bonferroni")$contrasts
}))
## $N400 VERB
## semantics = int:
## contrast estimate
                        SE
                              df lower.CL upper.CL t.ratio p.value
## neg - neu -0.181 0.157 141.4 -0.491 0.129 -1.152 0.2511
## semantics = vio:
                        SE
                              df lower.CL upper.CL t.ratio p.value
   contrast estimate
## neg - neu
               0.109 0.167 82.3 -0.223 0.440 0.654 0.5152
## semantics = mci:
                              df lower.CL upper.CL t.ratio p.value
   contrast estimate
                      SE
               0.112 0.166 53.5 -0.220 0.444 0.678 0.5006
## neg - neu
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## $N400 PICT
## semantics = int:
   contrast estimate
                        SE
                              df lower.CL upper.CL t.ratio p.value
   neg - neu -0.2701 0.164 35.4 -0.604 0.0635 -1.643 0.1092
## semantics = vio:
                        SE
                              df lower.CL upper.CL t.ratio p.value
   contrast estimate
   neg - neu -0.0515 0.143 188.4 -0.334 0.2308 -0.360 0.7195
## semantics = mci:
## contrast estimate
                        SE
                              df lower.CL upper.CL t.ratio p.value
## neg - neu 0.3004 0.170 42.4 -0.043 0.6438 1.765 0.0848
```

```
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
## $P600 VERB
## semantics = int:
                              df lower.CL upper.CL t.ratio p.value
   contrast estimate
                         SE
## neg - neu -0.102 0.172 130.2 -0.443 0.239 -0.593 0.5544
## semantics = vio:
## contrast estimate
                         SE
                               df lower.CL upper.CL t.ratio p.value
## neg - neu 0.170 0.176 83.5 -0.179 0.519 0.970 0.3350
## semantics = mci:
## contrast estimate
                         SE
                               df lower.CL upper.CL t.ratio p.value
   neg - neu -0.032 0.173 131.8 -0.374 0.310 -0.185 0.8535
## Degrees-of-freedom method: satterthwaite
## Confidence level used: 0.95
# Backup results
save(models, tests, means semantics, means context, means nested, means nested rev, file = "EEG/export/stats.RData")
# System specs and package versions
sessionInfo()
## R version 4.0.2 (2020-06-22)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS 10.16
## Matrix products: default
## LAPACK: /Library/Frameworks/R.framework/Versions/4.0/Resources/lib/libRlapack.dylib
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
## attached base packages:
                graphics grDevices datasets utils
## [1] stats
                                                       methods
                                                                 base
## other attached packages:
```

##	[1] magrittr_1.5	forcats_0.5.0 stringr_	1.4.0 dplyr_1.0.0	purrr_0.3.4	readr_1.3.1 ti	dyr_1.1.0 tibble_	3.0.3 gg
##	[14] lme4_1.1-23	Matrix_1.2-18 MASS_7.3	3-51.6				
##							
## loaded via a namespace (and not attached):							
##	$[1]$ minqa $_1.2.4$	colorspace_1.4-1	ellipsis_0.3.1	rio_0.5.16	estimability_1.3	fs_1.4.2	rstudioap
##	[12] lubridate_1.7.	9 xml2_1.3.2	codetools_0.2-16	splines_4.0.2	R.methodsS3_1.8.0	knitr_1.29	${\tt eegUtils}_$
##	[23] R.oo_1.23.0	shiny_1.5.0	compiler_4.0.2	httr_1.4.2	backports_1.1.8	$assertthat_0.2.1$	${\tt fastmap_1}$
##	[34] tools_4.0.2	gtable_0.3.0	glue_1.4.1	$reshape2_1.4.4$	Rcpp_1.0.5	carData_3.0-4	cellrange
##	[45] Rmisc_1.5	openxlsx_4.1.5	rvest_0.3.5	mime_0.9	$miniUI_0.1.1.1$	lifecycle_0.2.0	renv_0.12
##	[56] promises_1.1.1	parallel_4.0.2	yaml_2.2.1	curl_4.3	stringi_1.4.6	highr_0.8	boot_1.3-
##	[67] pracma_2.2.9	evaluate_0.14	lattice_0.20-41	htmlwidgets_1.5.1	tidyselect_1.1.0	plyr_1.8.6	R6_2.4.1
##	[78] pillar_1.4.6	haven_2.3.1	foreign_0.8-80	mgcv_1.8-31	$abind_1.4-5$	<pre>future.apply_1.6.0</pre>	modelr_0.
##	[89] grid_4.0.2	$readxl_1.3.1$	data.table_1.13.0	blob_1.2.1	reprex_0.3.0	digest_0.6.25	${\tt xtable_1}.$
##	[100] munsell_0.5.0	viridisLite_0.3.0					