

m21 LDT ERP analysis N250

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1 Load libraries

Load libraries

```
library(ez)
library(pander)
library(kableExtra)
library(afex)
library(gridExtra)
library(ggplot2)
library(emmeans)
library(tidyverse)
library(dplyr)
library(RColorBrewer)
library(wesanderson)
library(ggsci)
```

2 Set ggplot2 parameters

Before we begin, let's set some general parameters for `ggplot2`. We will set a general theme using the `theme_set()` function. We will use the 'classic' theme which gives us clean white background rather than the default grey with white grid lines. And we will position the legend at the top of the graph rather than at the right side which is the default.

```
theme_set(theme_minimal() + theme(legend.position = "bottom"))
```

#Define standard error of the mean function

```
sem <- function(x) sd(x)/sqrt(length(x))
```

3 Load and format data files

First we load the data files

Now we extract SubjID from the ERPset column

We then join the ERP data, and language into a single data frame

Divide into word, non-word and difference wave dataframes

```
n250_1_words <- n250_1 |> filter(bini %in% c(1:2))
n250_1_nonwords <- n250_1 |> filter(bini %in% c(3:6))
n250_1_diff <- n250_1 |> filter(bini %in% c(9:11))

n250_2_words <- n250_2 |> filter(bini %in% c(1:2))
n250_2_nonwords <- n250_2 |> filter(bini %in% c(3:6))
n250_2_diff <- n250_2 |> filter(bini %in% c(9:11))
```

Then we do some more formatting and cleanup of the dataframes. We create separate columns, one for each independent variable (anteriority, laterality, morphological family size). To do this we have to use `separate` function from the `stringr` package. Run `vignette("programming", package = "dplyr")` to see more about `tidy-selection` and `tidy-evaluation`.

Now we need to extract just the bins and channels that we intend to analyse. For this analysis we will use 9 channels: F3, Fz, F4, C3, Cz, C4, P3, Pz, P4 . We will use `thmutate` function from the `dplyr` package along with the `case_when` function. The `case_when` function is a sequence of two-sided formulas. The left hand side determines which values match this case. The right hand side provides the replacement value.

4 Now we can compute the ANOVA using ezANOVA and aov_ez

4.1 Group 1

```
anova_results.1a <- ezANOVA(n250_1_nonwords,
  dv = value,
  wid = SubjID,
  within = .(family_size, complexity, anteriority, laterality),
  between = .(lang_type_semantic, lang_type_ortho),
  type = 3)
anova_results.1a$ANOVA
```

```
||
|| 2                                     Effect
|| 3                                lang_type_semantic
|| 5                                lang_type_ortho
|| 9                                  family_size
|| 9                                  complexity
|| 13                               anteriority
|| 17                               laterality
|| 4 lang_type_semantic:lang_type_ortho
|| 6 lang_type_semantic:family_size
|| 7 lang_type_ortho:family_size
|| 10 lang_type_semantic:complexity
|| 11 lang_type_ortho:complexity
```

```

|| 14                                lang_type_semantic:anteriority
|| 15                                lang_type_ortho:anteriority
|| 18                                lang_type_semantic:laterality
|| 19                                lang_type_ortho:laterality
|| 21                                family_size:complexity
|| 25                                family_size:anteriority
|| 29                                complexity:anteriority
|| 33                                family_size:laterality
|| 37                                complexity:laterality
|| 41                                anteriority:laterality
|| 8                                lang_type_semantic:lang_type_ortho:family_size
|| 12                                lang_type_semantic:lang_type_ortho:complexity
|| 16                                lang_type_semantic:lang_type_ortho:anteriority
|| 20                                lang_type_semantic:lang_type_ortho:laterality
|| 22                                lang_type_semantic:family_size:complexity
|| 23                                lang_type_ortho:family_size:complexity
|| 26                                lang_type_semantic:family_size:anteriority
|| 27                                lang_type_ortho:family_size:anteriority
|| 30                                lang_type_semantic:complexity:anteriority
|| 31                                lang_type_ortho:complexity:anteriority
|| 34                                lang_type_semantic:family_size:laterality
|| 35                                lang_type_ortho:family_size:laterality
|| 38                                lang_type_semantic:complexity:laterality
|| 39                                lang_type_ortho:complexity:laterality
|| 42                                lang_type_semantic:anteriority:laterality
|| 43                                lang_type_ortho:anteriority:laterality
|| 45                                family_size:complexity:anteriority
|| 49                                family_size:complexity:laterality
|| 53                                family_size:anteriority:laterality
|| 57                                complexity:anteriority:laterality
|| 24                                lang_type_semantic:lang_type_ortho:family_size:complexity
|| 28                                lang_type_semantic:lang_type_ortho:family_size:anteriority
|| 32                                lang_type_semantic:lang_type_ortho:complexity:anteriority
|| 36                                lang_type_semantic:lang_type_ortho:family_size:laterality
|| 40                                lang_type_semantic:lang_type_ortho:complexity:laterality
|| 44                                lang_type_semantic:lang_type_ortho:anteriority:laterality
|| 46                                lang_type_semantic:family_size:complexity:anteriority
|| 47                                lang_type_ortho:family_size:complexity:anteriority
|| 50                                lang_type_semantic:family_size:complexity:laterality
|| 51                                lang_type_ortho:family_size:complexity:laterality
|| 54                                lang_type_semantic:family_size:anteriority:laterality
|| 55                                lang_type_ortho:family_size:anteriority:laterality
|| 58                                lang_type_semantic:complexity:anteriority:laterality
|| 59                                lang_type_ortho:complexity:anteriority:laterality
|| 61                                family_size:complexity:anteriority:laterality
|| 48                                lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority
|| 52                                lang_type_semantic:lang_type_ortho:family_size:complexity:laterality
|| 56                                lang_type_semantic:lang_type_ortho:family_size:anteriority:laterality
|| 60                                lang_type_semantic:lang_type_ortho:complexity:anteriority:laterality
|| 62                                lang_type_semantic:family_size:complexity:anteriority:laterality
|| 63                                lang_type_ortho:family_size:complexity:anteriority:laterality
|| 64 lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority:laterality
||   DFn DFd      F      p p<.05      ges
|| 2    1   56 5.923851e-01 4.447328e-01    5.231956e-03

```

3	1	56	2.527261e+00	1.175246e-01	2.194579e-02
5	1	56	3.114857e-01	5.789945e-01	4.723971e-04
9	1	56	1.851730e-01	6.686164e-01	2.910648e-04
13	2	112	1.005751e+01	9.613578e-05	* 2.260062e-02
17	2	112	1.476640e+00	2.328160e-01	5.760913e-04
4	1	56	4.806680e-01	4.909855e-01	4.249462e-03
6	1	56	2.766313e-01	6.009945e-01	4.195595e-04
7	1	56	7.028264e-01	4.053998e-01	1.065270e-03
10	1	56	6.279683e-01	4.314459e-01	9.863876e-04
11	1	56	1.317063e-01	7.180356e-01	2.070404e-04
14	2	112	6.825905e-02	9.340573e-01	1.569097e-04
15	2	112	1.395414e-01	8.699080e-01	3.207165e-04
18	2	112	9.288165e-02	9.113715e-01	3.625610e-05
19	2	112	5.560171e-01	5.750634e-01	2.170006e-04
21	1	56	9.548137e-01	3.326995e-01	1.429864e-03
25	2	112	1.153696e-01	8.911425e-01	2.470441e-05
29	2	112	1.471373e+00	2.340138e-01	3.174114e-04
33	2	112	1.952539e-02	9.806673e-01	2.254712e-06
37	2	112	5.222836e-01	5.946020e-01	7.478559e-05
41	4	224	9.837849e-01	4.171722e-01	3.482273e-04
8	1	56	4.495542e-01	5.053026e-01	6.816482e-04
12	1	56	1.562321e-01	6.941512e-01	2.455851e-04
16	2	112	2.805294e-01	7.559130e-01	6.445490e-04
20	2	112	4.075575e-01	6.662558e-01	1.590694e-04
22	1	56	3.931095e-04	9.842519e-01	5.895366e-07
23	1	56	7.572899e-02	7.841839e-01	1.135561e-04
26	2	112	2.350406e+00	1.000166e-01	5.030580e-04
27	2	112	2.830246e-01	7.540385e-01	6.060264e-05
30	2	112	2.909670e+00	5.862470e-02	6.274928e-04
31	2	112	4.075641e-02	9.600772e-01	8.794877e-06
34	2	112	2.419010e+00	9.364715e-02	2.792599e-04
35	2	112	2.332859e-01	7.923109e-01	2.693823e-05
38	2	112	5.445966e-01	5.816040e-01	7.798033e-05
39	2	112	1.827437e-01	8.332295e-01	2.616828e-05
42	4	224	2.311072e+00	5.867584e-02	8.176590e-04
43	4	224	6.565069e-01	6.228652e-01	2.324087e-04
45	2	112	3.667501e+00	2.865563e-02	* 7.442768e-04
49	2	112	7.668078e-01	4.669163e-01	7.357447e-05
53	4	224	7.567599e-01	5.544595e-01	9.316782e-05
57	4	224	1.417270e+00	2.290923e-01	1.929653e-04
24	1	56	3.367750e+00	7.179735e-02	5.025155e-03
28	2	112	1.173787e+00	3.129678e-01	2.512893e-04
32	2	112	6.587968e-01	5.194671e-01	1.421436e-04
36	2	112	1.582326e-01	8.538417e-01	1.827176e-05
40	2	112	4.942383e-01	6.113596e-01	7.077007e-05
44	4	224	1.134669e+00	3.409749e-01	4.016138e-04
46	2	112	3.195000e+00	4.472683e-02	* 6.484504e-04
47	2	112	1.404967e-01	8.690794e-01	2.853260e-05
50	2	112	9.516795e-01	3.891916e-01	9.131111e-05
51	2	112	6.072227e-01	5.466458e-01	5.826333e-05
54	4	224	4.402994e-01	7.794051e-01	5.420918e-05
55	4	224	1.657943e+00	1.607691e-01	2.040935e-04
58	4	224	9.222684e-01	4.517284e-01	1.255779e-04
59	4	224	1.935981e+00	1.054039e-01	2.635707e-04

```

|| 61 4 224 8.950801e-01 4.676429e-01 8.848617e-05
|| 48 2 112 4.218830e-01 6.568486e-01 8.567268e-05
|| 52 2 112 1.885387e+00 1.565557e-01 1.808817e-04
|| 56 4 224 1.118854e+00 3.483973e-01 1.377405e-04
|| 60 4 224 7.536243e-01 5.565349e-01 1.026174e-04
|| 62 4 224 2.953256e+00 2.091360e-02 * 2.918947e-04
|| 63 4 224 1.270423e+00 2.824838e-01 1.255873e-04
|| 64 4 224 4.912276e-01 7.421871e-01 4.856391e-05

```

```
anova_results.1a$`Sphericity Corrections`
```

```

||
||                                     Effect
|| 13                                anteriority
|| 14                   lang_type_semantic:anteriority
|| 15                   lang_type_ortho:anteriority
|| 16                   lang_type_semantic:lang_type_ortho:anteriority
|| 17                                laterality
|| 18                   lang_type_semantic:laterality
|| 19                   lang_type_ortho:laterality
|| 20                   lang_type_semantic:lang_type_ortho:laterality
|| 25                                family_size:anteriority
|| 26                   lang_type_semantic:family_size:anteriority
|| 27                   lang_type_ortho:family_size:anteriority
|| 28                   lang_type_semantic:lang_type_ortho:family_size:anteriority
|| 29                                complexity:anteriority
|| 30                   lang_type_semantic:complexity:anteriority
|| 31                   lang_type_ortho:complexity:anteriority
|| 32                   lang_type_semantic:lang_type_ortho:complexity:anteriority
|| 33                                family_size:laterality
|| 34                   lang_type_semantic:family_size:laterality
|| 35                   lang_type_ortho:family_size:laterality
|| 36                   lang_type_semantic:lang_type_ortho:family_size:laterality
|| 37                                complexity:laterality
|| 38                   lang_type_semantic:complexity:laterality
|| 39                   lang_type_ortho:complexity:laterality
|| 40                   lang_type_semantic:lang_type_ortho:complexity:laterality
|| 41                                anteriority:laterality
|| 42                   lang_type_semantic:anteriority:laterality
|| 43                   lang_type_ortho:anteriority:laterality
|| 44                   lang_type_semantic:lang_type_ortho:anteriority:laterality
|| 45                                family_size:complexity:anteriority
|| 46                   lang_type_semantic:family_size:complexity:anteriority
|| 47                   lang_type_ortho:family_size:complexity:anteriority
|| 48                   lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority
|| 49                                family_size:complexity:laterality
|| 50                   lang_type_semantic:family_size:complexity:laterality
|| 51                   lang_type_ortho:family_size:complexity:laterality
|| 52                   lang_type_semantic:lang_type_ortho:family_size:complexity:laterality
|| 53                                family_size:anteriority:laterality
|| 54                   lang_type_semantic:family_size:anteriority:laterality
|| 55                   lang_type_ortho:family_size:anteriority:laterality
|| 56                   lang_type_semantic:lang_type_ortho:family_size:anteriority:laterality
|| 57                                complexity:anteriority:laterality
|| 58                   lang_type_semantic:complexity:anteriority:laterality

```

59				lang_type_ortho:complexity:anteriority:laterality			
60				lang_type_semantic:lang_type_ortho:complexity:anteriority:laterality			
61				family_size:complexity:anteriority:laterality			
62				lang_type_semantic:family_size:complexity:anteriority:laterality			
63				lang_type_ortho:family_size:complexity:anteriority:laterality			
64	lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority:laterality						
	GGe	p[GG]	p[GG]<.05	HFe	p[HF]	p[HF]<.05	
13	0.5972431	0.001304555	*	0.6029112	0.001257218	*	
14	0.5972431	0.837518865		0.6029112	0.839672733		
15	0.5972431	0.754896282		0.6029112	0.757222816		
16	0.5972431	0.639408974		0.6029112	0.641585728		
17	0.9676624	0.233235915		1.0017031	0.232815957		
18	0.9676624	0.905858758		1.0017031	0.911371470		
19	0.9676624	0.569462988		1.0017031	0.575063438		
20	0.9676624	0.659564183		1.0017031	0.666255802		
25	0.7502068	0.833213400		0.7662772	0.837863971		
26	0.7502068	0.115418277		0.7662772	0.114395461		
27	0.7502068	0.689896024		0.7662772	0.694708068		
28	0.7502068	0.302915562		0.7662772	0.303770040		
29	0.8505385	0.235414362		0.8744316	0.235276279		
30	0.8505385	0.067715573		0.8744316	0.066178672		
31	0.8505385	0.940469926		0.8744316	0.944187226		
32	0.8505385	0.496166867		0.8744316	0.500145938		
33	0.8991522	0.972805451		0.9271236	0.975270917		
34	0.8991522	0.099780158		0.9271236	0.098048983		
35	0.8991522	0.768899059		0.9271236	0.775720472		
36	0.8991522	0.831791882		0.9271236	0.838278893		
37	0.7663538	0.547312040		0.7836293	0.551250444		
38	0.7663538	0.535769584		0.7836293	0.539585895		
39	0.7663538	0.774446773		0.7836293	0.779595581		
40	0.7663538	0.562273428		0.7836293	0.566364070		
41	0.8210441	0.406858787		0.8782850	0.410474245		
42	0.8210441	0.071959020		0.8782850	0.067398255		
43	0.8210441	0.593354269		0.8782850	0.603395685		
44	0.8210441	0.338364356		0.8782850	0.339424158		
45	0.7250612	0.043412643	*	0.7392957	0.042492026	*	
46	0.7250612	0.061596114		0.7392957	0.060593244		
47	0.7250612	0.800367251		0.7392957	0.804843519		
48	0.7250612	0.592315817		0.7392957	0.596207686		
49	0.8568951	0.448848857		0.8813107	0.452142415		
50	0.8568951	0.377678944		0.8813107	0.379811875		
51	0.8568951	0.522395952		0.8813107	0.526794692		
52	0.8568951	0.163015221		0.8813107	0.161935414		
53	0.6942320	0.510664271		0.7341031	0.517326907		
54	0.6942320	0.709437807		0.7341031	0.720314914		
55	0.6942320	0.181986571		0.7341031	0.179166456		
56	0.6942320	0.341071099		0.7341031	0.342510395		
57	0.6606241	0.242723922		0.6963514	0.241503347		
58	0.6606241	0.422055790		0.6963514	0.425962660		
59	0.6606241	0.133771459		0.6963514	0.130493008		
60	0.6606241	0.506442583		0.6963514	0.512766465		
61	0.9298523	0.461995624		1.0042270	0.467642933		
62	0.9298523	0.023983403	*	1.0042270	0.020913597	*	
63	0.9298523	0.283813807		1.0042270	0.282483787		

```
|| 64 0.9298523 0.728778628          1.0042270 0.742187144
```

```
anova_results.1b <- aov_ez(id = "SubjID",
                           dv = "value",
                           data = n250_1_nonwords,
                           within = c("family_size",
                                       "complexity",
                                       "anteriority",
                                       "laterality"),
                           between = c("lang_type_semantic", "lang_type_ortho"),
                           type = 3)

anova_results.1b
```

```
|| Anova Table (Type 3 tests)
```

```
||
```

```
|| Response: value
```

```
||
```

	Effect
1	lang_type_semantic
2	lang_type_ortho
3	lang_type_semantic:lang_type_ortho
4	family_size
5	lang_type_semantic:family_size
6	lang_type_ortho:family_size
7	lang_type_semantic:lang_type_ortho:family_size
8	complexity
9	lang_type_semantic:complexity
10	lang_type_ortho:complexity
11	lang_type_semantic:lang_type_ortho:complexity
12	anteriority
13	lang_type_semantic:anteriority
14	lang_type_ortho:anteriority
15	lang_type_semantic:lang_type_ortho:anteriority
16	laterality
17	lang_type_semantic:laterality
18	lang_type_ortho:laterality
19	lang_type_semantic:lang_type_ortho:laterality
20	family_size:complexity
21	lang_type_semantic:family_size:complexity
22	lang_type_ortho:family_size:complexity
23	lang_type_semantic:lang_type_ortho:family_size:complexity
24	family_size:anteriority
25	lang_type_semantic:family_size:anteriority
26	lang_type_ortho:family_size:anteriority
27	lang_type_semantic:lang_type_ortho:family_size:anteriority
28	complexity:anteriority
29	lang_type_semantic:complexity:anteriority
30	lang_type_ortho:complexity:anteriority
31	lang_type_semantic:lang_type_ortho:complexity:anteriority
32	family_size:laterality
33	lang_type_semantic:family_size:laterality
34	lang_type_ortho:family_size:laterality
35	lang_type_semantic:lang_type_ortho:family_size:laterality
36	complexity:laterality
37	lang_type_semantic:complexity:laterality


```

|| 38                                lang_type_ortho:complexity:laterality
|| 39                        lang_type_semantic:lang_type_ortho:complexity:laterality
|| 40                                anteriority:laterality
|| 41                        lang_type_semantic:anteriority:laterality
|| 42                                lang_type_ortho:anteriority:laterality
|| 43                        lang_type_semantic:lang_type_ortho:anteriority:laterality
|| 44                                family_size:complexity:anteriority
|| 45                        lang_type_semantic:family_size:complexity:anteriority
|| 46                                lang_type_ortho:family_size:complexity:anteriority
|| 47                        lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority
|| 48                                family_size:complexity:laterality
|| 49                        lang_type_semantic:family_size:complexity:laterality
|| 50                                lang_type_ortho:family_size:complexity:laterality
|| 51                        lang_type_semantic:lang_type_ortho:family_size:complexity:laterality
|| 52                                family_size:anteriority:laterality
|| 53                        lang_type_semantic:family_size:anteriority:laterality
|| 54                                lang_type_ortho:family_size:anteriority:laterality
|| 55                        lang_type_semantic:lang_type_ortho:family_size:anteriority:laterality
|| 56                                complexity:anteriority:laterality
|| 57                        lang_type_semantic:complexity:anteriority:laterality
|| 58                                lang_type_ortho:complexity:anteriority:laterality
|| 59                        lang_type_semantic:lang_type_ortho:complexity:anteriority:laterality
|| 60                                family_size:complexity:anteriority:laterality
|| 61                        lang_type_semantic:family_size:complexity:anteriority:laterality
|| 62                                lang_type_ortho:family_size:complexity:anteriority:laterality
|| 63 lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority:laterality
||
||      df      MSE      F      ges p.value
|| 1      1, 56 192.65    0.59 .005    .445
|| 2      1, 56 192.65    2.53 .022    .118
|| 3      1, 56 192.65    0.48 .004    .491
|| 4      1, 56  32.92    0.31 <.001    .579
|| 5      1, 56  32.92    0.28 <.001    .601
|| 6      1, 56  32.92    0.70 .001    .405
|| 7      1, 56  32.92    0.45 <.001    .505
|| 8      1, 56  34.12    0.19 <.001    .669
|| 9      1, 56  34.12    0.63 <.001    .431
|| 10     1, 56  34.12    0.13 <.001    .718
|| 11     1, 56  34.12    0.16 <.001    .694
|| 12  1.19, 66.89 41.76 10.06 ** .023    .001
|| 13  1.19, 66.89 41.76    0.07 <.001    .838
|| 14  1.19, 66.89 41.76    0.14 <.001    .755
|| 15  1.19, 66.89 41.76    0.28 <.001    .639
|| 16  1.94, 108.38  4.38    1.48 <.001    .233
|| 17  1.94, 108.38  4.38    0.09 <.001    .906
|| 18  1.94, 108.38  4.38    0.56 <.001    .569
|| 19  1.94, 108.38  4.38    0.41 <.001    .660
|| 20      1, 56  32.54    0.95 .001    .333
|| 21      1, 56  32.54    0.00 <.001    .984
|| 22      1, 56  32.54    0.08 <.001    .784
|| 23      1, 56  32.54    3.37 + .005    .072
|| 24  1.50, 84.02   3.10    0.12 <.001    .833
|| 25  1.50, 84.02   3.10    2.35 <.001    .115
|| 26  1.50, 84.02   3.10    0.28 <.001    .690
|| 27  1.50, 84.02   3.10    1.17 <.001    .303

```

```

|| 28 1.70, 95.26 2.75 1.47 <.001 .235
|| 29 1.70, 95.26 2.75 2.91 + <.001 .068
|| 30 1.70, 95.26 2.75 0.04 <.001 .940
|| 31 1.70, 95.26 2.75 0.66 <.001 .496
|| 32 1.80, 100.71 1.39 0.02 <.001 .973
|| 33 1.80, 100.71 1.39 2.42 + <.001 .100
|| 34 1.80, 100.71 1.39 0.23 <.001 .769
|| 35 1.80, 100.71 1.39 0.16 <.001 .832
|| 36 1.53, 85.83 2.03 0.52 <.001 .547
|| 37 1.53, 85.83 2.03 0.54 <.001 .536
|| 38 1.53, 85.83 2.03 0.18 <.001 .774
|| 39 1.53, 85.83 2.03 0.49 <.001 .562
|| 40 3.28, 183.91 2.34 0.98 <.001 .407
|| 41 3.28, 183.91 2.34 2.31 + <.001 .072
|| 42 3.28, 183.91 2.34 0.66 <.001 .593
|| 43 3.28, 183.91 2.34 1.13 <.001 .338
|| 44 1.45, 81.21 3.04 3.67 * <.001 .043
|| 45 1.45, 81.21 3.04 3.20 + <.001 .062
|| 46 1.45, 81.21 3.04 0.14 <.001 .800
|| 47 1.45, 81.21 3.04 0.42 <.001 .592
|| 48 1.71, 95.97 1.21 0.77 <.001 .449
|| 49 1.71, 95.97 1.21 0.95 <.001 .378
|| 50 1.71, 95.97 1.21 0.61 <.001 .522
|| 51 1.71, 95.97 1.21 1.89 <.001 .163
|| 52 2.78, 155.51 0.96 0.76 <.001 .511
|| 53 2.78, 155.51 0.96 0.44 <.001 .709
|| 54 2.78, 155.51 0.96 1.66 <.001 .182
|| 55 2.78, 155.51 0.96 1.12 <.001 .341
|| 56 2.64, 147.98 1.12 1.42 <.001 .243
|| 57 2.64, 147.98 1.12 0.92 <.001 .422
|| 58 2.64, 147.98 1.12 1.94 <.001 .134
|| 59 2.64, 147.98 1.12 0.75 <.001 .506
|| 60 3.72, 208.29 0.58 0.90 <.001 .462
|| 61 3.72, 208.29 0.58 2.95 * <.001 .024
|| 62 3.72, 208.29 0.58 1.27 <.001 .284
|| 63 3.72, 208.29 0.58 0.49 <.001 .729
|| ---
|| Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '+' 0.1 ' ' 1
||
|| Sphericity correction method: GG

```

4.2 Group 2

```

anova_results.2a <- ezANOVA(n250_2_nonwords,
  dv = value,
  wid = SubjID,
  within = .(family_size, complexity, anteriority, laterality),
  between = .(lang_type_semantic, lang_type_ortho),
  type = 3)
anova_results.2a$ANOVA

```

```

|| Effect

```

```

|| 2                                lang_type_semantic
|| 3                                lang_type_ortho
|| 5                                family_size
|| 9                                complexity
|| 13                               anteriority
|| 17                               laterality
|| 4                                lang_type_semantic:lang_type_ortho
|| 6                                lang_type_semantic:family_size
|| 7                                lang_type_ortho:family_size
|| 10                               lang_type_semantic:complexity
|| 11                               lang_type_ortho:complexity
|| 14                               lang_type_semantic:anteriority
|| 15                               lang_type_ortho:anteriority
|| 18                               lang_type_semantic:laterality
|| 19                               lang_type_ortho:laterality
|| 21                               family_size:complexity
|| 25                               family_size:anteriority
|| 29                               complexity:anteriority
|| 33                               family_size:laterality
|| 37                               complexity:laterality
|| 41                               anteriority:laterality
|| 8                                lang_type_semantic:lang_type_ortho:family_size
|| 12                               lang_type_semantic:lang_type_ortho:complexity
|| 16                               lang_type_semantic:lang_type_ortho:anteriority
|| 20                               lang_type_semantic:lang_type_ortho:laterality
|| 22                               lang_type_semantic:family_size:complexity
|| 23                               lang_type_ortho:family_size:complexity
|| 26                               lang_type_semantic:family_size:anteriority
|| 27                               lang_type_ortho:family_size:anteriority
|| 30                               lang_type_semantic:complexity:anteriority
|| 31                               lang_type_ortho:complexity:anteriority
|| 34                               lang_type_semantic:family_size:laterality
|| 35                               lang_type_ortho:family_size:laterality
|| 38                               lang_type_semantic:complexity:laterality
|| 39                               lang_type_ortho:complexity:laterality
|| 42                               lang_type_semantic:anteriority:laterality
|| 43                               lang_type_ortho:anteriority:laterality
|| 45                               family_size:complexity:anteriority
|| 49                               family_size:complexity:laterality
|| 53                               family_size:anteriority:laterality
|| 57                               complexity:anteriority:laterality
|| 24                               lang_type_semantic:lang_type_ortho:family_size:complexity
|| 28                               lang_type_semantic:lang_type_ortho:family_size:anteriority
|| 32                               lang_type_semantic:lang_type_ortho:complexity:anteriority
|| 36                               lang_type_semantic:lang_type_ortho:family_size:laterality
|| 40                               lang_type_semantic:lang_type_ortho:complexity:laterality
|| 44                               lang_type_semantic:lang_type_ortho:anteriority:laterality
|| 46                               lang_type_semantic:family_size:complexity:anteriority
|| 47                               lang_type_ortho:family_size:complexity:anteriority
|| 50                               lang_type_semantic:family_size:complexity:laterality
|| 51                               lang_type_ortho:family_size:complexity:laterality
|| 54                               lang_type_semantic:family_size:anteriority:laterality
|| 55                               lang_type_ortho:family_size:anteriority:laterality
|| 58                               lang_type_semantic:complexity:anteriority:laterality

```

```

|| 59                                lang_type_ortho:complexity:anteriority:laterality
|| 61                                family_size:complexity:anteriority:laterality
|| 48    lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority
|| 52    lang_type_semantic:lang_type_ortho:family_size:complexity:laterality
|| 56    lang_type_semantic:lang_type_ortho:family_size:anteriority:laterality
|| 60    lang_type_semantic:lang_type_ortho:complexity:anteriority:laterality
|| 62    lang_type_semantic:family_size:complexity:anteriority:laterality
|| 63    lang_type_ortho:family_size:complexity:anteriority:laterality
|| 64 lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority:laterality
||      DFn DFd      F      p p<.05      ges
|| 2      1  35 7.329838e-03 0.932260782      1.063476e-04
|| 3      1  35 4.992281e-01 0.484516974      7.191901e-03
|| 5      1  35 6.069609e-04 0.980484776      1.022126e-06
|| 9      1  35 1.527592e+00 0.224705301      2.688437e-03
|| 13     2  70 9.398105e-01 0.395575270      3.313964e-03
|| 17     2  70 5.812789e-01 0.561858796      7.189608e-04
|| 4      1  35 1.247201e+00 0.271695865      1.777570e-02
|| 6      1  35 1.452414e-04 0.990452871      2.445878e-07
|| 7      1  35 2.500552e-01 0.620165811      4.209180e-04
|| 10     1  35 3.349912e-02 0.855832466      5.911115e-05
|| 11     1  35 2.126507e+00 0.153686238      3.738539e-03
|| 14     2  70 3.005921e-01 0.741330592      1.062344e-03
|| 15     2  70 8.527887e-01 0.430604514      3.008030e-03
|| 18     2  70 4.433419e+00 0.015386019      * 5.457518e-03
|| 19     2  70 2.617716e+00 0.080102628      3.229614e-03
|| 21     1  35 1.073322e+01 0.002378345      * 9.243002e-03
|| 25     2  70 1.287183e-02 0.987212990      5.575702e-06
|| 29     2  70 7.773803e-01 0.463535207      3.785980e-04
|| 33     2  70 2.134919e+00 0.125890957      6.548924e-04
|| 37     2  70 3.334612e-01 0.717571518      1.710190e-04
|| 41     4 140 3.113302e+00 0.017244137      * 2.656395e-03
|| 8      1  35 6.219095e-01 0.435647897      1.046205e-03
|| 12     1  35 1.768415e+00 0.192180348      3.110947e-03
|| 16     2  70 1.243996e+00 0.294522900      4.381883e-03
|| 20     2  70 1.327439e-01 0.875909220      1.642768e-04
|| 22     1  35 7.885509e-01 0.380600793      6.849326e-04
|| 23     1  35 4.822117e-01 0.492006259      4.189589e-04
|| 26     2  70 2.299300e-01 0.795186947      9.958960e-05
|| 27     2  70 4.394230e-01 0.646173385      1.903100e-04
|| 30     2  70 3.062284e-01 0.737199657      1.491729e-04
|| 31     2  70 3.520506e-01 0.704482491      1.714904e-04
|| 34     2  70 1.173406e+00 0.315320767      3.600516e-04
|| 35     2  70 1.593629e+00 0.210469725      4.889313e-04
|| 38     2  70 1.544660e-01 0.857163881      7.922674e-05
|| 39     2  70 8.594705e-01 0.427805159      4.406694e-04
|| 42     4 140 1.556929e+00 0.189191814      1.330201e-03
|| 43     4 140 5.809358e-01 0.676939272      4.967511e-04
|| 45     2  70 1.100309e+00 0.338453836      4.834596e-04
|| 49     2  70 4.771489e-01 0.622553467      1.282944e-04
|| 53     4 140 7.001974e-01 0.593068044      4.144890e-04
|| 57     4 140 6.347405e-01 0.638540874      3.462414e-04
|| 24     1  35 2.094124e-01 0.650056577      1.819864e-04
|| 28     2  70 1.974503e+00 0.146487182      8.545707e-04
|| 32     2  70 2.957745e+00 0.058461184      1.438946e-03

```

36	2	70	3.931016e-01	0.676440733	1.206495e-04
40	2	70	3.401355e-01	0.712843521	1.744414e-04
44	4	140	1.448989e+00	0.221100232	1.238094e-03
46	2	70	1.547373e-01	0.856932424	6.801755e-05
47	2	70	1.564559e+00	0.216406188	6.873041e-04
50	2	70	5.580175e-02	0.945768632	1.500551e-05
51	2	70	4.958779e-01	0.611159086	1.333295e-04
54	4	140	1.409866e+00	0.233811338	8.342338e-04
55	4	140	7.012107e-01	0.592376763	4.150885e-04
58	4	140	1.281112e+00	0.280330380	6.985811e-04
59	4	140	8.890307e-01	0.472327353	4.848856e-04
61	4	140	1.332303e+00	0.260944893	6.993823e-04
48	2	70	2.489995e+00	0.090228414	1.093400e-03
52	2	70	5.628803e-01	0.572122532	1.513421e-04
56	4	140	1.054153e+00	0.381695453	6.238856e-04
60	4	140	6.911396e-01	0.599265642	3.769946e-04
62	4	140	1.220062e+00	0.305064389	6.405001e-04
63	4	140	1.545028e+00	0.192492435	8.109601e-04
64	4	140	1.017442e+00	0.400589826	5.341869e-04

anova_results.2a\$`Sphericity Corrections`

	Effect
13	anteriority
14	lang_type_semantic:anteriority
15	lang_type_ortho:anteriority
16	lang_type_semantic:lang_type_ortho:anteriority
17	laterality
18	lang_type_semantic:laterality
19	lang_type_ortho:laterality
20	lang_type_semantic:lang_type_ortho:laterality
25	family_size:anteriority
26	lang_type_semantic:family_size:anteriority
27	lang_type_ortho:family_size:anteriority
28	lang_type_semantic:lang_type_ortho:family_size:anteriority
29	complexity:anteriority
30	lang_type_semantic:complexity:anteriority
31	lang_type_ortho:complexity:anteriority
32	lang_type_semantic:lang_type_ortho:complexity:anteriority
33	family_size:laterality
34	lang_type_semantic:family_size:laterality
35	lang_type_ortho:family_size:laterality
36	lang_type_semantic:lang_type_ortho:family_size:laterality
37	complexity:laterality
38	lang_type_semantic:complexity:laterality
39	lang_type_ortho:complexity:laterality
40	lang_type_semantic:lang_type_ortho:complexity:laterality
41	anteriority:laterality
42	lang_type_semantic:anteriority:laterality
43	lang_type_ortho:anteriority:laterality
44	lang_type_semantic:lang_type_ortho:anteriority:laterality
45	family_size:complexity:anteriority
46	lang_type_semantic:family_size:complexity:anteriority
47	lang_type_ortho:family_size:complexity:anteriority

```

|| 48 lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority
|| 49 family_size:complexity:laterality
|| 50 lang_type_semantic:family_size:complexity:laterality
|| 51 lang_type_ortho:family_size:complexity:laterality
|| 52 lang_type_semantic:lang_type_ortho:family_size:complexity:laterality
|| 53 family_size:anteriority:laterality
|| 54 lang_type_semantic:family_size:anteriority:laterality
|| 55 lang_type_ortho:family_size:anteriority:laterality
|| 56 lang_type_semantic:lang_type_ortho:family_size:anteriority:laterality
|| 57 complexity:anteriority:laterality
|| 58 lang_type_semantic:complexity:anteriority:laterality
|| 59 lang_type_ortho:complexity:anteriority:laterality
|| 60 lang_type_semantic:lang_type_ortho:complexity:anteriority:laterality
|| 61 family_size:complexity:anteriority:laterality
|| 62 lang_type_semantic:family_size:complexity:anteriority:laterality
|| 63 lang_type_ortho:family_size:complexity:anteriority:laterality
|| 64 lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority:laterality
|| GGe p[GG] p[GG]<.05 HFe p[HF] p[HF]<.05
|| 13 0.6308240 0.35903290 0.6434714 0.36069691
|| 14 0.6308240 0.63919107 0.6434714 0.64365263
|| 15 0.6308240 0.38572862 0.6434714 0.38771716
|| 16 0.6308240 0.28201135 0.6434714 0.28273939
|| 17 0.8640514 0.53814043 0.9048432 0.54560946
|| 18 0.8640514 0.02028186 * 0.9048432 0.01866630 *
|| 19 0.8640514 0.08872018 0.9048432 0.08605364
|| 20 0.8640514 0.84674871 0.9048432 0.85625616
|| 25 0.6478637 0.95041101 0.6623223 0.95315841
|| 26 0.6478637 0.69640799 0.6623223 0.70161960
|| 27 0.6478637 0.56145311 0.6623223 0.56572751
|| 28 0.6478637 0.16353859 0.6623223 0.16291830
|| 29 0.6665744 0.41730550 0.6830659 0.42011736
|| 30 0.6665744 0.64775186 0.6830659 0.65321108
|| 31 0.6665744 0.61800216 0.6830659 0.62323242
|| 32 0.6665744 0.08089363 0.6830659 0.07963205
|| 33 0.6489182 0.14567255 0.6634902 0.14488702
|| 34 0.6489182 0.29923972 0.6634902 0.30024330
|| 35 0.6489182 0.21648304 0.6634902 0.21642768
|| 36 0.6489182 0.58768304 0.6634902 0.59222718
|| 37 0.7166390 0.64557509 0.7387973 0.65220570
|| 38 0.7166390 0.78475858 0.7387973 0.79182101
|| 39 0.7166390 0.39611074 0.7387973 0.39906767
|| 40 0.7166390 0.64120051 0.7387973 0.64779146
|| 41 0.4579444 0.05553246 0.4819673 0.05269779
|| 42 0.4579444 0.21975983 0.4819673 0.21878408
|| 43 0.4579444 0.54775894 0.4819673 0.55606682
|| 44 0.4579444 0.24252372 0.4819673 0.24213169
|| 45 0.6581659 0.31798606 0.6737382 0.31928694
|| 46 0.6581659 0.76438062 0.6737382 0.76996260
|| 47 0.6581659 0.22127363 0.6737382 0.22125664
|| 48 0.6581659 0.11275290 0.6737382 0.11167084
|| 49 0.6987432 0.55559897 0.7188378 0.56085571
|| 50 0.6987432 0.88826850 0.7188378 0.89367641
|| 51 0.6987432 0.54579544 0.7188378 0.55092370
|| 52 0.6987432 0.51261173 0.7188378 0.51727807

```

```

|| 53 0.3059840 0.43504432          0.3113277 0.43732895
|| 54 0.3059840 0.24785931          0.3113277 0.24817860
|| 55 0.3059840 0.43466584          0.3113277 0.43694665
|| 56 0.3059840 0.32540368          0.3113277 0.32653330
|| 57 0.3410440 0.47641257          0.3501500 0.48029066
|| 58 0.3410440 0.27644080          0.3501500 0.27718719
|| 59 0.3410440 0.38156099          0.3501500 0.38398291
|| 60 0.3410440 0.45298325          0.3501500 0.45651609
|| 61 0.3070405 0.26277739          0.3124929 0.26325183
|| 62 0.3070405 0.28619612          0.3124929 0.28691739
|| 63 0.3070405 0.22433727          0.3124929 0.22442385
|| 64 0.3070405 0.33524099          0.3124929 0.33649217

```

```

anova_results.2b <- aov_ez(id = "SubjID",
  dv = "value",
  data = n250_2_nonwords,
  within = c("family_size",
    "complexity",
    "anteriority",
    "laterality"),
  between = c("lang_type_semantic", "lang_type_ortho"),
  type = 3)
anova_results.2b

```

```

|| Anova Table (Type 3 tests)
||
|| Response: value
||
||                                     Effect
|| 1                                     lang_type_semantic
|| 2                                     lang_type_ortho
|| 3                                lang_type_semantic:lang_type_ortho
|| 4                                     family_size
|| 5                                lang_type_semantic:family_size
|| 6                                lang_type_ortho:family_size
|| 7                                lang_type_semantic:lang_type_ortho:family_size
|| 8                                     complexity
|| 9                                lang_type_semantic:complexity
|| 10                                lang_type_ortho:complexity
|| 11                                lang_type_semantic:lang_type_ortho:complexity
|| 12                                     anteriority
|| 13                                lang_type_semantic:anteriority
|| 14                                lang_type_ortho:anteriority
|| 15                                lang_type_semantic:lang_type_ortho:anteriority
|| 16                                     laterality
|| 17                                lang_type_semantic:laterality
|| 18                                lang_type_ortho:laterality
|| 19                                lang_type_semantic:lang_type_ortho:laterality
|| 20                                family_size:complexity
|| 21                                lang_type_semantic:family_size:complexity
|| 22                                lang_type_ortho:family_size:complexity
|| 23                                lang_type_semantic:lang_type_ortho:family_size:complexity
|| 24                                family_size:anteriority
|| 25                                lang_type_semantic:family_size:anteriority
|| 26                                lang_type_ortho:family_size:anteriority

```

```

|| 27          lang_type_semantic:lang_type_ortho:family_size:anteriority
|| 28                      complexity:anteriority
|| 29          lang_type_semantic:complexity:anteriority
|| 30          lang_type_ortho:complexity:anteriority
|| 31          lang_type_semantic:lang_type_ortho:complexity:anteriority
|| 32                      family_size:laterality
|| 33          lang_type_semantic:family_size:laterality
|| 34          lang_type_ortho:family_size:laterality
|| 35          lang_type_semantic:lang_type_ortho:family_size:laterality
|| 36                      complexity:laterality
|| 37          lang_type_semantic:complexity:laterality
|| 38          lang_type_ortho:complexity:laterality
|| 39          lang_type_semantic:lang_type_ortho:complexity:laterality
|| 40                      anteriority:laterality
|| 41          lang_type_semantic:anteriority:laterality
|| 42          lang_type_ortho:anteriority:laterality
|| 43          lang_type_semantic:lang_type_ortho:anteriority:laterality
|| 44                      family_size:complexity:anteriority
|| 45          lang_type_semantic:family_size:complexity:anteriority
|| 46          lang_type_ortho:family_size:complexity:anteriority
|| 47          lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority
|| 48                      family_size:complexity:laterality
|| 49          lang_type_semantic:family_size:complexity:laterality
|| 50          lang_type_ortho:family_size:complexity:laterality
|| 51          lang_type_semantic:lang_type_ortho:family_size:complexity:laterality
|| 52                      family_size:anteriority:laterality
|| 53          lang_type_semantic:family_size:anteriority:laterality
|| 54          lang_type_ortho:family_size:anteriority:laterality
|| 55          lang_type_semantic:lang_type_ortho:family_size:anteriority:laterality
|| 56                      complexity:anteriority:laterality
|| 57          lang_type_semantic:complexity:anteriority:laterality
|| 58          lang_type_ortho:complexity:anteriority:laterality
|| 59          lang_type_semantic:lang_type_ortho:complexity:anteriority:laterality
|| 60                      family_size:complexity:anteriority:laterality
|| 61          lang_type_semantic:family_size:complexity:anteriority:laterality
|| 62          lang_type_ortho:family_size:complexity:anteriority:laterality
|| 63 lang_type_semantic:lang_type_ortho:family_size:complexity:anteriority:laterality
||
||          df      MSE      F      ges p.value
|| 1          1, 35 276.17      0.01 <.001      .932
|| 2          1, 35 276.17      0.50 .007      .485
|| 3          1, 35 276.17      1.25 .018      .272
|| 4          1, 35  32.05      0.00 <.001      .980
|| 5          1, 35  32.05      0.00 <.001      .990
|| 6          1, 35  32.05      0.25 <.001      .620
|| 7          1, 35  32.05      0.62 .001      .436
|| 8          1, 35  33.59      1.53 .003      .225
|| 9          1, 35  33.59      0.03 <.001      .856
|| 10         1, 35  33.59      2.13 .004      .154
|| 11         1, 35  33.59      1.77 .003      .192
|| 12 1.26, 44.16 53.37      0.94 .003      .359
|| 13 1.26, 44.16 53.37      0.30 .001      .639
|| 14 1.26, 44.16 53.37      0.85 .003      .386
|| 15 1.26, 44.16 53.37      1.24 .004      .282
|| 16 1.73, 60.48 13.63      0.58 <.001      .538

```


17	1.73, 60.48	13.63	4.43 *	.005	.020
18	1.73, 60.48	13.63	2.62 +	.003	.089
19	1.73, 60.48	13.63	0.13 <	.001	.847
20	1, 35	16.54	10.73 **	.009	.002
21	1, 35	16.54	0.79 <	.001	.381
22	1, 35	16.54	0.48 <	.001	.492
23	1, 35	16.54	0.21 <	.001	.650
24	1.30, 45.35	6.36	0.01 <	.001	.950
25	1.30, 45.35	6.36	0.23 <	.001	.696
26	1.30, 45.35	6.36	0.44 <	.001	.561
27	1.30, 45.35	6.36	1.97 <	.001	.164
28	1.33, 46.66	6.96	0.78 <	.001	.417
29	1.33, 46.66	6.96	0.31 <	.001	.648
30	1.33, 46.66	6.96	0.35 <	.001	.618
31	1.33, 46.66	6.96	2.96 +	.001	.081
32	1.30, 45.42	4.50	2.13 <	.001	.146
33	1.30, 45.42	4.50	1.17 <	.001	.299
34	1.30, 45.42	4.50	1.59 <	.001	.216
35	1.30, 45.42	4.50	0.39 <	.001	.588
36	1.43, 50.16	6.81	0.33 <	.001	.646
37	1.43, 50.16	6.81	0.15 <	.001	.785
38	1.43, 50.16	6.81	0.86 <	.001	.396
39	1.43, 50.16	6.81	0.34 <	.001	.641
40	1.83, 64.11	8.89	3.11 +	.003	.056
41	1.83, 64.11	8.89	1.56	.001	.220
42	1.83, 64.11	8.89	0.58 <	.001	.548
43	1.83, 64.11	8.89	1.45	.001	.243
44	1.32, 46.07	6.36	1.10 <	.001	.318
45	1.32, 46.07	6.36	0.15 <	.001	.764
46	1.32, 46.07	6.36	1.56 <	.001	.221
47	1.32, 46.07	6.36	2.49	.001	.113
48	1.40, 48.91	3.66	0.48 <	.001	.556
49	1.40, 48.91	3.66	0.06 <	.001	.888
50	1.40, 48.91	3.66	0.50 <	.001	.546
51	1.40, 48.91	3.66	0.56 <	.001	.513
52	1.22, 42.84	9.21	0.70 <	.001	.435
53	1.22, 42.84	9.21	1.41 <	.001	.248
54	1.22, 42.84	9.21	0.70 <	.001	.435
55	1.22, 42.84	9.21	1.05 <	.001	.325
56	1.36, 47.75	7.61	0.63 <	.001	.476
57	1.36, 47.75	7.61	1.28 <	.001	.276
58	1.36, 47.75	7.61	0.89 <	.001	.382
59	1.36, 47.75	7.61	0.69 <	.001	.453
60	1.23, 42.99	8.14	1.33 <	.001	.263
61	1.23, 42.99	8.14	1.22 <	.001	.286
62	1.23, 42.99	8.14	1.55 <	.001	.224
63	1.23, 42.99	8.14	1.02 <	.001	.335

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '+' 0.1 ' ' 1					
Sphericity correction method: GG					

5 Examine and plot interactions

5.1 Group 1

5.1.1 Family Size by Complexity by Anteriority Interaction

```
# Examine the 2-way interaction between `family_size` and `complexity`  
# at each level of `Anteriority`  
(se_frontal_1.1 <-n250_1_nonwords |> filter(anteriority == "Frontal")|>  
  ezANOVA(dv = value,  
    wid = SubjID,  
    within = .(complexity, family_size)))
```

5.1.1.1 Simple Effects complexity| family_size * anteriority

```
|| $ANOVA  
||  
||           Effect DFn DFd           F           p p<.05           ges  
|| 2           complexity      1 66 1.0123262 0.3180208      0.0012870537  
|| 3           family_size      1 66 0.1344549 0.7150300      0.0001689185  
|| 4 complexity:family_size      1 66 0.1623686 0.6882874      0.0002476601
```

```
(se_central_1.1 <-n250_1_nonwords |> filter(anteriority == "Central")|>  
  ezANOVA(dv = value,  
    wid = SubjID,  
    within = .(complexity, family_size)))
```

```
|| $ANOVA  
||  
||           Effect DFn DFd           F           p p<.05           ges  
|| 2           complexity      1 66 0.01336488 0.9083154      2.478299e-05  
|| 3           family_size      1 66 0.22231793 0.6388344      4.309762e-04  
|| 4 complexity:family_size      1 66 0.73842072 0.3932777      1.132454e-03
```

```
(se_parietal_1.1 <-n250_1_nonwords |> filter(anteriority == "Parietal")|>  
  ezANOVA(dv = value,  
    wid = SubjID,  
    within = .(complexity, family_size)))
```

```
|| $ANOVA  
||  
||           Effect DFn DFd           F           p p<.05           ges  
|| 2           complexity      1 66 0.26773166 0.60658740      4.849084e-04  
|| 3           family_size      1 66 0.03579367 0.85052283      5.836017e-05  
|| 4 complexity:family_size      1 66 3.97705105 0.05025801      6.186163e-03
```

```
# Examine `complexity` at each level of `family_size` at parietal sites.  
(se_parietal_small_1.1 <-n250_1_nonwords |> filter(anteriority == "Parietal" &  
  family_size == "small")|>  
  ezANOVA(dv = value,  
    wid = SubjID,  
    within = complexity))
```

```

|| $ANOVA
||      Effect DFn DFd      F      p p<.05      ges
|| 2 complexity   1  66 2.621483 0.1101951      0.009714344

```

```

(se_parietal_large_1.1 <- n250_1_nonwords |> filter(anteriority == "Parietal" &
                                                    family_size == "large")|>

ezANOVA(dv = value,
        wid = SubjID,
        within = complexity))

```

```

|| $ANOVA
||      Effect DFn DFd      F      p p<.05      ges
|| 2 complexity   1  66 1.127293 0.2922231      0.003351804

```

There is a significant complexity by family size interaction only at parietal sites $F(1,66) = 3.97705105, p = 0.05025801$, where the complexity effect is a slightly more probable for small families $F(1,66) = 2.621483, p = 0.1101951$ than for large $F(1,66) = 1.127293, p = 0.2922231$.

```

emms <- emmeans(anova_results.1b, ~ complexity | family_size * anteriority )
pairwise_results <- pairs(emms, by = c("anteriority", "family_size"))
summary(pairwise_results)

```

5.1.1.2 Pairwise Comparisons complexity | family_size * anteriority

```

|| anteriority = Frontal, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple    0.181 0.368 56   0.492  0.6243
||
|| anteriority = Central, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.079 0.411 56  -0.192  0.8483
||
|| anteriority = Parietal, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.500 0.411 56  -1.215  0.2295
||
|| anteriority = Frontal, family_size = large:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple    0.334 0.349 56   0.957  0.3426
||
|| anteriority = Central, family_size = large:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple    0.252 0.358 56   0.704  0.4841
||
|| anteriority = Parietal, family_size = large:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple    0.464 0.357 56   1.300  0.1990
||

```

|| Results are averaged over the levels of: lang_type_semantic, lang_type_ortho, laterality

```
(nw_famsize_cmplx_ant_1 <- n250_1_nonwords |>
  na.omit()|>
  group_by(family_size, anteriority, complexity) |>
  summarise(mean = mean(value),
            se = sem(value),
            num_stim = n()))
```

5.1.1.3 Condition Means

```
|| # A tibble: 12 x 6
|| # Groups:   family_size, anteriority [6]
||   family_size anteriority complexity    mean    se num_stim
||   <chr>        <fct>      <chr>    <dbl> <dbl>    <int>
|| 1 large      Frontal    complex -0.752 0.253    180
|| 2 large      Frontal    simple  -1.05 0.237    180
|| 3 large      Central    complex -0.976 0.225    180
|| 4 large      Central    simple  -1.21 0.232    180
|| 5 large      Parietal   complex  0.248 0.234    180
|| 6 large      Parietal   simple  -0.171 0.253    180
|| 7 small      Frontal    complex -0.703 0.230    180
|| 8 small      Frontal    simple  -0.940 0.245    180
|| 9 small      Central    complex -0.951 0.232    180
|| 10 small     Central    simple  -0.931 0.268    180
|| 11 small     Parietal   complex -0.0755 0.215    180
|| 12 small     Parietal   simple   0.371 0.284    180
```

```
(difference_scores_1.1 <- nw_famsize_cmplx_ant_1 %>%
  pivot_wider(names_from = complexity, values_from = c(mean, se, num_stim)) %>%
  mutate(mean_diff = `mean_simple` - `mean_complex`,
         avg_se = mean(`se_complex`, `se_simple`),
         total_num_stim = sum(`num_stim_complex`, `num_stim_simple`)))
```

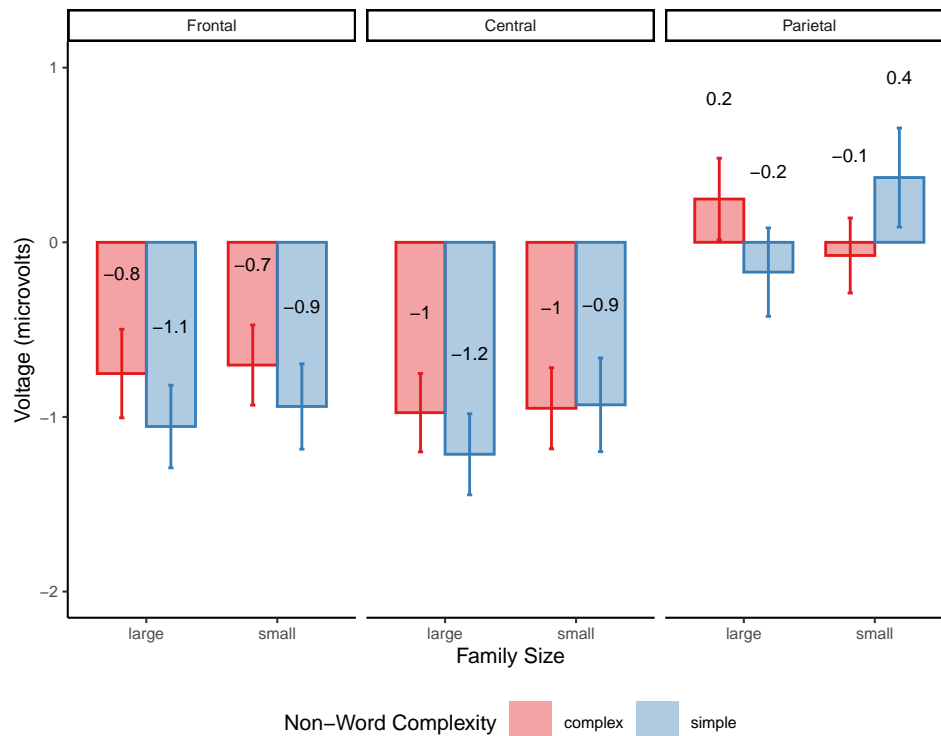
5.1.1.4 Diff Scores

```
|| # A tibble: 6 x 11
|| # Groups:   family_size, anteriority [6]
||   family_size anteriority mean_complex mean_simple se_complex se_simple
||   <chr>        <fct>      <dbl>      <dbl>    <dbl>    <dbl>
|| 1 large      Frontal    -0.752    -1.05    0.253    0.237
|| 2 large      Central    -0.976    -1.21    0.225    0.232
|| 3 large      Parietal    0.248    -0.171    0.234    0.253
|| 4 small      Frontal    -0.703    -0.940    0.230    0.245
|| 5 small      Central    -0.951    -0.931    0.232    0.268
|| 6 small      Parietal    -0.0755    0.371    0.215    0.284
|| # i 5 more variables: num_stim_complex <int>, num_stim_simple <int>,
|| #   mean_diff <dbl>, avg_se <dbl>, total_num_stim <int>
```

5.1.1.5 Plots First we plot the raw scores then the difference scores

```
# plot raw scores
# facet_wrap() wraps a 1d sequence of panels into 2d. Use vars() to supply faceting variables;
# Control the number of rows and columns with nrow and ncol.

p1.a <- nw_famsize_cmplx_ant_1 |> ggplot(aes(x= family_size, y=mean,
      fill = complexity, colour = complexity,
      ymin = mean - se, ymax = mean + se)) +
  facet_wrap(vars(anteriority), ncol = 3, labeller = "label_value") +
  coord_cartesian(xlim = NULL, ylim = c(-2, 1), expand = TRUE, default = FALSE, clip = "on") +
  geom_col(position = "dodge", width = 0.75, alpha = .4) +
  labs(y = "Voltage (microvolts)", x = "Family Size") +
  geom_errorbar(width = .08, position = position_dodge(0.75)) +
  theme_classic(base_size = 8) +
  geom_text(aes(label = round(mean, digits = 1)), colour = "black",
    size = 2.5, vjust = -7,
    position = position_dodge(.75))+
  guides(fill=guide_legend(title="Non-Word Complexity"),
    colour= "none") +
  theme(legend.position = "bottom")
p1.a + scale_fill_brewer(palette = "Set1")+
  scale_colour_brewer(palette = "Set1")
```

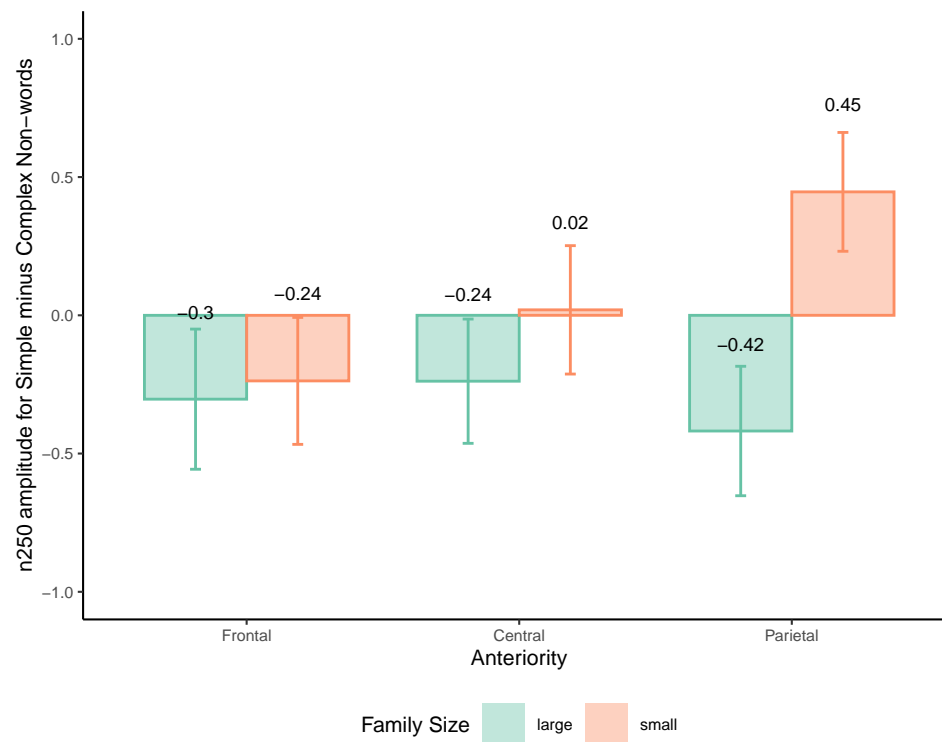


```
# plot diff scores
p1.b <- difference_scores_1.1 |> ggplot(aes(x = anteriority,
      y = mean_diff,
      fill = family_size,
```

```

    colour = family_size,
    ymin = mean_diff - avg_se,
    ymax = mean_diff + avg_se)) +
coord_cartesian(xlim = NULL,ylim = c(-1, 1), expand = TRUE,default = FALSE,clip = "on") +
geom_col(position = "dodge", width = 0.75, alpha = 0.4) +
labs(y = "n250 amplitude for Simple minus Complex Non-words", x = "Anteriority") +
geom_errorbar(width = .08, position = position_dodge(0.75)) +
theme_classic(base_size = 8) +
geom_text(aes(label = round(mean_diff, digits = 2)),colour = "black",size = 2.5, vjust = -6,
    position = position_dodge(.75))+
guides(fill=guide_legend(title="Family Size"),
    colour= "none") +
theme(legend.position = "bottom")
p1.b + scale_fill_brewer(palette = "Set2")+
    scale_colour_brewer(palette = "Set2")

```



5.1.2 Language Type Semantic by Complexity by Family Size by Anteriority x Laterality Interaction

```

# Examine the 4-way interaction between `anteriority`, `laterality`, `complexity`,
# and `lang_type_semantics` at each level of `family_size`
se_large_1.2 <- n250_1_nonwords |> filter(family_size == "small")|>
  ezANOVA(dv = value,
    wid = SubjID,
    within = .(complexity, anteriority, laterality),
    between = .(lang_type_semantic),

```

```

    type = 3)
se_large_1.2$`Sphericity Corrections`

```

5.1.2.1 Simple Effects complexity | lang_type_semantic * family_size * laterality * anteriority

	Effect	GGe	p[GG]
5	anteriority	0.6420185	0.0006468489
6	lang_type_semantic:anteriority	0.6420185	0.8969800170
7	laterality	0.9497291	0.2894057610
8	lang_type_semantic:laterality	0.9497291	0.6172478894
9	complexity:anteriority	0.8377386	0.0178795376
10	lang_type_semantic:complexity:anteriority	0.8377386	0.0072212665
11	complexity:laterality	0.7471483	0.3305037994
12	lang_type_semantic:complexity:laterality	0.7471483	0.5152148668
13	anteriority:laterality	0.8519658	0.7110158444
14	lang_type_semantic:anteriority:laterality	0.8519658	0.1056260982
15	complexity:anteriority:laterality	0.6938894	0.2714100076
16	lang_type_semantic:complexity:anteriority:laterality	0.6938894	0.3781100282

	p[GG]<.05	HFe	p[HF]	p[HF]<.05
5	*	0.6502418	0.0006103261	*
6		0.6502418	0.8993752116	
7		0.9809890	0.2899966187	
8		0.9809890	0.6234363611	
9	*	0.8597778	0.0170414357	*
10	*	0.8597778	0.0067500799	*
11		0.7624319	0.3316615514	
12		0.7624319	0.5185158604	
13		0.9115961	0.7235423754	
14		0.9115961	0.1007803592	
15		0.7322749	0.2708036651	
16		0.7322749	0.3806240337	

```

se_small_1.2 <-n250_1_nonwords |> filter(family_size == "large")|>
  ezANOVA(dv = value,
    wid = SubjID,
    within = .(complexity, anteriority, laterality),
    between = lang_type_semantic)
se_small_1.2$`Sphericity Corrections`

```

	Effect	GGe	p[GG]
5	anteriority	0.5954644	0.002232974
6	lang_type_semantic:anteriority	0.5954644	0.513478455
7	laterality	0.9564988	0.414803890
8	lang_type_semantic:laterality	0.9564988	0.389741357
9	complexity:anteriority	0.7614393	0.639974214
10	lang_type_semantic:complexity:anteriority	0.7614393	0.886320231
11	complexity:laterality	0.8649282	0.880900161
12	lang_type_semantic:complexity:laterality	0.8649282	0.332825187
13	anteriority:laterality	0.7843992	0.194956664
14	lang_type_semantic:anteriority:laterality	0.7843992	0.181594680
15	complexity:anteriority:laterality	0.8966788	0.321410726
16	lang_type_semantic:complexity:anteriority:laterality	0.8966788	0.013587141

	p[GG]<.05	HFe	p[HF]	p[HF]<.05
5	*	0.6008266	0.002165149	*
6		0.6008266	0.515087060	
7		0.9883472	0.417763501	
8		0.9883472	0.392224709	
9		0.7777471	0.644501670	
10		0.7777471	0.890363334	
11		0.8891174	0.886140731	
12		0.8891174	0.334174970	
13		0.8344431	0.191850888	
14		0.8344431	0.178202807	
15		0.9630745	0.321595863	
16	*	0.9630745	0.011485622	*

*# Examine the 3-way interaction between `complexity`, `anteriority` and `laterality`
at each level of `lang_type_semantics` for non-words from large families*

```
se_large_hisem_1.2 <-n250_1_nonwords |> filter(family_size == "large" &
                                                lang_type_semantic == "High Semantic")|>
  ezANOVA(dv = value,
    wid = SubjID,
    within = .(anteriority, laterality, complexity))
se_large_hisem_1.2$`Sphericity Corrections`
```

	Effect	GGe	p[GG]	p[GG]<.05	HFe
2	anteriority	0.6110393	0.005812212	*	0.6239192
3	laterality	0.7877555	0.791690570		0.8252721
5	anteriority:laterality	0.8015027	0.075135208		0.9128127
6	anteriority:complexity	0.7243769	0.676748688		0.7524635
7	laterality:complexity	0.8457392	0.460557823		0.8924752
8	anteriority:laterality:complexity	0.7942870	0.014072816	*	0.9034095

	p[HF]	p[HF]<.05
2	0.005492252	*
3	0.802157676	
5	0.066153967	
6	0.685275708	
7	0.467096302	
8	0.010357847	*

```
se_large_losem_1.2 <-n250_1_nonwords |> filter(family_size == "large" &
                                                lang_type_semantic == "Low Semantic")|>
  ezANOVA(dv = value,
    wid = SubjID,
    within = .(complexity, anteriority, laterality))
se_large_losem_1.2$`Sphericity Corrections`
```

	Effect	GGe	p[GG]	p[GG]<.05	HFe
3	anteriority	0.5835088	0.1058768		0.5930110
4	laterality	0.9887507	0.2035848		1.0606910
5	complexity:anteriority	0.7962426	0.8286265		0.8350731
6	complexity:laterality	0.8910445	0.6677000		0.9453824
7	anteriority:laterality	0.7255469	0.3300543		0.8148732
8	complexity:anteriority:laterality	0.8165460	0.3079664		0.9324843


```

||           p[HF] p[HF]<.05
|| 3 0.1051529
|| 4 0.2032465
|| 5 0.8388942
|| 6 0.6800471
|| 7 0.3316478
|| 8 0.3074230

```

*# Examine the 2-way interaction between complexity and anteriority
at each level of laterality for non-words from large families for high semantic readers*

```

# left
se_large_hisem_left_1.2 <-n250_1_nonwords |>
  filter(family_size == "large" &
         lang_type_semantic == "High Semantic" &
         laterality == "Left")|>
  ezANOVA(dv = value,
         wid = SubjID,
         within = .(complexity, anteriority))
se_large_hisem_left_1.2$`Sphericity Corrections`

```

```

||           Effect      GGe      p[GG] p[GG]<.05      HFe      p[HF]
|| 3           anteriority 0.6254110 0.03982631      * 0.6401030 0.03877314
|| 4 complexity:anteriority 0.8400725 0.09421697      0.8858823 0.09115917
|| p[HF]<.05
|| 3           *
|| 4

```

```

# midline
se_large_hisem_mid_1.2 <-n250_1_nonwords |>
  filter(family_size == "large" &
         lang_type_semantic == "High Semantic" &
         laterality == "Midline")|>
  ezANOVA(dv = value,
         wid = SubjID,
         within = .(complexity, anteriority))
se_large_hisem_mid_1.2$`Sphericity Corrections`

```

```

||           Effect      GGe      p[GG] p[GG]<.05      HFe      p[HF]
|| 3           anteriority 0.6878769 0.00893451      * 0.7108360 0.008255881
|| 4 complexity:anteriority 0.8055359 0.59499271      0.8458191 0.604317874
|| p[HF]<.05
|| 3           *
|| 4

```

```

# right
se_large_hisem_right_1.2 <-n250_1_nonwords |>
  filter(family_size == "large" &
         lang_type_semantic == "High Semantic" &
         laterality == "Right")|>
  ezANOVA(dv = value,
         wid = SubjID,

```

```

    within = .(complexity, anteriority))
se_large_hisem_right_1.2$`Sphericity Corrections`

```

```

||           Effect      GGe      p[GG] p[GG]<.05      HFe      p[HF]
|| 3           anteriority 0.594651 0.001884944      * 0.6055055 0.001763554
|| 4 complexity:anteriority 0.837690 0.572013472      0.8831120 0.581452330
|| p[HF]<.05
|| 3      *
|| 4

```

*# Finally we examine the simple effect of complexity at each level of anteriority
for non-words from large families for high semantic readers at left sites
Frontal*

```

se_large_hisem_left_frontal_1.2 <- n250_1_nonwords |>
  filter(family_size == "large" &
    lang_type_semantic == "High Semantic" &
    laterality == "Left" &
    anteriority == "Frontal" )|>
  ezANOVA(dv = value,
    wid = SubjID,
    within = .(complexity))
se_large_hisem_left_frontal_1.2$ANOVA

```

```

||           Effect DFn DFd      F      p p<.05      ges
|| 2 complexity    1  29 0.1946546 0.6623425      0.001206553

```

Central

```

se_large_hisem_left_central_1.2 <- n250_1_nonwords |>
  filter(family_size == "large" &
    lang_type_semantic == "High Semantic" &
    laterality == "Left" &
    anteriority == "Central" )|>
  ezANOVA(dv = value,
    wid = SubjID,
    within = .(complexity))
se_large_hisem_left_central_1.2$ANOVA

```

```

||           Effect DFn DFd      F      p p<.05      ges
|| 2 complexity    1  29 0.3014515 0.5871769      0.002813469

```

Parietal

```

se_large_hisem_left_parietal_1.2 <- n250_1_nonwords |>
  filter(family_size == "large" &
    lang_type_semantic == "High Semantic" &
    laterality == "Left" &
    anteriority == "Parietal" )|>
  ezANOVA(dv = value,
    wid = SubjID,
    within = .(complexity))
se_large_hisem_left_parietal_1.2$ANOVA

```

```

||           Effect DFn DFd      F      p p<.05      ges
|| 2 complexity    1  29 0.8494285 0.3643184      0.00653431

```

```
emms <- emmeans(anova_results.1b, ~complexity|lang_type_semantic*family_size*laterality*anteriority)
pairwise_results <- pairs(emms, by = c("laterality", "anteriority", "lang_type_semantic", "family_size"))
summary(pairwise_results)
```

5.1.2.2 Pairwise Comparisons complexity | lang_type_semantic * family_size * laterality * anteriority

```
|| laterality = Left, anteriority = Frontal, lang_type_semantic = High Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.281 0.525 56  -0.536  0.5944
||
|| laterality = Midline, anteriority = Frontal, lang_type_semantic = High Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.480 0.563 56  -0.854  0.3966
||
|| laterality = Right, anteriority = Frontal, lang_type_semantic = High Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.327 0.578 56  -0.565  0.5741
||
|| laterality = Left, anteriority = Central, lang_type_semantic = High Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.555 0.539 56  -1.029  0.3079
||
|| laterality = Midline, anteriority = Central, lang_type_semantic = High Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.213 0.621 56  -0.343  0.7330
||
|| laterality = Right, anteriority = Central, lang_type_semantic = High Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.357 0.763 56  -0.468  0.6419
||
|| laterality = Left, anteriority = Parietal, lang_type_semantic = High Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.278 0.574 56  -0.484  0.6303
||
|| laterality = Midline, anteriority = Parietal, lang_type_semantic = High Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.343 0.630 56  -0.545  0.5883
||
|| laterality = Right, anteriority = Parietal, lang_type_semantic = High Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple  -0.205 0.608 56  -0.337  0.7370
||
|| laterality = Left, anteriority = Frontal, lang_type_semantic = Low Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple   0.470 0.521 56   0.903  0.3706
||
|| laterality = Midline, anteriority = Frontal, lang_type_semantic = Low Semantic, family_size = small:
|| contrast      estimate      SE df t.ratio p.value
|| complex - simple   0.918 0.559 56   1.643  0.1059
||
|| laterality = Right, anteriority = Frontal, lang_type_semantic = Low Semantic, family_size = small:
```

contrast	estimate	SE	df	t.ratio	p.value
complex - simple	0.788	0.574	56	1.373	0.1752
lateralality = Left, anteriority = Central, lang_type_semantic = Low Semantic, family_size = small:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	-0.277	0.535	56	-0.518	0.6064
lateralality = Midline, anteriority = Central, lang_type_semantic = Low Semantic, family_size = small:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	0.257	0.617	56	0.416	0.6790
lateralality = Right, anteriority = Central, lang_type_semantic = Low Semantic, family_size = small:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	0.671	0.758	56	0.886	0.3796
lateralality = Left, anteriority = Parietal, lang_type_semantic = Low Semantic, family_size = small:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	-0.614	0.570	56	-1.077	0.2862
lateralality = Midline, anteriority = Parietal, lang_type_semantic = Low Semantic, family_size = small:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	-1.028	0.626	56	-1.641	0.1064
lateralality = Right, anteriority = Parietal, lang_type_semantic = Low Semantic, family_size = small:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	-0.530	0.604	56	-0.879	0.3833
lateralality = Left, anteriority = Frontal, lang_type_semantic = High Semantic, family_size = large:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	-0.138	0.539	56	-0.256	0.7987
lateralality = Midline, anteriority = Frontal, lang_type_semantic = High Semantic, family_size = large:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	0.408	0.552	56	0.739	0.4632
lateralality = Right, anteriority = Frontal, lang_type_semantic = High Semantic, family_size = large:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	0.252	0.475	56	0.531	0.5978
lateralality = Left, anteriority = Central, lang_type_semantic = High Semantic, family_size = large:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	-0.236	0.521	56	-0.452	0.6528
lateralality = Midline, anteriority = Central, lang_type_semantic = High Semantic, family_size = large:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	0.116	0.604	56	0.191	0.8489
lateralality = Right, anteriority = Central, lang_type_semantic = High Semantic, family_size = large:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	0.148	0.505	56	0.293	0.7708
lateralality = Left, anteriority = Parietal, lang_type_semantic = High Semantic, family_size = large:					
contrast	estimate	SE	df	t.ratio	p.value
complex - simple	0.548	0.480	56	1.141	0.2586

```

||
|| laterality = Midline, anteriority = Parietal, lang_type_semantic = High Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.401 0.555 56   0.722  0.4731
||
|| laterality = Right, anteriority = Parietal, lang_type_semantic = High Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple   -0.107 0.555 56  -0.193  0.8474
||
|| laterality = Left, anteriority = Frontal, lang_type_semantic = Low Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.707 0.535 56   1.320  0.1921
||
|| laterality = Midline, anteriority = Frontal, lang_type_semantic = Low Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.480 0.549 56   0.875  0.3851
||
|| laterality = Right, anteriority = Frontal, lang_type_semantic = Low Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.294 0.472 56   0.622  0.5364
||
|| laterality = Left, anteriority = Central, lang_type_semantic = Low Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.587 0.517 56   1.135  0.2612
||
|| laterality = Midline, anteriority = Central, lang_type_semantic = Low Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.260 0.600 56   0.433  0.6666
||
|| laterality = Right, anteriority = Central, lang_type_semantic = Low Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.640 0.501 56   1.276  0.2073
||
|| laterality = Left, anteriority = Parietal, lang_type_semantic = Low Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.523 0.477 56   1.098  0.2768
||
|| laterality = Midline, anteriority = Parietal, lang_type_semantic = Low Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.613 0.552 56   1.112  0.2709
||
|| laterality = Right, anteriority = Parietal, lang_type_semantic = Low Semantic, family_size = large:
|| contrast      estimate    SE df t.ratio p.value
|| complex - simple    0.804 0.551 56   1.459  0.1501
||
|| Results are averaged over the levels of: lang_type_ortho

```

```

(nw_sem_famsize_lat_ant_cmplx_1 <- n250_1_nonwords |>
  na.omit()|>
  group_by( lang_type_semantic, family_size, laterality, anteriority, complexity ) |>
  summarise(mean = mean(value),

```

```
se = sem(value),
num_stim = n()))
```

5.1.2.3 Condition Means complexity | lang_type_semantic * family_size * laterality * anteriority

```
|| # A tibble: 72 x 8
|| # Groups:   lang_type_semantic, family_size, laterality, anteriority [36]
||   lang_type_semantic family_size laterality anteriority complexity mean se
||   <chr>             <chr>      <fct>      <fct>      <chr>      <dbl> <dbl>
|| 1 High Semantic     large    Left      Frontal    complex   -0.557 0.507
|| 2 High Semantic     large    Left      Frontal    simple    -0.354 0.575
|| 3 High Semantic     large    Left      Central    complex   -0.789 0.454
|| 4 High Semantic     large    Left      Central    simple    -0.522 0.481
|| 5 High Semantic     large    Left      Parietal   complex    0.615 0.495
|| 6 High Semantic     large    Left      Parietal   simple     0.141 0.586
|| 7 High Semantic     large    Midline   Frontal    complex   -0.554 0.544
|| 8 High Semantic     large    Midline   Frontal    simple    -0.949 0.555
|| 9 High Semantic     large    Midline   Central    complex   -0.694 0.509
|| 10 High Semantic    large    Midline   Central    simple    -0.818 0.573
|| # i 62 more rows
|| # i 1 more variable: num_stim <int>
```

```
(difference_scores_1.2 <- nw_sem_famsize_lat_ant_cmplx_1 %>%
  pivot_wider(names_from = complexity, values_from = c(mean, se, num_stim)) %>%
  mutate(mean_diff = `mean_simple` - `mean_complex`,
         avg_se = mean(`se_complex`, `se_simple`),
         total_num_stim = sum(`num_stim_complex`, `num_stim_simple`)))
```

5.1.2.4 Diff Scores complexity | lang_type_semantic * family_size * laterality * anteriority

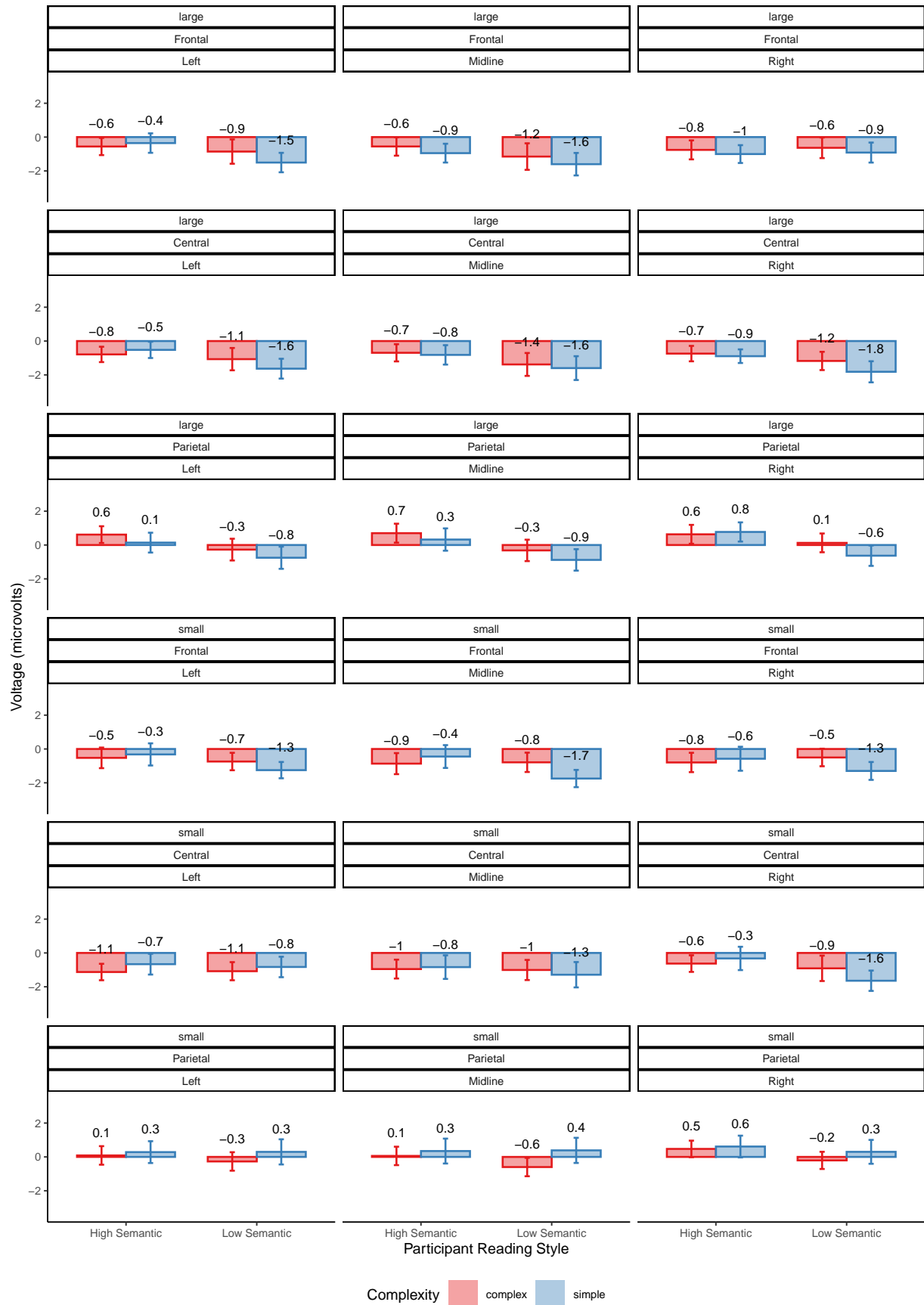
```
|| # A tibble: 36 x 13
|| # Groups:   lang_type_semantic, family_size, laterality, anteriority [36]
||   lang_type_semantic family_size laterality anteriority mean_complex
||   <chr>             <chr>      <fct>      <fct>      <dbl>
|| 1 High Semantic     large    Left      Frontal    -0.557
|| 2 High Semantic     large    Left      Central    -0.789
|| 3 High Semantic     large    Left      Parietal    0.615
|| 4 High Semantic     large    Midline   Frontal    -0.554
|| 5 High Semantic     large    Midline   Central    -0.694
|| 6 High Semantic     large    Midline   Parietal    0.700
|| 7 High Semantic     large    Right     Frontal    -0.756
|| 8 High Semantic     large    Right     Central    -0.742
|| 9 High Semantic     large    Right     Parietal    0.633
|| 10 High Semantic    small    Left      Frontal    -0.525
|| # i 26 more rows
|| # i 8 more variables: mean_simple <dbl>, se_complex <dbl>, se_simple <dbl>,
|| #   num_stim_complex <int>, num_stim_simple <int>, mean_diff <dbl>,
|| #   avg_se <dbl>, total_num_stim <int>
```

5.1.2.5 Plots complexity | lang_type_semantic * family_size * laterality * anteriority
First we plot the raw scores then the difference scores

Plot interaction complexity | lang_type_semantic * family_size * laterality * anteriority
Raw Scores facet_wrap() wraps a 1d sequence of panels into 2d. Use vars() to supply faceting variables; Control the number of rows and columns with nrow and ncol. labeller options are “label_value” and “label_both”. The latter prints the name of the variable & its value.

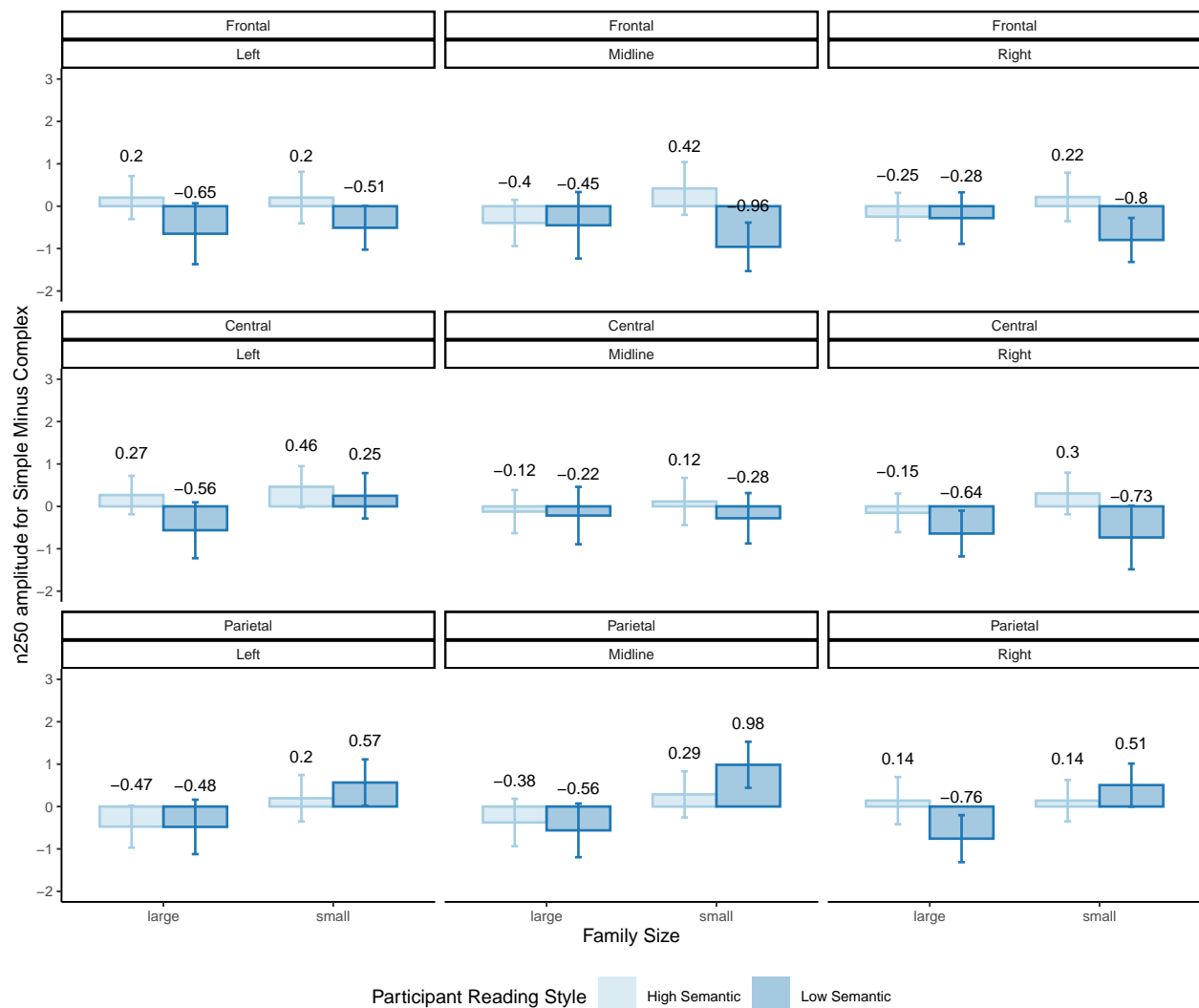
Plot raw scores

```
p2.a <- nw_sem_famsize_lat_ant_cmplx_1 |> ggplot(aes(x= lang_type_semantic, y=mean,
                                                    fill = complexity, colour = complexity,
                                                    ymin = mean - se, ymax = mean + se)) +
  facet_wrap(vars(family_size, anteriority, laterality),
            labeller = "label_value", ncol = 3) +
  coord_cartesian(xlim = NULL, ylim = c(-3.5, 3.5), expand=TRUE, default=FALSE, clip="on") +
  geom_col(position = "dodge", width = 0.75, alpha = 0.4) +
  labs(y = "Voltage (microvolts)", x = "Participant Reading Style") +
  geom_errorbar(width = .08, position = position_dodge(0.75)) +
  theme_classic(base_size = 8) +
  geom_text(aes(label = round(mean, digits = 1)),
            colour = "black",
            size = 2.5,
            vjust = -2,
            position = position_dodge(.75))+
  guides(fill=guide_legend(title="Complexity"),
         colour= "none") +
  theme(legend.position = "bottom")
p2.a + scale_fill_brewer(palette = "Set1")+
  scale_colour_brewer(palette = "Set1")
```



Plot diff scores

```
p2.b <- difference_scores_1.2 |> ggplot(aes(x = family_size, y = mean_diff,
      fill = lang_type_semantic, colour = lang_type_semantic,
      ymin = mean_diff - avg_se, ymax = mean_diff + avg_se)) +
  facet_wrap(vars(anteriority, laterality),
    labeller = "label_value", ncol = 3) +
  coord_cartesian(xlim = NULL, ylim = c(-2, 3), expand = TRUE, default = FALSE, clip = "on") +
  geom_col(position = "dodge", width = 0.75, alpha = 0.4) +
  labs(y = "n250 amplitude for Simple Minus Complex", x = "Family Size") +
  geom_errorbar(width = .08, position = position_dodge(0.75)) +
  theme_classic(base_size = 8) +
  geom_text(aes(label = round(mean_diff, digits = 2)), colour = "black", size = 2.5, vjust = -3,
    position = position_dodge(.75)) +
  guides(fill=guide_legend(title="Participant Reading Style"),
    colour= "none") +
  theme(legend.position = "bottom")
p2.b + scale_fill_brewer(palette = "Paired") +
  scale_colour_brewer(palette = "Paired")
```



5.2 Group 2

5.2.1 Language Type Semantic by Laterality Interaction

```
# Examine the difference between `lang_type_semantic` at each level of `laterality`  
se_left_2.1 <- n250_2_nonwords |> filter(laterality == "Left")|>  
  ezANOVA(dv = value,  
          wid = SubjID,  
          between = lang_type_semantic)  
se_left_2.1$ANOVA
```

5.2.1.1 Simple Effects lang_type_semantic | laterality

		Effect	DFn	DFd	F	p	p<.05	ges
	1	lang_type_semantic	1	37	0.2837876	0.5974145		0.007611555

```
se_mid_2.1 <- n250_2_nonwords |> filter(laterality == "Midline")|>  
  ezANOVA(dv = value,  
          wid = SubjID,  
          between = lang_type_semantic)  
se_mid_2.1$ANOVA
```

		Effect	DFn	DFd	F	p	p<.05	ges
	1	lang_type_semantic	1	37	0.02925513	0.8651233		0.0007900544

```
se_right_2.1 <- n250_2_nonwords |> filter(laterality == "Right")|>  
  ezANOVA(dv = value,  
          wid = SubjID,  
          between = lang_type_semantic)  
se_right_2.1$ANOVA
```

		Effect	DFn	DFd	F	p	p<.05	ges
	1	lang_type_semantic	1	37	0.6945411	0.4099719		0.01842551

No significant effect of `lang_type_semantic` at any laterality, but the effect is less probable at midline sites ($F(1,37) = 0.02925513, p = 0.8651233$) than at left ($F(1,37) = 0.2837876, p = 0.5974145$) or right ($F(1,37) = 0.6945411, p = 0.4099719$) sites

```
emms <- emmeans(anova_results.2b, ~ lang_type_semantic | laterality )  
pairwise_results <- pairs(emms, by = c("laterality"))  
summary(pairwise_results)
```

5.2.1.2 Pairwise Comparisons lang_type_semantic | laterality

```

|| laterality = Left:
|| contrast estimate SE df t.ratio p.value
|| High Semantic - Low Semantic -0.619 0.868 35 -0.713 0.4806
||
|| laterality = Midline:
|| contrast estimate SE df t.ratio p.value
|| High Semantic - Low Semantic -0.288 1.000 35 -0.287 0.7755
||
|| laterality = Right:
|| contrast estimate SE df t.ratio p.value
|| High Semantic - Low Semantic 0.677 0.920 35 0.736 0.4668
||
|| Results are averaged over the levels of: lang_type_ortho, anteriority, complexity, family_size

```

```

(nw_langsem_lat_2 <- n250_2_nonwords |>
  na.omit())>
group_by(laterality, lang_type_semantic) |>
summarise(mean = mean(value),
  se = sem(value),
  num = n())

```

5.2.1.3 Condition Means

```

|| # A tibble: 6 x 5
|| # Groups:   laterality [3]
||   laterality lang_type_semantic mean se num
||   <fct>      <chr>          <dbl> <dbl> <int>
|| 1 Left      High Semantic      2.74 0.180 252
|| 2 Left      Low Semantic       3.21 0.266 216
|| 3 Midline   High Semantic      3.07 0.196 252
|| 4 Midline   Low Semantic       3.24 0.304 216
|| 5 Right     High Semantic      3.53 0.191 252
|| 6 Right     Low Semantic       2.78 0.367 216

```

```

(difference_scores_2.1 <- nw_langsem_lat_2 %>%
  pivot_wider(names_from = lang_type_semantic, values_from = c(mean, se, num)) %>%
  mutate(mean_diff = `mean_Low Semantic` - `mean_High Semantic`,
    avg_se = mean(`se_Low Semantic`, `se_High Semantic`),
    total_num_stim = sum(`num_Low Semantic`, `num_High Semantic`)))

```

5.2.1.4 Diff Scores

```

|| # A tibble: 3 x 10
|| # Groups:   laterality [3]
||   laterality 'mean_High Semantic' 'mean_Low Semantic' 'se_High Semantic'
||   <fct>      <dbl>          <dbl>          <dbl>
|| 1 Left      2.74          3.21          0.180
|| 2 Midline   3.07          3.24          0.196

```

```

|| 3 Right          3.53          2.78          0.191
|| # i 6 more variables: 'se_Low Semantic' <dbl>, 'num_High Semantic' <int>,
|| #   'num_Low Semantic' <int>, mean_diff <dbl>, avg_se <dbl>,
|| #   total_num_stim <int>

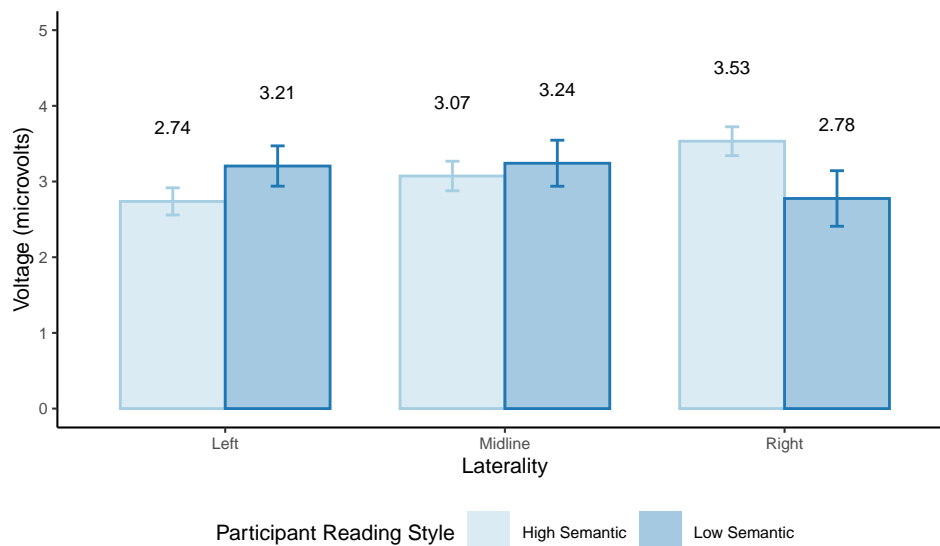
```

5.2.1.5 Plotslang_type_ortho | anteriority First we plot the raw scores then the difference scores

```

# plot raw scores
p3.a <- nw_langsem_lat_2 |> ggplot(aes(x=laterality,
                                     y=mean,
                                     fill = lang_type_semantic,
                                     colour = lang_type_semantic,
                                     ymin = mean - se,
                                     ymax = mean + se)) +
  coord_cartesian(xlim = NULL,ylim = c(0, 5), expand = TRUE,default = FALSE,clip = "on") +
  geom_col(position = "dodge", width = .75, alpha = 0.4) +
  labs(y = "Voltage (microvolts)", x = "Laterality") +
  geom_errorbar(width = .1, position = position_dodge(0.75)) +
  theme_classic(base_size = 8) +
  geom_text(aes(label = round(mean, digits = 2)),colour = "black",size = 2.5, vjust = -5,
            position = position_dodge(.75))+
  guides(fill=guide_legend(title="Participant Reading Style"),
         colour= "none") +
  theme(legend.position = "bottom")
p3.a + scale_fill_brewer(palette = "Paired")+
  scale_colour_brewer(palette = "Paired")

```



```

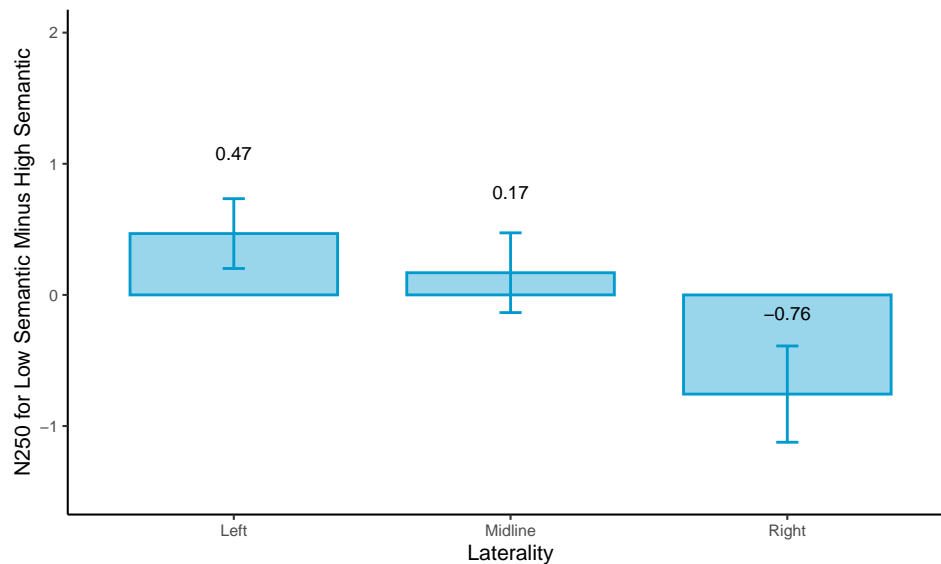
# plot diff scores
p3.b <- difference_scores_2.1 |> ggplot(aes(x = laterality,
                                     y = mean_diff,
                                     ymin = mean_diff - avg_se,
                                     ymax = mean_diff + avg_se)) +
  coord_cartesian(xlim = NULL,ylim = c(-1.5, 2), expand = TRUE,default = FALSE,clip = "on") +
  geom_col(position = "dodge", width = 0.75, alpha = 0.4,
          colour = "deepskyblue3", fill= "deepskyblue3") +

```

```

labs(y = "N250 for Low Semantic Minus High Semantic", x = "Laterality") +
geom_errorbar(width = .08, position = position_dodge(0.75), colour = "deepskyblue3") +
theme_classic(base_size = 8) +
geom_text(aes(label = round(mean_diff, digits = 2)), colour = "black", size = 2.5, vjust = -5.5,
          position = position_dodge(.75)) +
guides(fill=guide_legend(title="Laterality"),
       colour= "none") +
theme(legend.position = "bottom")
p3.b

```



```

# grid.arrange(p1.a, p1.b, nrow = 1)

```

5.2.2 Family Size by Complexity Interaction

```

# Examine the difference between `complexity` at each level of `family_size`

se_smlfam_2.2 <- n250_2_nonwords |> filter(family_size == "small") |>
  ezANOVA(dv = value,
          wid = SubjID,
          within = complexity)
se_smlfam_2.2$ANOVA

```

5.2.2.1 Simple Effects complexity | family_size

	Effect	DFn	DFd	F	p	p<.05	ges
2	complexity	1	39	6.963152	0.01189552	*	0.02683827

```

se_lrgfam_2.2 <- n250_2_nonwords |> filter(family_size == "large") |>
  ezANOVA(dv = value,
          wid = SubjID,
          within = complexity)
se_lrgfam_2.2$ANOVA

```

```
||           Effect DFn Dfd           F           p p<.05           ges
|| 2 complexity    1   39 0.9821524 0.3277772           0.003372297
```

No significant effect of `lang_type_semantic` at any laterality, but the effect is less probable at midline sites ($F(1, 37) = 0.02925513, p = 0.8651233$) than at left ($F(1, 37) = 0.2837876, p = 0.5974145$) or right ($F(1, 37) = 0.6945411, p = 0.4099719$) sites

```
emms <- emmeans(anova_results.2b, ~ complexity | family_size )
pairwise_results <- pairs(emms, by = c("family_size"))
summary(pairwise_results)
```

5.2.2.2 Pairwise Comparisons complexity | family_size

```
|| family_size = small:
|| contrast          estimate      SE df t.ratio p.value
|| complex - simple    1.102 0.407 35   2.708 0.0104
||
|| family_size = large:
|| contrast          estimate      SE df t.ratio p.value
|| complex - simple   -0.331 0.353 35  -0.939 0.3540
||
|| Results are averaged over the levels of: lang_type_semantic, lang_type_ortho, laterality, anteriority
```

```
(nw_cmplx_famsize_2 <- n250_2_nonwords |>
  na.omit())>
group_by(family_size, complexity) |>
summarise(mean = mean(value),
           se = sem(value),
           num_stim = n()))
```

5.2.2.3 Condition Means complexity | family_size

```
|| # A tibble: 4 x 5
|| # Groups:   family_size [2]
||   family_size complexity mean      se num_stim
||   <chr>         <chr>   <dbl> <dbl>   <int>
|| 1 large         complex    2.95 0.189    351
|| 2 large         simple    3.25 0.186    351
|| 3 small         complex    3.62 0.194    351
|| 4 small         simple    2.56 0.243    351
```

```
(difference_scores_2.2 <- nw_cmplx_famsize_2 %>%
  pivot_wider(names_from = complexity, values_from = c(mean, se, num_stim)) %>%
  mutate(mean_diff = `mean_simple` - `mean_complex`,
         avg_se = mean(`se_simple`, `se_complex`),
         total_num_stim = sum(`num_stim_simple`, `num_stim_complex`)))
```

5.2.2.4 Diff Scores complexity | family_size

```

|| # A tibble: 2 x 10
|| # Groups:   family_size [2]
||   family_size mean_complex mean_simple se_complex se_simple num_stim_complex
||   <chr>          <dbl>      <dbl>    <dbl>    <dbl>          <int>
|| 1 large          2.95        3.25    0.189    0.186           351
|| 2 small          3.62        2.56    0.194    0.243           351
|| # i 4 more variables: num_stim_simple <int>, mean_diff <dbl>, avg_se <dbl>,
|| #   total_num_stim <int>

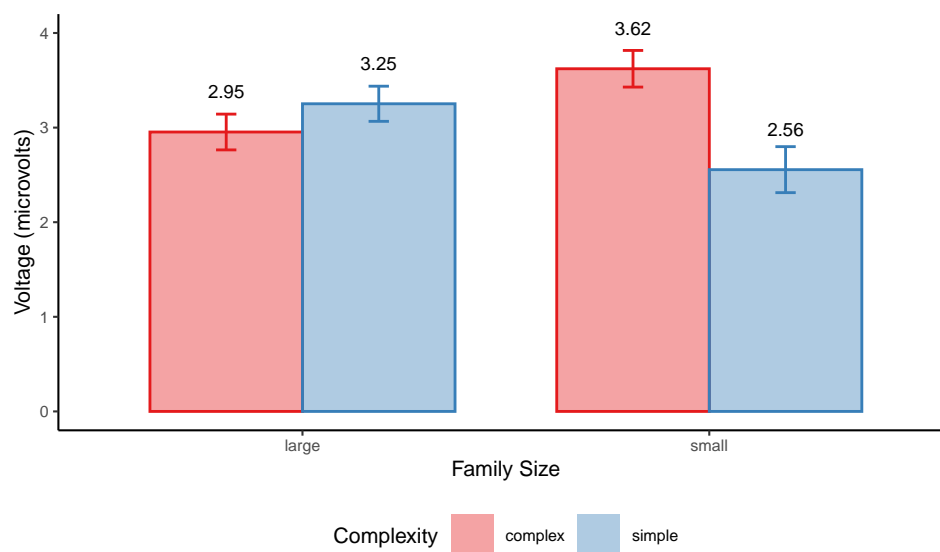
```

5.2.2.5 Plotscomplexity | family_size First we plot the raw scores then the difference scores

```

# plot raw scores
p4.a <- nw_cmplx_famsize_2 |> ggplot(aes(x=family_size,
                                         y=mean,
                                         fill = complexity,
                                         colour = complexity,
                                         ymin = mean - se,
                                         ymax = mean + se)) +
  coord_cartesian(xlim = NULL,ylim = c(0, 4), expand = TRUE,default = FALSE,clip = "on") +
  geom_col(position = "dodge", width = .75, alpha = .4) +
  labs(y = "Voltage (microvolts)", x = "Family Size") +
  geom_errorbar(width = .1, position = position_dodge(0.75)) +
  theme_classic(base_size = 8) +
  geom_text(aes(label = round(mean, digits = 2)),colour = "black",size = 2.5, vjust = -2.5,
            position = position_dodge(.75))+
  guides(fill=guide_legend(title="Complexity"),
         colour= "none") +
  theme(legend.position = "bottom")
p4.a + scale_fill_brewer(palette = "Set1")+
  scale_colour_brewer(palette = "Set1")

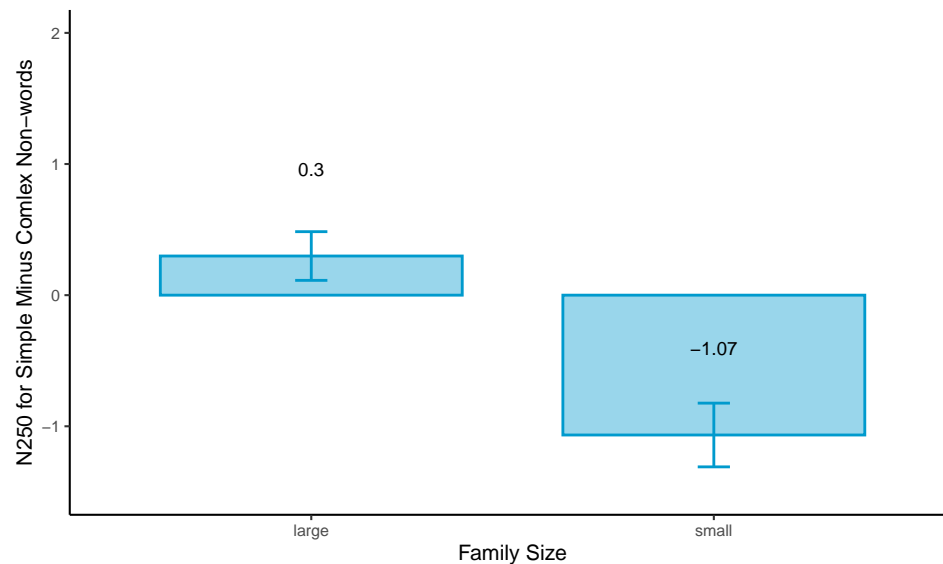
```



```

# plot diff scores
p4.b <- difference_scores_2.2 |> ggplot(aes(x = family_size,
      y = mean_diff,
      ymin = mean_diff - avg_se,
      ymax = mean_diff + avg_se)) +
  coord_cartesian(xlim = NULL,ylim = c(-1.5, 2), expand = TRUE,default = FALSE,clip = "on") +
  geom_col(position = "dodge", width = 0.75, alpha = 0.4,
    colour = "deepskyblue3", fill= "deepskyblue3") +
  labs(y = "N250 for Simple Minus Complex Non-words", x = "Family Size") +
  geom_errorbar(width = .08, position = position_dodge(0.75), colour = "deepskyblue3") +
  theme_classic(base_size = 8) +
  geom_text(aes(label = round(mean_diff, digits = 2)),colour = "black",size = 2.5, vjust = -6,
    position = position_dodge(.75))+
  guides(fill=guide_legend(title="Complexity Effect"),
    colour= "none") +
  theme(legend.position = "bottom")
p4.b

```



```

# grid.arrange(p1.a, p1.b, nrow = 1)

```