

M21 202303 n250 lme

Joanna Morris

2023-05-29

This R script contains the code for analysing the morph 21 erp data for the 200-300 ms time window.

1. First we load the libraries we need

```
library(readr)
library(psych)
library(dplyr)
library(tidyr)
library(ggplot2)
```

Let's load the N250 erp data file and the spelling and vocab data, then we join the files. We will use the `inner_join` rather than the `full_join` function in order to eliminate rows with missing data.

```
sv_202303.na <- read_csv("m21_spell_vocab_raw_z_pca.csv", show_col_types = FALSE)
n250 <- read_csv("S101-177_n250.csv", show_col_types = FALSE)
n250 <- inner_join(sv_202303.na, n250, by = "SubjID") #join subject PCA data
```

Let's save a .csv file with the data from the combined dataset

```
write_csv(n250, "202303_sv_n250_rmna.csv")
```

We will create a subset with only the electrode sites we will be analysing—F3, Fz, F4, C3, Cz, C4, P3, Pz, P4

```
sites = c(3, 2, 25, 7, 20, 21, 12, 11, 16)
n250_9 <- dplyr::filter(n250, chindex %in% sites)
```

7. We then create separate columns, one for each independent variable (anteriority, laterality, morphological family size). To do this we have to use the `mutate` 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.

```
n250_9 <- dplyr::mutate(n250_9,
  anteriority = case_when(grepl("F", chlabel) ~ "Frontal",
    grepl("C", chlabel) ~ "Central",
    grepl("P", chlabel) ~ "Parietal"))

n250_9 <- dplyr::mutate(n250_9,
  laterality = case_when(grepl("3", chlabel) ~ "Left",
```

```

grepl("z", chlabel) ~ "Midline",
grepl("Z", chlabel) ~ "Midline",
grepl("4", chlabel) ~ "Right"))

n250_9 <- dplyr::mutate(n250_9,
  fam_size = case_when(grepl("small", binlabel) ~ "Small",
    grepl("large", binlabel) ~ "Large"))

```

8. We then create a smaller dataset with only the columns we need

```

n250_9b <- dplyr::select(n250_9,
  SubjID,
  lang_type,
  anteriority,
  laterality,
  fam_size,
  value,
  chlabel,
  binlabel)

```

9. We then divide dataset into 3 separate ones—for “words”, “simple nonwords” and “complex nonwords”

```

n250_words <- dplyr::filter(n250_9b, grepl("Critical_word", binlabel))
n250_nwsmpl <- dplyr::filter(n250_9b, grepl("simple", binlabel))
n250_nwcplx <- dplyr::filter(n250_9b, grepl("complex", binlabel))

```

#Plot Means

Get condition means

#Define standard error of the mean function

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

```

(cw.cond.means <- n250_words |>
  group_by(fam_size, lang_type) |>
  summarise(mean = mean(value),
    se = sem(value),
    num_stim = n()))

```

```

# A tibble: 4 x 5
# Groups:   fam_size [2]
  fam_size lang_type      mean    se num_stim
  <chr>    <chr>      <dbl> <dbl>   <int>
1 Large   Orthographic  0.0725 0.213    270
2 Large   Semantic     -0.577 0.175    234
3 Small   Orthographic  0.298  0.219    270
4 Small   Semantic     -0.338 0.196    234

```

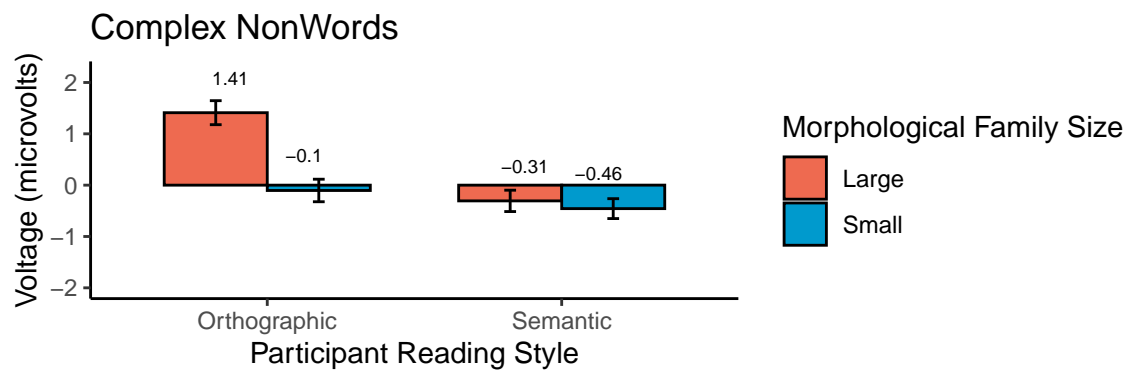
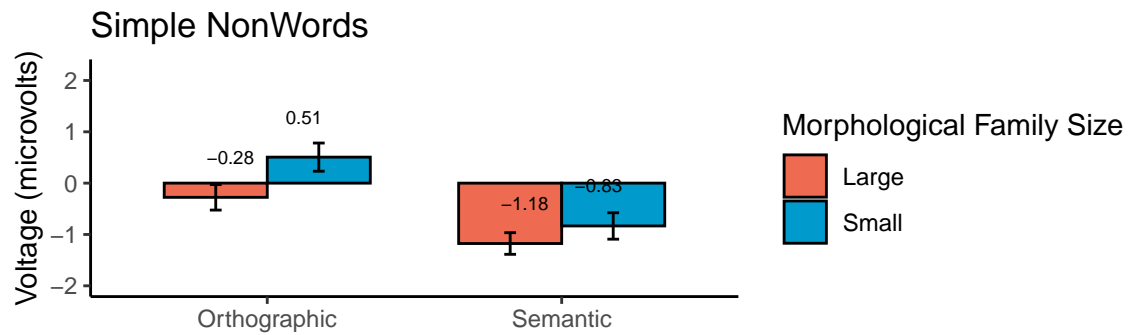
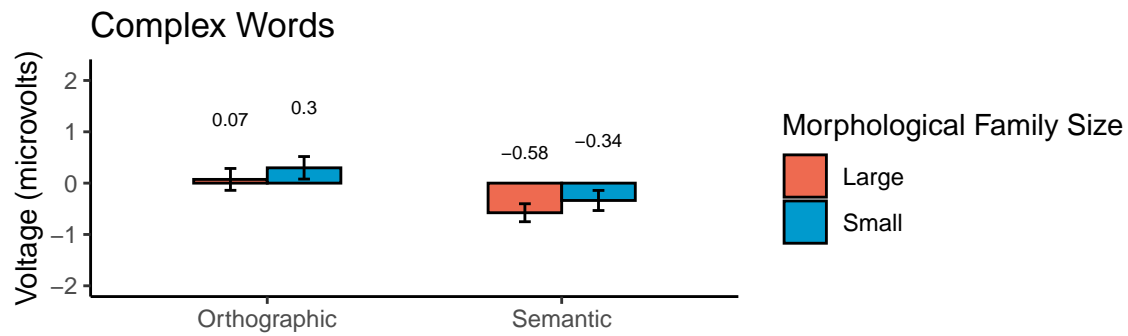
```
(nw_smp.cond.means <- n250_nwsmpl |>
  group_by(fam_size, lang_type) |>
  summarise(mean = mean(value),
            se = sem(value),
            num_stim = n()))
```

```
# A tibble: 4 x 5
# Groups:   fam_size [2]
  fam_size lang_type      mean    se num_stim
  <chr>    <chr>      <dbl> <dbl>   <int>
1 Large   Orthographic -0.277 0.248    270
2 Large   Semantic     -1.18 0.211    234
3 Small   Orthographic  0.506 0.275    270
4 Small   Semantic     -0.834 0.257    234
```

```
(nw_cpx.cond.means <- n250_nwcplx |>
  group_by(fam_size, lang_type) |>
  summarise(mean = mean(value),
            se = sem(value),
            num_stim = n()))
```

```
# A tibble: 4 x 5
# Groups:   fam_size [2]
  fam_size lang_type      mean    se num_stim
  <chr>    <chr>      <dbl> <dbl>   <int>
1 Large   Orthographic  1.41 0.234    270
2 Large   Semantic     -0.306 0.208    234
3 Small   Orthographic -0.104 0.219    270
4 Small   Semantic     -0.457 0.194    234
```

Barplots



LME

```
library(lme4)
```

COMPLEX WORDS

```
cw_null.model = lmer(value ~ 1 + (1|SubjID) ,
                      data= n250_words, REML=FALSE)
summary(cw_null.model)
```

Linear mixed model fit by maximum likelihood ['lmerMod']

Formula: value ~ 1 + (1 | SubjID)
 Data: n250_words

AIC	BIC	logLik	deviance	df.resid
4555.6	4570.4	-2274.8	4549.6	1005

Scaled residuals:

Min	1Q	Median	3Q	Max
-4.2085	-0.6434	-0.0517	0.5876	3.5597

Random effects:

Groups	Name	Variance	Std.Dev.
SubjID	(Intercept)	5.713	2.390
Residual		4.490	2.119

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	-0.2065	0.3292	-0.627

Main effects models with random intercepts

```

cw_main.model = lmer(value ~ lang_type + fam_size + (1 + fam_size|SubjID) ,
                      data= n250_words, REML=FALSE)
summary(cw_main.model)

```

Linear mixed model fit by maximum likelihood ['lmerMod']

Formula: value ~ lang_type + fam_size + (1 + fam_size | SubjID)
 Data: n250_words

AIC	BIC	logLik	deviance	df.resid
4356.7	4391.1	-2171.4	4342.7	1001

Scaled residuals:

Min	1Q	Median	3Q	Max
-4.2594	-0.5430	-0.0442	0.5305	3.1336

Random effects:

Groups	Name	Variance	Std.Dev.	Corr
SubjID	(Intercept)	6.286	2.507	
	fam_sizeSmall	4.843	2.201	-0.32
Residual		3.212	1.792	

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	-0.08767	0.46429	-0.189
lang_typeSemantic	-0.47973	0.65174	-0.736
fam_sizeSmall	0.21459	0.31765	0.676

Correlation of Fixed Effects:

	(Intr)	lng_tS
lng_typSmnt	-0.663	
fam_sizSml1	-0.262	0.000

```
# Interaction effects models with random intercepts
cw_inter.model = lmer(value ~ lang_type * fam_size + (1 + fam_size|SubjID) ,
                      data= n250_words, REML=FALSE)
summary(cw_inter.model)
```

```
Linear mixed model fit by maximum likelihood ['lmerMod']
Formula: value ~ lang_type * fam_size + (1 + fam_size | SubjID)
Data: n250_words
```

AIC	BIC	logLik	deviance	df.resid
4358.7	4398.1	-2171.4	4342.7	1000

Scaled residuals:

Min	1Q	Median	3Q	Max
-4.2590	-0.5432	-0.0442	0.5296	3.1340

Random effects:

Groups	Name	Variance	Std.Dev.	Corr
SubjID	(Intercept)	6.285	2.507	
	fam_sizeSmall	4.842	2.200	-0.32
Residual		3.212	1.792	

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	-0.07928	0.47835	-0.166
lang_typeSemantic	-0.49750	0.69591	-0.715
fam_sizeSmall	0.19271	0.43719	0.441
lang_typeSemantic:fam_sizeSmall	0.04635	0.63626	0.073

Correlation of Fixed Effects:

	(Intr)	lng_tS	fm_szS
lng_typSmnt	-0.687		
fam_sizSml1	-0.350	0.241	
lng_typS:_S	0.241	-0.351	-0.687

```
anova(cw_null.model,cw_main.model)
```

Data: n250_words

Models:

cw_null.model: value ~ 1 + (1 | SubjID)

cw_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)

	npar	AIC	BIC	logLik	deviance	Chisq	Df	Pr(>Chisq)
cw_null.model	3	4555.6	4570.4	-2274.8	4549.6			
cw_main.model	7	4356.7	4391.1	-2171.4	4342.7	206.91	4	< 2.2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
anova(cw_main.model,cw_inter.model)
```

Data: n250_words

Models:

```

cw_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)
cw_inter.model: value ~ lang_type * fam_size + (1 + fam_size | SubjID)
               npar    AIC    BIC logLik deviance Chisq Df Pr(>Chisq)
cw_main.model    7 4356.7 4391.1 -2171.4  4342.7
cw_inter.model   8 4358.7 4398.1 -2171.4  4342.7 0.0053  1    0.9419

```

SIMPLE NONWORDS

```

nw.smpl_null.model = lmer(value ~ 1 + (1|SubjID) ,
                           data= n250_nwsmpl, REML=FALSE)
summary(nw.smpl_null.model)

```

```

Linear mixed model fit by maximum likelihood ['lmerMod']
Formula: value ~ 1 + (1 | SubjID)
Data: n250_nwsmpl

```

AIC	BIC	logLik	deviance	df.resid
5144.8	5159.6	-2569.4	5138.8	1005

Scaled residuals:

Min	1Q	Median	3Q	Max
-4.6813	-0.5297	0.0084	0.5082	5.1466

Random effects:

Groups	Name	Variance	Std.Dev.
SubjID	(Intercept)	8.104	2.847
Residual		8.159	2.856

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	-0.4552	0.3944	-1.154

Main effects models with random intercepts

```

nw.smpl_main.model = lmer(value ~ lang_type + fam_size + (1 + fam_size|SubjID) ,
                           data= n250_nwsmpl, REML=FALSE)
summary(nw.smpl_main.model)

```

```

Linear mixed model fit by maximum likelihood ['lmerMod']
Formula: value ~ lang_type + fam_size + (1 + fam_size | SubjID)
Data: n250_nwsmpl

```

AIC	BIC	logLik	deviance	df.resid
4706.9	4741.3	-2346.4	4692.9	1001

Scaled residuals:

Min	1Q	Median	3Q	Max
-4.0230	-0.5421	-0.0158	0.5006	4.6008

Random effects:

Groups	Name	Variance	Std.Dev.	Corr
--------	------	----------	----------	------

```

SubjID   (Intercept)    9.734   3.120
          fam_sizeSmall 12.482   3.533   -0.44
Residual                4.403   2.098
Number of obs: 1008, groups:  SubjID, 55

```

Fixed effects:

```

                Estimate Std. Error t value
(Intercept)      -0.1740     0.5641  -0.308
lang_typeSemantic -1.0279     0.7701  -1.335
fam_sizeSmall      0.4073     0.4945   0.824

```

Correlation of Fixed Effects:

```

              (Intr) lng_tS
lng_typSmnt  -0.645
fam_sizSml1  -0.345  0.000

```

Interaction effects models with random intercepts

```

nw.smpl_inter.model = lmer(value ~ lang_type * fam_size + (1 + fam_size|SubjID) ,
                           data= n250_nwsmpl, REML=FALSE)
summary(nw.smpl_inter.model)

```

```

Linear mixed model fit by maximum likelihood ['lmerMod']
Formula: value ~ lang_type * fam_size + (1 + fam_size | SubjID)
Data: n250_nwsmpl

```

```

      AIC      BIC  logLik deviance df.resid
4708.8  4748.2  -2346.4  4692.8     1000

```

Scaled residuals:

```

      Min      1Q  Median      3Q      Max
-4.0225 -0.5416 -0.0147  0.5016  4.6003

```

Random effects:

```

Groups   Name          Variance Std.Dev. Corr
SubjID   (Intercept)    9.734   3.120
          fam_sizeSmall 12.477   3.532   -0.44
Residual                4.403   2.098
Number of obs: 1008, groups:  SubjID, 55

```

Fixed effects:

```

                Estimate Std. Error t value
(Intercept)      -0.1970     0.5935  -0.332
lang_typeSemantic -0.9791     0.8634  -1.134
fam_sizeSmall      0.4659     0.6807   0.684
lang_typeSemantic:fam_sizeSmall -0.1239     0.9904  -0.125

```

Correlation of Fixed Effects:

```

              (Intr) lng_tS fm_szS
lng_typSmnt  -0.687
fam_sizSml1  -0.452  0.311
lng_typS:_S   0.311 -0.452 -0.687

```



```
anova(nw.smpl_null.model,nw.smpl_main.model)
```

Data: n250_nwsmpl

Models:

nw.smpl_null.model: value ~ 1 + (1 | SubjID)

nw.smpl_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)

	npar	AIC	BIC	logLik	deviance	Chisq	Df	Pr(>Chisq)
nw.smpl_null.model	3	5144.8	5159.6	-2569.4	5138.8			
nw.smpl_main.model	7	4706.9	4741.3	-2346.4	4692.9	445.95	4	< 2.2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
anova(nw.smpl_main.model,nw.smpl_inter.model)
```

Data: n250_nwsmpl

Models:

nw.smpl_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)

nw.smpl_inter.model: value ~ lang_type * fam_size + (1 + fam_size | SubjID)

	npar	AIC	BIC	logLik	deviance	Chisq	Df	Pr(>Chisq)
nw.smpl_main.model	7	4706.9	4741.3	-2346.4	4692.9			
nw.smpl_inter.model	8	4708.8	4748.2	-2346.4	4692.8	0.0157	1	0.9004

COMPLEX NONWORDS

```
nw.cplx_null.model = lmer(value ~ 1 + (1|SubjID) ,  
                           data= n250_nwcplx, REML=FALSE)  
summary(nw.cplx_null.model)
```

Linear mixed model fit by maximum likelihood ['lmerMod']

Formula: value ~ 1 + (1 | SubjID)

Data: n250_nwcplx

	AIC	BIC	logLik	deviance	df.resid
	4850.0	4864.8	-2422.0	4844.0	1005

Scaled residuals:

	Min	1Q	Median	3Q	Max
	-3.7570	-0.6195	-0.0018	0.5511	4.5243

Random effects:

Groups	Name	Variance	Std.Dev.
SubjID	(Intercept)	5.766	2.401
Residual		6.106	2.471

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	0.07054	0.33308	0.212

```
# Main effects models with random intercepts
nw.cplx_main.model = lmer(value ~ lang_type + fam_size + (1 + fam_size|SubjID) ,
                           data= n250_nwcplx, REML=FALSE)
summary(nw.cplx_main.model)
```

```
Linear mixed model fit by maximum likelihood ['lmerMod']
Formula: value ~ lang_type + fam_size + (1 + fam_size | SubjID)
Data: n250_nwcplx
```

AIC	BIC	logLik	deviance	df.resid
4582.8	4617.2	-2284.4	4568.8	1001

Scaled residuals:

Min	1Q	Median	3Q	Max
-3.6097	-0.5449	-0.0332	0.4656	4.5605

Random effects:

Groups	Name	Variance	Std.Dev.	Corr
SubjID	(Intercept)	7.256	2.694	
	fam_sizeSmall	6.884	2.624	-0.47
Residual		4.044	2.011	

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	0.8926	0.4860	1.837
lang_typeSemantic	-0.8805	0.6563	-1.342
fam_sizeSmall	-0.8138	0.3759	-2.165

Correlation of Fixed Effects:

	(Intr)	lng_tS
lng_typSmnt	-0.638	
fam_sizSmll	-0.372	0.000

```
# Interaction effects models with random intercepts
nw.cplx_inter.model = lmer(value ~ lang_type * fam_size + (1 + fam_size|SubjID),
                             data= n250_nwcplx, REML=FALSE)
summary(nw.cplx_inter.model)
```

```
Linear mixed model fit by maximum likelihood ['lmerMod']
Formula: value ~ lang_type * fam_size + (1 + fam_size | SubjID)
Data: n250_nwcplx
```

AIC	BIC	logLik	deviance	df.resid
4581.9	4621.2	-2283.0	4565.9	1000

Scaled residuals:

Min	1Q	Median	3Q	Max
-3.5973	-0.5473	-0.0524	0.4676	4.5481

Random effects:

Groups	Name	Variance	Std.Dev.	Corr
SubjID	(Intercept)	7.161	2.676	

```

          fam_sizeSmall 6.487    2.547    -0.45
Residual              4.044    2.011
Number of obs: 1008, groups:  SubjID, 55

```

Fixed effects:

```

                                Estimate Std. Error t value
(Intercept)                    1.1777    0.5120    2.300
lang_typeSemantic               -1.4841    0.7449   -1.992
fam_sizeSmall                  -1.4071    0.5041   -2.791
lang_typeSemantic:fam_sizeSmall  1.2561    0.7336    1.712

```

Correlation of Fixed Effects:

```

          (Intr) lng_tS fm_szS
lng_typSmnt -0.687
fam_sizSml1 -0.473  0.325
lng_typS:_S  0.325 -0.473 -0.687

```

```
anova(nw.cplx_null.model,nw.cplx_main.model)
```

Data: n250_nwcplx

Models:

nw.cplx_null.model: value ~ 1 + (1 | SubjID)

nw.cplx_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)

```

                                npar    AIC    BIC logLik deviance Chisq Df Pr(>Chisq)
nw.cplx_null.model              3 4850.0 4864.8 -2422.0   4844.0
nw.cplx_main.model              7 4582.8 4617.2 -2284.4   4568.8 275.25  4 < 2.2e-16 ***
---

```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
anova(nw.cplx_main.model,nw.cplx_inter.model)
```

Data: n250_nwcplx

Models:

nw.cplx_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)

nw.cplx_inter.model: value ~ lang_type * fam_size + (1 + fam_size | SubjID)

```

                                npar    AIC    BIC logLik deviance Chisq Df Pr(>Chisq)
nw.cplx_main.model              7 4582.8 4617.2 -2284.4   4568.8
nw.cplx_inter.model            8 4581.9 4621.2 -2282.9   4565.9 2.8546  1  0.09111 .
---

```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Model Comparisons

```
anova(cw_null.model,cw_main.model)
```

Data: n250_words

Models:

cw_null.model: value ~ 1 + (1 | SubjID)

cw_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)

```

                                npar    AIC    BIC logLik deviance Chisq Df Pr(>Chisq)

```

```

cw_null.model      3 4555.6 4570.4 -2274.8   4549.6
cw_main.model      7 4356.7 4391.1 -2171.4   4342.7 206.91  4 < 2.2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
anova(cw_main.model,cw_inter.model)
```

Data: n250_words

Models:

```

cw_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)
cw_inter.model: value ~ lang_type * fam_size + (1 + fam_size | SubjID)
               npar    AIC    BIC logLik deviance Chisq Df Pr(>Chisq)
cw_main.model      7 4356.7 4391.1 -2171.4   4342.7
cw_inter.model     8 4358.7 4398.1 -2171.4   4342.7 0.0053  1    0.9419

```

```
anova(nw.smpl_null.model,nw.smpl_main.model)
```

Data: n250_nwsmpl

Models:

```

nw.smpl_null.model: value ~ 1 + (1 | SubjID)
nw.smpl_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)
                  npar    AIC    BIC logLik deviance Chisq Df Pr(>Chisq)
nw.smpl_null.model   3 5144.8 5159.6 -2569.4   5138.8
nw.smpl_main.model   7 4706.9 4741.3 -2346.4   4692.9 445.95  4 < 2.2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
anova(nw.smpl_main.model,nw.smpl_inter.model)
```

Data: n250_nwsmpl

Models:

```

nw.smpl_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)
nw.smpl_inter.model: value ~ lang_type * fam_size + (1 + fam_size | SubjID)
                  npar    AIC    BIC logLik deviance Chisq Df Pr(>Chisq)
nw.smpl_main.model   7 4706.9 4741.3 -2346.4   4692.9
nw.smpl_inter.model  8 4708.8 4748.2 -2346.4   4692.8 0.0157  1    0.9004

```

```
anova(nw.cplx_null.model,nw.cplx_main.model)
```

Data: n250_nwcplx

Models:

```

nw.cplx_null.model: value ~ 1 + (1 | SubjID)
nw.cplx_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)
                  npar    AIC    BIC logLik deviance Chisq Df Pr(>Chisq)
nw.cplx_null.model   3 4850.0 4864.8 -2422.0   4844.0
nw.cplx_main.model   7 4582.8 4617.2 -2284.4   4568.8 275.25  4 < 2.2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
anova(nw.cplx_main.model,nw.cplx_inter.model)
```

Data: n250_nwcplx

Models:

nw.cplx_main.model: value ~ lang_type + fam_size + (1 + fam_size | SubjID)

nw.cplx_inter.model: value ~ lang_type * fam_size + (1 + fam_size | SubjID)

	npars	AIC	BIC	logLik	deviance	Chisq	Df	Pr(>Chisq)
nw.cplx_main.model	7	4582.8	4617.2	-2284.4	4568.8			
nw.cplx_inter.model	8	4581.9	4621.2	-2282.9	4565.9	2.8546	1	0.09111

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

COMPLEX WORDS

```
cw_null.model = lmer(value ~ 1 + (1|SubjID) ,  
                      data= n250_words, REML=FALSE)  
summary(cw_null.model)
```

Linear mixed model fit by maximum likelihood ['lmerMod']

Formula: value ~ 1 + (1 | SubjID)

Data: n250_words

AIC	BIC	logLik	deviance	df.resid
4555.6	4570.4	-2274.8	4549.6	1005

Scaled residuals:

Min	1Q	Median	3Q	Max
-4.2085	-0.6434	-0.0517	0.5876	3.5597

Random effects:

Groups	Name	Variance	Std.Dev.
SubjID	(Intercept)	5.713	2.390
Residual		4.490	2.119

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	-0.2065	0.3292	-0.627

Main effects models with random intercepts

```
cw_main.model = lmer(value ~ lang_type + (1 |SubjID) ,  
                      data= n250_words, REML=FALSE)  
summary(cw_main.model)
```

Linear mixed model fit by maximum likelihood ['lmerMod']

Formula: value ~ lang_type + (1 | SubjID)

Data: n250_words

AIC	BIC	logLik	deviance	df.resid
4557.1	4576.8	-2274.6	4549.1	1004

Scaled residuals:

	Min	1Q	Median	3Q	Max
	-4.2137	-0.6430	-0.0487	0.5896	3.5544

Random effects:

Groups	Name	Variance	Std.Dev.
SubjID	(Intercept)	5.656	2.378
Residual		4.490	2.119

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	0.0182	0.4511	0.040
lang_typeSemantic	-0.4755	0.6562	-0.725

Correlation of Fixed Effects:

	(Intr)
lng_typSmnt	-0.687

```
anova(cw_null.model,cw_main.model)
```

Data: n250_words

Models:

cw_null.model: value ~ 1 + (1 | SubjID)

cw_main.model: value ~ lang_type + (1 | SubjID)

	npars	AIC	BIC	logLik	deviance	Chisq	Df	Pr(>Chisq)
cw_null.model	3	4555.6	4570.4	-2274.8	4549.6			
cw_main.model	4	4557.1	4576.8	-2274.6	4549.1	0.5224	1	0.4698

COMPLEX WORDS

```

cw_null.model = lmer(value ~ 1 + (1|SubjID) ,
                      data= n250_words, REML=FALSE)
summary(cw_null.model)
```

Linear mixed model fit by maximum likelihood ['lmerMod']

Formula: value ~ 1 + (1 | SubjID)

Data: n250_words

AIC	BIC	logLik	deviance	df.resid
4555.6	4570.4	-2274.8	4549.6	1005

Scaled residuals:

	Min	1Q	Median	3Q	Max
	-4.2085	-0.6434	-0.0517	0.5876	3.5597

Random effects:

Groups	Name	Variance	Std.Dev.
SubjID	(Intercept)	5.713	2.390
Residual		4.490	2.119

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	-0.2065	0.3292	-0.627

```
# Main effects models with random intercepts
cw_main.model = lmer(value ~ fam_size + (1 | SubjID) ,
                     data= n250_words, REML=FALSE)
summary(cw_main.model)
```

```
Linear mixed model fit by maximum likelihood ['lmerMod']
Formula: value ~ fam_size + (1 | SubjID)
Data: n250_words
```

AIC	BIC	logLik	deviance	df.resid
4554.6	4574.3	-2273.3	4546.6	1004

Scaled residuals:

Min	1Q	Median	3Q	Max
-4.1601	-0.6427	-0.0446	0.6039	3.5108

Random effects:

Groups	Name	Variance	Std.Dev.
SubjID	(Intercept)	5.714	2.390
Residual		4.476	2.116

Number of obs: 1008, groups: SubjID, 55

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	-0.3225	0.3359	-0.96
fam_sizeSmall	0.2319	0.1333	1.74

Correlation of Fixed Effects:

	(Intr)
fam_sizSml1	-0.198

```
anova(cw_null.model,cw_main.model)
```

Data: n250_words

Models:

cw_null.model: value ~ 1 + (1 | SubjID)

cw_main.model: value ~ fam_size + (1 | SubjID)

	npar	AIC	BIC	logLik	deviance	Chisq	Df	Pr(>Chisq)
cw_null.model	3	4555.6	4570.4	-2274.8	4549.6			
cw_main.model	4	4554.6	4574.3	-2273.3	4546.6	3.024	1	0.08204 .

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1