

Data analysis

MD

1/24/23

First import, tidy, ...

```
library(tidyverse)
```

```
-- Attaching packages ----- tidyverse 1.3.2 --
v ggplot2 3.4.0      v purrr   1.0.1
v tibble  3.1.8      v dplyr   1.0.10
v tidyr   1.2.1      v stringr 1.5.0
v readr   2.1.3      v forcats 0.5.2
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()     masks stats::lag()
```

```
filename1 <- "exp_april_2022_RESULTS_2022-05-02-0622.csv"
filename2 <- "exp_april_2022_RESULTS_2022-05-09-0636.csv"

data1 <- read_csv2(filename1)
```

```
i Using "','" as decimal and "'.'" as grouping mark. Use `read_delim()` for more control.
Rows: 2273 Columns: 9-- Column specification -----
Delimiter: ";"
chr (4): materials, is_test_trial, condition, content
dbl (5): participant, item, position, question order, rating1
i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
data2 <- read_csv2(filename2)
```

```
i Using "','" as decimal and "'.'" as grouping mark. Use `read_delim()` for more control.
Rows: 4162 Columns: 9-- Column specification -----
Delimiter: ";"
chr (4): materials, is_test_trial, condition, content
dbl (5): participant, item, position, question order, rating1
i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
summary(data1)
```

materials	participant	is_test_trial	item
Length:2273	Min. : 4.00	Length:2273	Min. : 1.000
Class :character	1st Qu.:15.00	Class :character	1st Qu.: 3.000
Mode :character	Median :27.00	Mode :character	Median : 5.000
	Mean :26.65		Mean : 6.101
	3rd Qu.:39.00		3rd Qu.: 8.000
	Max. :48.00		Max. :18.000

condition	position	question order	rating1
Length:2273	Min. : 1.00	Min. :1	Min. :1.000
Class :character	1st Qu.:16.00	1st Qu.:1	1st Qu.:1.000
Mode :character	Median :32.00	Median :1	Median :4.000
	Mean :32.11	Mean :1	Mean :3.996
	3rd Qu.:48.00	3rd Qu.:1	3rd Qu.:7.000
	Max. :64.00	Max. :1	Max. :7.000

content
Length:2273
Class :character
Mode :character

```
max(data1$participant)
```

```
[1] 48
```

```
summary(data2)
```

```

materials      participant      is_test_trial      item
Length:4162    Min.       : 1.00    Length:4162      Min.       : 1.000
Class :character 1st Qu.:22.00    Class :character 1st Qu.: 3.000
Mode  :character Median :40.00    Mode  :character Median : 5.000
                        Mean  :41.85                        Mean  : 6.107
                        3rd Qu.:62.00                        3rd Qu.: 8.000
                        Max.  :84.00                        Max.  :18.000

condition      position      question order      rating1
Length:4162    Min.       : 1.00    Min.       :1      Min.       :1.000
Class :character 1st Qu.:16.00    1st Qu.:1      1st Qu.:1.000
Mode  :character Median :32.00    Median :1      Median :3.000
                        Mean  :32.49    Mean  :1      Mean  :3.862
                        3rd Qu.:48.00    3rd Qu.:1      3rd Qu.:7.000
                        Max.  :64.00    Max.  :1      Max.  :7.000

content
Length:4162
Class :character
Mode  :character

```

```

data2.tmp <- data2 %>%
  mutate(data2, participant = participant + max(data1$participant))

data2 <- data2.tmp

summary(data1)

```

```

materials      participant      is_test_trial      item
Length:2273    Min.       : 4.00    Length:2273      Min.       : 1.000
Class :character 1st Qu.:15.00    Class :character 1st Qu.: 3.000
Mode  :character Median :27.00    Mode  :character Median : 5.000
                        Mean  :26.65                        Mean  : 6.101
                        3rd Qu.:39.00                        3rd Qu.: 8.000
                        Max.  :48.00                        Max.  :18.000

condition      position      question order      rating1
Length:2273    Min.       : 1.00    Min.       :1      Min.       :1.000

```

```

Class :character  1st Qu.:16.00  1st Qu.:1  1st Qu.:1.000
Mode  :character  Median :32.00  Median :1  Median :4.000
                Mean  :32.11  Mean  :1  Mean  :3.996
                3rd Qu.:48.00  3rd Qu.:1  3rd Qu.:7.000
                Max.   :64.00  Max.   :1  Max.   :7.000

```

```

  content
Length:2273
Class :character
Mode  :character

```

```
summary(data2)
```

```

materials      participant      is_test_trial      item
Length:4162    Min.   : 49.00  Length:4162    Min.   : 1.000
Class :character 1st Qu.: 70.00  Class :character 1st Qu.: 3.000
Mode  :character Median : 88.00  Mode  :character Median : 5.000
                Mean  : 89.85  Mean  : 6.107
                3rd Qu.:110.00  3rd Qu.: 8.000
                Max.   :132.00  Max.   :18.000

```

```

condition      position      question order      rating1
Length:4162    Min.   : 1.00  Min.   :1  Min.   :1.000
Class :character 1st Qu.:16.00  1st Qu.:1  1st Qu.:1.000
Mode  :character Median :32.00  Median :1  Median :3.000
                Mean  :32.49  Mean  :1  Mean  :3.862
                3rd Qu.:48.00  3rd Qu.:1  3rd Qu.:7.000
                Max.   :64.00  Max.   :1  Max.   :7.000

```

```

  content
Length:4162
Class :character
Mode  :character

```

```
# data2$demographic1 <- as.numeric(data2$demographic1)
```

```
data3 <- bind_rows(data1, data2)
```

```

data3.tmp <- data3

data3.tmp$participant <- as.factor(data3.tmp$participant)

levels(data3.tmp$participant) <- paste(1:length(levels(data3.tmp$participant)), sep="")

data3 <- data3.tmp

summary(data3)

```

```

materials      participant  is_test_trial      item
Length:6435    1      : 64  Length:6435    Min.   : 1.000
Class :character 3      : 64  Class :character 1st Qu.: 3.000
Mode  :character 4      : 64  Mode  :character Median : 5.000
                    5      : 64                    Mean  : 6.105
                    6      : 64                    3rd Qu.: 8.000
                    7      : 64                    Max.   :18.000
                    (Other):6051

condition      position    question order    rating1
Length:6435    Min.   : 1.00  Min.   :1      Min.   :1.00
Class :character 1st Qu.:16.00  1st Qu.:1      1st Qu.:1.00
Mode  :character Median :32.00  Median :1      Median :4.00
                    Mean  :32.35  Mean  :1      Mean  :3.91
                    3rd Qu.:48.00  3rd Qu.:1      3rd Qu.:7.00
                    Max.   :64.00  Max.   :1      Max.   :7.00

content
Length:6435
Class :character
Mode  :character

```

```

write_csv2(data3, "merged_results.csv")

```

```

data <- data3

```

- now we add demographics
- TODO: some graphs, etc.

```
data_demographics <- read_csv2("merged_demographics_cleaned.csv")
```

i Using '"','" as decimal and "'.'" as grouping mark. Use `read_delim()` for more control.

Rows: 88 Columns: 14

-- Column specification -----

Delimiter: ";"

chr (4): demographic2, demographic4, demographic5, demographic6

dbl (4): participant, time taken sec, demographic1, demographic3

lgl (4): id, region_Morava, age_under_27, reading_time_over_60_minutes

dtm (2): trial start utc, trial end utc

i Use `spec()` to retrieve the full column specification for this data.

i Specify the column types or set `show_col_types = FALSE` to quiet this message.

```
summary(data_demographics)
```

participant	id	trial start utc	
Min. : 1.00	Mode:logical	Min. :2022-04-28 11:52:48	
1st Qu.:22.75	NA's:88	1st Qu.:2022-04-29 15:05:01	
Median :44.50		Median :2022-05-02 08:47:37	
Mean :44.50		Mean :2022-05-01 12:19:56	
3rd Qu.:66.25		3rd Qu.:2022-05-02 18:32:05	
Max. :88.00		Max. :2022-05-04 07:28:26	
trial end utc	time taken sec	demographic1	
Min. :2022-04-28 12:00:07	Min. : 269	Min. :19.00	
1st Qu.:2022-04-29 15:21:27	1st Qu.: 780	1st Qu.:21.00	
Median :2022-05-02 09:00:52	Median : 1005	Median :23.00	
Mean :2022-05-01 13:00:20	Mean : 2423	Mean :25.59	
3rd Qu.:2022-05-02 18:50:27	3rd Qu.: 1614	3rd Qu.:25.00	
Max. :2022-05-04 07:41:37	Max. :72552	Max. :71.00	
	NA's :1		
demographic2	demographic3	demographic4	demographic5
Length:88	Min. : 0.000	Length:88	Length:88
Class :character	1st Qu.: 1.000	Class :character	Class :character
Mode :character	Median : 1.000	Mode :character	Mode :character
	Mean : 1.425		
	3rd Qu.: 2.000		
	Max. :10.000		

```

demographic6      region_Morava  age_under_27
Length:88         Mode :logical  Mode :logical
Class :character  FALSE:59      FALSE:16
Mode :character   TRUE :29      TRUE :72

```

```

reading_time_over_60_minutes
Mode :logical
FALSE:46
TRUE :42

```

```

data_demographics$participant <- as.factor(data_demographics$participant)

data_demographics2 <- data_demographics %>%
  select(participant, region_Morava, age_under_27, reading_time_over_60_minutes, demographic6)

data.tmp <- data %>%
  left_join(data_demographics2, by = "participant")

summary(data.tmp)

```

```

materials      participant  is_test_trial      item
Length:6435    1      : 64  Length:6435        Min.   : 1.000
Class :character 3      : 64  Class :character    1st Qu.: 3.000
Mode :character  4      : 64  Mode :character     Median : 5.000
                    5      : 64                    Mean   : 6.105
                    6      : 64                    3rd Qu.: 8.000
                    7      : 64                    Max.   :18.000
                    (Other):6051

condition      position    question order    rating1
Length:6435    Min.   : 1.00  Min.   :1        Min.   :1.00
Class :character 1st Qu.:16.00  1st Qu.:1        1st Qu.:1.00
Mode :character  Median :32.00  Median :1        Median :4.00
                    Mean   :32.35  Mean   :1        Mean   :3.91

```

	3rd Qu.:48.00	3rd Qu.:1	3rd Qu.:7.00
	Max. :64.00	Max. :1	Max. :7.00
content	region_Morava	age_under_27	
Length:6435	Mode :logical	Mode :logical	
Class :character	FALSE:3714	FALSE:972	
Mode :character	TRUE :1633	TRUE :4375	
	NA's :1088	NA's :1088	

reading_time_over_60_minutes	demographic1
Mode :logical	Min. :19.00
FALSE:2831	1st Qu.:21.00
TRUE :2516	Median :22.00
NA's :1088	Mean :25.34
	3rd Qu.:25.00
	Max. :71.00
	NA's :1152

```
data <- data.tmp
```

- here we continue with filtering

```
items <- filter(data, grepl("^Accept", materials))
practice <- filter(data, grepl("^Practice", materials))
```

```
fillers_good <- filter(data, grepl("^Filler.*Good", materials))
fillers_bad <- filter(data, grepl("^Filler.*Bad", materials))
```

```
fillers <- filter(data, grepl("^Filler.*Bad", materials) | grepl("^Filler.*Good", materials))

nrow(fillers_good) + nrow (fillers_bad) == nrow (fillers)
```

```
[1] TRUE
```

```
nrow(items) + nrow(practice) + nrow(fillers) == nrow(data)
```

```
[1] TRUE
```

- fillers


```
group_by(fillers, condition) %>%
  summarise(mean(rating1), median = median(rating1))
```

```
# A tibble: 2 x 3
  condition `mean(rating1)` median
  <chr>      <dbl>    <dbl>
1 bad      2.06      1
2 good     5.92      7
```

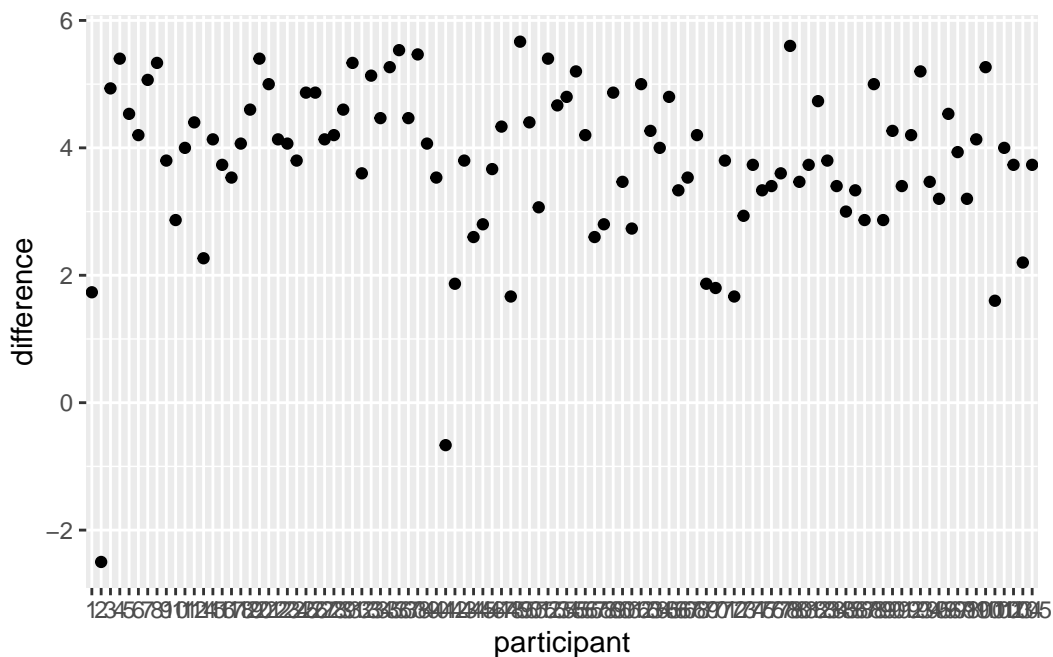
```
by_participant_and_condition <- group_by(fillers, participant, condition)
```

```
diff_fillers <- summarise(by_participant_and_condition, mean = mean(rating1, na.rm=TRUE),
```

`summarise()` has grouped output by 'participant'. You can override using the
`.groups` argument.

```
diff_fillers.tmp <- summarise(diff_fillers, difference = mean[2]-mean[1])
```

```
ggplot(data = diff_fillers.tmp) +
  geom_point(mapping = aes(x = participant, y = difference))
```



- erase all under the difference ≤ 2

```
vector_for_removing <- filter(diff_fillers.tmp, difference <= 3)
vector_for_removing$participant
```

```
[1] 1 2 10 14 41 42 44 45 48 57 58 61 69 70 72 73 85 87 89
[20] 101 104
105 Levels: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 ... 105
```

```
clean_items <- filter(items, !(participant %in% vector_for_removing$participant))
clean_fillers <- filter(fillers, !(participant %in% vector_for_removing$participant))
```

- check

```
((select(items, participant) %>%
  unique %>%
  nrow)-(select(clean_items, participant) %>%
  unique %>%
  nrow)==nrow(vector_for_removing))
```

```
[1] TRUE
```

- all clean

```
items <- clean_items
fillers <- clean_fillers

items <- group_by(items, participant, condition)
items
```

```
# A tibble: 2,406 x 13
```

```
# Groups:   participant, condition [804]
```

	mater~1	parti~2	is_te~3	item	condi~4	posit~5	quest~6	rating1	content	regio~7
	<chr>	<fct>	<chr>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>	<chr>	<lgl>
1	Accept~	3	no	1	ATop	43	1	1	"*Kont~	TRUE
2	Accept~	3	no	2	ZBott	45	1	1	"*Kont~	TRUE
3	Accept~	3	no	3	ZTop	47	1	1	"*Kont~	TRUE
4	Accept~	3	no	4	ABott	61	1	1	"*Kont~	TRUE
5	Accept~	3	no	5	ATop	55	1	7	"*Kont~	TRUE
6	Accept~	3	no	6	ZBott	51	1	1	"*Kont~	TRUE

```

7 Accept~ 3      no      7 ZTop      41      1      1 "*Kont~ TRUE
8 Accept~ 3      no      8 ABott     53      1      7 "*Kont~ TRUE
9 Accept~ 3      no      9 ATop      63      1      3 "*Kont~ TRUE
10 Accept~ 3     no     10 ZBott     59      1      2 "*Kont~ TRUE
# ... with 2,396 more rows, 3 more variables: age_under_27 <lgl>,
#   reading_time_over_60_minutes <lgl>, demographic1 <dbl>, and abbreviated
#   variable names 1: materials, 2: participant, 3: is_test_trial,
#   4: condition, 5: position, 6: `question order`, 7: region_Morava

```

```
summarise(items, mean_cond_accept = mean(rating1))
```

`summarise()` has grouped output by 'participant'. You can override using the `.groups` argument.

```

# A tibble: 804 x 3
# Groups:   participant [82]
  participant condition mean_cond_accept
  <fct>         <chr>         <dbl>
1 3           ABott           4.33
2 3           ATop           3.67
3 3           basA           6.67
4 3           basZ           5.33
5 3           EqA            1
6 3           EqZ            7
7 3           NRA            3
8 3           NRZ            1
9 3           ZBott          1.33
10 3          ZTop           1.33
# ... with 794 more rows

```

```

items <- group_by(items, condition)
summarise(items, mean_cond_accept = mean(rating1), median_cond_accept = median(rating1), s

```

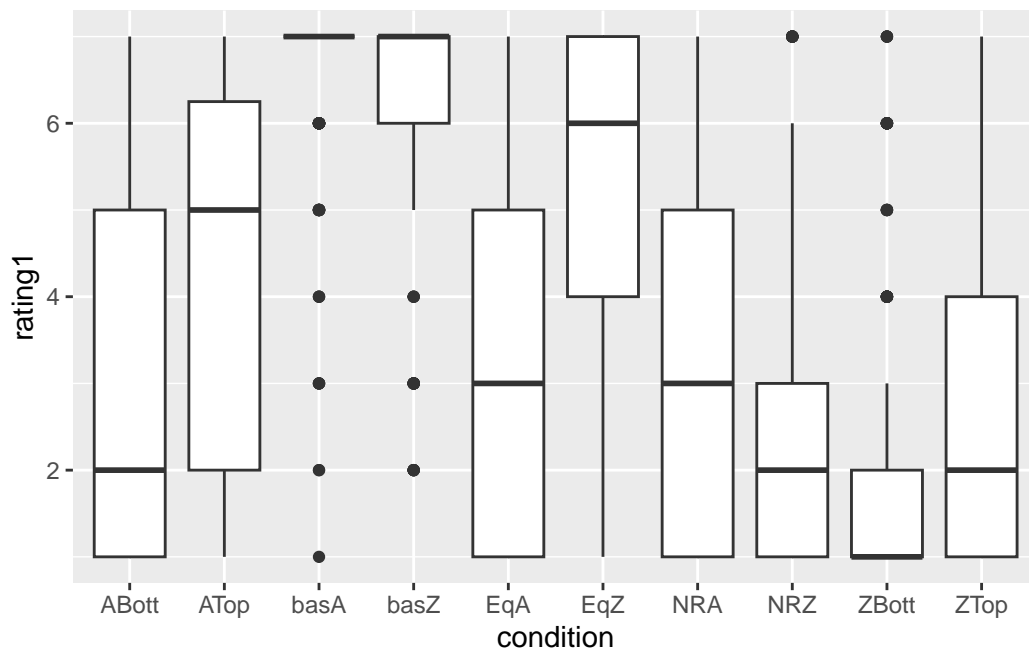
```

# A tibble: 10 x 4
  condition mean_cond_accept median_cond_accept sd_cond_accept
  <chr>         <dbl>         <dbl>         <dbl>
1 ABott           3.12             2           2.22
2 ATop            4.33             5           2.30
3 basA            6.68             7           0.931

```

4	basZ	6.45	7	1.16
5	EqA	3.39	3	2.11
6	EqZ	5.40	6	1.89
7	NRA	3.09	3	1.97
8	NRZ	2.37	2	1.59
9	ZBott	1.95	1	1.50
10	ZTop	2.7	2	1.97

```
ggplot(items, aes(x = condition, y = rating1)) +
  geom_boxplot()
```



- better graph

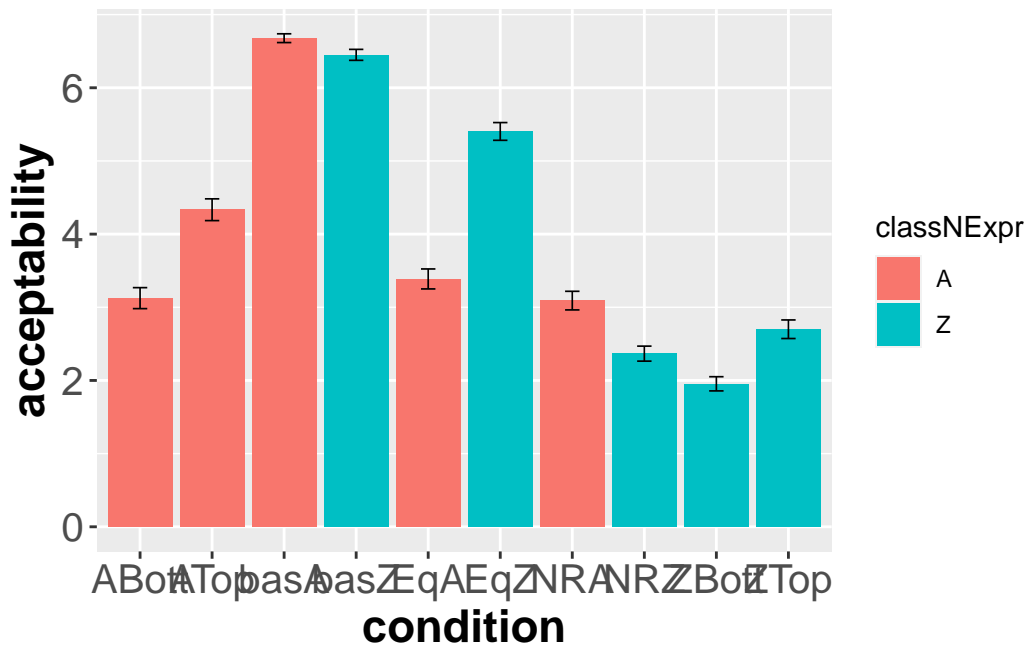
```
items$classNExpr <- "Z"
items$classNExpr[items$condition == "ABott"|items$condition == "ATop"|items$condition == "
basA"|items$condition == "basZ"|items$condition == "EqA"|items$condition == "EqZ"|items$condition == "NRA"|items$condition == "NRZ"|items$condition == "ZBott"|items$condition == "ZTop"] <- "A"

p <- ggplot(items, aes(condition, rating1, fill = classNExpr)) +
  stat_summary(geom = "bar", fun.y = mean, position = "dodge") +
  stat_summary(geom = "errorbar", fun.data = mean_se, size=.3,
    width=.2,
    position=position_dodge(.9))
```

Warning: The `fun.y` argument of `stat_summary()` is deprecated as of ggplot2 3.3.0.
i Please use the `fun` argument instead.

Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
i Please use `linewidth` instead.

```
p + labs(y = "acceptability") +  
  theme(axis.text=element_text(size=15),  
        axis.title=element_text(size=17,face="bold"))
```



```
ggsave("error_bar.png", p)
```

Saving 5.5 x 3.5 in image

And some models

- first main effects

```
items$condition <- as.factor(items$condition)
```

```
levels(items$condition)
```

```
[1] "ABott" "ATop"  "basA"  "basZ"  "EqA"   "EqZ"   "NRA"   "NRZ"   "ZBott"  
[10] "ZTop"
```

```
library("lmerTest")
```

Loading required package: lme4

Loading required package: Matrix

Attaching package: 'Matrix'

The following objects are masked from 'package:tidyr':

expand, pack, unpack

Attaching package: 'lmerTest'

The following object is masked from 'package:lme4':

lmer

The following object is masked from 'package:stats':

step

```
items$condition <- relevel(items$condition, ref="basZ")
```

```
m1 <- lmer(as.numeric(rating1) ~ condition + (1|participant) + (1|item), data=items)  
summary(m1)
```

Linear mixed model fit by REML. t-tests use Satterthwaite's method [
 lmerModLmerTest]
 Formula: as.numeric(rating1) ~ condition + (1 | participant) + (1 | item)
 Data: items

REML criterion at convergence: 9607.9

Scaled residuals:

Min	1Q	Median	3Q	Max
-3.0810	-0.7028	-0.0378	0.5735	3.0915

Random effects:

Groups	Name	Variance	Std.Dev.
participant	(Intercept)	0.27127	0.5208
item	(Intercept)	0.03288	0.1813
Residual		3.00284	1.7329

Number of obs: 2406, groups: participant, 82; item, 18

Fixed effects:

	Estimate	Std. Error	df	t value	Pr(> t)
(Intercept)	6.4574	0.1332	350.7494	48.493	< 2e-16 ***
conditionABott	-3.3131	0.1598	2278.0524	-20.732	< 2e-16 ***
conditionATop	-2.0934	0.1598	2278.3344	-13.100	< 2e-16 ***
conditionbasA	0.2222	0.1581	2308.1992	1.405	0.16
conditionEqA	-3.0587	0.1584	2305.6950	-19.313	< 2e-16 ***
conditionEqZ	-1.0522	0.1579	2309.1970	-6.662	3.37e-11 ***
conditionNRA	-3.3679	0.1581	2300.1125	-21.309	< 2e-16 ***
conditionNRZ	-4.0815	0.1584	2305.6949	-25.772	< 2e-16 ***
conditionZBott	-4.4740	0.1598	2278.0829	-27.996	< 2e-16 ***
conditionZTop	-3.7350	0.1598	2278.2993	-23.372	< 2e-16 ***

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

Correlation of Fixed Effects:

	(Intr)	cn dtAB	cn dtAT	cn dt nA	cn dtEA	cn dtEZ	cn dNRA	cn dNRZ	cn dtZB
conditnABtt	-0.589								
conditinATp	-0.589	0.510							
conditinbsA	-0.597	0.496	0.496						
conditinEqA	-0.595	0.496	0.495	0.501					
conditinEqZ	-0.598	0.497	0.497	0.504	0.501				
conditinNRA	-0.595	0.496	0.496	0.501	0.500	0.502			
conditinNRZ	-0.595	0.496	0.496	0.501	0.501	0.501	0.500		
conditnZBtt	-0.589	0.509	0.510	0.496	0.496	0.497	0.496	0.496	

conditinZTp -0.589 0.510 0.509 0.496 0.495 0.497 0.496 0.496 0.510

- better base model: with interactions
- let's start with the part without probability

`summary(items)`

```

materials      participant  is_test_trial      item
Length:2406    3          : 30  Length:2406      Min.   : 1.000
Class :character 4          : 30  Class :character 1st Qu.: 4.000
Mode  :character 5          : 30  Mode  :character Median : 8.000
                        6          : 30      Mean  : 8.307
                        7          : 30      3rd Qu.:12.000
                        8          : 30      Max.   :18.000
                        (Other):2226

```

```

condition      position    question order    rating1
EqZ   :243      Min.      : 5.00  Min.      :1      Min.      :1.000
basA   :242      1st Qu.:19.00  1st Qu.:1      1st Qu.:2.000
NRA    :241      Median   :34.00  Median :1      Median :4.000
basZ    :240      Mean     :34.44  Mean     :1      Mean     :3.953
ABott   :240      3rd Qu.:49.00  3rd Qu.:1      3rd Qu.:7.000
ATop    :240      Max.      :64.00  Max.      :1      Max.      :7.000
(Other):960

```

```

content      region_Morava  age_under_27
Length:2406   Mode :logical  Mode :logical
Class :character FALSE:1290   FALSE:390
Mode  :character TRUE :696    TRUE :1596
                        NA's :420    NA's :420

```

```

reading_time_over_60_minutes  demographic1  classNExpr
Mode :logical                 Min.      :19.00  Length:2406
FALSE:960                     1st Qu.:21.00  Class :character
TRUE :1026                     Median :23.00  Mode  :character
NA's :420                      Mean   :25.89
                                3rd Qu.:25.00
                                Max.   :71.00
                                NA's   :450

```

`(2406/10)*6`


```
[1] 1443.6
```

```
items_without_probability <- items %>%  
  filter((condition != "ABott" & condition != "ATop" & condition != "ZBott" & condition  
    group_by(participant)
```

```
items_without_probability$condition <- as.factor(items_without_probability$condition)  
items_without_probability$classNExpr <- as.factor(items_without_probability$classNExpr)
```

```
levels(items_without_probability$condition)
```

```
[1] "basZ" "ABott" "ATop" "basA" "EqA" "EqZ" "NRA" "NRZ" "ZBott"  
[10] "ZTop"
```

```
items_without_probability$condition <- relevel(items_without_probability$condition, ref="basZ")
```

```
items_without_probability$condition2 <- "NA"
```

```
items_without_probability$condition2[items_without_probability$condition == "basA" | items_
```

```
items_without_probability$condition2[items_without_probability$condition == "NRA" | items_
```

```
items_without_probability$condition2[items_without_probability$condition == "EqA" | items_
```

```
items_without_probability$condition <- items_without_probability$condition2
```

```
items_without_probability <- select(items_without_probability, -condition2)
```

```
m1 <- lmer(as.numeric(rating1) ~ condition * classNExpr + (1|participant) + (1|item), data=items_without_probability,  
summary(m1)
```

```
Linear mixed model fit by REML. t-tests use Satterthwaite's method [  
lmerModLmerTest]
```

```
Formula: as.numeric(rating1) ~ condition * classNExpr + (1 | participant) +  
  (1 | item)
```

```
Data: items_without_probability
```

REML criterion at convergence: 5487.5

Scaled residuals:

Min	1Q	Median	3Q	Max
-3.8249	-0.6080	0.0416	0.5281	3.4818

Random effects:

Groups	Name	Variance	Std.Dev.
participant	(Intercept)	0.31155	0.5582
item	(Intercept)	0.09736	0.3120
Residual		2.37733	1.5419

Number of obs: 1446, groups: participant, 82; item, 18

Fixed effects:

	Estimate	Std. Error	df	t value	Pr(> t)
(Intercept)	6.6674	0.1383	109.1613	48.226	< 2e-16 ***
conditionEq	-3.2603	0.1409	1347.5412	-23.147	< 2e-16 ***
conditionNR	-3.5707	0.1407	1346.7314	-25.379	< 2e-16 ***
classNExprZ	-0.1964	0.1409	1347.6423	-1.394	0.16355
conditionEq:classNExprZ	2.1833	0.1994	1350.9627	10.949	< 2e-16 ***
conditionNR:classNExprZ	-0.5255	0.1995	1348.0250	-2.634	0.00853 **

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

Correlation of Fixed Effects:

	(Intr)	cndtnE	cndtnNR	clsNEZ	cE:NEZ
conditionEq	-0.507				
conditionNR	-0.508	0.499			
classNExprZ	-0.507	0.498	0.501		
cndtnEq:NEZ	0.359	-0.708	-0.356	-0.709	
cndtnNR:NEZ	0.358	-0.350	-0.707	-0.708	0.500

- now let's do the Bayesian analysis on the data without probability

```
library(rstanarm)
```

Loading required package: Rcpp

This is rstanarm version 2.21.3

- See <https://mc-stan.org/rstanarm/articles/priors> for changes to default priors!

- Default priors may change, so it's safest to specify priors, even if equivalent to the default
- For execution on a local, multicore CPU with excess RAM we recommend calling

```
options(mc.cores = parallel::detectCores())
```

```
library(bayestestR)
```

```
# m1 <- lmer(as.numeric(rating1) ~ condition * classNExpr + (1|participant) + (1|item), data =
```

```
items_without_probability$item <- as.factor(items_without_probability$item)
```

```
items_without_probability$condition <- as.factor(items_without_probability$condition)
```

```
items_without_probability$condition <- relevel(items_without_probability$condition, ref="b")
```

```
items_without_probability$classNExpr <- relevel(items_without_probability$classNExpr, ref="b")
```

```
# m1 <- lmer(as.numeric(rating1) ~ condition * classNExpr + (1|participant) + (1|item), data =
```

```
# full random effects:
```

```
# model_bayes <- stan_glmer(as.numeric(rating1) ~ condition * classNExpr + (1 + participant | item), data =
```

```
# partial random effects
```

```
# model_bayes <- stan_glmer(as.numeric(rating1) ~ condition * classNExpr + (1|item) + (1|participant), data =
```

```
model_bayes <- stan_glmer(as.numeric(rating1) ~ condition * classNExpr + (1|item) + (1|participant), data =
```

SAMPLING FOR MODEL 'continuous' NOW (CHAIN 1).

Chain 1:

Chain 1: Gradient evaluation took 0.001147 seconds

Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 11.47 seconds.

Chain 1: Adjust your expectations accordingly!

Chain 1:

Chain 1:

Chain 1: Iteration: 1 / 2000 [0%] (Warmup)

Chain 1: Iteration: 200 / 2000 [10%] (Warmup)

Chain 1: Iteration: 400 / 2000 [20%] (Warmup)

Chain 1: Iteration: 600 / 2000 [30%] (Warmup)

Chain 1: Iteration: 800 / 2000 [40%] (Warmup)

```

Chain 1: Iteration: 1000 / 2000 [ 50%] (Warmup)
Chain 1: Iteration: 1001 / 2000 [ 50%] (Sampling)
Chain 1: Iteration: 1200 / 2000 [ 60%] (Sampling)
Chain 1: Iteration: 1400 / 2000 [ 70%] (Sampling)
Chain 1: Iteration: 1600 / 2000 [ 80%] (Sampling)
Chain 1: Iteration: 1800 / 2000 [ 90%] (Sampling)
Chain 1: Iteration: 2000 / 2000 [100%] (Sampling)
Chain 1:
Chain 1: Elapsed Time: 20.8581 seconds (Warm-up)
Chain 1:                7.1482 seconds (Sampling)
Chain 1:                28.0063 seconds (Total)
Chain 1:

```

SAMPLING FOR MODEL 'continuous' NOW (CHAIN 2).

```

Chain 2:
Chain 2: Gradient evaluation took 0.000205 seconds
Chain 2: 1000 transitions using 10 leapfrog steps per transition would take 2.05 seconds.
Chain 2: Adjust your expectations accordingly!
Chain 2:
Chain 2:
Chain 2: Iteration:    1 / 2000 [  0%] (Warmup)
Chain 2: Iteration:   200 / 2000 [ 10%] (Warmup)
Chain 2: Iteration:   400 / 2000 [ 20%] (Warmup)
Chain 2: Iteration:   600 / 2000 [ 30%] (Warmup)
Chain 2: Iteration:   800 / 2000 [ 40%] (Warmup)
Chain 2: Iteration:  1000 / 2000 [ 50%] (Warmup)
Chain 2: Iteration:  1001 / 2000 [ 50%] (Sampling)
Chain 2: Iteration:  1200 / 2000 [ 60%] (Sampling)
Chain 2: Iteration:  1400 / 2000 [ 70%] (Sampling)
Chain 2: Iteration:  1600 / 2000 [ 80%] (Sampling)
Chain 2: Iteration:  1800 / 2000 [ 90%] (Sampling)
Chain 2: Iteration:  2000 / 2000 [100%] (Sampling)
Chain 2:
Chain 2: Elapsed Time: 17.8535 seconds (Warm-up)
Chain 2:                6.67252 seconds (Sampling)
Chain 2:                24.526 seconds (Total)
Chain 2:

```

SAMPLING FOR MODEL 'continuous' NOW (CHAIN 3).

```

Chain 3:
Chain 3: Gradient evaluation took 0.00021 seconds
Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 2.1 seconds.
Chain 3: Adjust your expectations accordingly!

```

```

Chain 3:
Chain 3:
Chain 3: Iteration:    1 / 2000 [  0%] (Warmup)
Chain 3: Iteration:   200 / 2000 [ 10%] (Warmup)
Chain 3: Iteration:   400 / 2000 [ 20%] (Warmup)
Chain 3: Iteration:   600 / 2000 [ 30%] (Warmup)
Chain 3: Iteration:   800 / 2000 [ 40%] (Warmup)
Chain 3: Iteration:  1000 / 2000 [ 50%] (Warmup)
Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)
Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
Chain 3:
Chain 3: Elapsed Time: 21.8002 seconds (Warm-up)
Chain 3:                6.77625 seconds (Sampling)
Chain 3:                28.5765 seconds (Total)
Chain 3:

```

SAMPLING FOR MODEL 'continuous' NOW (CHAIN 4).

```

Chain 4:
Chain 4: Gradient evaluation took 0.000263 seconds
Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 2.63 seconds.
Chain 4: Adjust your expectations accordingly!
Chain 4:
Chain 4:
Chain 4: Iteration:    1 / 2000 [  0%] (Warmup)
Chain 4: Iteration:   200 / 2000 [ 10%] (Warmup)
Chain 4: Iteration:   400 / 2000 [ 20%] (Warmup)
Chain 4: Iteration:   600 / 2000 [ 30%] (Warmup)
Chain 4: Iteration:   800 / 2000 [ 40%] (Warmup)
Chain 4: Iteration:  1000 / 2000 [ 50%] (Warmup)
Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
Chain 4: Iteration: 1200 / 2000 [ 60%] (Sampling)
Chain 4: Iteration: 1400 / 2000 [ 70%] (Sampling)
Chain 4: Iteration: 1600 / 2000 [ 80%] (Sampling)
Chain 4: Iteration: 1800 / 2000 [ 90%] (Sampling)
Chain 4: Iteration: 2000 / 2000 [100%] (Sampling)
Chain 4:
Chain 4: Elapsed Time: 17.8366 seconds (Warm-up)
Chain 4:                4.43728 seconds (Sampling)
Chain 4:                22.2739 seconds (Total)

```

Chain 4:

```
print(model_bayes, digits = 3)
```

```
stan_glmer
family:      gaussian [identity]
formula:     as.numeric(rating1) ~ condition * classNExpr + (1 | item) + (1 |
              participant)
observations: 1446
```

```
-----
                        Median MAD_SD
(Intercept)           6.669  0.145
conditionEq           -3.266  0.137
conditionNR           -3.573  0.140
classNExprZ           -0.197  0.142
conditionEq:classNExprZ  2.184  0.199
conditionNR:classNExprZ -0.527  0.202
```

Auxiliary parameter(s):

```
      Median MAD_SD
sigma 1.543  0.029
```

Error terms:

```
Groups      Name      Std.Dev.
participant (Intercept) 0.5666
item        (Intercept) 0.3454
Residual                    1.5429
Num. levels: participant 82, item 18
```

```
-----
* For help interpreting the printed output see ?print.stanreg
* For info on the priors used see ?prior_summary.stanreg
```

```
describe_posterior(model_bayes)
```

Summary of Posterior Distribution

Parameter	Median	95% CI	pd	ROPE	% in ROPE	Rhat
(Intercept)	6.67	[6.38, 6.95]	100%	[-0.10, 0.10]	0%	1.00

conditionEq		-3.27		[-3.53, -3.00]		100%		[-0.10, 0.10]		0%		1.00
conditionNR		-3.57		[-3.86, -3.30]		100%		[-0.10, 0.10]		0%		1.00
classNExprZ		-0.20		[-0.47, 0.08]		92.30%		[-0.10, 0.10]		23.08%		1.00
conditionEq:classNExprZ		2.18		[1.81, 2.58]		100%		[-0.10, 0.10]		0%		1.00
conditionNR:classNExprZ		-0.53		[-0.91, -0.14]		99.67%		[-0.10, 0.10]		0%		1.00

```
bfactors <- bayesfactor(model_bayes)
```

Sampling priors, please wait...

Warning: Bayes factors might not be precise.

For precise Bayes factors, sampling at least 40,000 posterior samples is recommended.

```
print(bfactors, digits=3)
```

Bayes Factor (Savage-Dickey density ratio)

Parameter		BF
(Intercept)		1.14e+42
conditionEq		1.32e+20
conditionNR		4.20e+21
classNExprZ		0.033
conditionEq:classNExprZ		5.95e+08
conditionNR:classNExprZ		0.540

* Evidence Against The Null: 0

```
library(easystats)
```

```
# Attaching packages: easystats 0.6.0
✓ correlation 0.8.3    ✓ datawizard 0.6.5
✓ effectsize 0.8.2    ✓ insight     0.18.8
✓ modelbased 0.8.6    ✓ performance 0.10.2
✓ parameters 0.20.1   ✓ report      0.5.5
✓ see         0.7.4
```

```
exp(bfactors$log_BF)
```

```
[1] 1.137489e+42 1.315427e+20 4.196189e+21 3.298754e-02 5.952184e+08  
[6] 5.404608e-01
```

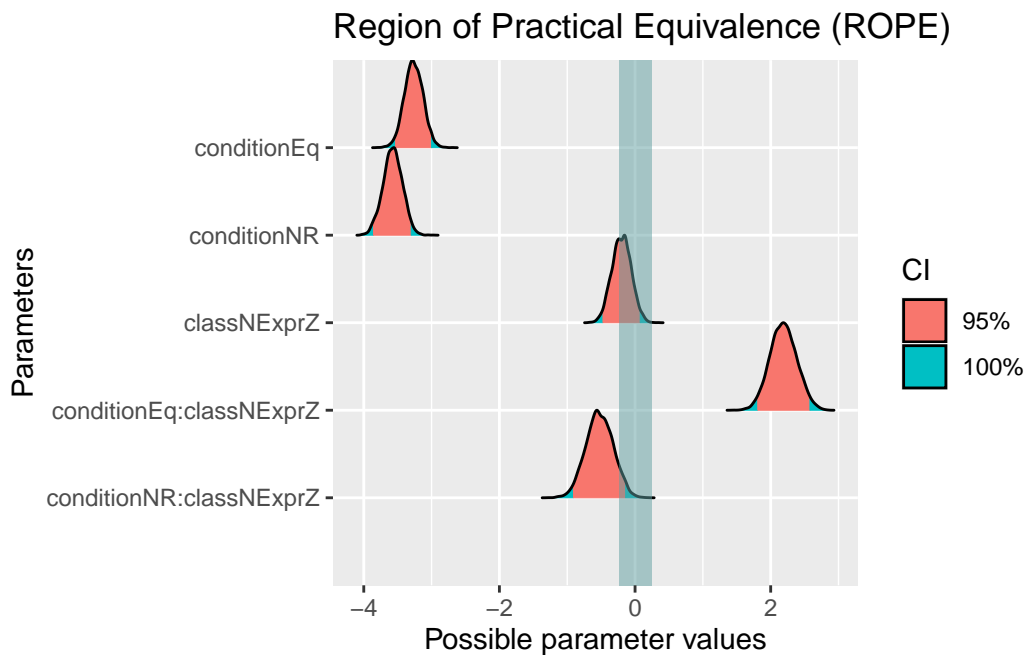
```
interpret_bf(exp(bfactors$log_BF), include_value = TRUE)
```

```
[1] "extreme evidence (BF = 1.14e+42) in favour of"  
[2] "extreme evidence (BF = 1.32e+20) in favour of"  
[3] "extreme evidence (BF = 4.20e+21) in favour of"  
[4] "very strong evidence (BF = 1/30.31) against"  
[5] "extreme evidence (BF = 5.95e+08) in favour of"  
[6] "anecdotal evidence (BF = 1/1.85) against"  
(Rules: jeffreys1961)
```

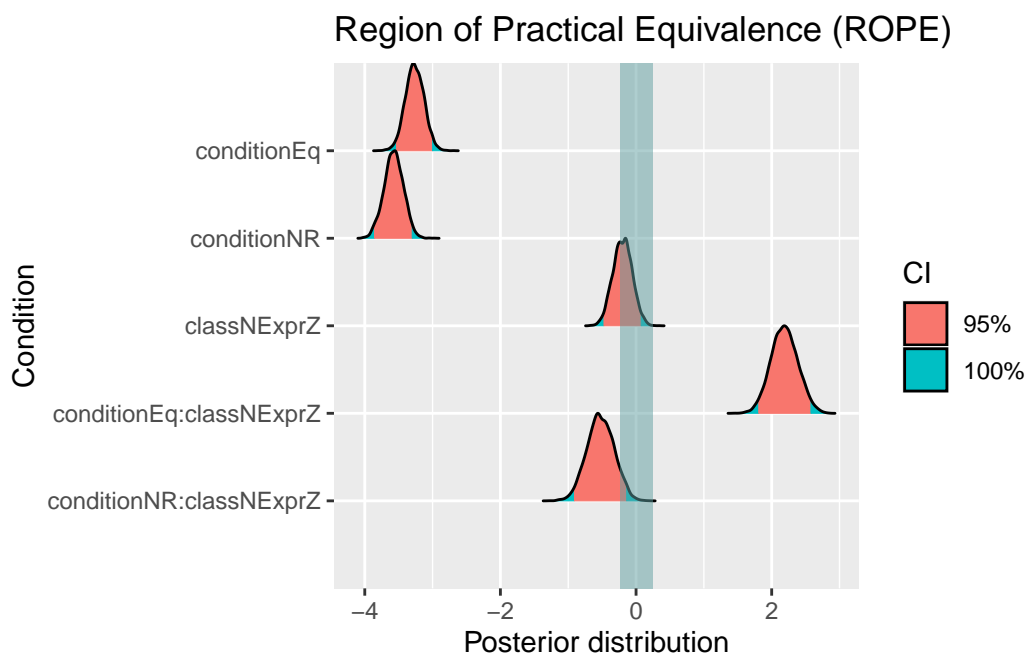
```
result <- rope(model_bayes, ci = c(0.95))
```

Possible multicollinearity between conditionNR:classNExprZ and conditionNR ($r = 0.72$), conditionNR

```
plot(result)
```




```
p <- plot(result)
p <- p + ggplot2::labs(x = "Posterior distribution", y = "Condition")
p
```



```
#ggsave("rope_final_graph.png")

library(bayestestR)

equivalence_test(model_bayes)
```

Possible multicollinearity between conditionNR:classNExprZ and conditionNR ($r = 0.72$), conditionNR:classNExprZ

Test for Practical Equivalence

ROPE: [-0.24 0.24]

Parameter		H0	inside ROPE	95% HDI
(Intercept)		Rejected	0.00 %	[6.38 6.95]
conditionEq		Rejected	0.00 %	[-3.53 -3.00]
conditionNR		Rejected	0.00 %	[-3.86 -3.30]

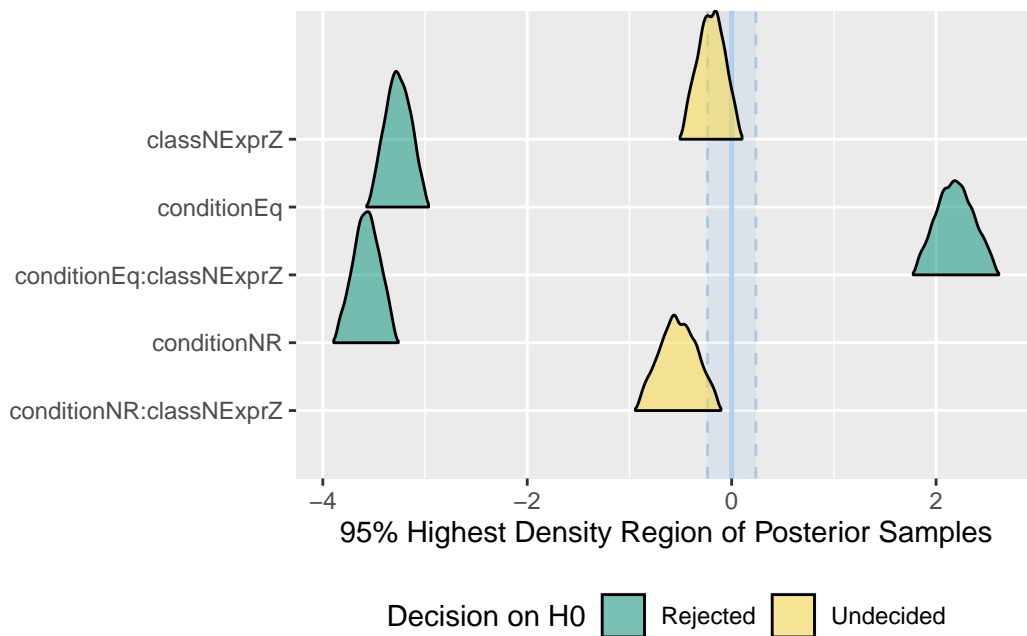
classNExprZ	Undecided	60.76 % [-0.47 0.08]
conditionEq:classNExprZ	Rejected	0.00 % [1.81 2.58]
conditionNR:classNExprZ	Undecided	5.71 % [-0.91 -0.14]

```
plot(equivalence_test(model_bayes))
```

Possible multicollinearity between conditionNR:classNExprZ and conditionNR ($r = 0.72$), conditionNR

Picking joint bandwidth of 0.0247

Warning: Removed 1000 rows containing non-finite values
(`stat_density_ridges()`).



```
# library(emmeans)
# model_bayes.em.s <- emmeans(model_bayes, c("GroupedCondition", "Age"))
# pairs(model_bayes.em.s)
# equivalence_test(pairs(model_bayes.em.s))
```

- here we look at demographics
- first region

- and check the Bayesian version too
- nothing
- then age
- only strange interaction effect: `conditionProb:age_under_27TRUE` 0.84099 0.41727 1494.67832 2.015 0.044 *
- also try z-transformation and linear regression for years
- first z-transformation
- finally reading time:
- again maybe interesting observation: people with higher reading time seem to accept NR more generally (`conditionNR:reading_time_over_60_minutesTRUE` 0.6803 0.3317 1497.0754 2.051 0.0405 *)
- end of demographics
- in the part without probability everything works as it should
- now adding step by step probability
- first bottom (more expected)
- adding clmm package
- now top
- adding clmm package
- now let's try to remove outliers who treat *ani* as neg-word
- it seems it doesn't help
- graphs for all

Correlations

- z-transformation
- but first better descriptive stats
- NR vs. baseline
- now by subject
- graphs
- graph for all conditions

- subjects consistently rating NR high but nothing like that happens with baseline
- no subject ranks baseline bad (consistently or not)
- great variation between speakers but only in some environments
- but it's not general acceptance of *ani*
- adding equatives (graph only for *ani*)
- because people who accept *ani* with equative are different people than those who accept *ani* with NRs
- only 3 subjects appear in both groups
- now correlations
- first z-transformation
- correlations
- first equatives with NR
- next ATop and NR
- checking against baseline

old analysis follows

- not run
- next pairwise comparison

```
““{r}{r eval=FALSE, include=FALSE} library(emmeans)
emmeans(m1, list(pairwise ~ condition), adjust = “tukey”)
““
```

- probability part