

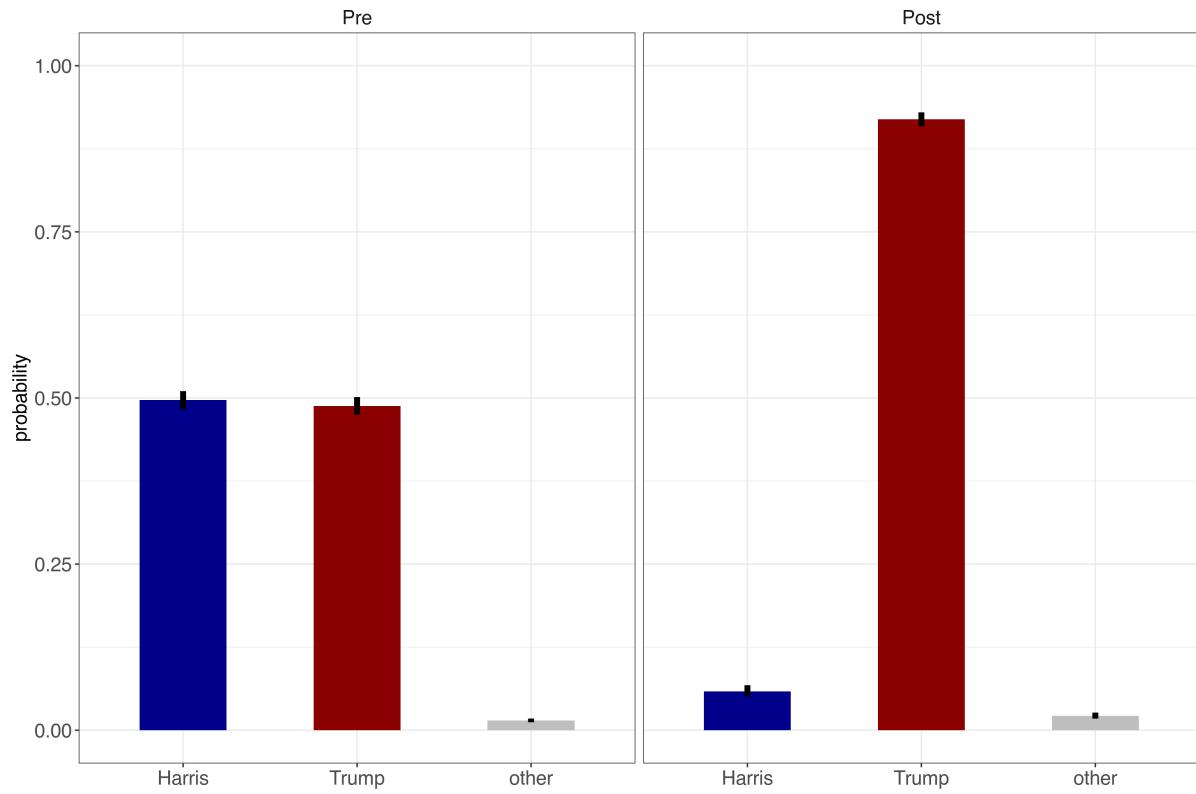
Pre- and post-election EDA

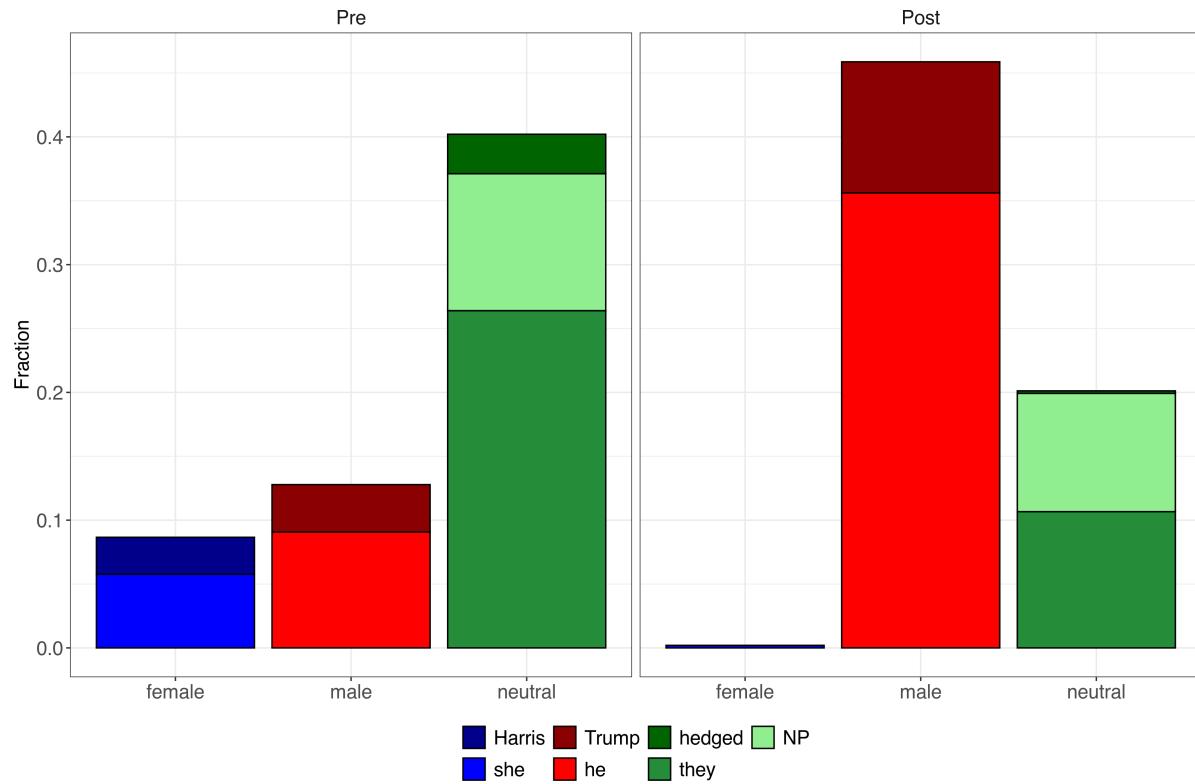
Setup and read in data

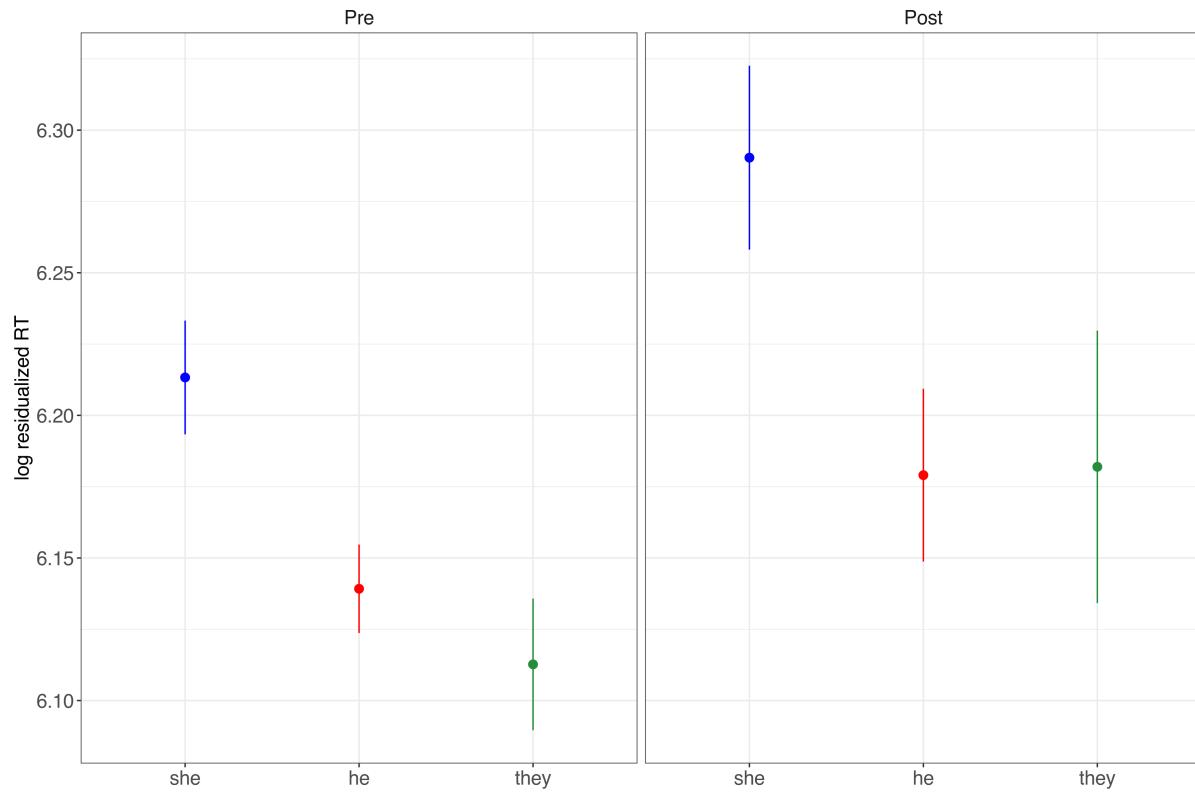
HSP abstract

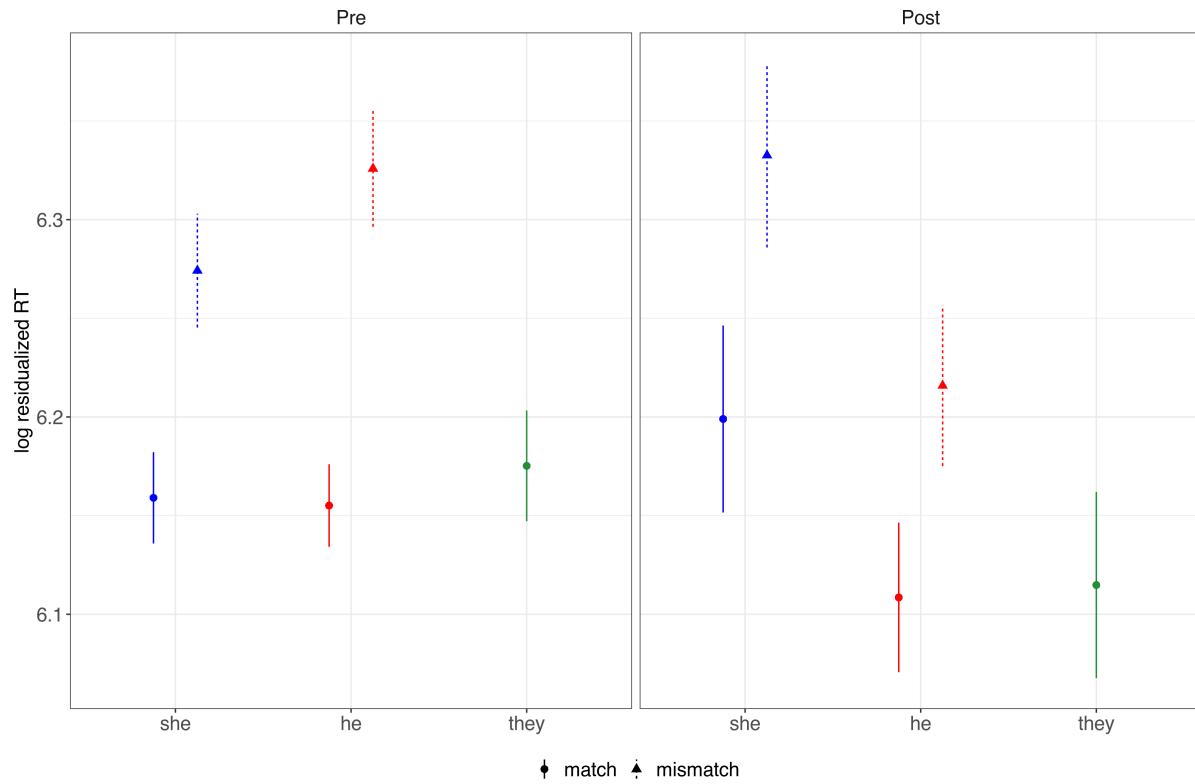
```
# A tibble: 12 x 3
# Groups:   condition [6]
  condition   batch     n
  <chr>        <fct> <int>
1 cloze-event  pre     333
2 cloze-event  post    354
3 event-cloze pre     161
4 event-cloze post    155
5 event-maze   pre     102
6 event-maze   post    99
7 event-spr    pre     163
8 event-spr    post    151
9 maze-event   pre     189
10 maze-event  post    181
11 spr-event   pre     331
12 spr-event   post    333

# A tibble: 2 x 2
  batch     n
  <fct> <int>
1 pre     494
2 post    484
```









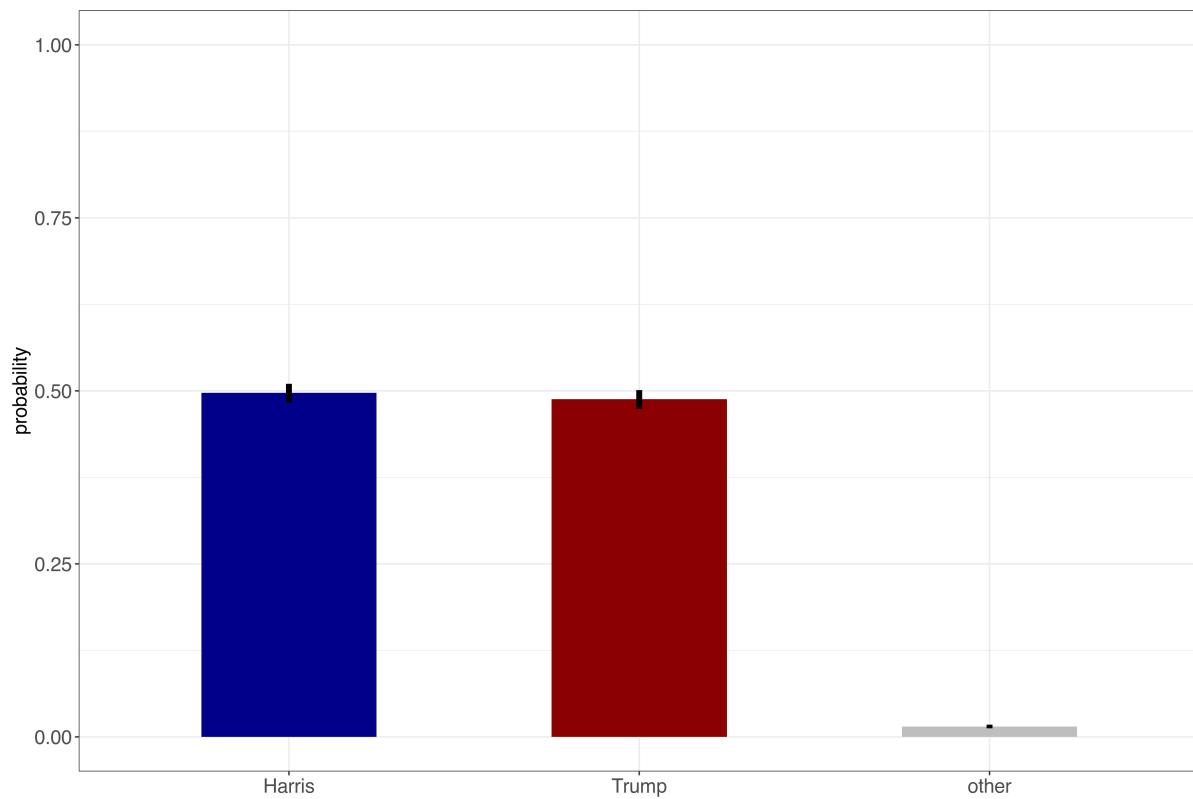
HSP poster

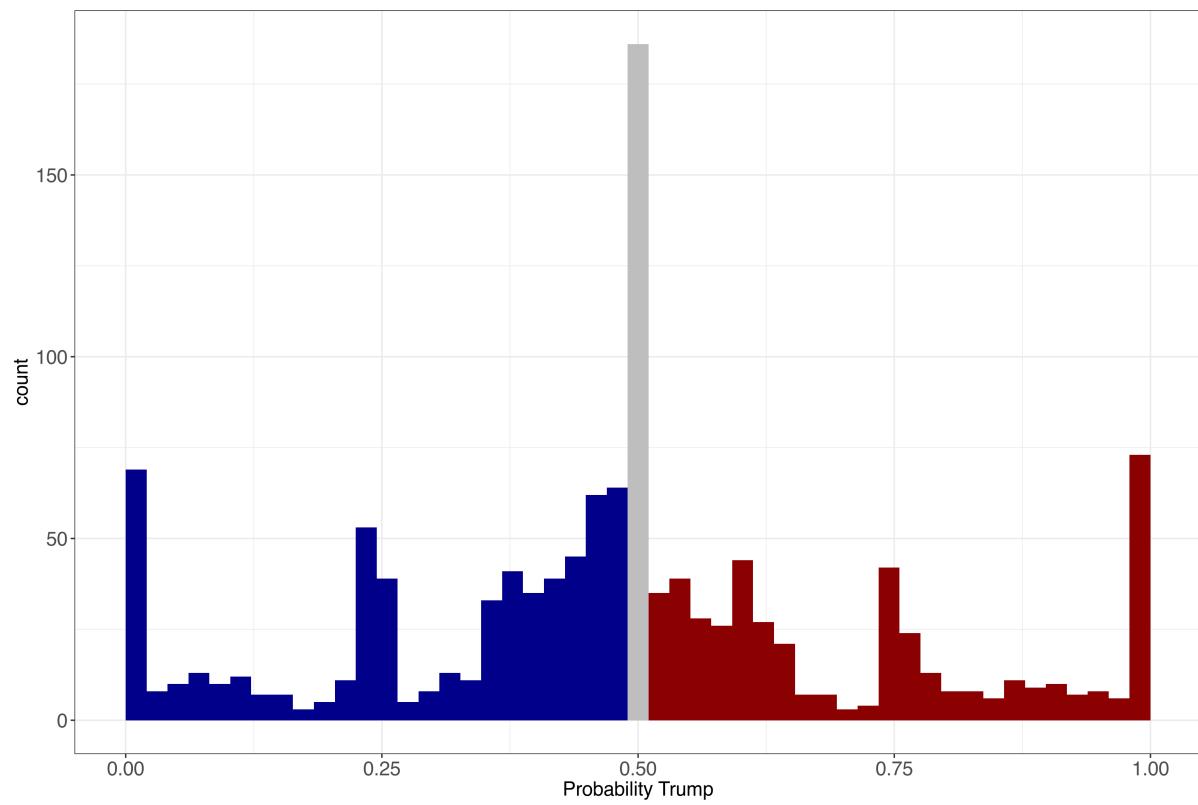
Need only run the `setup` and `read` chunks before below.

```
# A tibble: 12 x 4
  batch task  condition     n
  <fct> <fct> <chr>      <int>
1 pre   Cloze cloze-event  333
2 pre   Cloze event-cloze 161
3 pre   Maze  event-maze   102
4 pre   Maze  maze-event   189
5 pre   SPR   event-spr    163
6 pre   SPR   spr-event    331
7 post  Cloze cloze-event  354
8 post  Cloze event-cloze 155
9 post  Maze  event-maze   99
10 post  Maze  maze-event  181
11 post  SPR   event-spr   151
12 post  SPR   spr-event   333
```

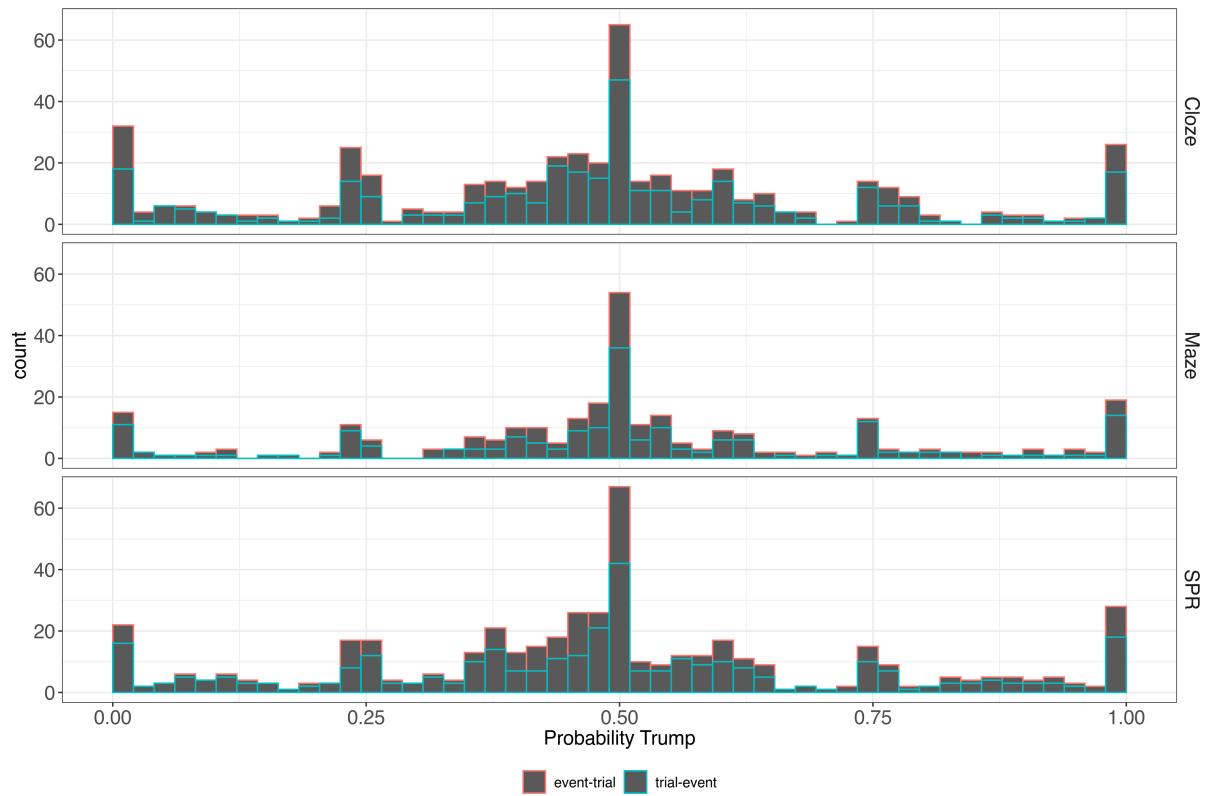
----- Batch: pre-election

Event expectations



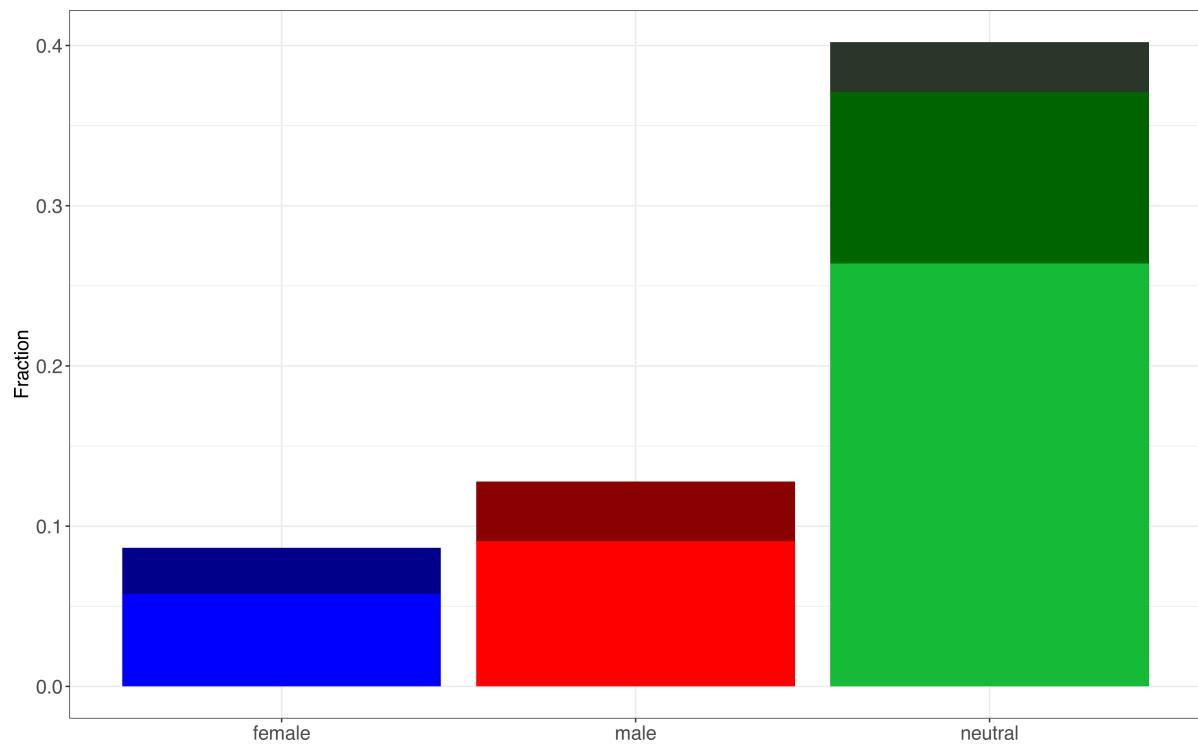


Event expectations histogram

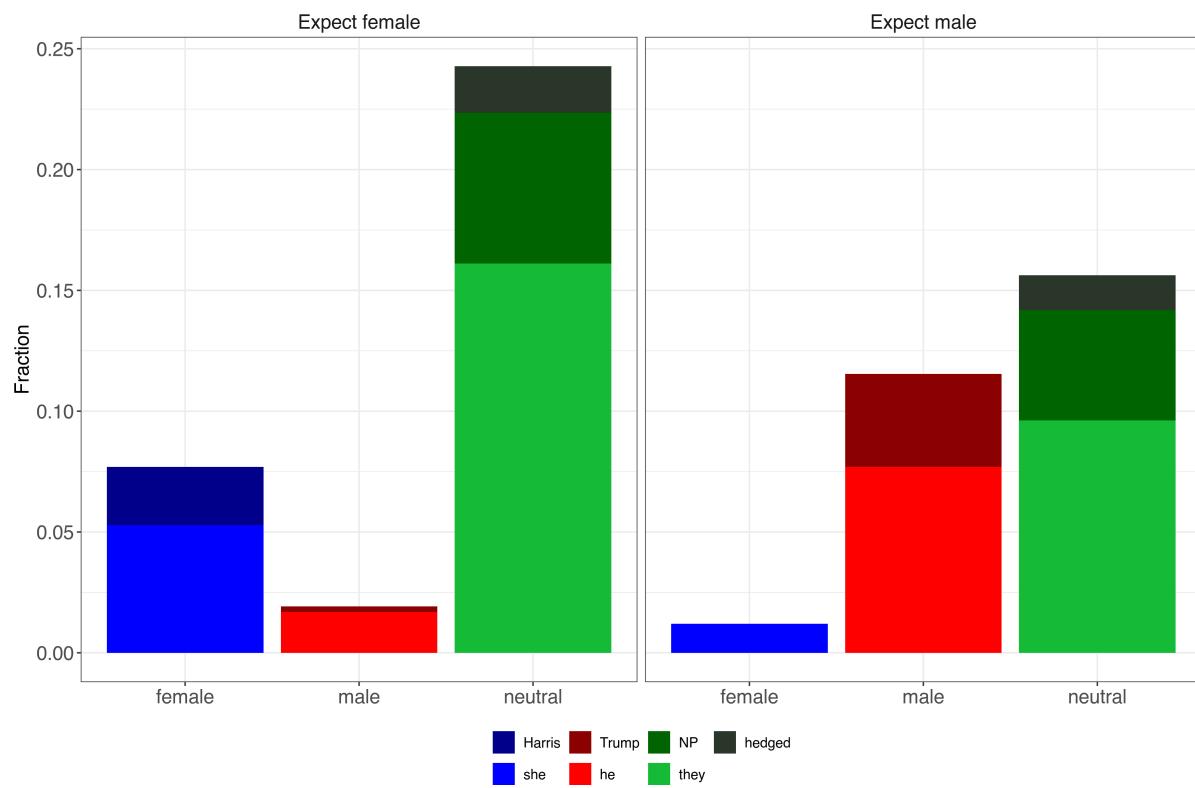


Event expectations histogram by task

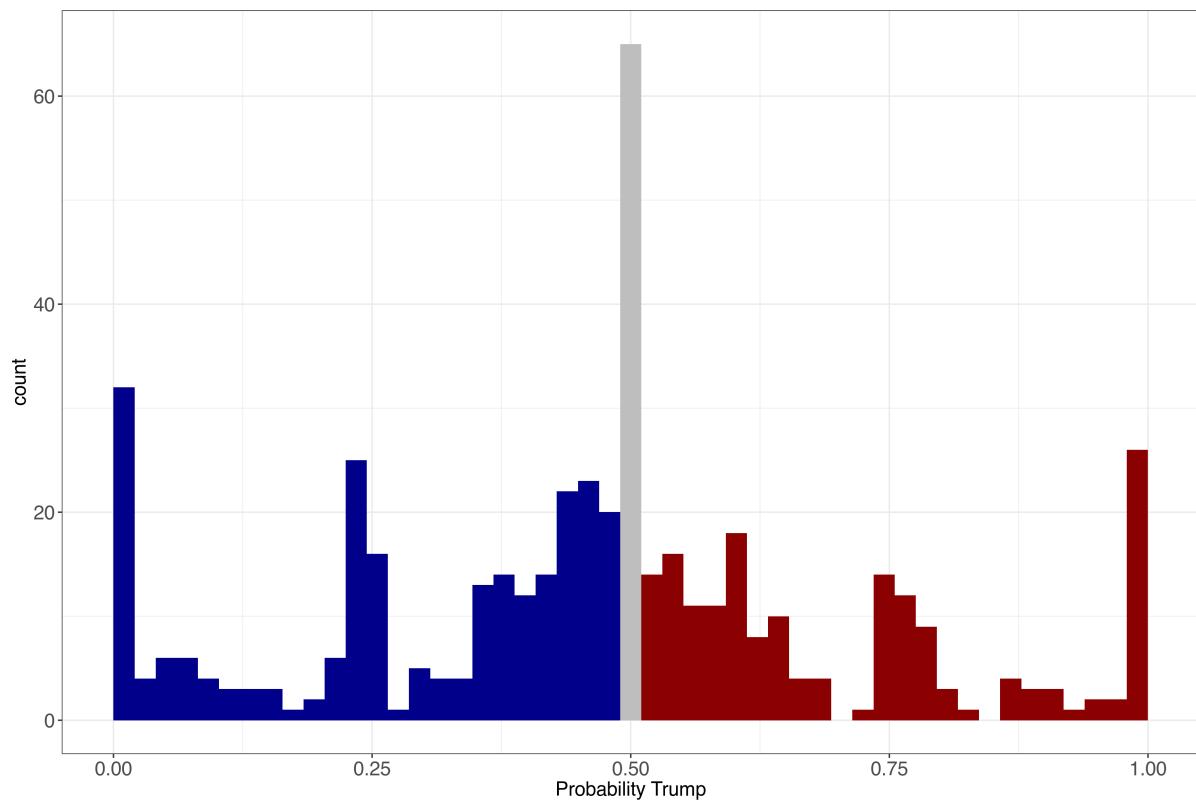
Cloze



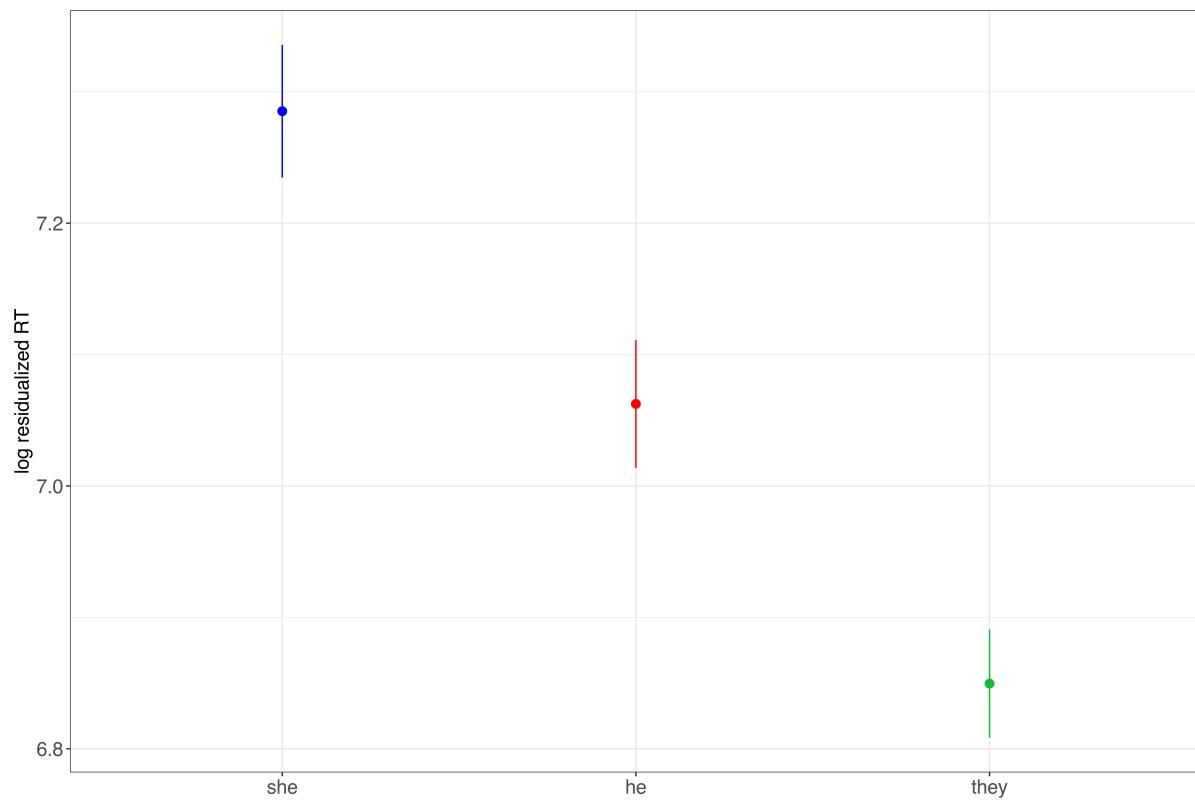
Cloze by expectations



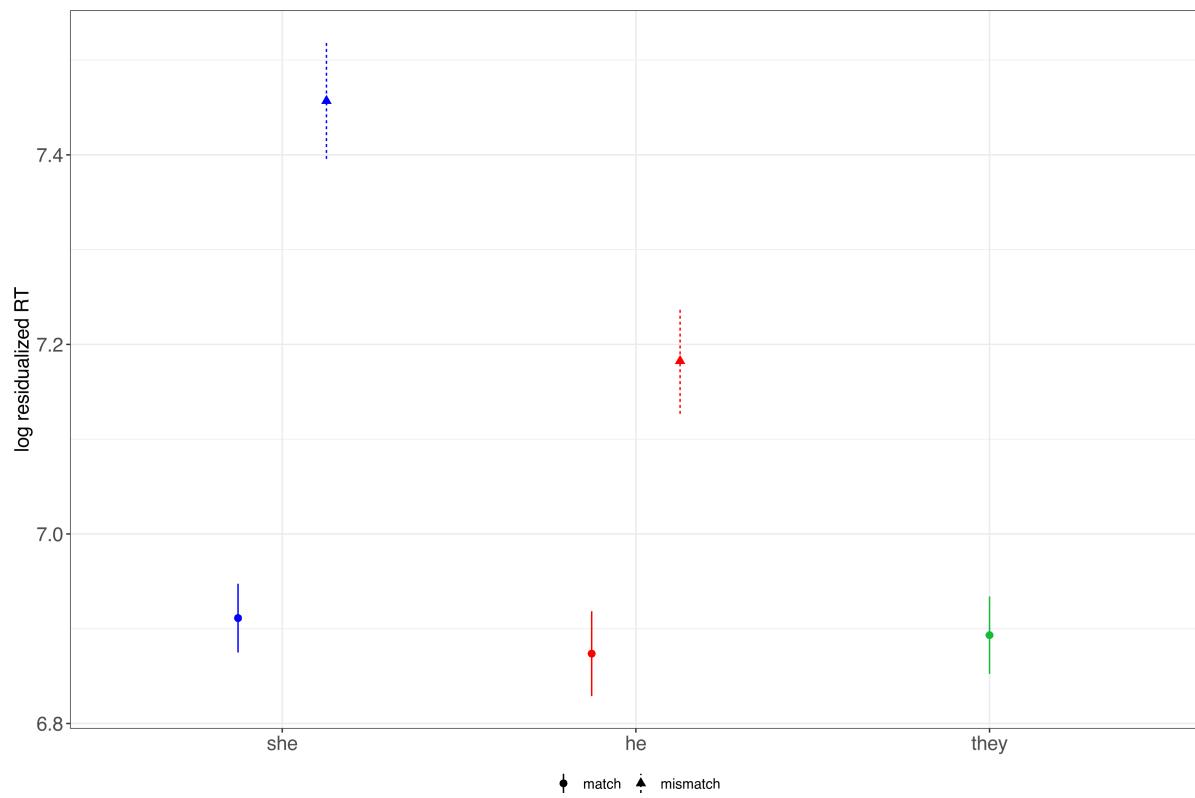
Cloze task expectations histogram



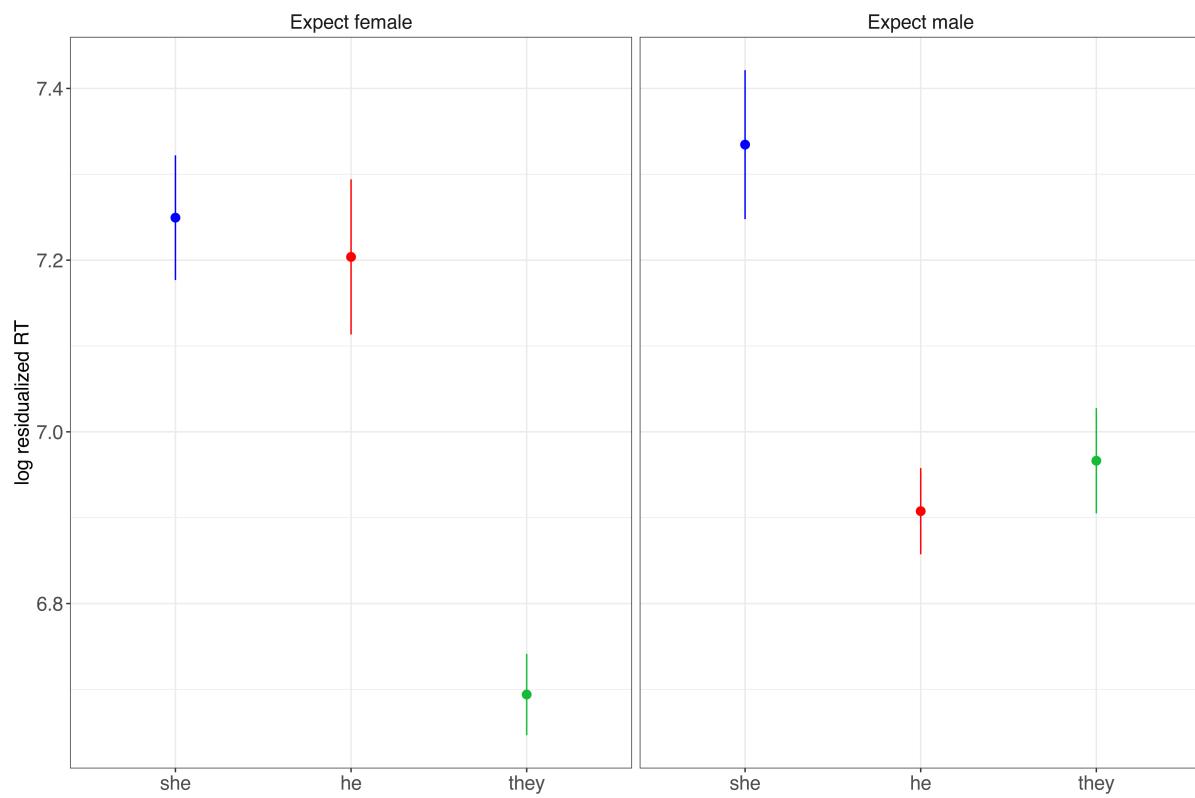
Maze



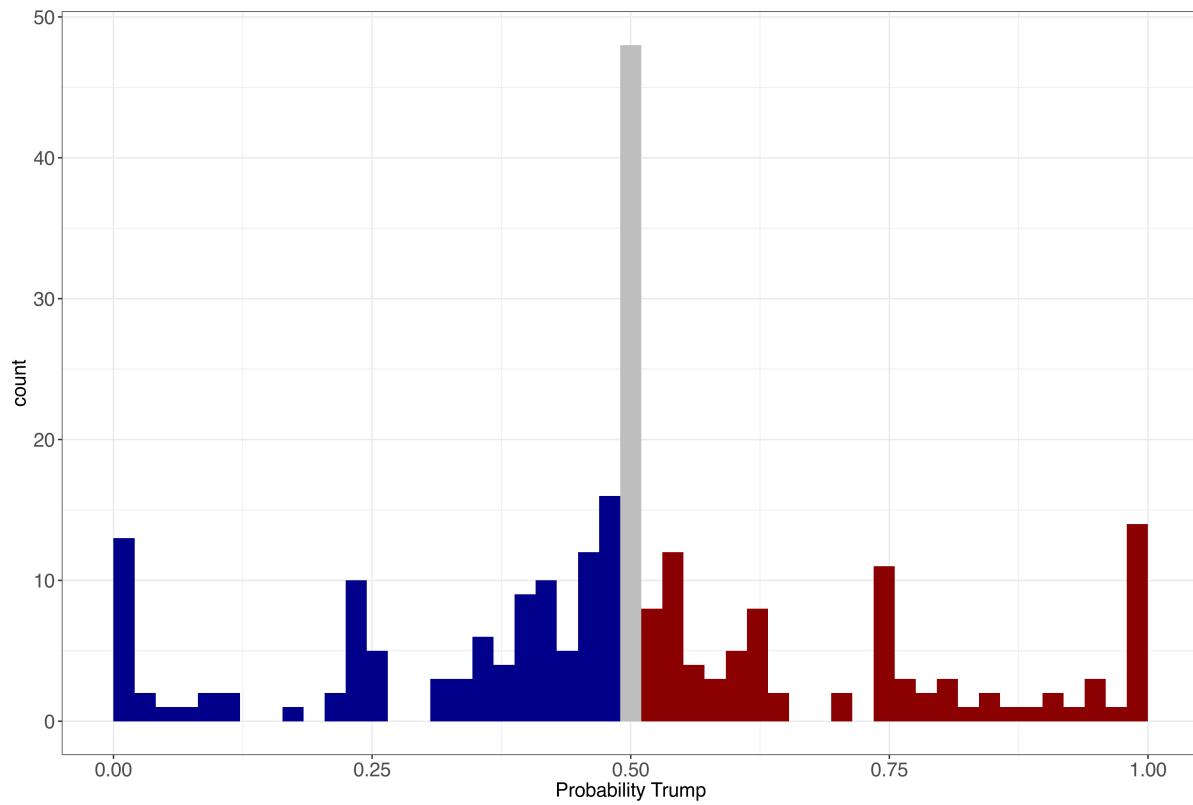
Maze second pronoun



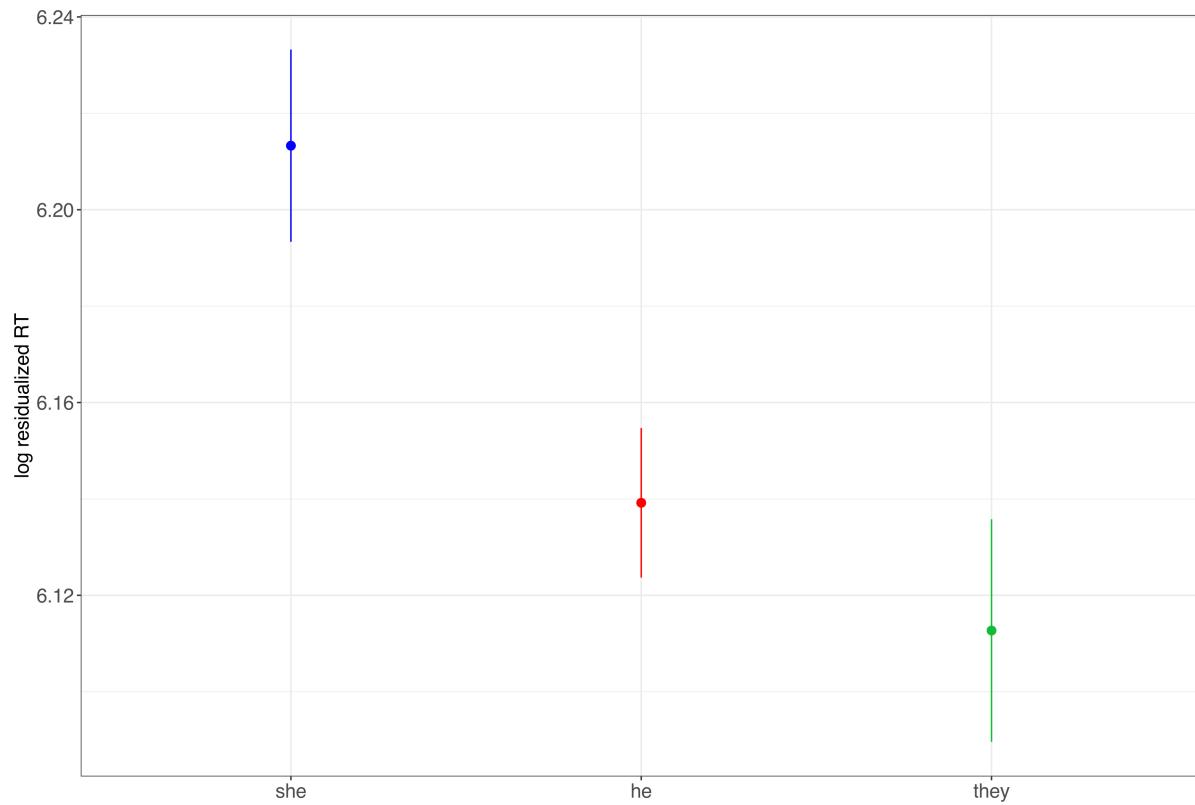
Maze by expectation



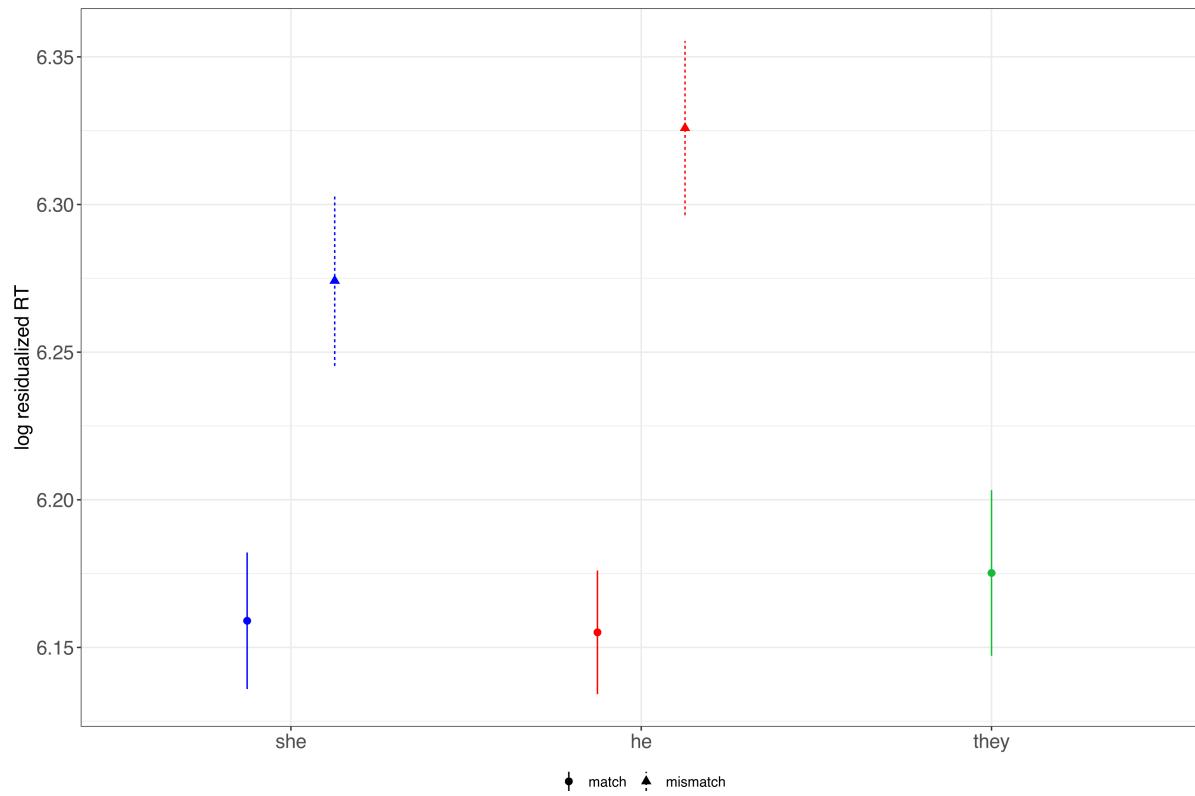
Maze task expectations histogram



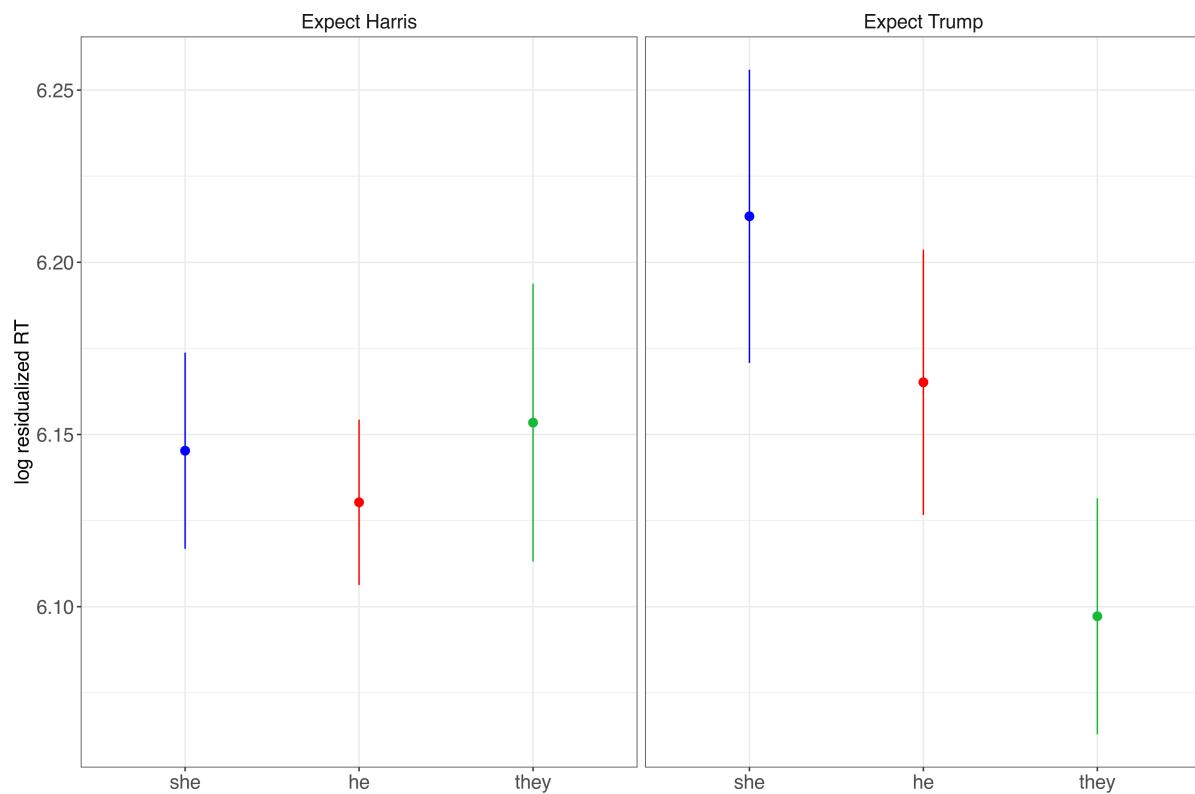
SPR



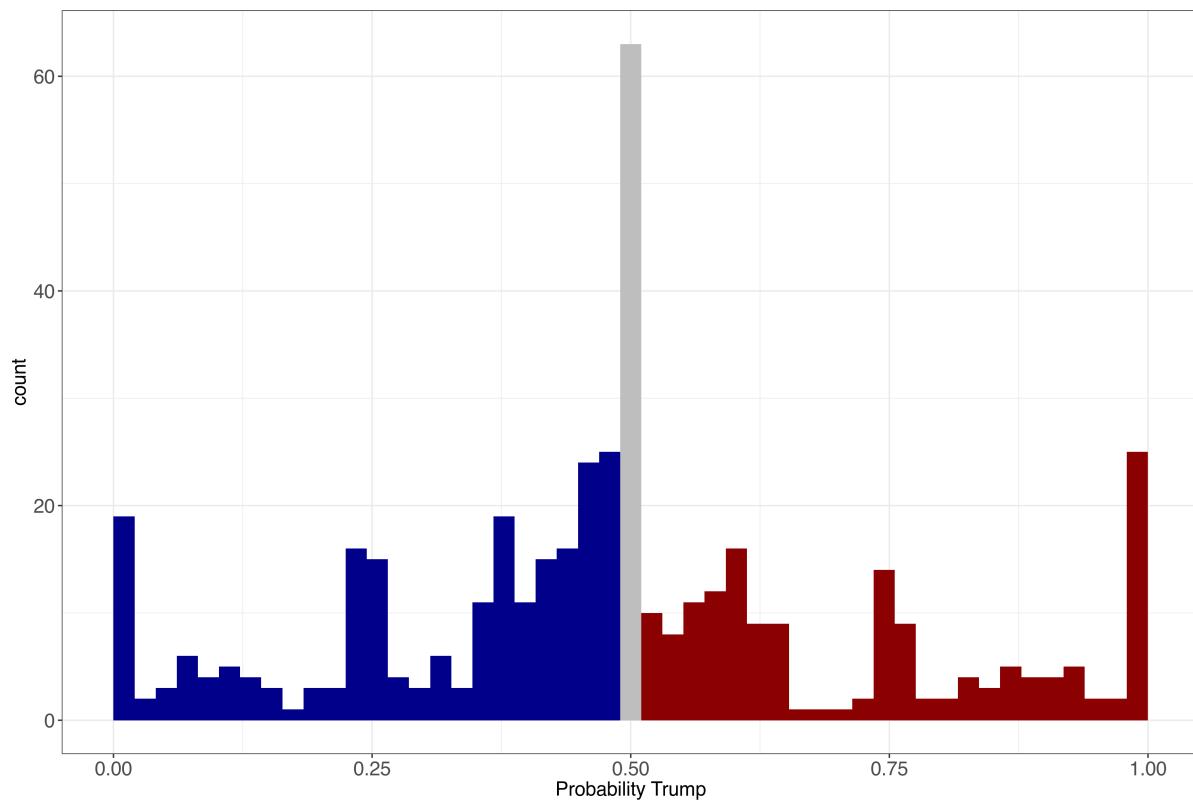
SPR second pronoun



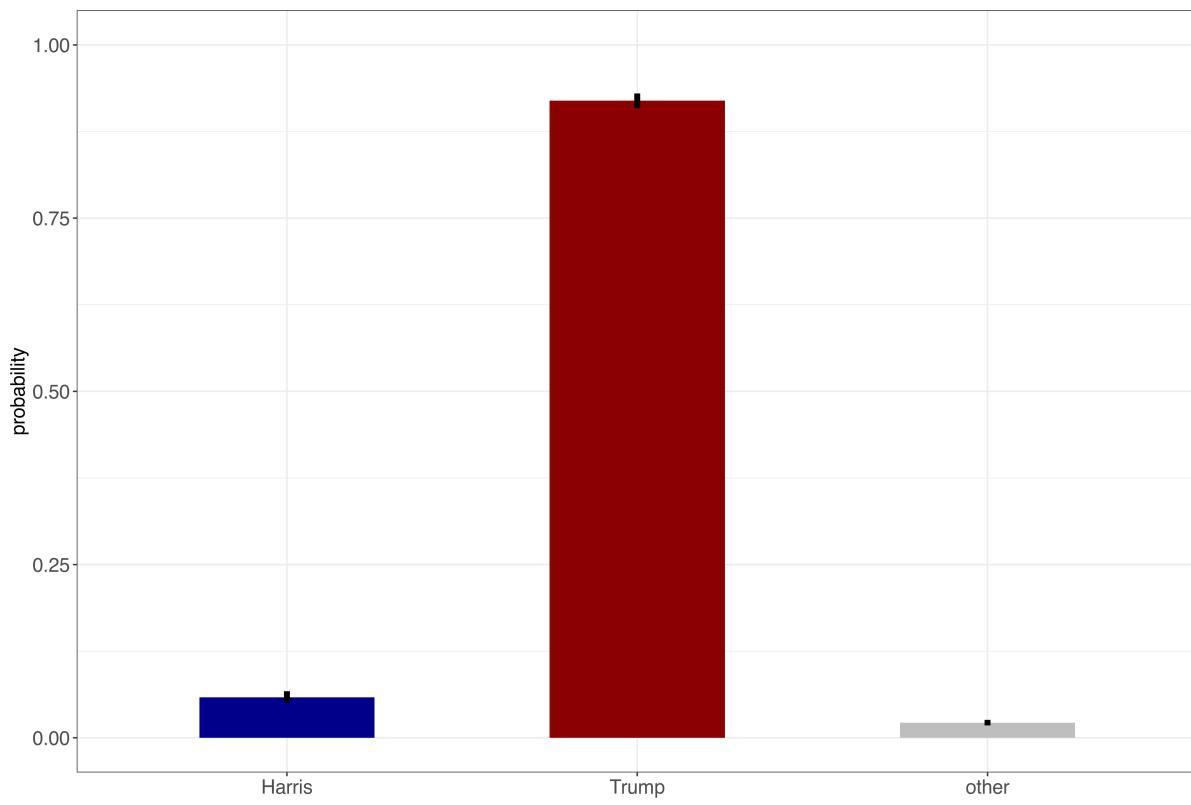
SPR by expectations

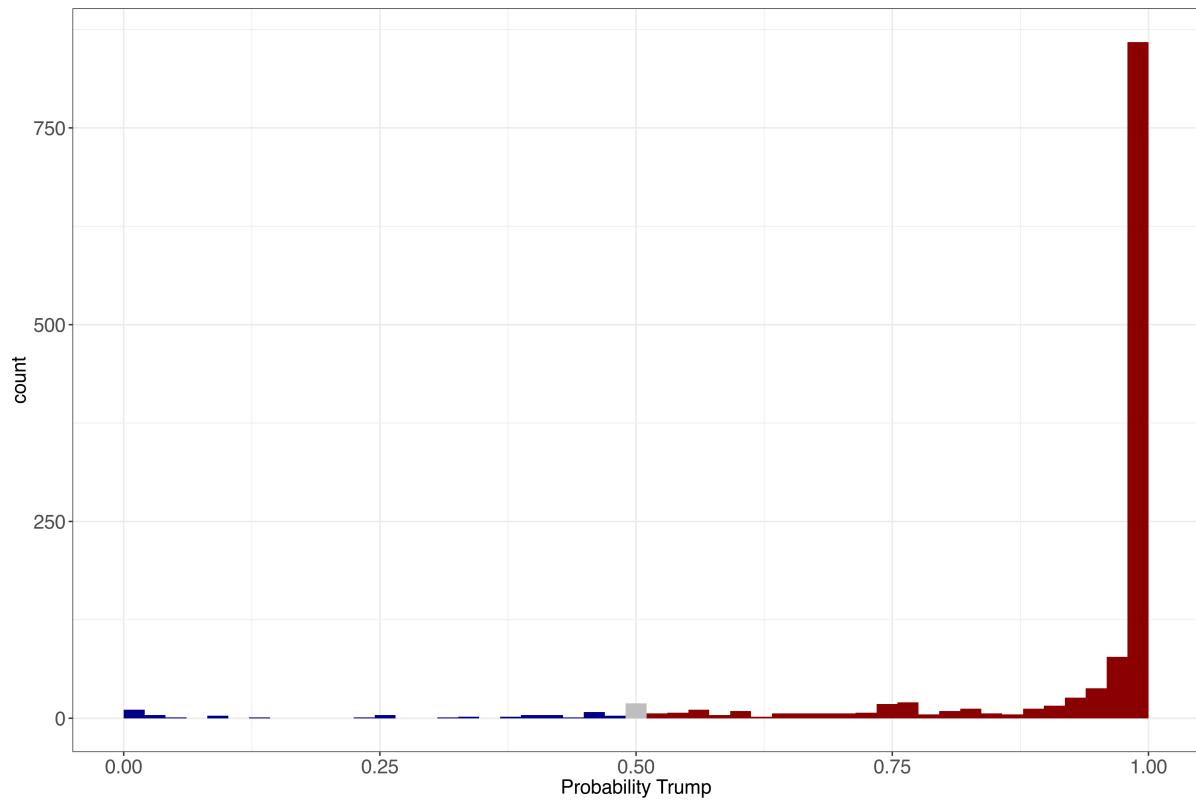


SPR expectations histogram
----- Batch: post-election



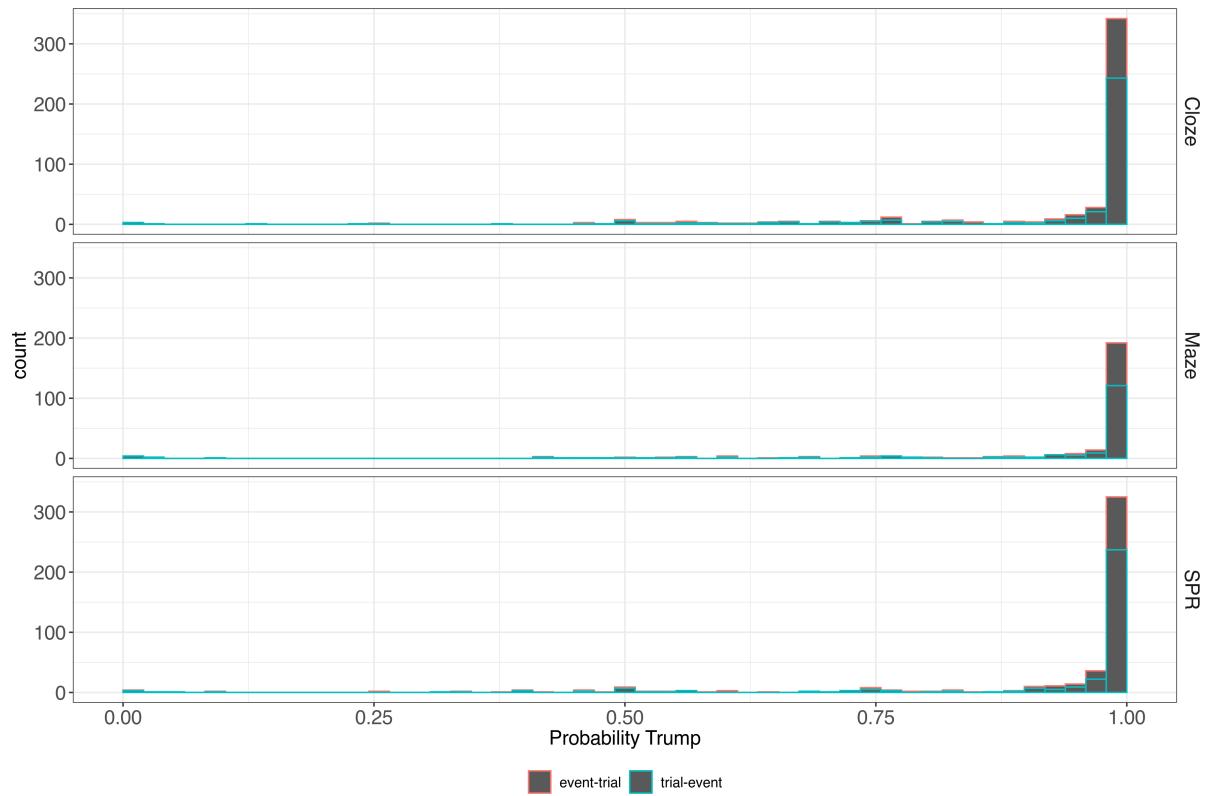
Event expectations



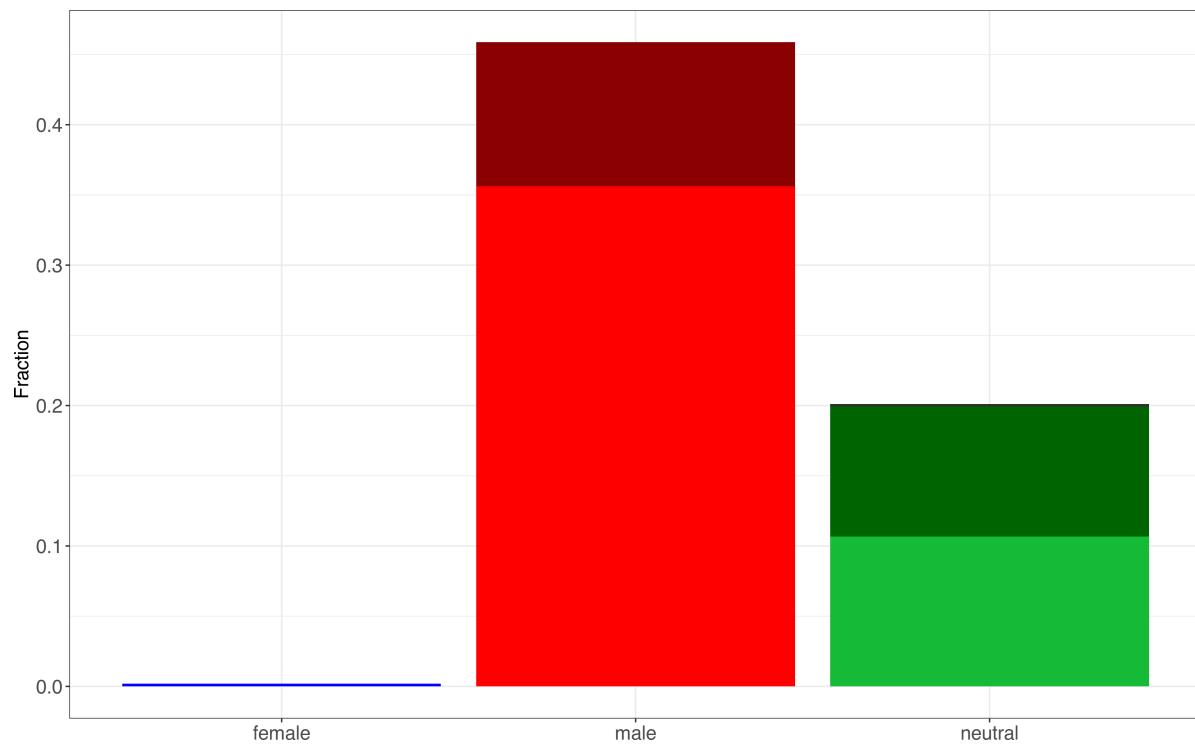


Event expectations histogram

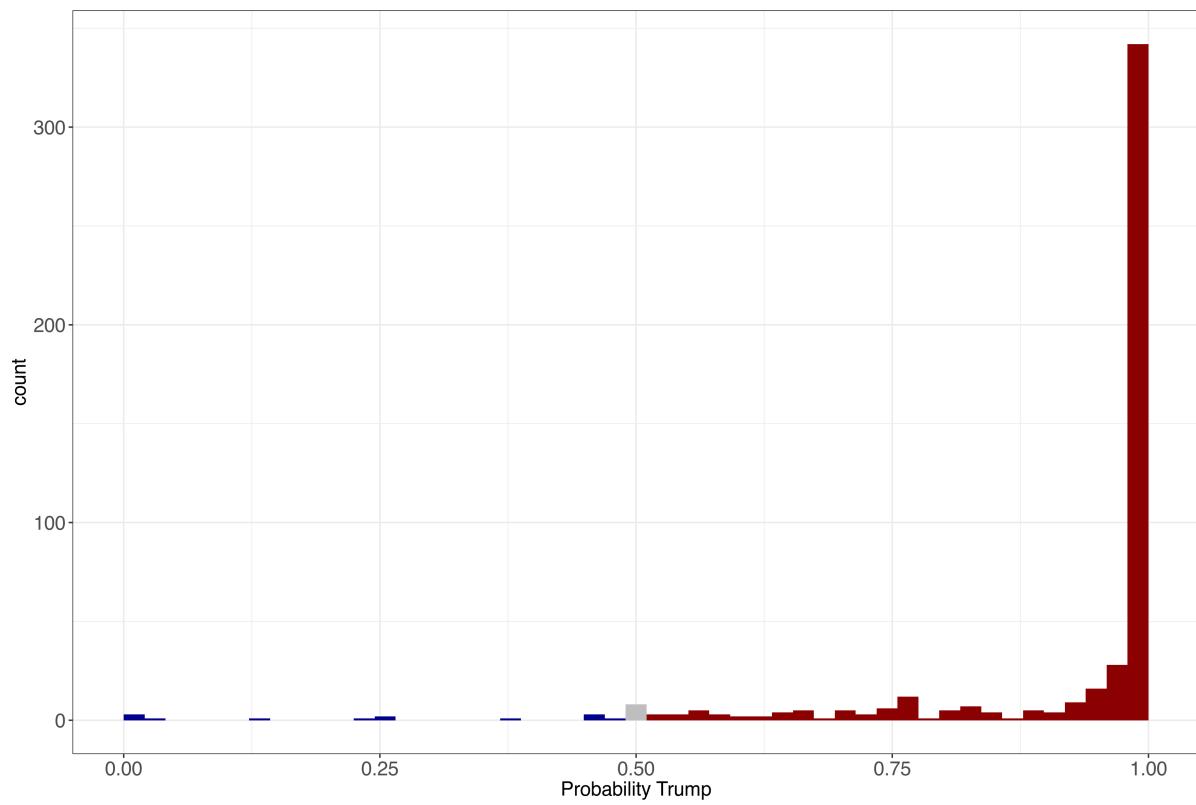
Event expectations histogram by task



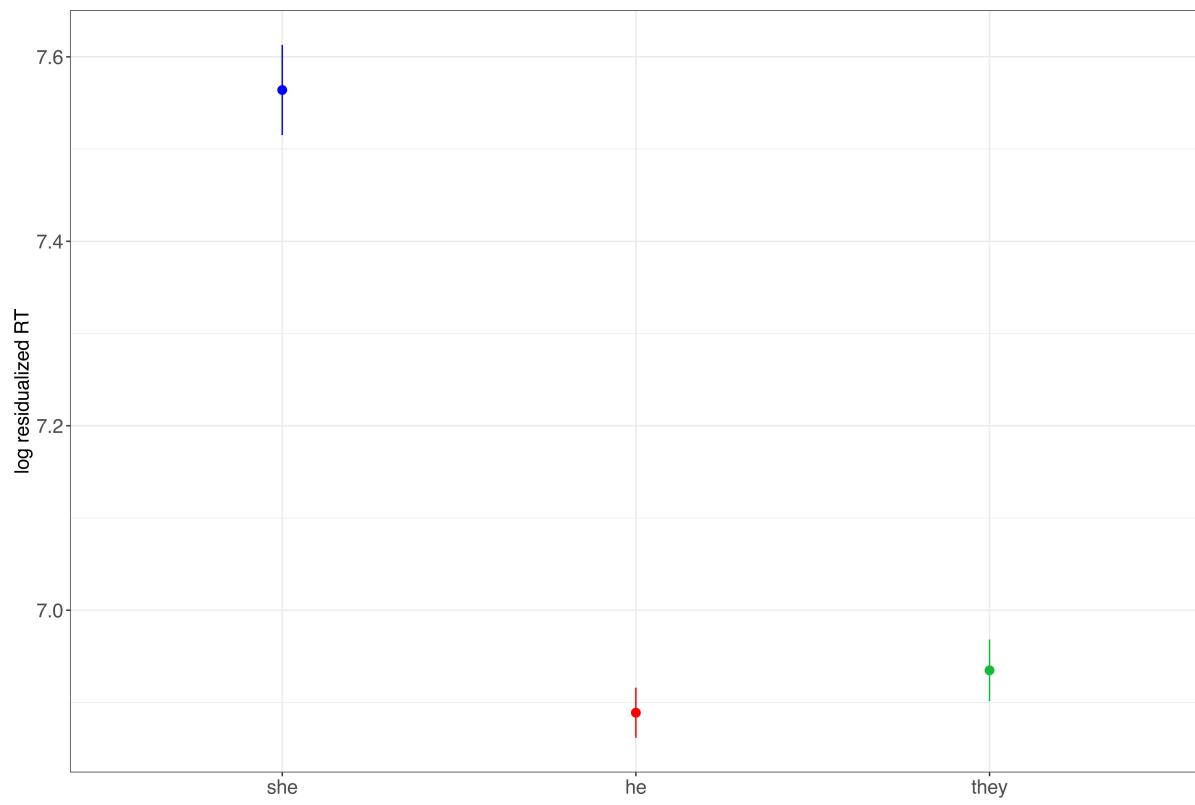
Cloze



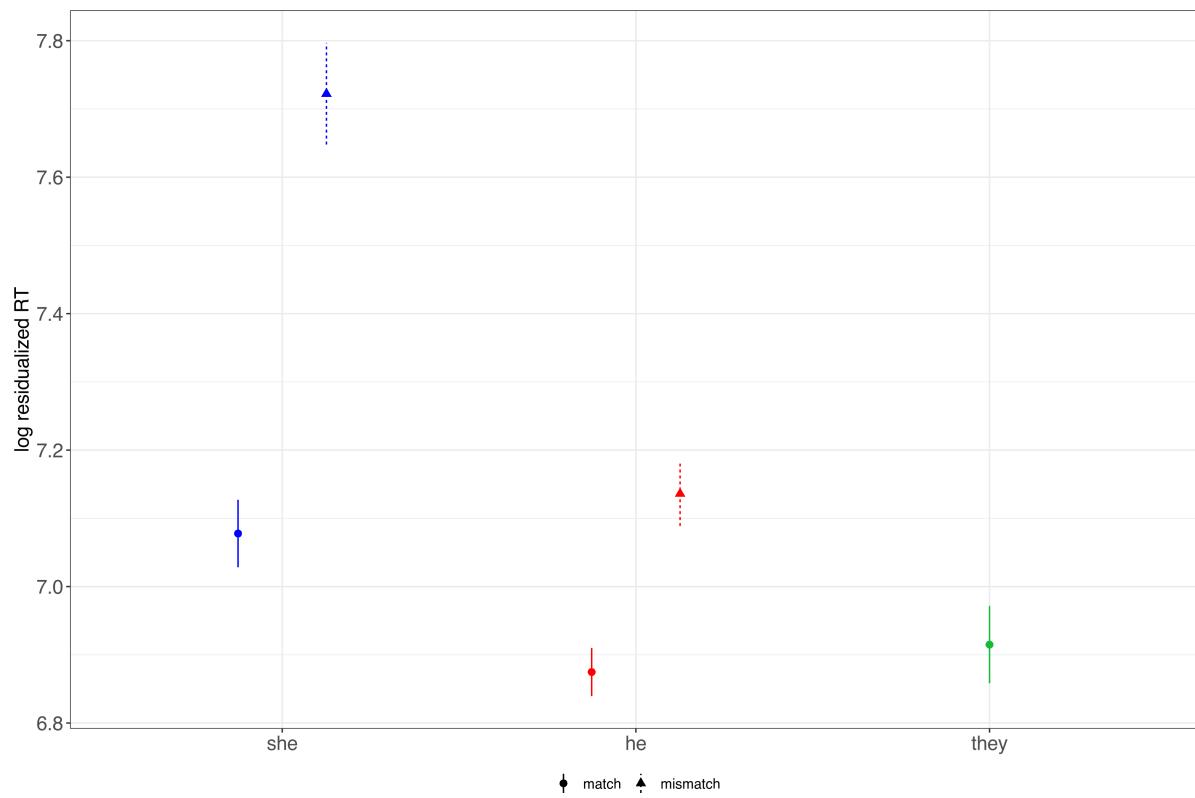
Cloze task expectations histogram



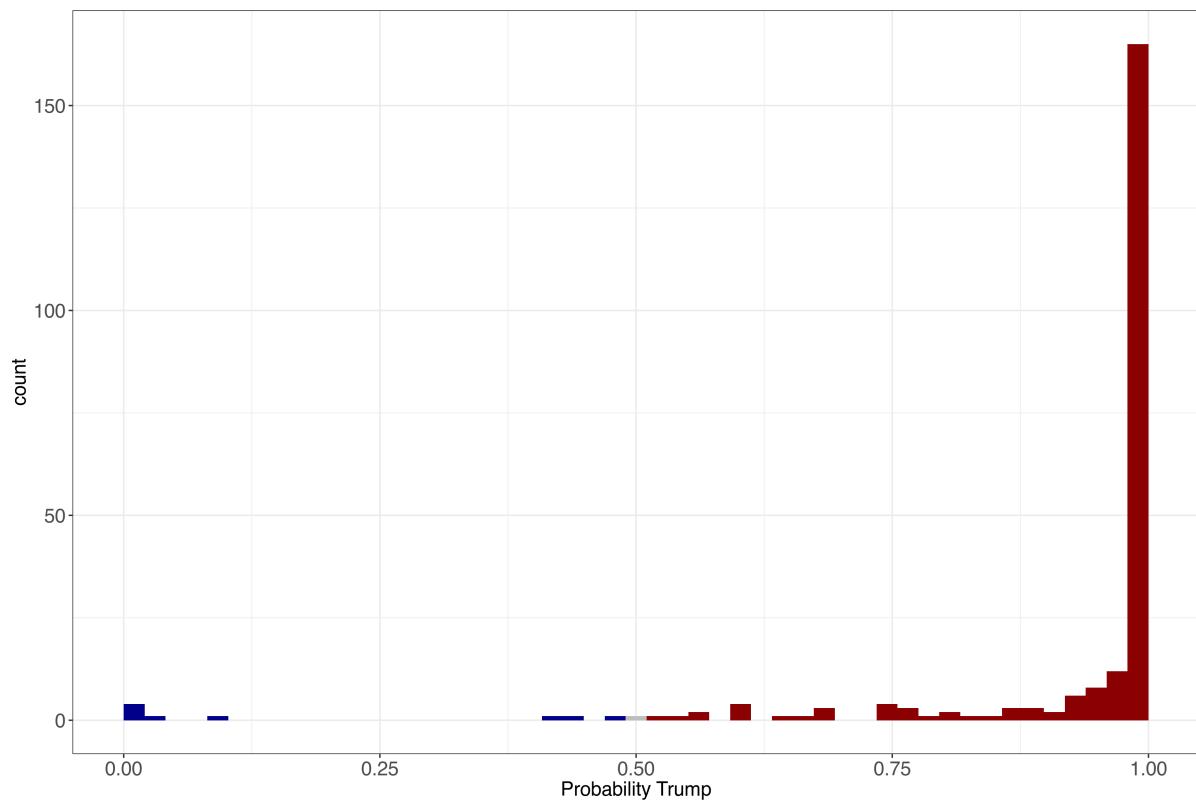
Maze



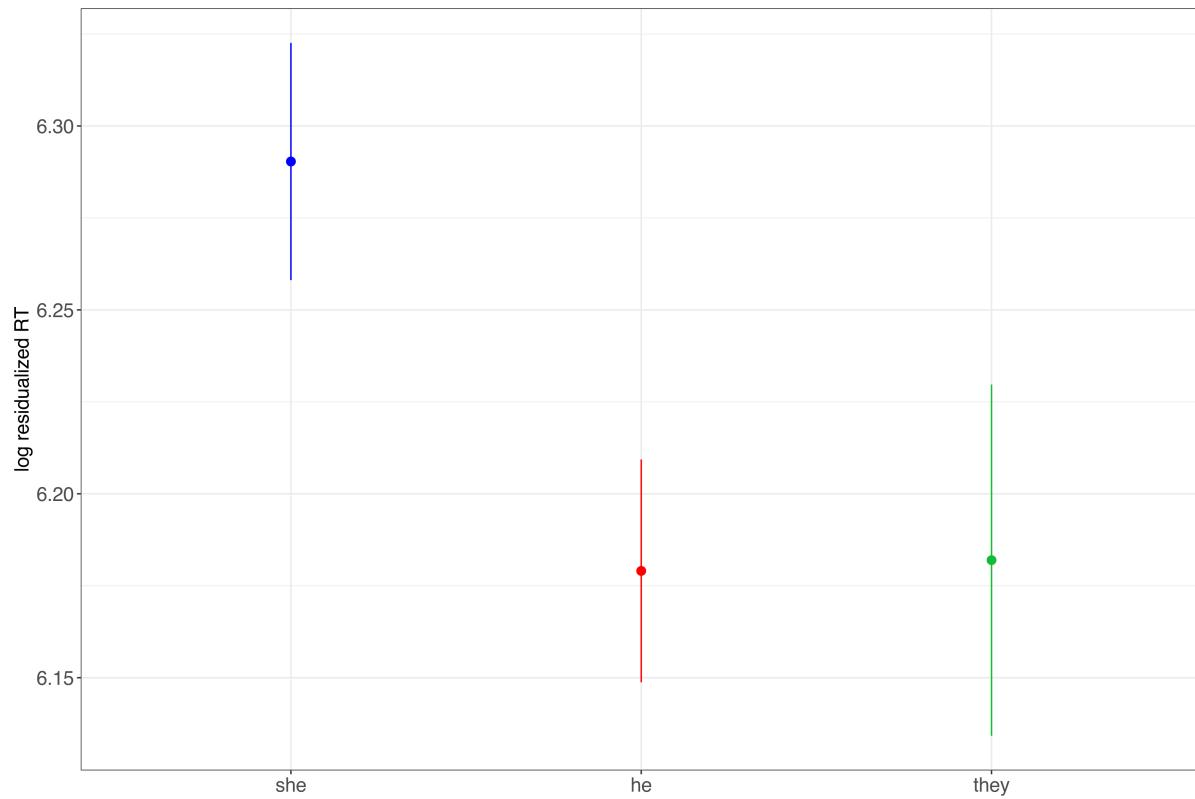
Maze second pronoun



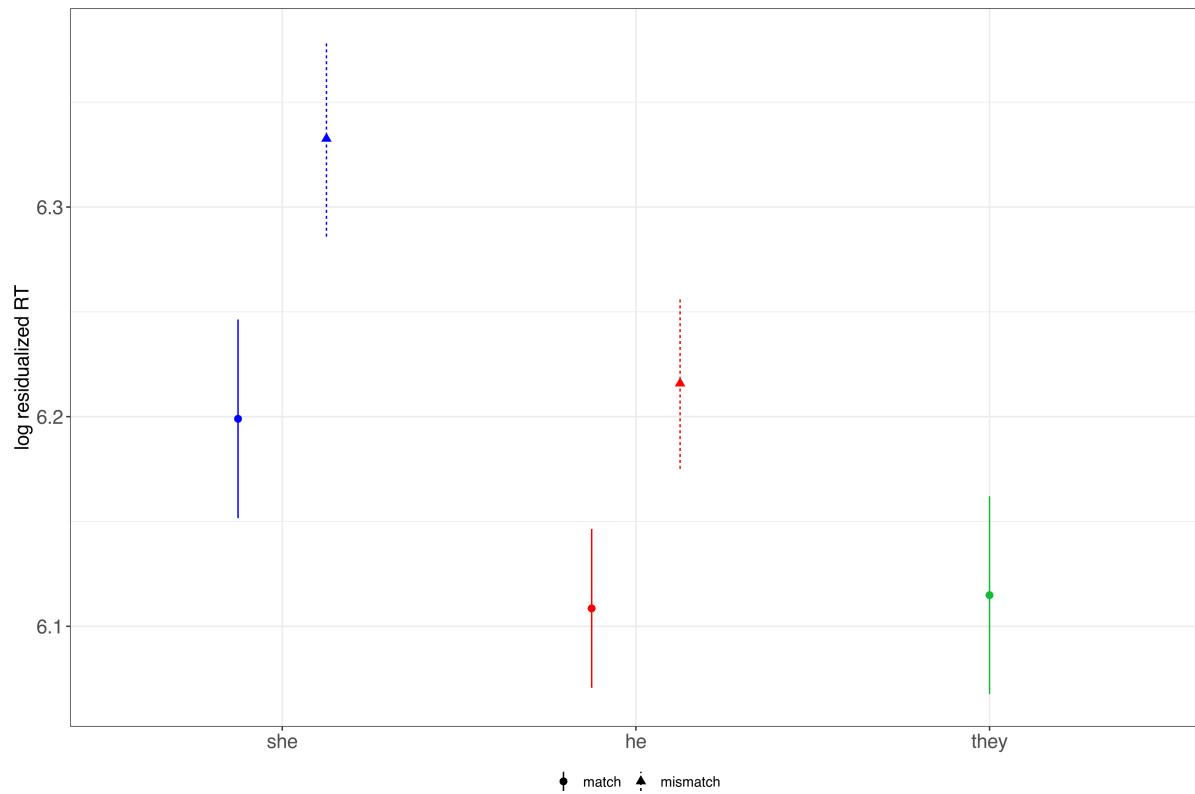
Maze task expectations histogram

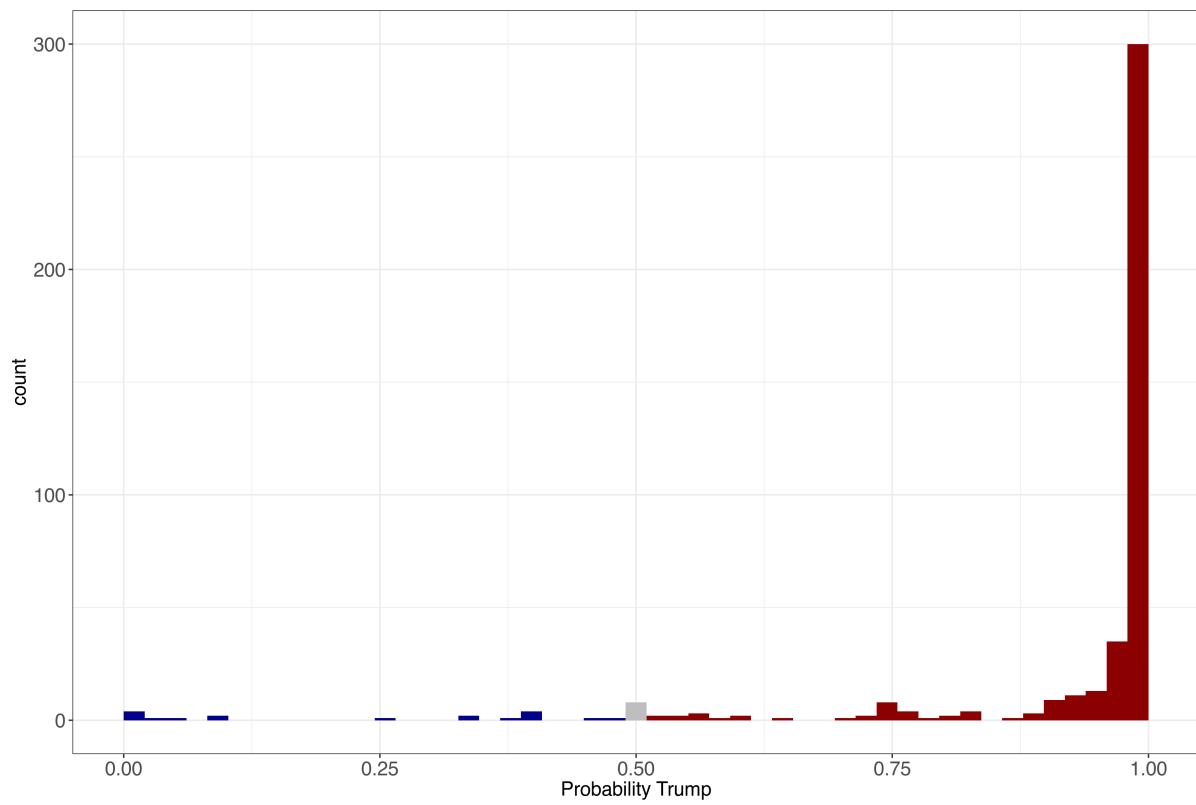


SPR



SPR second pronoun





SPR expectations histogram

Exploratory Data analysis

Other exploratory analysis below.

Event expectation

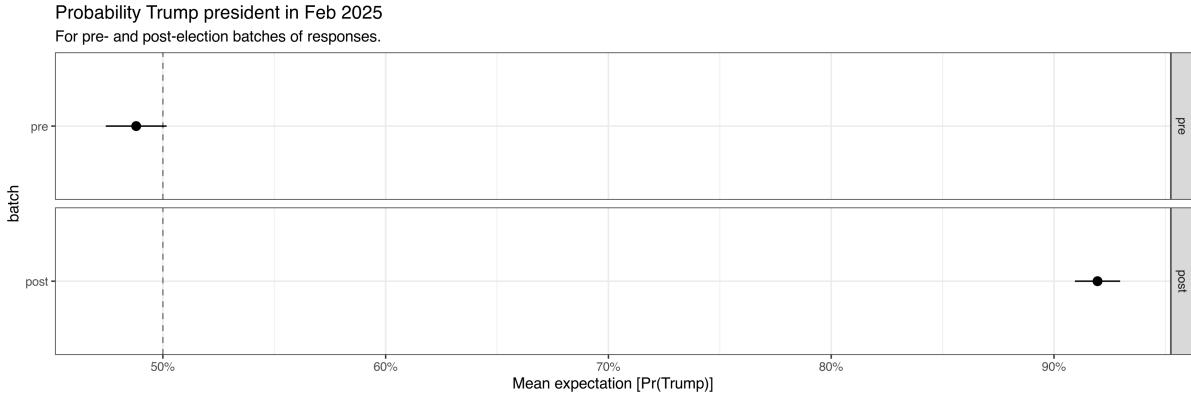


Table 1: One-sample t-tests comparing Trump probability to 50%

^ In the pre-election batch, mean expectations were 48.801%, and this was not significantly different from 50%. In the post-election batch, mean expectations were 91.954%, and this was significantly different from 50%.

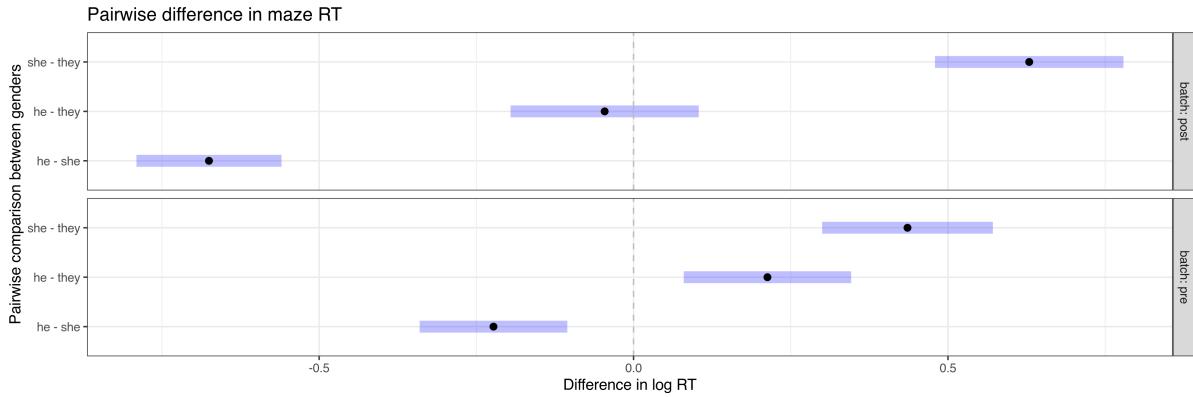
Maze

Pairwise comparisons between genders (first pronoun only): Within each batch {pre-election, post-election}, is there a significant difference in reading times for each pairwise comparison among {male,female,neutral} pronouns?

```
batch = post:
pronoun_gender_pairwise estimate      SE  df t.ratio p.value
he - she                 -0.675 0.0587 452 -11.506 <.0001
he - they                -0.046 0.0761 452  -0.604  0.5464
she - they                 0.629 0.0763 452   8.252 <.0001

batch = pre:
pronoun_gender_pairwise estimate      SE  df t.ratio p.value
he - she                 -0.223 0.0597 452  -3.729  0.0002
he - they                  0.213 0.0677 452   3.143  0.0018
she - they                 0.436 0.0691 452   6.308 <.0001
```

Results are given on the log (not the response) scale.



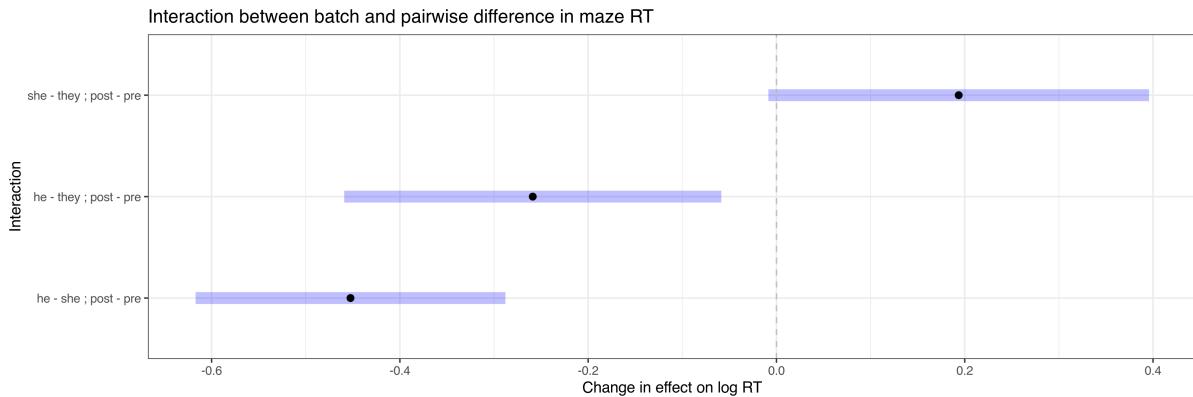
^ Pairwise comparisons between genders show, for Maze RTs:

- in pre-election data, RT of female pronouns is larger than that of male pronouns, which in turn are slower than gender-neutral pronouns. All differences are significant at 0.05 level.
- in post-election data, RT of female pronouns is significantly larger than that of either male or neutral pronouns respectively, and difference between male and neutral pronouns is not significant.

Now, looking at interaction between each pairwise contrast, pre-vs-post election:

pronoun_gender_pairwise	batch_pairwise	estimate	SE	df	t.ratio	p.value
he - she	post - pre	-0.452	0.0837	452	-5.404	< .0001
he - they	post - pre	-0.259	0.1020	452	-2.540	0.0114
she - they	post - pre	0.194	0.1030	452	1.882	0.0605

Results are given on the log (not the response) scale.



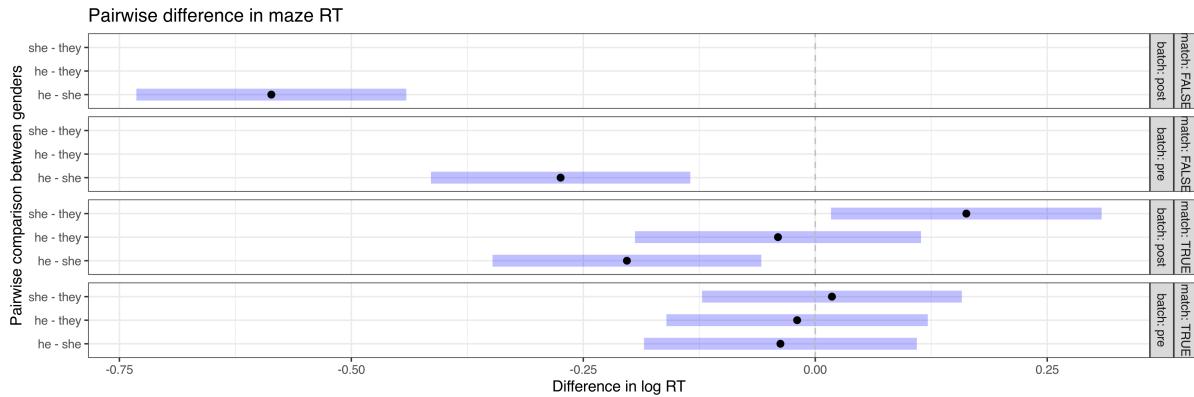
^ interaction between each pairwise contrast among pronouns and experiment batch (pre- vs post-election).

- Comparing post-election minus pre-election
 - log RT difference he-she is negative — that is, the bias for reading male faster than female pronouns increased.
 - log RT difference he-they is negative but smaller, and
 - log RT difference she-they is positive but not significant.

And looking at the second pronoun:

```
match = FALSE, batch = post:  
pronoun_gender_pairwise estimate      SE  df t.ratio p.value  
he - she                 -0.5862 0.0740 456  -7.919  <.0001  
he - they                nonEst   NA   NA       NA       NA  
she - they                nonEst   NA   NA       NA       NA  
  
match = TRUE, batch = post:  
pronoun_gender_pairwise estimate      SE  df t.ratio p.value  
he - she                 -0.2030 0.0737 456  -2.754  0.0061  
he - they                -0.0402 0.0784 456  -0.512  0.6088  
she - they                0.1628 0.0742 456   2.194  0.0287  
  
match = FALSE, batch = pre:  
pronoun_gender_pairwise estimate      SE  df t.ratio p.value  
he - she                 -0.2745 0.0711 456  -3.859  0.0001  
he - they                nonEst   NA   NA       NA       NA  
she - they                nonEst   NA   NA       NA       NA  
  
match = TRUE, batch = pre:  
pronoun_gender_pairwise estimate      SE  df t.ratio p.value  
he - she                 -0.0375 0.0748 456  -0.502  0.6161  
he - they                -0.0195 0.0717 456  -0.273  0.7852  
she - they                0.0180 0.0712 456   0.253  0.8007
```

Results are given on the log (not the response) scale.



```
batch = post:
```

Pronoun Comparison	Match	Estimate	SE	df	t.ratio	p.value
he - she	FALSE - TRUE	-0.383	0.104	456	-3.668	0.0003
he - they	FALSE - TRUE	nonEst	NA	NA	NA	NA
she - they	FALSE - TRUE	nonEst	NA	NA	NA	NA

```
batch = pre:
```

Pronoun Comparison	Match	Estimate	SE	df	t.ratio	p.value
he - she	FALSE - TRUE	-0.237	0.103	456	-2.295	0.0222
he - they	FALSE - TRUE	nonEst	NA	NA	NA	NA
she - they	FALSE - TRUE	nonEst	NA	NA	NA	NA

Results are given on the log (not the response) scale.

SPR

Pairwise comparisons between genders (first pronoun only): Within each batch {pre-election, post-election}, is there a significant difference in reading times for each pairwise comparison among {male,female,neutral} pronouns?

```
batch = post:
```

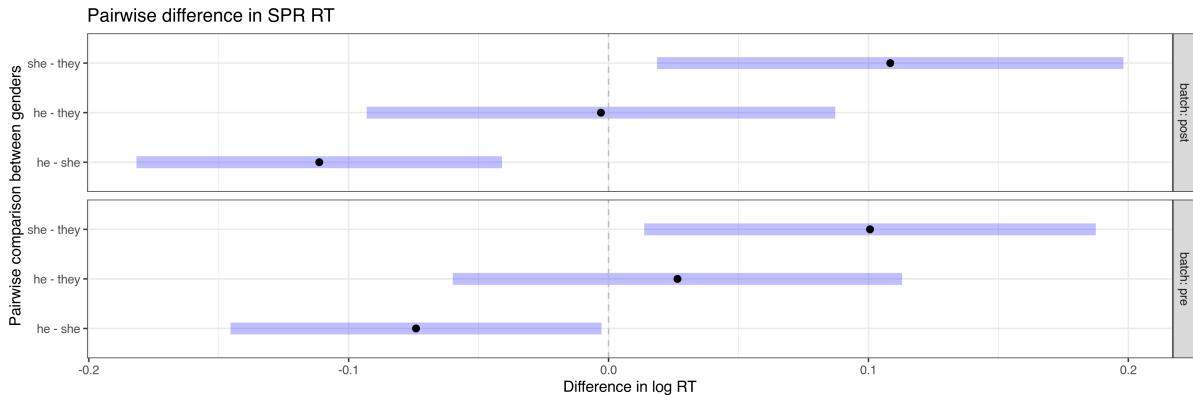
Pronoun Comparison	Estimate	SE	df	t.ratio	p.value
he - she	-0.11131	0.0358	836	-3.106	0.0020
he - they	-0.00292	0.0459	836	-0.064	0.9492
she - they	0.10839	0.0457	836	2.371	0.0180

```
batch = pre:
```

Pronoun Comparison	Estimate	SE	df	t.ratio	p.value
he - she	-0.07409	0.0364	836	-2.037	0.0420

he - they	0.02651	0.0440	836	0.602	0.5474
she - they	0.10060	0.0443	836	2.273	0.0233

Results are given on the log (not the response) scale.



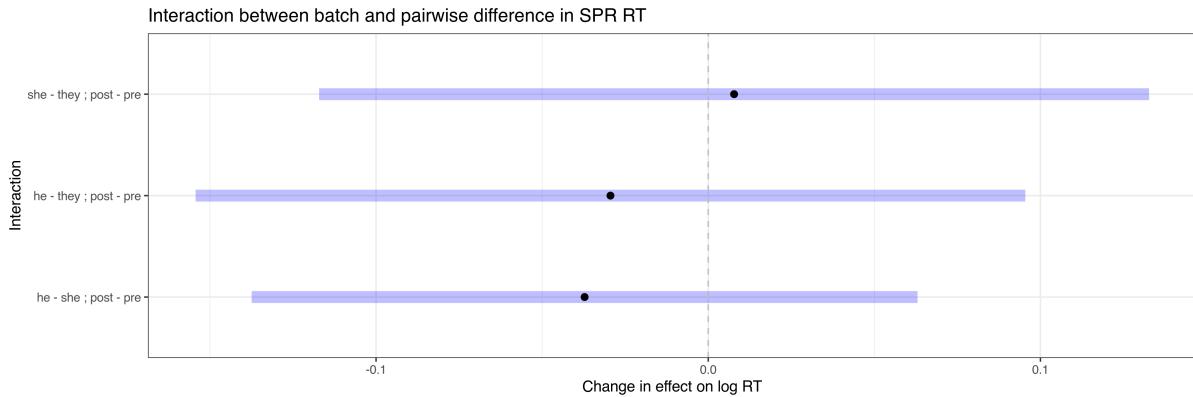
^ Pairwise comparisons between genders show, for SPR RTs:

- in pre-election data,
 - RT of female pronouns is slower than that of neutral pronouns ($p=0.0233$) and
 - RT of female pronouns is slower than that of male pronouns ($p=0.0420$).
 - Male-neutral comparison is not significant ($p=0.5474$).
- in post-election data,
 - RT of female pronouns is slower than that of neutral pronouns ($p=0.0180$) and
 - RT of female pronouns is slower than that of male pronouns ($p=0.0020$).
 - Male-neutral comparison is not significant ($p=0.5474$).

Next, looking at interaction between each pairwise contrast, pre-vs-post election:

pronoun_gender_pairwise	batch_pairwise	estimate	SE	df	t.ratio	p.value
he - she	post - pre	-0.03722	0.0511	836	-0.729	0.4663
he - they	post - pre	-0.02944	0.0636	836	-0.463	0.6438
she - they	post - pre	0.00779	0.0636	836	0.122	0.9027

Results are given on the log (not the response) scale.



^ interaction between each pairwise contrast among pronouns and experiment batch (pre- vs post-election).

- Comparing post-election minus pre-election
 - numerical speedup for he versus they and he versus she, and slowdown for she versus they, but no contrasts are significant.

And looking at the second pronoun:

```

match = FALSE, batch = post:
pronoun_gender_pairwise estimate      SE  df t.ratio p.value
he - she                  -0.11672 0.0508 827  -2.297  0.0219
he - they                 nonEst    NA   NA      NA      NA
she - they                 nonEst    NA   NA      NA      NA

match = TRUE, batch = post:
pronoun_gender_pairwise estimate      SE  df t.ratio p.value
he - she                  -0.09042 0.0509 827  -1.775  0.0763
he - they                 -0.00628 0.0524 827  -0.120  0.9046
she - they                  0.08414 0.0521 827   1.614  0.1068

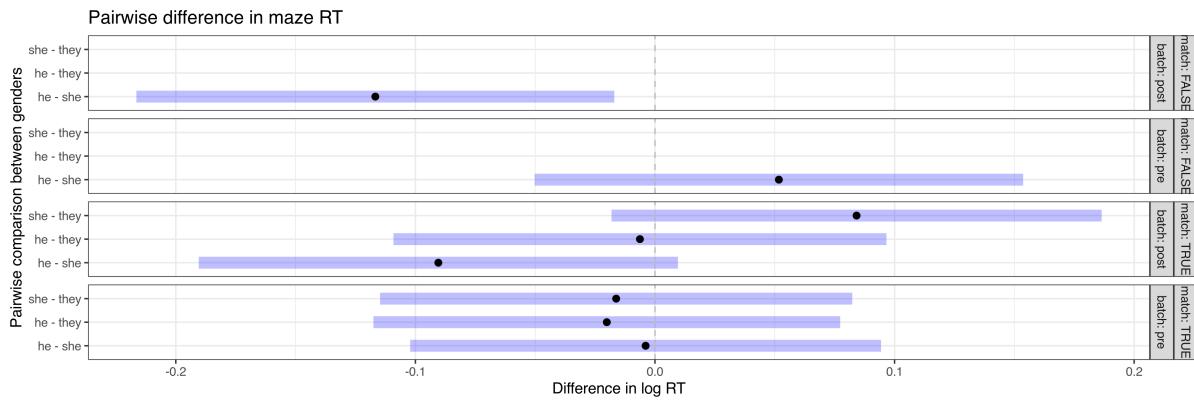
match = FALSE, batch = pre:
pronoun_gender_pairwise estimate      SE  df t.ratio p.value
he - she                  0.05171 0.0519 827   0.996  0.3197
he - they                 nonEst    NA   NA      NA      NA
she - they                 nonEst    NA   NA      NA      NA

match = TRUE, batch = pre:
pronoun_gender_pairwise estimate      SE  df t.ratio p.value
he - she                  -0.00391 0.0501 827  -0.078  0.9378
he - they                 -0.02010 0.0496 827  -0.405  0.6855

```

she - they -0.01620 0.0502 827 -0.323 0.7471

Results are given on the log (not the response) scale.



batch = post:

	pronoun_gender_pairwise	match_pairwise	estimate	SE	df	t.ratio	p.value
he - she	FALSE	- TRUE	-0.0263	0.0720	827	-0.366	0.7148
he - they	FALSE	- TRUE	nonEst	NA	NA	NA	NA
she - they	FALSE	- TRUE	nonEst	NA	NA	NA	NA

batch = pre:

	pronoun_gender_pairwise	match_pairwise	estimate	SE	df	t.ratio	p.value
he - she	FALSE	- TRUE	0.0556	0.0721	827	0.771	0.4409
he - they	FALSE	- TRUE	nonEst	NA	NA	NA	NA
she - they	FALSE	- TRUE	nonEst	NA	NA	NA	NA

Results are given on the log (not the response) scale.

Other exploratory plotting

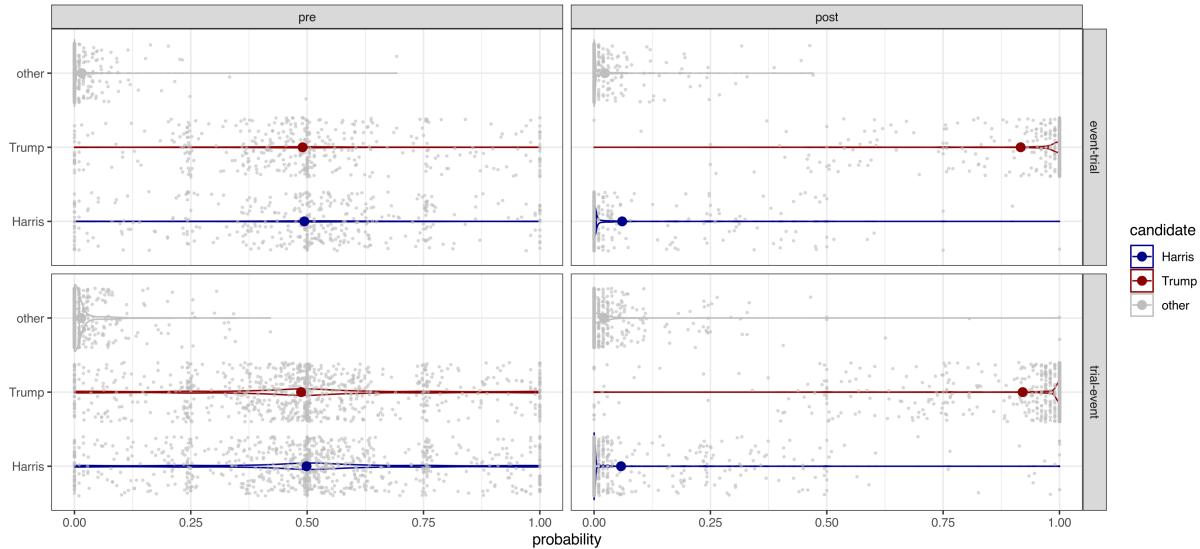
Number of participants per task pre- and post- election.

batch	cloze	maze	spr	total
pre	485	244	445	1174
post	497	235	432	1164

Expectations

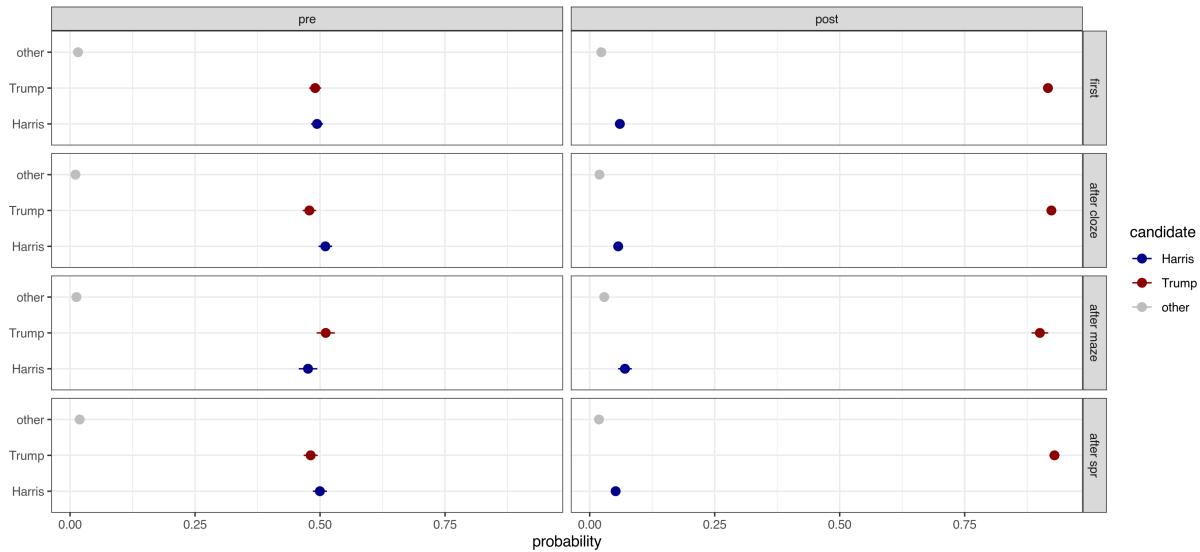
Check for order effect on belief estimation results, in pre- and post- election data.

"Who do you think will be the US president in February 2025?"
responses normalized per participant

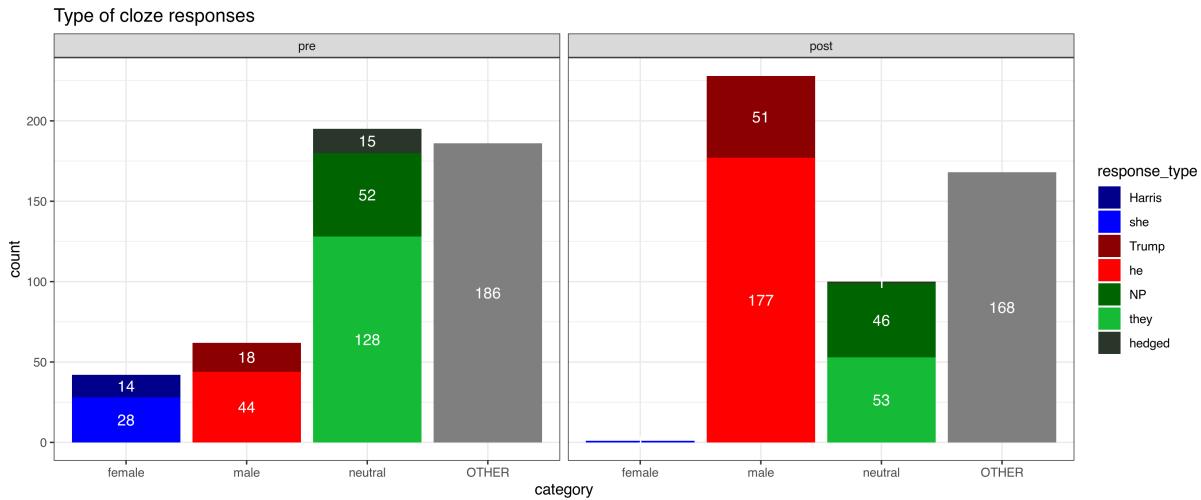


Looking for order effects, (combining all orders that start with “event” to one, for this plot, since all are identical wrt this data)

"Who do you think will be the US president in February 2025?"
responses normalized per participant, faceted by order

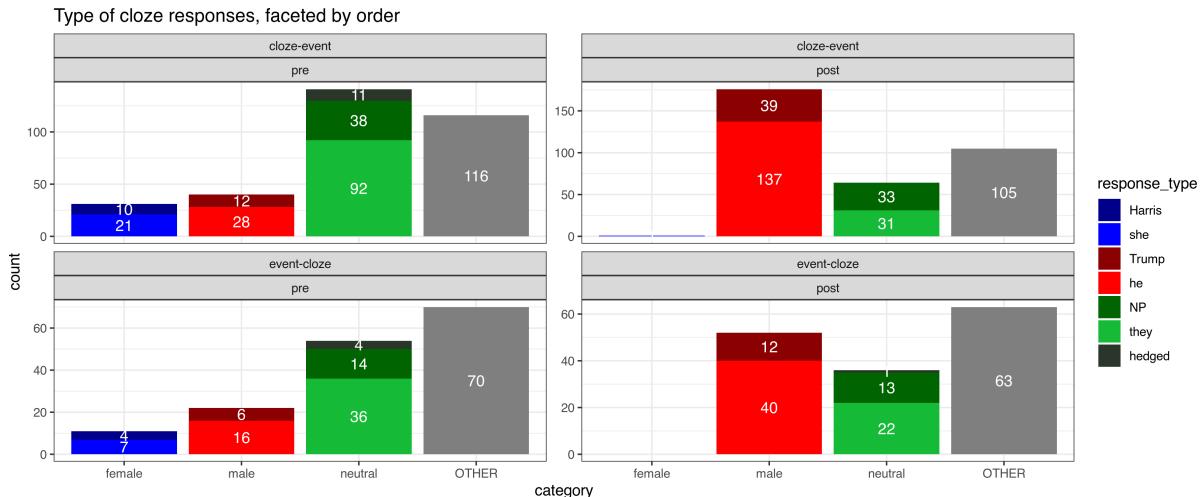


Cloze



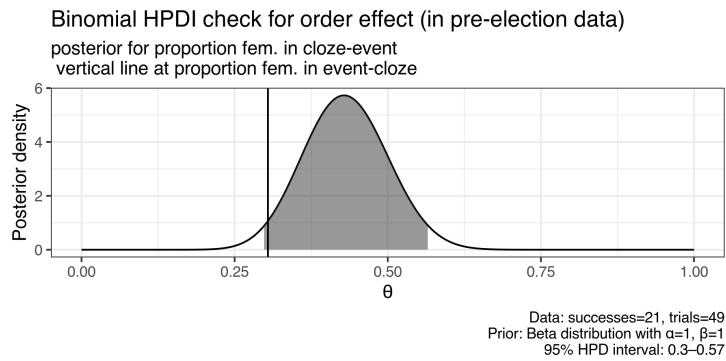
```
prop.test(F, M+F) p-value  
0.07709987
```

Faceting by order:



In the event-cloze order there seems to be a somewhat larger bias toward producing male pronouns or male referents/Trump.

But it seems the difference between orders here is actually not significant, per Titus' binomial 95%-HPDI test:



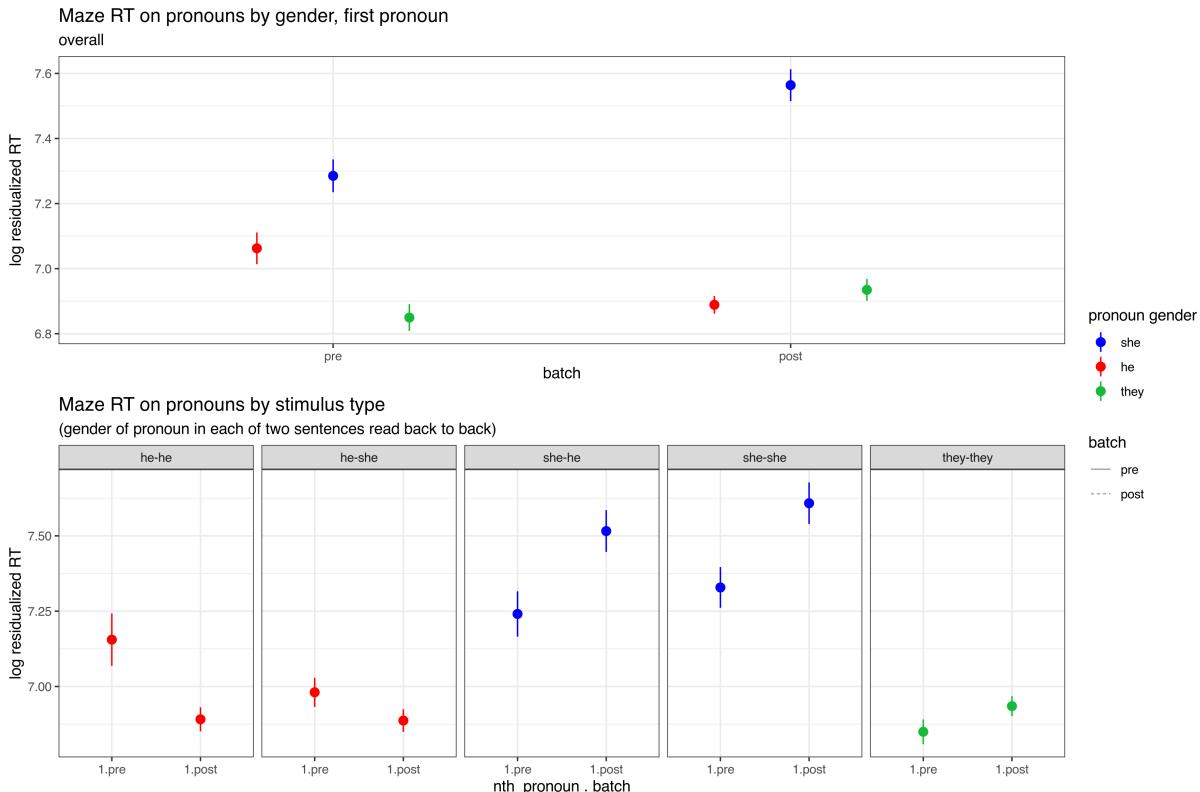
By item cloze breakdown

Type of cloze responses, by item
(OTHER responses not shown)

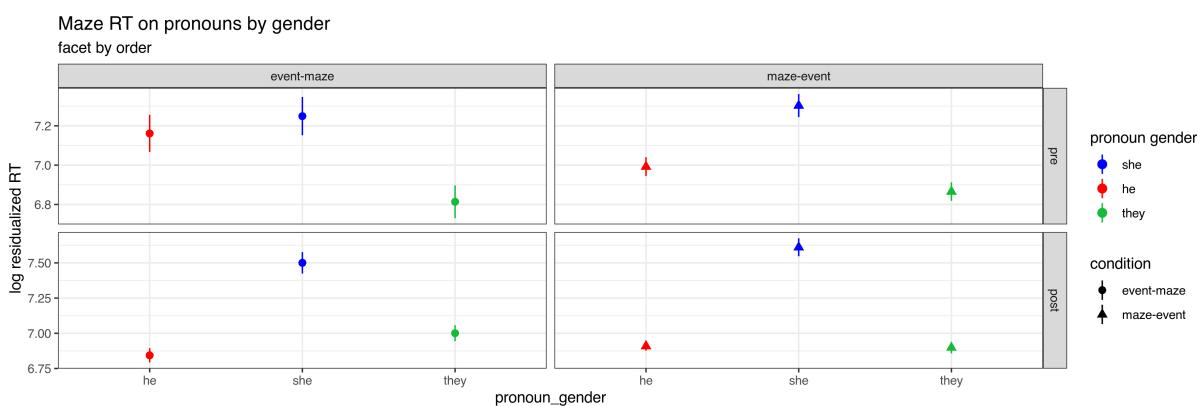


Maze

Look at average Maze RT by gender of pronoun, and for each condition of two sentences presented. Shows an effect of gender (male pronoun is read faster than female, neutral is perhaps even faster than male). These Maze RTs are residualized as with the SPR (control for participant mean reading speed and eg word length, punctuation, and by item effects).

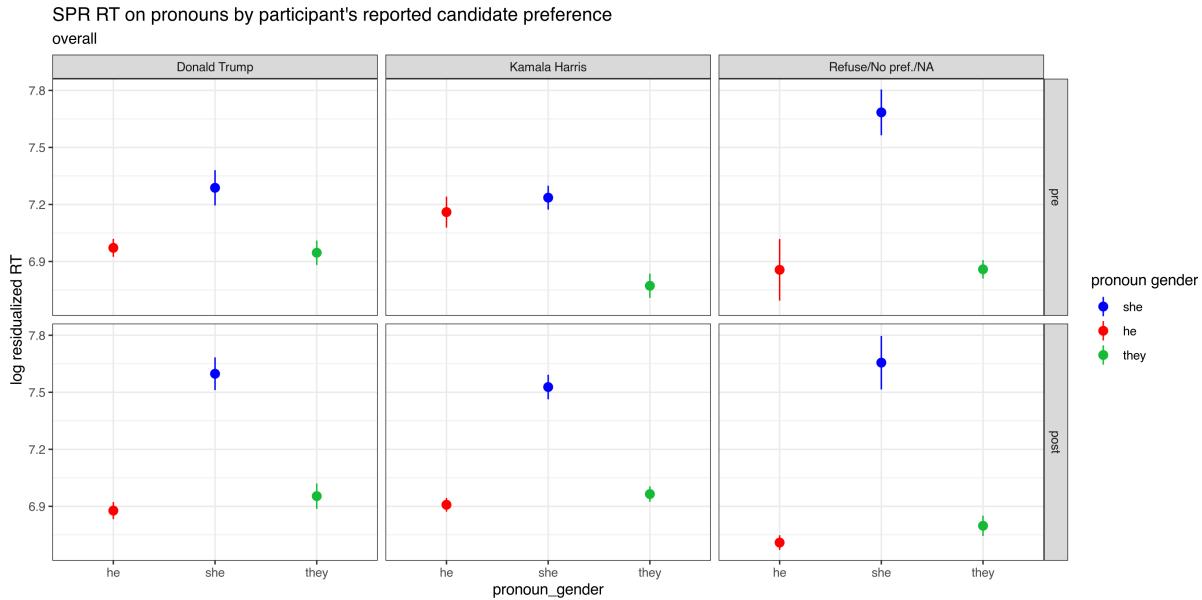


Faceting by order



^ In the pre-election data (top facets above), there may be a slightly less clear male pronoun bias for participants who did the maze task following the event estimation (event-maze) versus before (maze-event). In the post-election data, the bias was strong in either order.

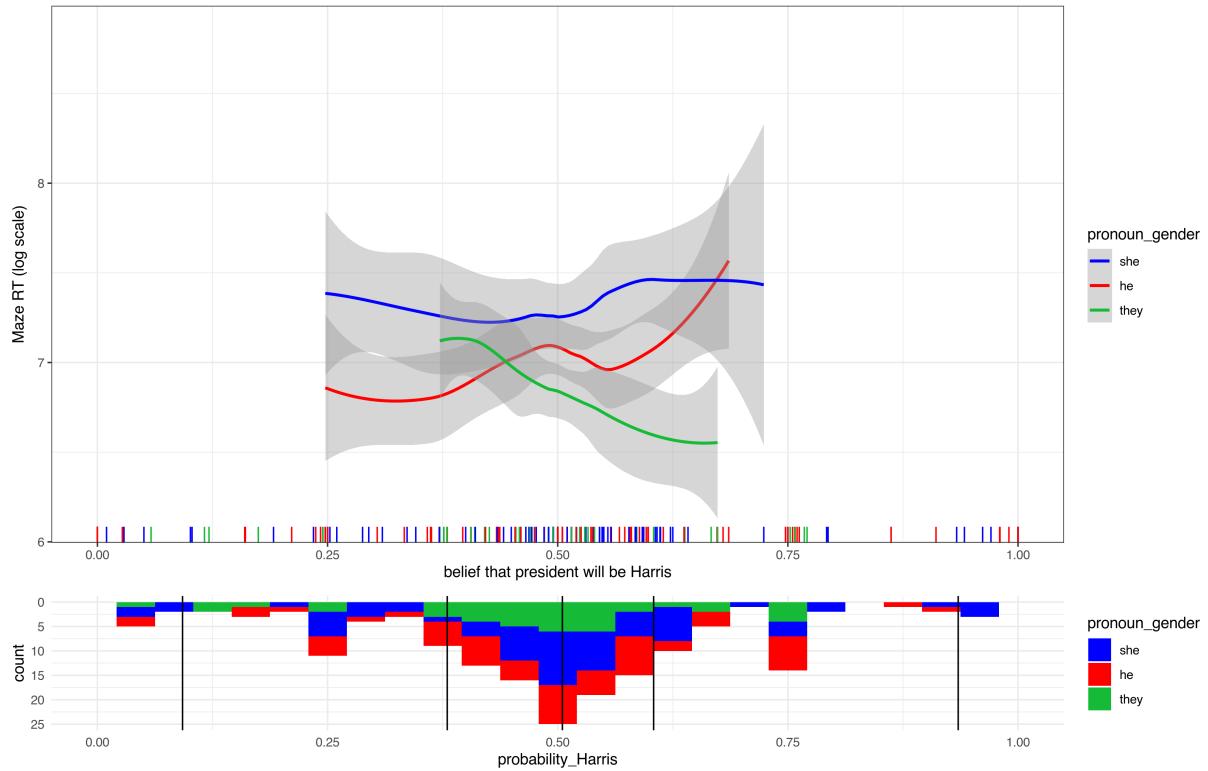
And broken down by participant reported preference



^ The pre-election bias is least strong among participants whose preferred candidate was Harris.

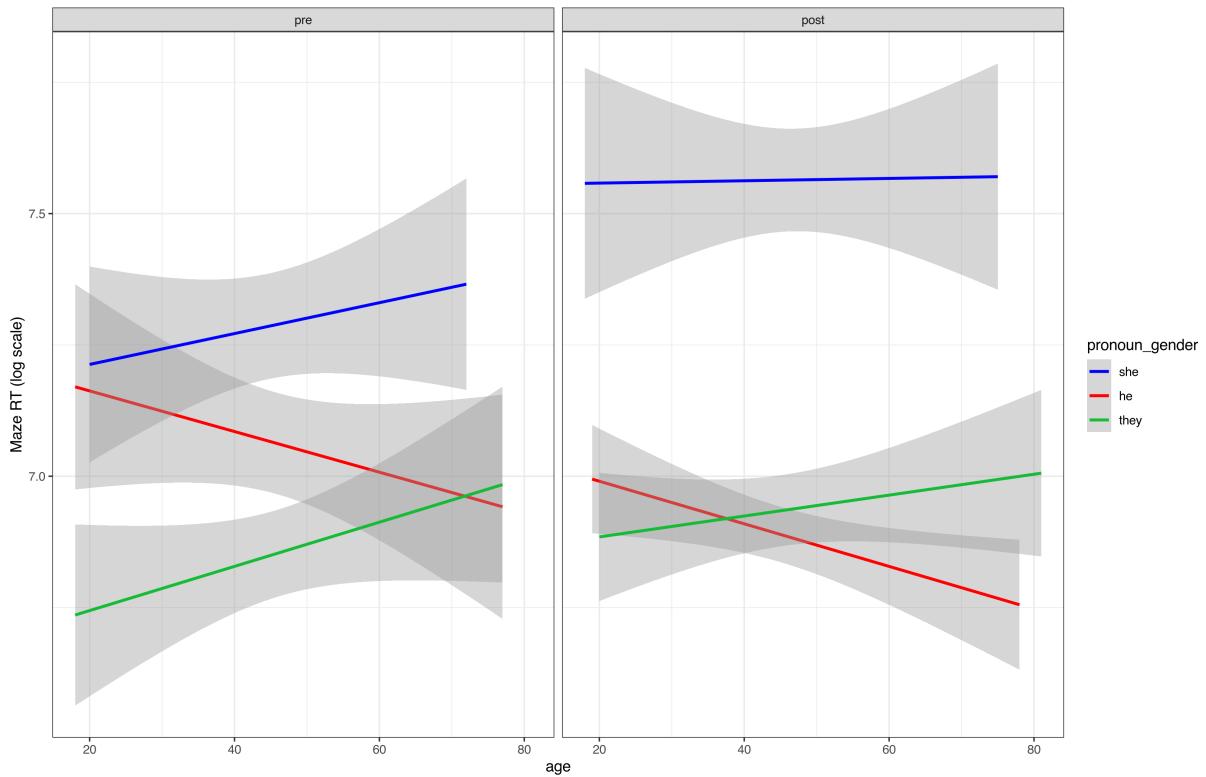
We could look at a related effect as a continuous interaction plotting event-probability (who the participant thinks will be president) on the x-axis versus RT on the y-axis:

Maze RT on pronouns, plotted against probability of female president



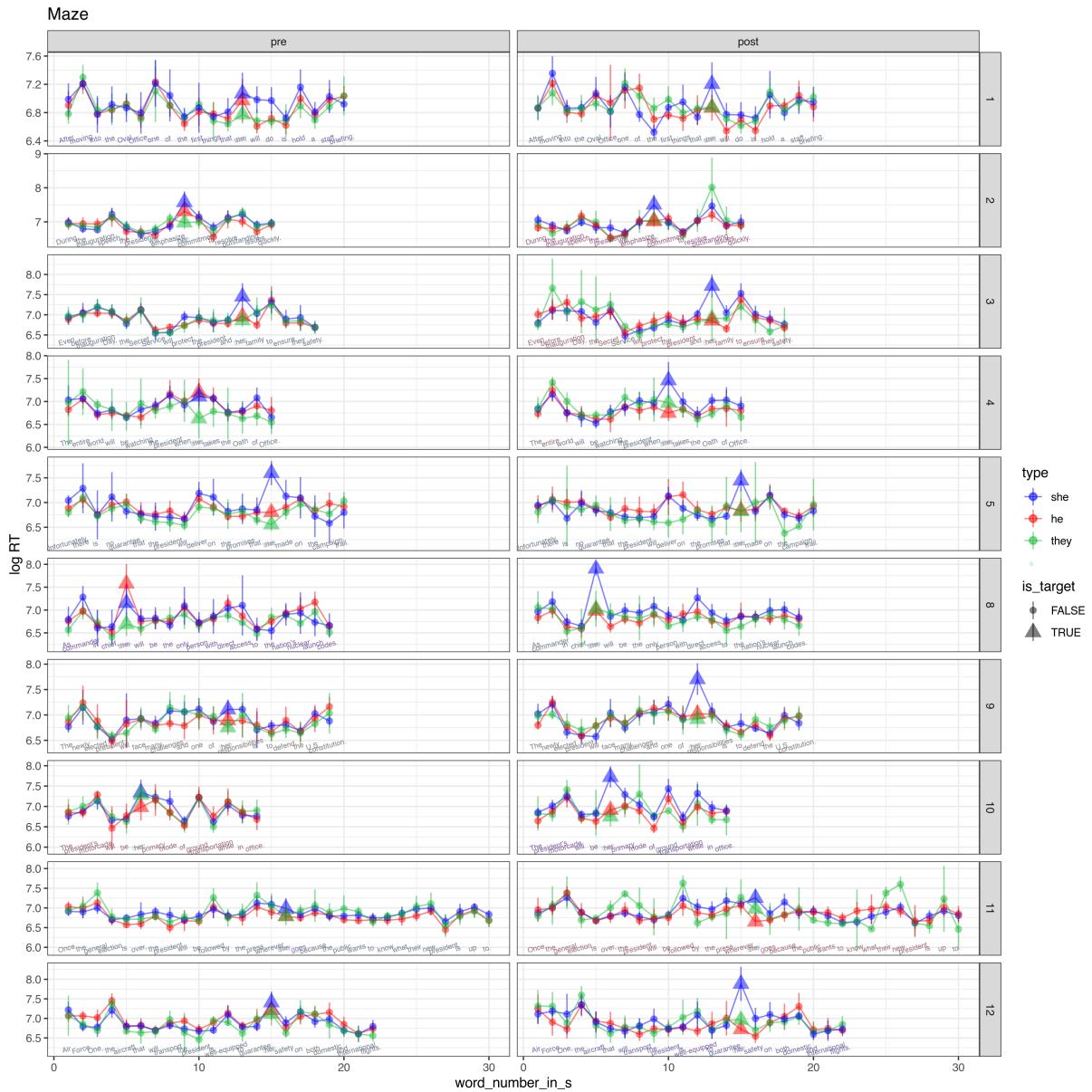
^ It looks like the male-pronoun processing bias (/ female-pronoun processing penalty) gets smaller as belief that the president would be female is higher. But note the majority of participants are in the middle of the distribution.

Maze RT on pronouns, plotted against participant age



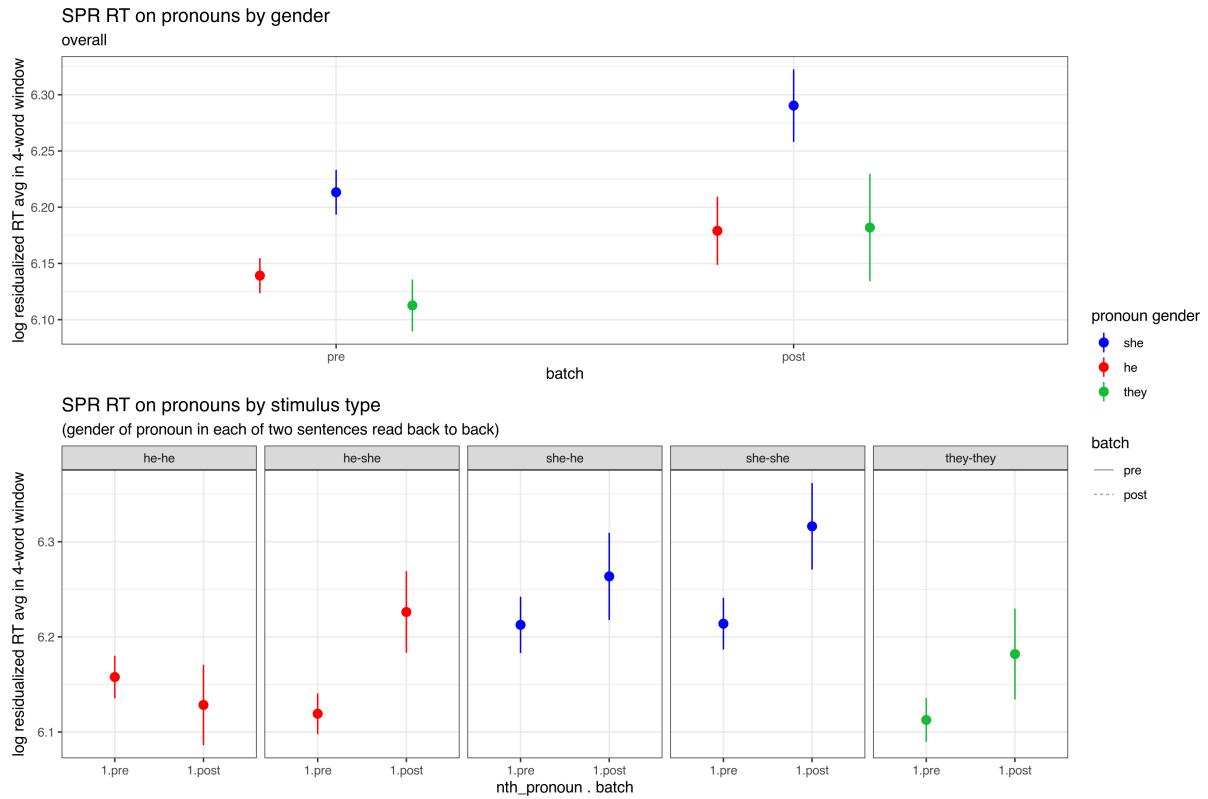
^Looking at participant age, there also could be something interesting going on, at least with the decreasing processing cost of the neutral pronoun for younger readers.

Maze reading time tracks on each of the target sentences:

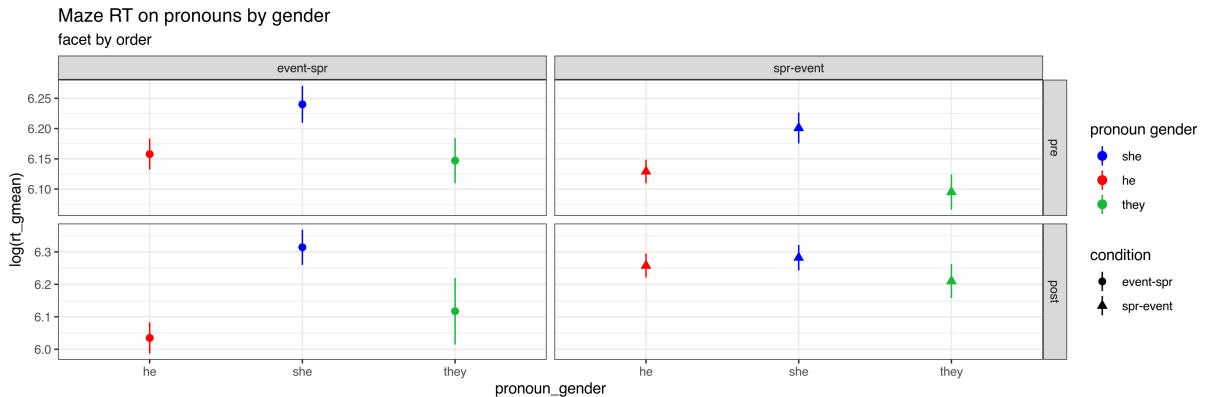


SPR

Likewise look at average SPR RT by gender of pronoun, and for each condition of two sentences presented. These RTs are residualized.



Faceting by order

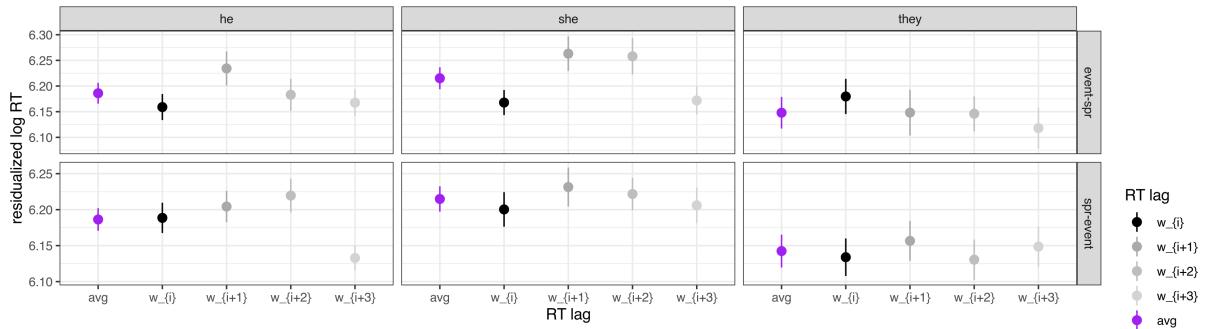


^ Probably not enough power to detect, but it may be true that the bias is stronger in the spr-event order, as with Maze data.

Pre-election data

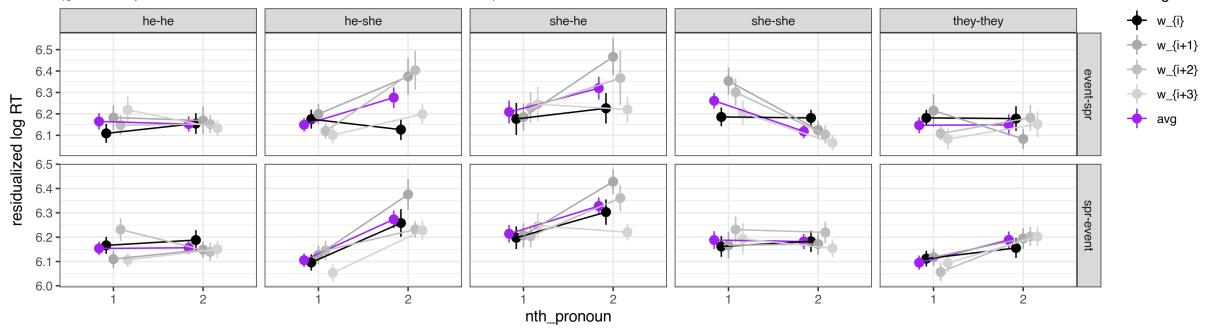
SPR by gender

Looking at RT with different lags w_{-i+n} , where w_{-i} = target pronoun.
Also showing geometric_mean(w_{-i+n}) for $n = \{0, \dots, 3\}$



Maze RT on pronouns by stimulus type

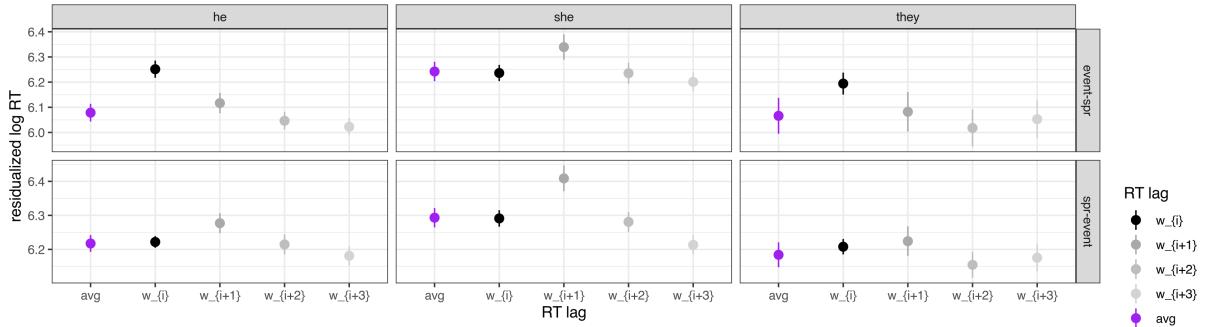
(gender of pronoun in each of two sentences read back to back)



Post-election data

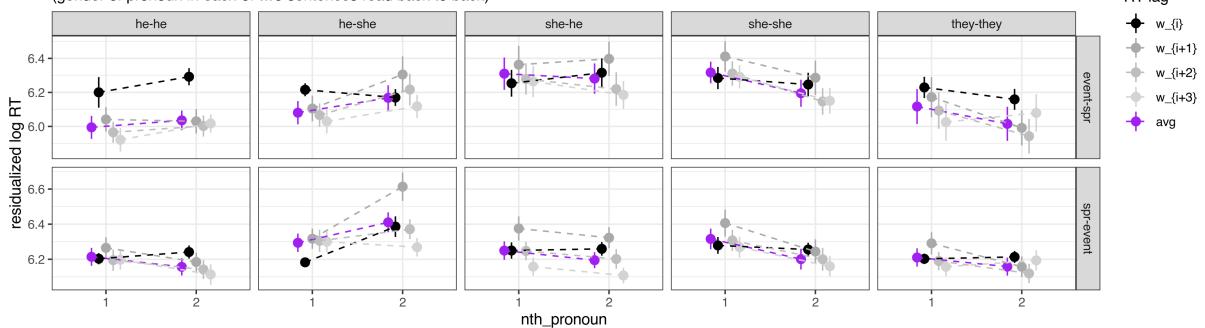
SPR by gender

Looking at RT with different lags $w_{\{i+n\}}$, where w_i = target pronoun.
Also showing geometric_mean($w_{\{i+n\}}$) for $n = \{0, \dots, 3\}$

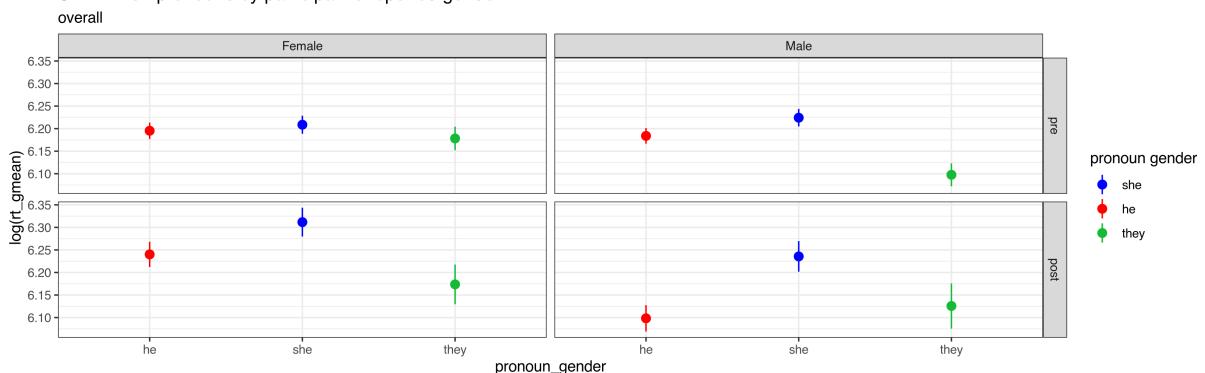


Maze RT on pronouns by stimulus type

(gender of pronoun in each of two sentences read back to back)

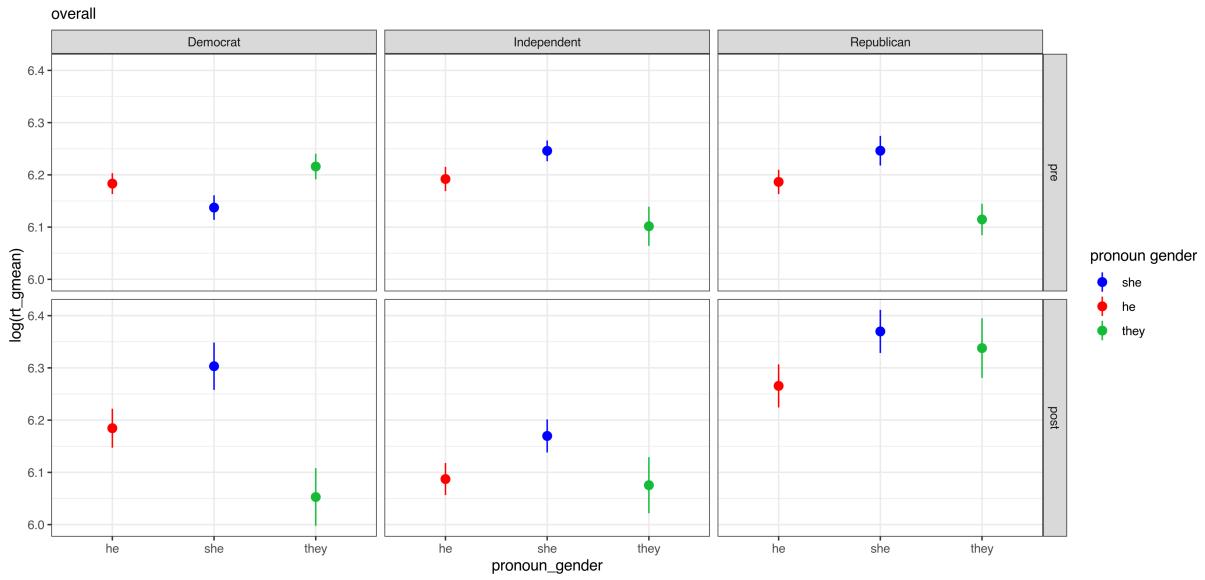


SPR RT on pronouns by participant's reported gender overall



^ SPR averages broken down by participant gender may be useful if we're interested in the difference that shows up in Maze as well: bias seems larger in males.

SPR RT on pronouns by participant's reported political affili.



^ SPR averages broken down by participant political affiliation, as with Maze above.

SPR tracks on each of the target sentences:

SPR

