## **Proof of Concept**

CorporaCoCo v1.0-2 (2017-03-31)

### Anthony Hennessey

Statistics and Probability, School of Mathematical Sciences University of Nottingham anthony.hennessey@nottingham.ac.uk

# Viola Wiegand

Centre for Corpus Research, College of Arts and Law University of Birmingham v.wiegand@bham.ac.uk

Michaela Mahlberg Centre for Corpus Research, College of Arts and Law University of Birmingham m.a.mahlberg@bham.ac.uk

Christopher R. Tench Division of Clinical Neurosciences, School of Medicine

University of Nottingham christopher.tench@nottingham.ac.uk

Jamie Lentin Shuttle Thread Manchester jm@ravingmantis.com

Load the CorporaCoCo package.

```
library(CorporaCoCo)
```

Create tokenized copies of 'Great Expectations' and 'A Tale of Two Cities' novels. The texts are available in the CorporaCorpus package which is available from github at https://github.com/ravingmantis/CorporaCorpus, there are installation instructions on the front page. (The CorporaCorpus package is not available on CRAN as at 17MB it exceeds the CRAN data package size limit of 5MB). The tokenization we use here is very simplistic, but it will do for our purposes. The  ${\tt stringi}$  package has a solid implementation of UTF-8 word boundaries so although this is simple tokenization it should do a reasonable job for text in any language.

```
library(CorporaCorpus)
library(stringi)
GE <- unlist(stri_extract_all_words(stri_trans_tolower(readLines(corpus_filepaths('DNov', 'GE'))))
TTC <- unlist(stri_extract_all_words(stri_trans_tolower(readLines(corpus_filepaths('DNov', 'TTC')))))
```

Choose the set of nodes.

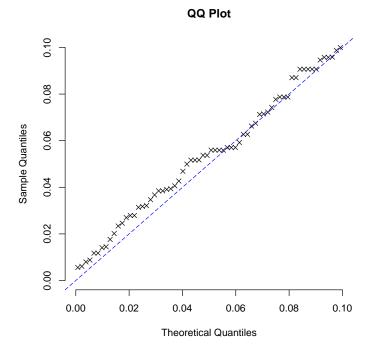
```
nodes <- c('back', 'eye', 'eyes', 'forehead', 'hand', 'hands', 'head', 'shoulder')</pre>
```

First we want to check that there are no significant results under the null. We create two corpora from alternate chunks of 1000 tokens of the two novels and check that there are no significant co-occurrence differences between our two sets of chunks.

```
chunks <- split(c(GE, TTC), ceiling(seq_along(c(GE, TTC)) / 1000))</pre>
corpus_a <- unlist( chunks[seq(1, length(chunks), 2)] )
corpus_b <- unlist( chunks[seq(2, length(chunks), 2)] )
corpus_a_c <- surface(corpus_a, span = '5LR')</pre>
corpus_b_c <- surface(corpus_b, span = '5LR')</pre>
results <- coco(corpus_a_c, corpus_b_c, nodes = nodes, fdr = 0.01)
results
Empty data.table (0 rows) of 11 cols: x,y,H_A,M_A,H_B,M_B...
```

This gives us the opportunity to check an assumption of FDR that the p-values are uniformly distributed.

```
results_all <- coco(corpus_a_c, corpus_b_c, nodes = nodes, fdr = 1.0)
test_p_values <- results_all$p_value[results_all$p_value <= 0.1]</pre>
plot(
     qunif(ppoints(test_p_values), min = 0, max = 0.1),
    sort(test_p_values),
bty = 'n', pch = 4, xlim = c(0.0, 0.1), ylim = c(0.0, 0.1),
main = "QQ Plot", xlab = "Theoretical Quantiles", ylab = "Sample Quantiles"
abline(a = 0, b = 1, col = 'blue', lty = 5)
```



Next we check that if we make some changes to one of our corpora that the method can spot them. Let us change about 90% of the 'my' tokens to 'CHIMERA' tokens in corpus\_a and comfirm that the method notices

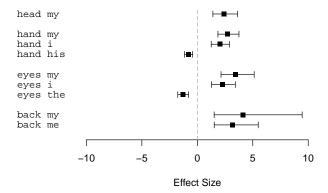
```
corpus_a_mod <- corpus_a
mys <- which(corpus_a_mod == 'my')</pre>
corpus\_a\_mod[sample(mys, floor(length(mys)*0.9))] <- \ 'CHIMERA')
corpus_a_mod_c <- surface(corpus_a_mod, span = '5LR')</pre>
results <- coco(corpus_a_mod_c, corpus_b_c, nodes = nodes, fdr = 0.01)
                                                                         p_value
               y H_A M_A H_B M_B effect_size CI_lower CI_upper
                                                                                    p_adjusted
    back CHIMERA
                                                     -Inf -2.281123 1.555671e-06 1.504334e-03
                                           -Inf
1:
                  18 1757
                             0 1947
                             0 1776
                                                     -Inf -2.865221 3.981028e-09 3.658564e-06
2:
    eves CHIMERA
                  26 1595
3:
                   5 1616
                           32 1744
                                       2.567485 1.192179
                                                          4.287936 1.695794e-05 7.792175e-03
    eyes
              my
    hand
         CHIMERA
                  48 2558
                             0 2493
                                           -Inf
                                                     -Inf -3.605990 9.176708e-15 1.051651e-11
5:
    hand
              mу
                   6 2600
                            43 2450
                                       2.926561
                                                  .684443
                                                          4.452541 1.854431e-08 1.062589e-05
6: hands CHIMERA
                  24 1387
                             0 1489
                                           -Inf
                                                     -Inf -2.687507 2.799724e-08 2.264976e-05
7: hands
              my
                   1 1410
                           24 1465
                                       4.528933 1.908262 9.886466 1.690464e-06 6.837928e-04
   head CHIMERA
                  26 2059
                             0 1937
                                           -Inf
                                                     -Inf -2.622318 4.050615e-08 2.171130e-05
8:
                   5 2080
                           40 1897
                                       3.132249 1.786789 4.834933 1.033929e-08 1.108372e-05
9:
    head
              mγ
```

Next a more realistic example (and the reason we chose that set of nodes). Here we check that the results indicate the different narative voice, third and first person, used in the two novels; the body part nouns are expected to be found in suspensions (Mahlberg, 2013).

```
results <- surface_coco(TTC, GE, span = '5LR', nodes = nodes, fdr = 0.01)
results
                                                       CI_upper p_value p_adjusted 5.4917238 9.754793e-07 9.423130e-04
         y H_A M_A H_B M_B effect_size
                                            CI lower
              3 1316
                     48 2355
1: back
                                  3.159998
                                            1.521928
        me
                      31 2372
                                  4.105901
2: back
        my
              1 1318
                                            1.517363
                                                       9.4521419 1.987134e-05 9.597855e-03
             10 1611
                      52 1724
                                  2.280107
                                            1.281850
                                                       3.4267531 2.247538e-07 6.869976e-05
3: eyes
4: eyes
              5 1616
                      58 1718
                                  3.446625
                                            2.137003
                                                       5.1270592 1.061195e-11 9.731159e-09
5: eyes the 120 1501
                                 -1.269288 -1.761782 -0.7909003 4.323172e-08 1.982175e-05
                      57 1719
6: hand his 175 2267 114 2543
                                 -0.783898 -1.147324 -0.4250235 1.158348e-05 4.413307e-03
             17 2425
                      74 2583
                                  2.030509
                                            1.250655
                                                       2.8889719 7.519299e-09 4.297280e-06
  hand
         i
             12 2430
                      85 2572
                                  2.742060
                                            1.858321
                                                       3.7535208 1.043073e-13 1.192232e-10
8: hand
        my
             9 1732
                      62 2219
                                  2.426331
                                            1.404175
                                                      3.6251454 3.575486e-08 3.822194e-05
```

and plot of the results (TTC is on the left)

```
plot(results)
```

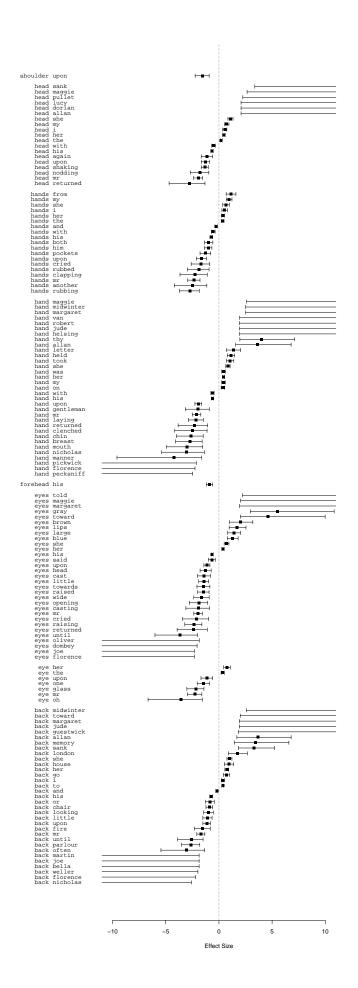


Finally we compare all of Dickens' novels against a set of 19th century novels to check if we can reproduce the observations from Mahlberg (2013) about Dickensian body language patterns. Practically we see this in terms such as *rubbing* co-occurring more frequently with *hands* in Dickens than the other 19th century novels.

```
DICKENS <- unlist(stri_extract_all_words(stri_trans_tolower(do.call(c, lapply(corpus_filepaths('DNov'), readLines)))))
NCNB <- unlist(stri_extract_all_words(stri_trans_tolower(do.call(c, lapply(corpus_filepaths('19C'), readLines)))))
results <- surface_coco(DICKENS, NCNB, span = '5LR', nodes = nodes, fdr = 0.01)
```

Here is a plot of the results; Dickens is on the left.

plot(results)



# References

Mahlberg, M. (2013). Corpus Stylistics and Dickens's Fiction. London: Routledge.