## **Proof of Concept**

CorporaCoCo v1.0-1 (2017-03-26)

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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 <a href="https://github.com/ravingmantis/CorporaCorpus">https://github.com/ravingmantis/CorporaCorpus</a>, 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 <a href="https://srtngi.ng/stringi.ng/

```
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')) ) ))</pre>
```

Choose the set of nodes.

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

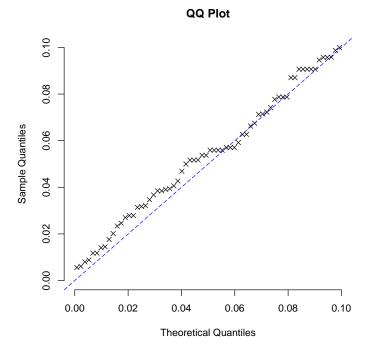
First we want to check that there are no significant results under the null. We create two corpora from alternate chunks of 1000 tokens of the two novels and check that there are no significant co-occurrence differences between our two sets of chunks.

```
chunks <- split(c(GE, TTC), ceiling(seq_along(c(GE, TTC)) / 1000))
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')
corpus_b_c <- surface(corpus_b, span = '5LR')
results <- coco(corpus_a_c, corpus_b_c, nodes = nodes, fdr = 0.01)
results</pre>
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]

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)</pre>
```



9: shoulder CHIMERA

my

19 425 0 334

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')
results <- coco(corpus_a_mod_c, corpus_b_c, nodes = nodes, fdr = 0.01)
results
                    y H_A M_A H_B M_B effect_size CI_lower CI_upper
                                                                                 p_value
                                                                                            p_adjusted
       back CHIMERA
                       17 1758
                                 0 1947
                                                 -Inf
                                                           -Inf -2.188961 3.278590e-06 3.170396e-03
       eyes CHIMERA
                       26 1595
                                  0 1776
                                                           -Inf -2.865221 3.981028e-09 3.658564e-06
2:
                                                 -Inf
3:
                        5 1616
                                32 1744
                                             2.567485 1.192179
                                                                 4.287936 1.695794e-05 7.792175e-03
       eves
                  mv
                                                           -Inf -3.541348 3.611265e-14 4.138510e-11
4:
       hand CHIMERA
                       46 2560
                                 0 2493
                                                 -Inf
       hand
                        8 2598
                                 43 2450
                                             2.510414
                                                      1.401183
                                                                 3.813061 2.570385e-07 1.472831e-04
                  my
      hands CHIMERA
                       21 1390
                                  0 1489
                                                 -Inf
                                                           -Inf -2.475588 2.490468e-07 2.014789e-04
                                                 -Inf
       head CHIMERA
                       28 2057
                                  0 1937
                                                           -Inf -2.738046 1.056226e-08 5.661373e-06
8:
                        3 2082
                                             3.871049 2.215923 6.209972 3.295074e-10 3.532319e-07
       head
                  mν
                                 40 1897
```

-Inf

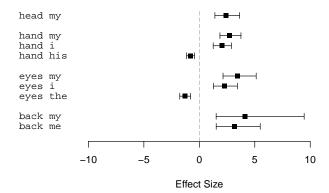
Next a more realistic example (and the reason we chose that set of nodes). Here we check that the results indicate the different narative voice, third and first person, used in the two novels; the body part nouns are expected to be found in suspensions (Mahlberg, 2013).

-Inf -1.859235 2.203334e-05 5.993069e-03

```
results <- surface_coco(TTC, GE, span = '5LR', nodes = nodes, fdr = 0.01)
results
                                                                                 p_adjusted
          y H_A
                M_A H_B M_B effect_size
                                            CI lower
                                                       CI_upper p_value p_adjusted 5.4917238 9.754793e-07 9.423130e-04
                      48 2355
              3 1316
1: back
                                  3.159998
                                            1.521928
        me
2: back
              1 1318
                      31 2372
                                  4.105901
                                            1.517363
                                                       9.4521419 1.987134e-05 9.597855e-03
        my
3: eyes
             10 1611
                      52 1724
                                  2.280107
                                            1.281850
                                                       3.4267531 2.247538e-07 6.869976e-05
         i
              5 1616
                      58 1718
                                  3.446625
                                            2.137003
                                                       5.1270592 1.061195e-11 9.731159e-09
  eyes
         my
  eyes the 120 1501
                      57 1719
                                 -1.269288 -1.761782
                                                      -0.7909003 4.323172e-08 1.982175e-05
6: hand his 175 2267 114 2543
                                 -0.783898 -1.147324 -0.4250235 1.158348e-05 4.413307e-03
                                                      2.8889719 7.519299e-09 4.297280e-06
             17 2425
                      74 2583
                                  2.030509
                                            1.250655
7: 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)

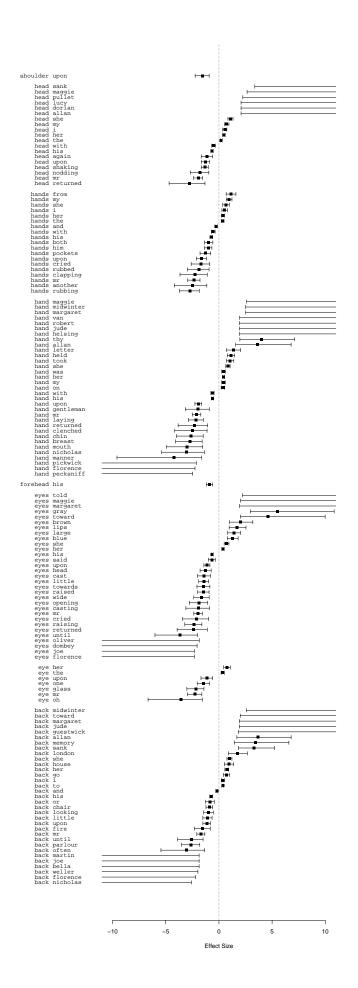


Finally we compare all of Dickens novels against a set of 19th century novels to check if we can reproduce the observations from Mahlberg (2013) that Dickens uses descriptions of body language more frequently that other authors of the time.

```
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. Routledge.