Using the koRpus Package for Corpus Analysis

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The R package tm.plugin.koRpus is an extension to the koRpus package, enhancing its usability for actual corpus analysis. It adds new classes and methods to koRpus, which are designed to work with complete text corpora in both koRpus and tm formats. This vignette gives you a quick overview.

1 What is tm.plugin.koRpus?

While the koRpus package focuses mostly on analysis steps of individual texts, tm.plugin.koRpus adds several new object classes and respective methods, which can be used to analyse complete text corpora in a single step. These classes are also a first step to combine object classes of both, the koRpus and tm packages.

There are three basic classes, which are hierarchically nested:

- class kRp.topicCorpus holds a list (named by topics) of objects of
 - class $\mathtt{kRp.sourcesCorpus},$ which in its sources slot holds a list of objects of
 - * class kRp.corpus, which in turn contains objects of both koRpus and

The idea behind this is to be able to categorize corpora on at least two levels. The default assumes that these levels are different *sources* and different *topics*, but apart from this naming (which is coded into the classes) you can actually use this for whatever levels you like.

If you don't need these levels, you can just use the function simpleCorpus() to create objects of class kRp.corpus. It represents a flat corpus of texts. To distinguish texts which came from different sources, use the function sourcesCorpus(), which will generate sub-corpora for each source given. And one level higher up, use the function topicCorpus(), to sort kRp.sourcesCorpus objects by different topics. Objects of this class will only be valid if there are texts of each topic from each source.

2 Tokenizing corpora

As with koRpus, the first step for text analysis is tokenizing and possibly POS tagging. This step is performed by the functions mentioned above, simpleCorpus(), sourcesCorpus(), or topicCorpus(), respectively. The package includes four sample texts taken from Wikipedia¹ in its tests directory we can use for an elaborate demonstration:

```
> library(tm.plugin.koRpus)
> # set tha root path to the sample files
> sampleRoot <- file.path(path.package("tm.plugin.koRpus"), "tests",
    "testthat", "samples")
> # now we can define the topics (names of the vector elements)
> # and their main path
> samplePaths <- c(
    C3S=file.path(sampleRoot, "C3S"),
    GEMA=file.path(sampleRoot, "GEMA")
> )
> # we also define the sources
> sampleSources <- c(
    wpa="Wikipedia_alt",
    wpn="Wikipedia_neu"
> )
> # and finally, we can tokenize all texts
> sampleTexts <- topicCorpus(paths=samplePaths, sources=sampleSources,
    tagger="tokenize", lang="de")
processing topic "C3S", source "Wikipedia_alt", 1 texts...
processing topic "C3S", source "Wikipedia_neu", 1 texts...
processing topic "GEMA", source "Wikipedia_alt", 1 texts...
processing topic "GEMA", source "Wikipedia_neu", 1 texts...
Should you need to get hold of the nested objects inside kRp.sourcesCorpus or
kRp.topicCorpus class objects, or replace them with updated ones, you can do so
by using the methods corpusTagged(), corpusSources(), or corpusTopics():
> allC3SSources <- corpusSources(corpusTopics(sampleTexts, "C3S"))</pre>
> names(allC3SSources)
[1] "wpa" "wpn"
```

3 Analysing corpora

After the initial tokenizing, we can analyse the corpus by calling the provided methods, for instance lexical diversity:

 $^{^{1}\}mathrm{see}\ \mathrm{the}\ \mathrm{file}\ \mathtt{tests/testthat/samples/License_of_sample_texts.txt}\ \mathrm{for}\ \mathrm{details}$

```
> sampleTexts <- lex.div(sampleTexts, char=FALSE, quiet=TRUE)
```

> corpusSummary(sampleTexts)

```
text topic source CTTR HD-D (vocd-D) Herdan's C Maas a Maas 1gV0
                                                                       6.21
wpaC3S01
          wpaC3S01
                    C3S
                           wpa 6.13
                                           38.14
                                                      0.95
                                                             0.16
wpnC3S01
          wpnC3S01
                    C3S
                           wpn 6.82
                                           38.05
                                                      0.94
                                                             0.17
                                                                       6.10
wpaGEMA01 wpaGEMA01 GEMA
                           wpa 7.07
                                           37.61
                                                      0.94
                                                             0.17
                                                                       6.11
wpnGEMA01 wpnGEMA01 GEMA
                           wpn 7.13
                                           37.87
                                                      0.94
                                                             0.16
                                                                       6.24
         MATTR MSTTR MTLD MTLD-MA Root TTR Summer TTR Uber index Yule's K
wpaC3S01
          0.81 0.79 100.16
                            NA 8.68 0.93 0.78
                                                           39.92
                                                                    49.92
wpnC3S01
          0.82 0.76 123.01
                               NA
                                     9.65 0.92 0.73
                                                           36.46
                                                                    54.88
wpaGEMA01 0.80 0.78 106.94
                               192
                                     10.00
                                             0.92 0.71
                                                           35.96
                                                                    65.08
                                                           37.47
                                                                    60.14
wpnGEMA01 0.81 0.79 111.64
                               NA
                                     10.08
                                             0.92 0.73
```

As you can see, corpusSummary() fetches a data frame from the object which contains the summarised results of all texts below the given object level. That is, if wou are only interested in the results for texts of the first topic, simply apply corpusSummary() on the result of corpusTopics():

> corpusSummary(corpusTopics(sampleTexts, "C3S"))

```
text topic source CTTR HD-D (vocd-D) Herdan's C Maas a Maas 1gV0
wpaC3S01 wpaC3S01
                  C3S
                         wpa 6.13
                                         38.14
                                                     0.95 0.16
                                                                     6.21
wpnC3S01 wpnC3S01
                  C3S
                         wpn 6.82
                                         38.05
                                                     0.94
                                                           0.17
                                                                     6.10
        MATTR MSTTR MTLD MTLD-MA Root TTR Summer TTR Uber index Yule's K
wpaC3S01 0.81 0.79 100.16
                               NA 8.68 0.93 0.78
                                                           39.92
                                                                   49.92
wpnC3S01 0.82 0.76 123.01
                                      9.65
                                            0.92 0.73
                                                           36.46
                                                                   54.88
                               NA
```

There are quite a number of corpus*() getter/setter methods for slots of these objects, e.g., corpusReadability() to get the readability() results from objects of class kRp.corpus.

Using the summary data.frame, you could now perform ANOVAs or plot interactions:

```
> library(sciplot)
> lineplot.CI(
+    x.factor=corpusSummary(sampleTexts)[["source"]],
+    response=corpusSummary(sampleTexts)[["MTLD"]],
+    group=corpusSummary(sampleTexts)[["topic"]],
+    type="l",
+    main="MTLD",
+    xlab="Media source",
+    ylab="Lexical diversity score",
+    col=c("grey", "black"),
+    lwd=2
+ )
```

MTLD

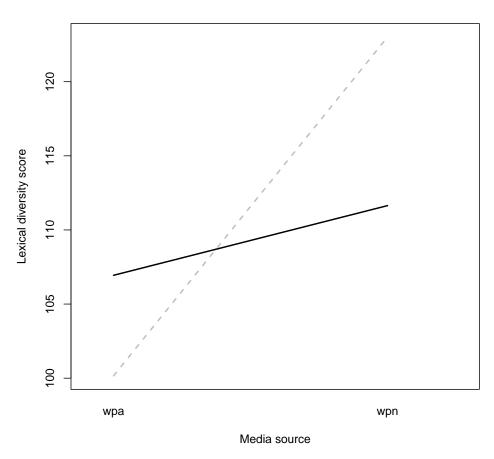


Figure 1: Well, the example texts aren't so impressive here, as there's not much variance in one text per source and topic.

3.1 Frequency analysis

The object classes make it quite comfortable to analyse type frequencies of corpora. There is a method read.corp.custom() for these classes, that will do this analysis recursively on all levels:

```
> sampleTexts <- read.corp.custom(sampleTexts, caseSens=FALSE)
```

> sampleTextsWordFreq <- query(corpusFreq(sampleTexts), var="wclass",

⁺ query=kRp.POS.tags(lang="de", list.classes=TRUE, tags="words"))

> head(sampleTextsWordFreq, 10)

	num	word	lemma	tag	wclass	lttr	freq	pct	pmio	log10
3	3	die		word.kRp	word	3	30	0.037220844	37220	4.570776
4	4	der		word.kRp	word	3	21	0.026054591	26054	4.415874
5	5	gema		${\tt word.kRp}$	word	4	17	0.021091811	21091	4.324097
6	6	und		${\tt word.kRp}$	word	3	17	0.021091811	21091	4.324097
7	7	einer		${\tt word.kRp}$	word	5	12	0.014888337	14888	4.172836
8	8	von		${\tt word.kRp}$	word	3	12	0.014888337	14888	4.172836
11	11	ist		${\tt word.kRp}$	word	3	10	0.012406948	12406	4.093632
12	12	bei		${\tt word.kRp}$		3		0.011166253	11166	4.047898
13	13	das		${\tt word.kRp}$	word	3	8	0.009925558	9925	3.996731
14	14 ι	ırheber		${\tt word.kRp}$	word	7	8	0.009925558	9925	3.996731
	${\tt rank}$.avg ran	nk.min	rank.rel	avg ran	nk.rel	l.min	inDocs		idf
3	263.0		263	99.24528		99.24528		4		0
4	26	32.0	262	98.86	5792	98.8	36792	4		0
5	26	30.5	260	98.30	189	98.3	11321	4		0
6	26	30.5	260	98.30	189	98.3	11321	4		0
7	25	58.5	258	97.54	1717	97.3	35849	4		0
8	25	58.5	258	97.54	1717	97.3	35849	4		0
11	25	56.0	255	96.60	377	96.2	22642	4		0
12	25	54.0	254	95.84	1906	95.8	34906	4		0
13	2	52.5	252	95.28	3302	95.0	09434	4		0
14	25	52.5	252	95.28	3302	95.0	09434	2 0.30	1029995	663981

In combination with the wordcloud package, this can directly be used to plot word clouds:

```
> require(wordcloud)
> colors <- brewer.pal(8, "RdGy")
> wordcloud(
+    head(sampleTextsWordFreq[["word"]], 200),
+    head(sampleTextsWordFreq[["freq"]], 200),
+    random.color=TRUE,
+    colors=colors
+ )
```



Figure 2: The 200 most frequent words in the example corpus