# Package 'stylo'

October 19, 2014

2 assign.plot.colors

Index		47
	zeta.eder	45
	zeta.craig	
	zeta.chisquare	
	txt.to.words.ext	
	txt.to.words	
	txt.to.features	
	stylo.default.settings	
	stylo	
	rolling.delta	
	rolling.classify	
	perform.svm	29
	perform.nsc	
	perform.naivebayes	
	perform.knn	
	perform.delta	
	oppose	
	make.table.of.frequencies	
	make.samples	
	make.ngrams	

assign.plot.colors Assign

Assign colors to samples

### Description

Function that assigns unique colors to each class represented in a corpus: used for graph auto-coloring.

### Usage

```
assign.plot.colors(labels, col = "colors", opacity = 1)
```

### **Arguments**

labels

a vector containing the names of the samples in a corpus; it is obligatory to use an underscore as a class delimiter. Consider the following examples: c("Sterne\_Tristram", "Sterne\_Sentimental", "Fielding\_Tom", ...), where the classes are the authors' names, and c("M\_Joyce\_Dubliners", "F\_Woolf\_Night\_and\_day", "M\_Conrad\_Lord\_Jim", ...), where the classes are M(ale) and F(emale) according to authors' gender. Note that only the part up to the first underscore in the sample's name will be included in the class label.

col

an optional argument specifying the color palette to be used: "colors" for full-color output (default), "greyscale" for greyscale (useful for preparing publishable pictures), and "black", if no colors should be used.

classify

opacity

optional argument to set transparency/opacity of the colors. 0 means full transparency, 1 means full opacity (default).

### **Details**

Function for graph auto-coloring; depending on the user's choice it assigns either colors or greyscale tones to matching strings of characters which stand for class identifiers. These metadata will typically be encoded in the texts' filenames. (As class delimiter, the underscore character should be used). Alternatively, all labels can be plotted in black.

#### Value

The function returns a vector of colors, using their conventional names (e.g. red, maroon4, mediumturquoise, gold4, deepskyblue, ...), or numeric values if the greyscale option was chosen (e.g. #000000, #000000, #595959, #B2B2B2, ...).

### Author(s)

Maciej Eder

### **Examples**

classify

Machine-learning classification

### **Description**

Function that performs a number of machine-learning methods for classification used in computational stylistics: Delta (Burrows, 2002), k-Nearest Neighbors, Support Vector Machines, Naive Bayes, and Nearest Shrunken Centroids (Jockers and Witten, 2010). Most of the options are derived from the stylo function.

### Usage

4 classify

#### **Arguments**

gui

an optional argument; if switched on, a simple yet effective graphical user interface (GUI) will appear. Default value is TRUE.

training.frequencies

using this optional argument, one can load a custom table containing frequencies/counts for several variables, e.g. most frequent words, across a number of text samples (for the training set). It can be either an R object (matrix or data frame), or a filename containing tab-delimited data. If you use an R object, make sure that the rows contain samples, and the columns – variables (words). If you use an external file, the variables should go vertically (i.e. in rows): this is because files containing vertically-oriented tables are far more flexible and easily editable using, say, Excel or any text editor. To flip your table horizontally/vertically use the generic function t().

test.frequencies

using this optional argument, one can load a custom table containing frequencies/counts for the test set. Further details: immediately above.

training.corpus

another option is to pass a pre-processed corpus as an argument (here: the training set). It is assumed that this object is a list, each element of which is a vector containing one tokenized sample. The example shown below will give you some hints how to prepare such a corpus. Also, refer to help(load.corpus.and.parse)

test.corpus

if training.corpus is used, then you should also prepare a similar R object containing the test set.

features

usually, a number of the most frequent features (words, word n-grams, character n-grams) are extracted automatically from the corpus, and they are used as variables for further analysis. However, in some cases it makes sense to use a set of tailored features, e.g. the words that are associated with emotions or, say, a specific subset of function words. This optional argument allows to pass either a filename containing your custom list of features, or a vector (R object) of features to be assessed.

path

if not specified, the current directory will be used for input/output procedures (reading files, outputting the results).

training.corpus.dir

the subdirectory (within the current working directory) that contains the training set, or the collection of texts used to exemplify the differences between particular classes (e.g. authors or genres). The discriminating features extracted from this training material will be used during the testing procedure (see below). If not specified, the default subdirectory primary\_set will be used.

test.corpus.dir

the subdirectory (within the working directory) that contains the test set, or the collection of texts that are used to test the effectiveness of the discriminative features extracted from the training set. In the case of authorship attribution e.g., this set might contain works of non-disputed authorship, in order to check whether a classification procedure attribute the tets texts to their correct author. This set contains 'new' or 'unseen' data (e.g. anonymous samples or samples of disputed authorship in the case of authorship studies). If not specified, the default subdirectory secondary\_set will be used.

classify 5

... any variable as produced by stylo.default.settings() can be set here to overwrite the default values.

#### **Details**

There are numerous additional options that are passed to this function; so far, they are all loaded when stylo.default.settings() is executed (it will be invoked automatically from inside this function); the user can set/change them in the GUI.

#### Value

The function returns an object of the class stylo.results: a list of variables, including tables of word frequencies, vector of features used, a distance table and some more stuff. Additionally, depending on which options have been chosen, the function produces a number of files used to save the results, features assessed, generated tables of distances, etc.

#### Author(s)

Maciej Eder, Mike Kestemont, Jan Rybicki

#### References

Eder, M. Kestemont, M. and Rybicki, J. (2013). Stylometry with R: a suite of tools. In: "Digital Humanities 2013: Conference Abstracts". University of Nebraska-Lincoln, Lincoln, NE, pp. 487-89.

Burrows, J. F. (2002). "Delta": a measure of stylistic difference and a guide to likely authorship. "Literary and Linguistic Computing", 17(3): 267-87.

Jockers, M. L. and Witten, D. M. (2010). A comparative study of machine learning methods for authorship attribution. "Literary and Linguistic Computing", 25(2): 215-23.

Argamon, S. (2008). Interpreting Burrows's Delta: geometric and probabilistic foundations. "Literary and Linguistic Computing", 23(2): 131-47.

### See Also

```
stylo, rolling.delta, oppose
```

### **Examples**

6 define.plot.area

```
txt2 = c("what", "do", "you", "read", "my", "lord")
  setTRAIN = list(txt1, txt2)
  names(setTRAIN) = c("hamlet_sample1", "polonius_sample1")
txt4 = c("to", "be", "or", "not", "to", "be")
txt5 = c("though", "this", "be", "madness", "yet", "there", "is", "method")
txt6 = c("the", "rest", "is", "silence")
  setTEST = list(txt4, txt5, txt6)
  names(setTEST) = c("hamlet_sample2", "polonius_sample2", "uncertain_1")
classify(training.corpus = setTRAIN, test.corpus = setTEST)
# using a custom set of features (words, n-grams) to be analyzed:
my.selection.of.function.words = c("the", "and", "of", "in", "if", "into",
                                   "within", "on", "upon", "since")
classify(features = my.selection.of.function.words)
# loading a custom set of features (words, n-grams) from a file:
classify(features = "wordlist.txt")
# batch mode, custom name of corpus directories:
my.test = classify(gui = FALSE, training.corpus.dir = "TrainingSet",
       test.corpus.dir = "TestSet")
summary(my.test)
# batch mode, character 3-grams requested:
classify(gui = FALSE, analyzed.features = "c", ngram.size = 3)
## End(Not run)
```

define.plot.area

Define area for scatterplots

### Description

Function that determines the size of a scatterplot, taking into consideration additional margin to fit longer labels appearing on a graph (if applicable), optional margin defined by user, and some space to offset scatterplot labels from points (if applicable).

### Usage

```
define.plot.area(x.coord, y.coord, xymargins = 2, v.offset = 0)
```

### Arguments

x.coord

a vector of x coordinates, optionally with names.

delete.markup 7

y.coord a vector of y coordinates.

xymargins additional margins (expressed as a % of the actual plot area).

v.offset label offset (expressed as a % of the actual plot area).

#### **Details**

Function that finds out the coordinates of scatterplots: it computes the extreme x and y values, adds margins, and optionally extends the top margin if a plot uses sample labels. Automatic margin extension will only take place if the x coordinates are supplemented by their names (i.e. labels of points to be shown on scatterplot).

### Author(s)

Maciej Eder

#### See Also

```
assign.plot.colors, stylo
```

### **Examples**

```
# to determine the plotting area for 4 points: define.plot.area( c(1,2,3,4), c(-0.001,0.11,-0.023,0.09))

# to determine plot coordinates, taking into consideration

# the objects' names

my.points = cbind(c(1,2,3,4),c(-0.001,0.11,-0.023,0.09))

rownames(my.points) = c("first","second","third","very_long_fourth")

define.plot.area(my.points[,1], my.points[,2])
```

delete.markup

Delete HTML or XML tags

### **Description**

Function for removing markup tags (e.g. HTML, XML) from a string of characters. All XML markup is assumed to be compliant with the TEI guidelines (http://www.tei-c.org/).

### Usage

```
delete.markup(input.text, markup.type = "plain")
```

8 delete.stop.words

### Arguments

input.text any string of characters (e.g. vector) containing markup tags that have to be

deleted.

markup.type any of the following values: plain (nothing will happen), html (all <tags>

will be deleted as well as HTML header), xml (TEI header, all strings between <note> </note> tags, and all the tags will be deleted), xml.drama (as above; but, additionally, speaker's names will be deleted, or strings within each the <speaker> </speaker> tags), xml.notitles (as above; but, additionally, all the chapter/section (sub)titles will be deleted, or strings within each the <head>

</head> tags).

### **Details**

This function needs to be used carefully: while a document formatted in compliance with the TEI guidelines will be parsed flawlessly, the cleaning up of an HTML page harvested randomly on the web might cause some side effects, e.g. the footers, disclaimers, etc. will not be removed.

### Author(s)

Maciej Eder, Mike Kestemont

#### See Also

```
load.corpus, txt.to.words, txt.to.words.ext, txt.to.features
```

### **Examples**

delete.stop.words

Exclude stop words (e.g. pronouns, particles, etc.) from a dataset

### Description

Function for removing custom words from a dataset: it can be the so-called stop words (frequent words without much meaning), or personal pronouns, or other custom elements of a dataset. It can be used to cull certain words from a vector containing tokenized text (particular words as elements of the vector), or to exclude unwanted columns (variables) from a table with frequencies. See examples below.

delete.stop.words 9

### Usage

```
delete.stop.words(input.data, stop.words = NULL)
```

### **Arguments**

either a vector containing words (actually, any countable features), or a data matrix/frame. The former in case of culling stop words from running text, the latter for culling them from tables of frequencies (then particular columns are excluded). The table should be oriented to contain samples in rows, variables in columns, and variables' names should be accessible via colnames(input.table). stop.words

a vector of words to be excluded.

#### Details

This function might be usefull to perform culling, or automatic deletion of the words that are too characteristic for particular texts. See help(culling) for further details.

### Author(s)

Maciej Eder

#### See Also

```
stylo.pronouns, perform.culling
```

### **Examples**

```
# (i) excluding stop words from a vector
my.text = c("omnis", "homines", "qui", "sese", "student", "praestare",
        "ceteris", "animalibus", "summa", "ope", "niti", "decet", "ne",
        "vitam", "silentio", "transeant", "veluti", "pecora", "quae",
        "natura", "prona", "atque", "ventri", "oboedientia", "finxit")
delete.stop.words(my.text, stop.words = c("qui", "quae", "ne", "atque"))
# (ii) excluding stop words from tabular data
# assume there is a matrix containing some frequencies
# (be aware that these counts are fictional):
t1 = c(2, 1, 0, 8, 9, 5, 6, 3, 4, 7)
t2 = c(7, 0, 5, 9, 1, 8, 6, 4, 2, 3)
t3 = c(5, 9, 2, 1, 6, 7, 8, 0, 3, 4)
t4 = c(2, 8, 6, 3, 0, 5, 9, 4, 7, 1)
my.data.table = rbind(t1, t2, t3, t4)
# names of the samples:
rownames(my.data.table) = c("text1", "text2", "text3", "text4")
# names of the variables (e.g. words):
colnames(my.data.table) = c("the", "of", "in", "she", "me", "you",
                                    "them", "if", "they", "he")
# the table looks as follows
```

10 gui.classify

gui.classify

GUI for the function classify

### **Description**

Graphical user interface for classify. Via the GUI, this function can set most of the variables needed for classify.

### Usage

```
gui.classify(...)
```

### **Arguments**

any variable as produced by stylo.default.settings can be set here to overwrite the default values.

### **Details**

The function calls stylo.default.settings to initialize a number of default variables. Then it reads the file classify\_config.txt (if the file exists and can be found in the current directory) to overwrite any default values. Then a GUI box appears, allowing the variables' customization by the user. Refer to HOWTO available at https://sites.google.com/site/computationalstylistics/for a detailed explanation what the particular variables are for and how to use them.

### Value

The function returns a list containing ca. 100 variables.

### Author(s)

Jan Rybicki, Maciej Eder

```
classify, gui.stylo
```

gui.oppose 11

### **Examples**

```
## Not run:
gui.classify()

my.variables = gui.classify()
summary(my.variables)

## End(Not run)
```

gui.oppose

GUI for the function oppose

### **Description**

Graphical user interface for oppose. This function sets most of the variables needed for oppose.

### Usage

```
gui.oppose(...)
```

### **Arguments**

... any variable as produced by stylo.default.settings can be set here to overwrite the default values.

### **Details**

The function calls stylo.default.settings to initialize a number of default variables. Then it reads the file oppose\_config.txt (if the file exists and can be found in the current directory) to overwrite any default values. Then a GUI box appears, allowing the variables' customization by the user. Refer to HOWTO available at https://sites.google.com/site/computationalstylistics/for a detailed explanation what the particular variables are for and how to use them.

### Value

The function returns a list containing ca. 100 variables.

### Author(s)

```
Jan Rybicki, Maciej Eder
```

```
oppose, stylo.default.settings
```

12 gui.stylo

### **Examples**

```
## Not run:
gui.oppose()

my.variables = gui.oppose()
summary(my.variables)

## End(Not run)
```

gui.stylo

GUI for stylo

### **Description**

Graphical user interface for the function stylo. This function sets most of the variables needed for stylo.

### Usage

```
gui.stylo(...)
```

### Arguments

any variable as produced by stylo.default.settings can be set here to overwrite the default values.

### **Details**

The function calls stylo.default.settings to initialize a number of default variables. Then it reads the file stylo\_config.txt (if the file exists and can be found in the current directory) to overwrite any default values. Then a GUI box appears, allowing the variables' customization by the user. Refer to HOWTO available at https://sites.google.com/site/computationalstylistics/for a detailed explanation what the particular variables are for and how to use them.

### Value

The function returns a list containing ca. 100 variables.

### Author(s)

Jan Rybicki, Maciej Eder

```
stylo, stylo.default.settings
```

load.corpus 13

### **Examples**

```
## Not run:
gui.stylo()

my.variables = gui.stylo()
summary(my.variables)

## End(Not run)
```

load.corpus

Load text files

### **Description**

Function for loading text files from a specified directory.

### **Usage**

```
load.corpus(files, corpus.dir = "", encoding = "native.enc")
```

### **Arguments**

files a vector of file names.

corpus.dir a directory containing the text files to be loaded; if not specified, the current

working directory will be used.

encoding useful if you use Windows and non-ASCII alphabets: French, Polish, Hebrew,

etc. In such a situation, it is quite convenient to convert your text files into Unicode and to set this option to encoding = "UTF-8". In Linux and Mac, you are always expected to use Unicode, thus you don't need to set anything.

### Value

The function returns an object of the class stylo.corpus. It is a list containing as elements the texts loaded.

### Author(s)

Maciej Eder

```
stylo, classify, rolling.delta, oppose, txt.to.words
```

### **Examples**

load.corpus.and.parse Load text files and perform pre-processing

### Description

A high-level function that controls a number of other functions responsible for loading texts from files, deleting markup, sampling from texts, converting samples to n-grams, etc. It is build on top of a number of functions and thus it requires a large number of arguments. The only obligatory argument, however, is a vector containing the names of the files to be loaded.

### Usage

#### **Arguments**

files a vector of file names.

corpus.dir the directory containing the text files to be loaded; if not specified, the current

directory will be used.

markup. type choose one of the following values: plain (nothing will happen), html (all tags

will be deleted as well as HTML header), xml (TEI header, any text between <note> </note> tags, and all the tags will be deleted), xml.drama (as above; additionally, speaker's names will be deleted, or strings within the <speaker> </speaker> tags), xml.notitles (as above; but, additionally, all the chapter/section (sub)titles will be deleted, or strings within each the <head> </head> tags); see

delete.markup for further details.

language an optional argument indicating the language of the texts analyzed; the val-

ues that will affect the function's behavior are: English.contr, English.all, Latin.corr (type help(txt.to.words.ext) for explanation). The default

value is English.

load.corpus.and.parse 15

splitting.rule if you are not satisfied with the default language settings (or your input string

of characters is not a regular text, but a sequence of, say, dance movements represented using symbolic signs), you can indicate your custom splitting regular expression here. This option will overwrite the above language settings. For

further details, refer to help(txt.to.words).

sample.size desired size of samples, expressed in number of words; default value is 10,000.

sampling one of three values: no.sampling (default), normal.sampling, random.sampling.

See make.samples for explanation.

sample.overlap if this opion is used, a reference text is segmented into consecutive, equal-sized

samples that are allowed to partially overlap. If one specifies the sample.size parameter of 5,000 and the sample.overlap of 1,000, for example, the first sample of a text contains words 1-5,000, the second 4001-9,000, the third sam-

ple 8001–13,000, and so forth.

number.of.samples

optional argument which will be used only if random. sampling was chosen; it

is self-evident.

sampling.with.replacement

optional argument which will be used only if random. sampling was chosen; it

specifies the method used to randomly harvest words from texts.

features an option for specifying the desired type of features: w for words, c for characters

(default: w). See txt.to.features for further details.

ngram. size an optional argument (integer) specifying the value of n, or the size of n-grams

to be produced. If this argument is missing, the default value of 1 is used. See

txt.to.features for further details.

preserve.case whether of not to lowercase all characters in the corpus (default = F).

encoding useful if you use Windows and non-ASCII alphabets: French, Polish, Hebrew,

etc. In such a situation, it is quite convenient to convert your text files into Unicode and to set this option to encoding = "UTF-8". In Linux and Mac, you

are always expected to use Unicode, thus you don't need to set anything.

#### Value

The function returns an object of the class stylo.corpus. It is a list containing as elements the samples (entire texts or sampled subsets) split into words/characters and combined into n-grams (if applicable).

### Author(s)

Maciej Eder

### See Also

load.corpus, delete.markup, txt.to.words, txt.to.words.ext, txt.to.features, make.samples

16 make.ngrams

### **Examples**

make.ngrams

Make text n-grams

### **Description**

Function that combines a vector of text units (words, characters, POS-tags, other features) into pairs, triplets, or longer sequences, commonly referred to as n-grams.

### Usage

```
make.ngrams(input.text, ngram.size = 1)
```

### **Arguments**

input.text a vector containing words or characters to be parsed into n-grams.

ngram.size an optional argument (integer) indicating the value of n, or the size of n-grams

to be produced. If this argument is missing, default value of 1 is used.

### **Details**

Function for combining series of items (e.g. words or characters) into n-grams, or strings of n elements. E.g. character 2-grams of the sentence "This is a sentence" are as follows: "th", "hi", "is", "s ", "i", "is", "s ", "a", "a ", "s", "se", "en", "nt", "te", "en", "nc", "ce". Character 4-grams would be, of course: "this", "his ", "is a", "s a ", " a s", etc. Word 2-grams: "this is", "is a", "a sentence". The issue whether using n-grams of items increases the accuracy of stylometric procedures has been heavily debated in the secondary literature (see the reference section for further reading). Eder (2013) e.g. shows that character n-grams are suprisingly robust for dealing with noisy corpora (in terms of a high number of misspelled characters).

### Author(s)

Maciej Eder

make.samples 17

#### References

Alexis, A., Craig, H., and Elliot, J. (2014). Language chunking, data sparseness, and the value of a long marker list: explorations with word n-grams and authorial attribution. "Literary and Linguistic Computing", 29, advanced access (doi: 10.1093/llc/fqt028).

Eder, M. (2011). Style-markers in authorship attribution: a cross-language study of the authorial fingerprint. "Studies in Polish Linguistics", 6: 99-114. http://www.wuj.pl/page,art,artid, 1923.html.

Eder, M. (2013). Mind your corpus: systematic errors in authorship attribution. "Literary and Linguistic Computing", 28(4): 603-14.

Hoover, D. L. (2002). Frequent word sequences and statistical stylistics. "Literary and Linguistic Computing", 17: 157-80.

Hoover, D. L. (2003). Frequent collocations and authorial style. "Literary and Linguistic Computing", 18: 261-86.

Hoover, D. L. (2012). The rarer they are, the more they are, the less they matter. In: Digital Humanities 2012: Conference Abstracts, Hamburg University, Hamburg, pp. 218-21.

Koppel, M., Schler, J. and Argamon, S. (2009). Computational methods in authorship attribution. "Journal of the American Society for Information Science and Technology", 60(1): 9-26.

Stamatatos, E. (2009). A survey of modern authorship attribution methods. "Journal of the American Society for Information Science and Technology", 60(3): 538-56.

#### See Also

```
txt.to.words, txt.to.words.ext, txt.to.features
```

### **Examples**

```
# Consider the string my.text:
my.text = "Quousque tandem abutere, Catilina, patientia nostra?"
# which can be split into a vector of consecutive words:
my.vector.of.words = txt.to.words(my.text)
# now, we create a vector of word 2-grams:
make.ngrams(my.vector.of.words, ngram.size = 2)
# similarly, you can produce character n-grams:
my.vector.of.chars = txt.to.features(my.vector.of.words, features = "c")
make.ngrams(my.vector.of.chars, ngram.size = 4)
```

make.samples

Split text to samples

### Description

Function that either splits an input text (a vector of linguistic items, such as words, word n-grams, character n-grams, etc.) into equal-sized samples of a desired length (expressed in words), or excerpts randomly a number of words from the original text.

18 make.samples

### Usage

### **Arguments**

tokenized.text input textual data stored either in a form of vector (single text), or as a list of

vectors (whole corpus); particular vectors should contain tokenized data, i.e.

words, word n-grams, or other features, as elements.

sample.size desired size of sample expressed in number of words; default value is 10,000.

sampling one of three values: no.sampling (default), normal.sampling, random.sampling.

sample.overlap if this opion is used, a reference text is segmented into consecutive, equal-sized

samples that are allowed to partially overlap. If one specifies the sample.size parameter of 5,000 and the sample.overlap of 1,000, for example, the first sample of a text contains words 1-5,000, the second 4001-9,000, the third sam-

ple 8001-13,000, and so forth.

number.of.samples

optional argument which will be used only if random.sampling was chosen; it is self-evident.

sampling.with.replacement

optional argument which will be used only if random. sampling was chosen; it specifies the method to randomly harvest words from texts.

### Details

Normal sampling is probably a good choice when the input texts are long: the advantage is that one gets a bigger number of samples which, in a way, validate the results (when several independent samples excerpted from one text are clustered together). When the analyzed texts are significantly unequal in length, it is not a bad idea to prepare samples as randomly chosen "bags of words". For this, set the sampling variable to random.sampling. The desired size of the sample should be specified via the sample.size variable. Sampling with and without replacement is also available. It has been shown by Eder (2010) that harvesting random samples from original texts improves the performance of authorship attribution methods.

### Author(s)

Mike Kestemont, Maciej Eder

#### References

Eder, M. (2014). Does size matter? Authorship attribution, short samples, big problem. "Literary and Linguistic Computing", 29, advanced access (doi:10.1093/llc/fqt066).

```
txt.to.words, txt.to.words.ext, txt.to.features, make.ngrams
```

### **Examples**

```
my.text = "Arma virumque cano, Troiae qui primus ab oris
          Italiam fato profugus Laviniaque venit
          litora, multum ille et terris iactatus et alto
          vi superum, saevae memorem Iunonis ob iram,
          multa quoque et bello passus, dum conderet urbem
           inferretque deos Latio; genus unde Latinum
           Albanique patres atque altae moenia Romae.
           Musa, mihi causas memora, quo numine laeso
           quidve dolens regina deum tot volvere casus
           insignem pietate virum, tot adire labores
           impulerit. tantaene animis caelestibus irae?"
my.words = txt.to.words(my.text)
# split the above text into samples of 20 words:
make.samples(my.words, sampling = "normal.sampling", sample.size = 20)
# excerpt randomly 50 words from the above text:
make.samples(my.words, sampling = "random.sampling", sample.size = 50)
# excerpt 5 random samples from the above text:
make.samples(my.words, sampling = "random.sampling", sample.size = 50,
             number.of.samples = 5)
```

make.table.of.frequencies

Prepare a table of (relative) word frequencies

### **Description**

Function that collects several frequency lists and combines them into a single frequency table. To this end a number of rearrangements inside particular lists are carried out. The table is produced using a reference list of words/features (passed as an argument).

### Usage

### Arguments

corpus

textual data: either a corpus (represented as a list), or a single text (represented as a vector); the data have to be split into words (or other features, such as character n-grams or word pairs).

features

a vector containing a reference feature list that will be used to build the table of frequencies (it is assumed that the reference list contains the same type of features as the corpus list, e.g. words, character n-grams, word pairs, etc.; otherwise, an empty table will be build).

absent.sensitive

this optional argument is used to prevent building tables of words/features that never occur in the corpus. When switched on (default), variables containing 0 values across all samples, will be excluded. However, in some cases this is important to keep all the variables regardless of their values. This is e.g. the case when comparing two corpora: even if a given word did not occur in corpus A, it might be present in corpus B. In short: whenever you perform any analysis involving two or multiple sets of texts, switch this option to FALSE.

relative

when this argument is switched to TRUE (default), relative frequencies are computed instead of raw frequencies.

#### Author(s)

Maciej Eder

#### See Also

load.corpus, load.corpus.and.parse

### **Examples**

```
# to get frequencies of the words "a", "the" and "of" from a text:
sample.txt = txt.to.words("My father had a small estate
             in Nottinghamshire: I was the third of five sons.")
make.table.of.frequencies(sample.txt, c("a", "the", "of"))
# to get a table of frequencies across several texts:
txt.1 = "Gallia est omnis divisa in partes tres, quarum unam incolunt
   Belgae, aliam Aquitani, tertiam qui ipsorum lingua Celtae, nostra
    Galli appellantur."
txt.2 = "Si quis antea, iudices, mirabatur quid esset quod, pro tantis
    opibus rei publicae tantaque dignitate imperi, nequaquam satis multi
    cives forti et magno animo invenirentur qui auderent se et salutem
    suam in discrimen offerre pro statu civitatis et pro communi
   libertate, ex hoc tempore miretur potius si quem bonum et fortem
    civem viderit, quam si quem aut timidum aut sibi potius quam rei
   publicae consulentem."
txt.3 = "Nam mores et instituta vitae resque domesticas ac familiaris
   nos profecto et melius tuemur et lautius, rem vero publicam nostri
   maiores certe melioribus temperaverunt et institutis et legibus."
my.corpus.raw = list(txt.1, txt.2, txt.3)
my.corpus.clean = lapply(my.corpus.raw, txt.to.words)
my.favorite.words = c("et", "in", "se", "rara", "avis")
make.table.of.frequencies(my.corpus.clean, my.favorite.words)
# to include all words in the reference list, no matter if they
# occurred in the corpus:
```

oppose 21

oppose

Contrastive analysis of texts

### **Description**

Function that performs a contrastive analysis between two given sets of texts. It generates a list of words significantly preferred by a tested author (or, a collection of authors), and another list containing the words significantly avoided by the former when compared to another set of texts. Some visualizations are available.

### Usage

### **Arguments**

gui

an optional argument; if switched on, a simple yet effective graphical interface (GUI) will appear. Default value is TRUE.

path

if not specified, the current working directory will be used for input/output procedures (reading files, outputting the results, etc.).

primary.corpus.dir

the subdirectory (within the current working directory) that contains one or more texts to be compared to a comparison corpus. These texts can e.g. be the oeuvre by author A (to be compared to the oeuvre of another author B) or a collection of texts by female authors (to be contrasted with texts by male authors). If not specified, the default subdirectory primary\_set will be used.

22 oppose

secondary.corpus.dir

the subdirectory (within the current working directory) that contains a comparison corpus: a pool of texts to be contrasted with texts from the primary.corpus. If not specified, the default subdirectory secondary\_set will be used.

test.corpus.dir

the subdirectory (within the current working directory) that contains texts to verify the discriminatory strength of the features extracted from the primary.set and secondary.sets. Ideally, the test.corpus.dir should contain texts known to belong to both classes (e.g. texts written by female and male authors in the case of a gender-oriented study). If not specified, the default subdirectory test\_set will be used. If the default subdirectory does not exist or does not contain any texts, the validation test will not be performed.

primary.corpus

another option is to pass a pre-processed corpus as an argument (here: the primary set). It is assumed that this object is a list, each element of which is a vector containing one tokenized sample. Refer to help(load.corpus.and.parse) to get some hints how to prepare such a corpus.

secondary.corpus

if primary.corpus is used, then you should also prepare a similar R object containing the secondary set.

test.corpus

if you decide to use test corpus, you can pass it as a pre-processed R object using this argument.

. . .

any variable produced by stylo.default.settings can be set here, in order to overwrite the default values.

### **Details**

This function performs a contrastive analysis between two given sets of texts, using Burrows's Zeta (2007) in its different flavors, including Craig's extensions (Craig and Kinney, 2009). Also, the Whitney-Wilcoxon procedure as introduced by Kilgariff (2001) is available. The function generates a vector of words significantly preferred by a tested author, and another vector containing the words significantly avoided.

### Value

The function returns an object of the class stylo.results: a list of variables, including a list of words significantly preferred in the primary set, words significantly avoided (or, preferred in the secondary set), and possibly some other results, if applicable.

#### Author(s)

Maciej Eder, Mike Kestemont, Jan Rybicki

### References

Eder, M. Kestemont, M. and Rybicki, J. (2013). Stylometry with R: a suite of tools. In: "Digital Humanities 2013: Conference Abstracts". University of Nebraska-Lincoln, Lincoln, NE, pp. 487-89.

perform.culling 23

Burrows, J. F. (2007). All the way through: testing for authorship in different frequency strata. "Literary and Linguistic Computing", 22(1): 27-48.

Craig, H. and Kinney, A. F., eds. (2009). Shakespeare, Computers, and the Mystery of Authorship. Cambridge: Cambridge University Press.

Hoover, D. (2010). Teasing out authorship and style with t-tests and Zeta. In: "Digital Humanities 2010: Conference Abstracts". King's College London, pp. 168-170.

Kilgariff A. (2001). Comparing Corpora. "International Journal of Corpus Linguistics" 6(1): 1-37.

#### See Also

```
stylo, classify, rolling.delta
```

### **Examples**

perform.culling

Exclude variables (e.g. words, n-grams) from a frequency table that are too characteristic for some samples

### Description

Culling refers to the automatic manipulation of the wordlist (proposed by Hoover 2004a, 2004b). The culling values specify the degree to which words that do not appear in all the texts of a corpus will be removed. A culling value of 20 indicates that words that appear in at least 20% of the texts in the corpus will be considered in the analysis. A culling setting of 0 means that no words will be removed; a culling setting of 100 means that only those words will be used in the analysis that appear in all texts of the corpus at least once.

### Usage

```
perform.culling(input.table, culling.level = 0)
```

### **Arguments**

input.table a matrix or data frame containing frequencies of words or any other countable features; the table should be oriented to contain samples in rows, variables in

columns, and variables' names should be accessible via colnames (input.table).

culling.level percentage of samples that need to have a given word in order to prevent this

word from being culled (see the description above).

24 perform.delta

### Author(s)

Maciej Eder

#### References

```
Hoover, D. (2004a). Testing Burrows's Delta. "Literary and Linguistic Computing", 19(4): 453-75. Hoover, D. (2004b). Delta prime. "Literary and Linguistic Computing", 19(4): 477-95.
```

### See Also

```
delete.stop.words, stylo.pronouns
```

### **Examples**

```
# assume there is a matrix containing some frequencies
# (be aware that these counts are entirely fictional):
t1 = c(2, 1, 0, 2, 9, 1, 0, 0, 2, 0)
t2 = c(1, 0, 4, 2, 1, 0, 3, 0, 1, 3)
t3 = c(5, 2, 2, 0, 6, 0, 1, 0, 0, 0)
t4 = c(1, 4, 1, 0, 0, 0, 0, 3, 0, 1)
my.data.table = rbind(t1, t2, t3, t4)
# names of the samples:
rownames(my.data.table) = c("text1", "text2", "text3", "text4")
# names of the variables (e.g. words):
colnames(my.data.table) = c("the", "of", "in", "she", "me", "you",
                                    "them", "if", "they", "he")
# the table looks as follows
print(my.data.table)
# selecting the words that appeared in at laest 50% of samples:
perform.culling(my.data.table, 50)
```

perform.delta

Distance-based classifier

### **Description**

Delta: a simple yet effective machine-learning method of supervised classification, introduced by Burrows (2002). It computes a table of distances between samples, and compares each sample from the test set against training samples, in order to find its nearest neighbor. Apart from classic Delta, a number of alternative distance measures are supported by this function.

perform.delta 25

### Usage

#### **Arguments**

training.set

a table containing frequencies/counts for several variables – e.g. most frequent words – across a number of text samples (for the training set). Make sure that the rows contain samples, and the columns – variables (words, n-grams, or whatever needs to be analyzed).

test.set

a table containing frequencies/counts for the training set. The variables used (i.e. columns) must match the columns of the training set.

classes.training.set

a vector containing class identifiers for the training set. When missing, the row names of the training set table will be used; the assumed classes are the strings of characters followed by the first underscore. Consider the following examples: c("Sterne\_Tristram", "Sterne\_Sentimental", "Fielding\_Tom", ...), where the classes are the authors' names, and c("M\_Joyce\_Dubliners", "F\_Woolf\_Night\_and\_day", "M\_Conrad\_Lord\_Jim", ...), where the classes are M(ale) and F(emale) according to authors' gender. Note that only the part up to the first underscore in the sample's name will be included in the class label.

classes.test.set

a vector containing class identifiers for the test set. When missing, the row names of the test set table will be used (see above).

distance

a kernel (i.e. a distance measure) used for computing similarities between texts. Available options so far: "CD" (Burrow's Delta, default), "AD" (Argamon's Linear Delta), "ED" (Eder's Delta), "ES" (Eder's Simple distance), "CB" (Canberra Distance), "MH" (Manhattan Distance), "EU" (Euclidean Distance).

no.of.candidates

how many nearest neighbors will be computed for each test sample (default = 3).

z.scores.both.sets

many distance measures convert input variables into z-scores before computing any distances. Such a variable weighting is highly dependent on the number of input texts. One might choose either training set only to scale the variables, or the entire corpus (both sets). The latter is default.

### Value

The function returns a vector of "guessed" classes: each test sample is linked with one of the classes represented in the training set. Additionally, final scores and final rankings of candidates are returned as attributes.

26 perform.knn

### Author(s)

Maciej Eder

### References

Argamon, S. (2008). Interpreting Burrows's Delta: geometric and probabilistic foundations. "Literary and Linguistic Computing", 23(2): 131-47.

Burrows, J. F. (2002). "Delta": a measure of stylistic difference and a guide to likely authorship. "Literary and Linguistic Computing", 17(3): 267-87.

Jockers, M. L. and Witten, D. M. (2010). A comparative study of machine learning methods for authorship attribution. "Literary and Linguistic Computing", 25(2): 215-23.

### See Also

```
perform.svm, perform.nsc, perform.knn, perform.naivebayes
```

### **Examples**

```
## Not run:
perform.delta(training.set, test.set)
## End(Not run)
```

perform.knn

k-Nearest Neighbor classifier

### **Description**

A machine-learning supervised classifier; this function is a wrapper for the k-NN procedure provided by the package class.

### Usage

```
perform.knn(training.set, test.set, k.value = 1)
```

### **Arguments**

training.set	a table containing frequencies/counts for several variables – e.g. most frequent
	words – across a number of text samples (for the training set). Make sure that the
	rows contain samples, and the columns – variables (words, n-grams, or whatever needs to be analyzed).
test.set	a table containing frequencies/counts for the training set. The variables used (i.e. columns) must match the columns of the training set.

k.value number of nearest neighbors considered.

perform.naivebayes 27

### Value

The function returns a vector of "guessed" classes: each test sample is linked with one of the classes represented in the training set.

### Author(s)

Maciej Eder

### See Also

```
perform.svm, perform.nsc, perform.delta, perform.naivebayes
```

### **Examples**

```
## Not run:
perform.knn(training.set, test.set)
## End(Not run)
```

perform.naivebayes

Naive Bayes classifier

### Description

A machine-learning supervised classifier; this function is a wrapper for the Naive Bayes procedure provided by the package e1071.

### Usage

```
perform.naivebayes(training.set, test.set)
```

### **Arguments**

training.set

a table containing frequencies/counts for several variables – e.g. most frequent words – across a number of text samples (for the training set). Make sure that the rows contain samples, and the columns – variables (words, n-grams, or whatever

needs to be analyzed).

test.set

a table containing frequencies/counts for the training set. The variables used

(i.e. columns) must match the columns of the training set.

### Value

The function returns a vector of "guessed" classes: each test sample is linked with one of the classes represented in the training set.

### Author(s)

Maciej Eder

28 perform.nsc

### See Also

```
perform.svm, perform.nsc, perform.delta, perform.knn
```

### **Examples**

```
## Not run:
perform.naivebayes(training.set, test.set)
## End(Not run)
```

perform.nsc

Nearest Shrunken Centroids classifier

### **Description**

A machine-learning supervised classifier; this function is a wrapper for the Nearest Shrunken Centroids procedure provided by the package pamr.

### Usage

### **Arguments**

training.set

a table containing frequencies/counts for several variables – e.g. most frequent words – across a number of text samples (for the training set). Make sure that the rows contain samples, and the columns – variables (words, n-grams, or whatever needs to be analyzed).

test.set

a table containing frequencies/counts for the training set. The variables used (i.e. columns) must match the columns of the training set.

classes.training.set

a vector containing class identifiers for the training set. When missing, the row names of the training set table will be used; the assumed classes are the strings of characters followed by the first underscore. Consider the following examples: c("Sterne\_Tristram", "Sterne\_Sentimental", "Fielding\_Tom", ...), where the classes are the authors' names, and c("M\_Joyce\_Dubliners", "F\_Woolf\_Night\_and\_day", "M\_Conrad\_Lord\_Jim", ...), where the classes are M(ale) and F(emale) according to authors' gender. Note that only the part up to the first underscore in the sample's name will be included in the class label.

classes.test.set

a vector containing class identifiers for the test set. When missing, the row names of the test set table will be used (see above).

perform.svm 29

```
no.of.candidates
```

how many nearest neighbors will be computed for each test sample (default = 3).

### Value

The function returns a vector of "guessed" classes: each test sample is linked with one of the classes represented in the training set. Additionally, final scores and final rankings of candidates are returned as attributes.

### Author(s)

Maciej Eder

### See Also

perform.delta, perform.svm, perform.knn, perform.naivebayes

### **Examples**

```
## Not run:
perform.nsc(training.set, test.set)
## End(Not run)
```

perform.svm

Support Vector Machines classifier

### **Description**

A machine-learning supervised classifier; this function is a wrapper for the Support Vector Machines procedure provided by the package e1071.

### Usage

```
perform.svm(training.set,
    test.set,
    classes.training.set = NULL,
    classes.test.set = NULL,
    no.of.candidates = 3,
    tune.parameters = FALSE,
    svm.kernel = "linear",
    svm.degree = 3,
    svm.coef0 = 0,
    svm.cost = 1)
```

30 perform.svm

### **Arguments**

training.set a table containing frequencies/counts for several variables – e.g. most frequent

words – across a number of text samples (for the training set). Make sure that the rows contain samples, and the columns – variables (words, n-grams, or whatever

needs to be analyzed).

test.set a table containing frequencies/counts for the training set. The variables used

(i.e. columns) must match the columns of the training set.

classes.training.set

a vector containing class identifiers for the training set. When missing, the row names of the training set table will be used; the assumed classes are the strings of characters followed by the first underscore. Consider the following examples: c("Sterne\_Tristram", "Sterne\_Sentimental", "Fielding\_Tom", ...), where the classes are the authors' names, and c("M\_Joyce\_Dubliners", "F\_Woolf\_Night\_and\_day",

 $"M\_Conrad\_Lord\_Jim", ...), where the classes are M(ale) and F(emale) according to authors' gender. Note that only the part up to the first underscore in the$ 

sample's name will be included in the class label.

classes.test.set

a vector containing class identifiers for the test set. When missing, the row names of the test set table will be used (see above).

no.of.candidates

how many nearest neighbors will be computed for each test sample (default = 3).

tune.parameters

if this argument is used, two parameters, namely gamma and cost, are tuned

using a bootstrap procedure, and then used to build a SVM model.

svm.kernel SVM kernel. Available values: "linear", which is probably the best choice in

stylometry, since the number of variables (e.g. MFWs) is many times bigger

than the number of classes; "polynomial", and "radial".

svm. degree parameter needed for kernel of type "polynomial" (default: 3).

svm.coef0 parameter needed for kernel of type "polynomial" (default: 0).

svm.cost cost of constraints violation (default: 1); it is the C-constant of the regularization

term in the Lagrange formulation.

### Value

The function returns a vector of "guessed" classes: each test sample is linked with one of the classes represented in the training set. Additionally, final scores and final rankings of candidates are returned as attributes.

### Author(s)

Maciej Eder

#### See Also

perform.delta, perform.nsc, perform.knn, perform.naivebayes

rolling.classify 31

### **Examples**

```
## Not run:
perform.svm(training.set, test.set)
## End(Not run)
```

rolling.classify

Sequential machine-learning classification

### Description

Function that splits a text into equal-sized consecutive blocks (slices) and performs a supervised classification of these blocks against a training set. A number of machine-learning methods for classification used in computational stylistics are available: Delta, k-Nearest Neighbors, Support Vector Machines, Naive Bayes, and Nearest Shrunken Centroids.

### Usage

```
rolling.classify(gui = FALSE, training.corpus.dir = "reference_set",
    test.corpus.dir = "test_set", training.frequencies = NULL,
    test.frequencies = NULL, training.corpus = NULL,
    test.corpus = NULL, features = NULL, path = NULL,
    slice.size = 5000, slice.overlap = 4500,
    training.set.sampling = "no.sampling", mfw = 100, culling = 0,
    milestone.points = NULL, add.ticks = FALSE, ...)
```

### Arguments

gui

an optional argument; if switched on, a simple yet effective graphical user interface (GUI) will appear. Default value is FALSE so far, since GUI is still under development.

### training.frequencies

using this optional argument, one can load a custom table containing frequencies/counts for several variables, e.g. most frequent words, across a number of text samples (for the training set). It can be either an R object (matrix or data frame), or a filename containing tab-delimited data. If you use an R object, make sure that the rows contain samples, and the columns – variables (words). If you use an external file, the variables should go vertically (i.e. in rows): this is because files containing vertically-oriented tables are far more flexible and easily editable using, say, Excel or any text editor. To flip your table horizontally/vertically use the generic function t().

#### test.frequencies

using this optional argument, one can load a custom table containing frequencies/counts for the test set. Further details: immediately above.

32 rolling.classify

training.corpus

another option is to pass a pre-processed corpus as an argument (here: the training set). It is assumed that this object is a list, each element of which is a vector containing one tokenized sample. The example shown below will give you some hints how to prepare such a corpus. Also, refer to help(load.corpus.and.parse)

test.corpus

if training.corpus is used, then you should also prepare a similar R object containing the test set.

features

usually, a number of the most frequent features (words, word n-grams, character n-grams) are extracted automatically from the corpus, and they are used as variables for further analysis. However, in some cases it makes sense to use a set of tailored features, e.g. the words that are associated with emotions or, say, a specific subset of function words. This optional argument allows to pass either a filename containing your custom list of features, or a vector (R object) of features to be assessed.

path

if not specified, the current directory will be used for input/output procedures (reading files, outputting the results).

training.corpus.dir

the subdirectory (within the current working directory) that contains the training set, or the collection of texts used to exemplify the differences between particular classes (e.g. authors or genres). The discriminating features extracted from this training material will be used during the testing procedure (see below). If not specified, the default subdirectory reference\_set will be used.

test.corpus.dir

the subdirectory (within the working directory) that contains a test to be assessed, long enough to be split automatically into equal-sized slices, or blocks. If not specified, the default subdirectory test\_set will be used.

slice.size

a text to be analyzed is segmented into consecutive, equal-sized samples (slices, windows, or blocks); the slice size is set using this parameter: default is 5,000 words. The samples are allowed to partially overlap (see the next parameter).

slice.overlap

if we specifies a slice.size of 5,000 and a slice.overlap of 4,500 (which is default), then the first sample of a text contains words 1–5,000, the second 501–5,500, the third sample 1001–6,000, and so forth.

training.set.sampling

sometimes, it makes sense to split training set texts into smaller samples. Available options: "no.sampling" (default), "normal.sampling", "random.sampling". See help(make.samples) for further details.

mfw

number of the most frequent words (MFWs) to be analyzed.

culling

culling level; see help(perform.culling) to get some help on the culling procedure principles.

milestone.points

sometimes, there is a need to mark one or more passages in an analyzed text (e.g. when external evidence suggests an authorial takeover at a certain point) to compare if the a priori knowledge is confirmed by stylometric evidence. To this end, one should add into the test file a string "xmilestone" (when input texts are loaded directly from files), or specify the break points using this parameter. E.g.,

to add two lines at 10,000 words and 15,000 words, use milestone.points = c(10000, 15000).

rolling.delta 33

add.ticks	a graphical parameter: consider adding tiny ticks (short horizontal lines) to see the density of sampling. Default: FALSE.
	any variable as produced by stylo.default.settings() can be set here to overwrite the default values.

### **Details**

There are numerous additional options that are passed to this function; so far, they are all loaded when stylo.default.settings() is executed (it will be invoked automatically from inside this function); the user can set/change them in the GUI.

#### Value

The function returns an object of the class stylo.results: a list of variables, including tables of word frequencies, vector of features used, a distance table and some more stuff. Additionally, depending on which options have been chosen, the function produces a number of files used to save the results, features assessed, generated tables of distances, etc.

### Author(s)

Maciej Eder

#### See Also

```
classify, rolling.delta
```

### **Examples**

```
## Not run:
# standard usage (it builds a corpus from a collection of text files):
rolling.classify()

rolling.classify(training.frequencies = "freqs_train.txt",
    test.frequencies = "freqs_test.txt", write.png.file = TRUE,
    classification.method = "nsc")

## End(Not run)
```

rolling.delta

Sequential stylometric analysis

### **Description**

Function that analyses collaborative works and tries to determine the authorship of their fragments.

### Usage

34 rolling.delta

### **Arguments**

gui an optional argument; if switched on, a simple yet effective graphical user inter-

face (GUI) will appear. Default value is TRUE.

path if not specified, the current working directory will be used for input/output pro-

cedures (reading files, outputting the results).

primary.corpus.dir

the subdirectory (within the current working directory) that contains a collection of texts written by the authorial candidates, likely to have been involved in the collaborative work analyzed. If not specified, the default subdirectory

primary\_set will be used.

secondary.corpus.dir

the subdirectory (within the current working directory) that contains the collaborative work to be analyzed. If not specified, the default subdirectory secondary\_set will be used.

#### **Details**

The procedure provided by this function analyses collaborative works and tries to determine the authorship of their fragments. The first step involves a "windowing" procedure (Dalen-Oskam and Zundert, 2007) in which each reference text is segmented into consecutive, equal-sized samples or windows. After "rolling" through the test text, we can plot the resulting series of Delta scores for each reference text in a graph.

#### Value

The function returns an object of the class stylo.results, and produces a final plot.

### Author(s)

Mike Kestemont, Maciej Eder, Jan Rybicki

#### References

Eder, M. Kestemont, M. and Rybicki, J. (2013). Stylometry with R: a suite of tools. In: "Digital Humanities 2013: Conference Abstracts". University of Nebraska-Lincoln, Lincoln, NE, pp. 487-89.

van Dalen-Oskam, K. and van Zundert, J. (2007). Delta for Middle Dutch: author and copyist distinction in Walewein. "Literary and Linguistic Computing", 22(3): 345-62.

Hoover, D. (2011). The Tutor's Story: a case study of mixed authorship. In: "Digital Humanities 2011: Conference Abstracts". Stanford University, Stanford, CA, pp. 149-51.

Rybicki, J., Kestemont, M. and Hoover D. (2014). Collaborative authorship: Conrad, Ford and rolling delta. "Literary and Linguistic Computing", 29(3): 422-31.

#### See Also

stylo, classify, oppose

stylo 35

### **Examples**

stylo

Stylometric multidimensional analyses

### **Description**

It is quite a long story what this function does. Basically, it is an all-in-one tool for a variety of experiments in computational stylistics. For a more detailed description, refer to HOWTO available at: https://sites.google.com/site/computationalstylistics/

### Usage

#### **Arguments**

gui

an optional argument; if switched on, a simple yet effective graphical interface (GUI) will appear. Default value is TRUE.

frequencies

using this optional argument, one can load a custom table containing frequencies/counts for several variables, e.g. most frequent words, across a number of text samples. It can be either an R object (matrix or data frame), or a filename containing tab-delimited data. If you use an R object, make sure that the rows contain samples, and the columns – variables (words). If you use an external file, the variables should go vertically (i.e. in rows): this is because files containing vertically-oriented tables are far more flexible and easily editable using, say, Excel or any text editor. To flip your table horizontally/vertically use the generic function t().

generic function

parsed.corpus

another option is to pass a pre-processed corpus as an argument. It is assumed that this object is a list, each element of which is a vector containing one to-kenized sample. The example shown below will give you some hints how to prepare such a corpus.

features

usually, a number of the most frequent features (words, word n-grams, character n-grams) are extracted automatically from the corpus, and they are used as variables for further analysis. However, in some cases it makes sense to use a set of tailored features, e.g. the words that are associated with emotions or, say,

36 stylo

a specific subset of function words. This optional argument allows to pass either a filename containing your custom list of features, or a vector (R object) of

features to be assessed.

path if not specified, the current directory will be used for input/output procedures

(reading files, outputting the results).

corpus.dir the subdirectory (within the current working directory) that contains the corpus

text files. If not specified, the default subdirectory corpus will be used. This option is immaterial when an external corpus and/or external table with frequen-

cies is loaded.

... any variable produced by stylo.default.settings can be set here, in order to

overwrite the default values. An example of such a variable is network = TRUE (switched off as default) for producing stylometric bootstrap consensus networks (Eder, forthcoming); the function saves a csv file, containing a list of

nodes that can be loaded into, say, Gephi.

#### **Details**

If no additional argument is passed, then the functions tries to load text files from the default subdirectory corpus. There are a lot of additional options that should be passed to this function; they are all loaded when stylo.default.settings is executed (which is typically called automatically from inside the stylo function).

#### Value

The function returns an object of the class stylo.results: a list of variables, including a table of word frequencies, vector of features used, a distance table and some more stuff. Additionally, depending on which options have been chosen, the function produces a number of files containing results, plots, tables of distances, etc.

### Author(s)

Maciej Eder, Jan Rybicki, Mike Kestemont

### References

Eder, M. Kestemont, M. and Rybicki, J. (2013). Stylometry with R: a suite of tools. In: "Digital Humanities 2013: Conference Abstracts". University of Nebraska-Lincoln, Lincoln, NE, pp. 487-80

Eder, M. (forthcoming). Visualization in stylometry: some problems and solutions.

#### See Also

```
classify, oppose, rolling.delta
```

### **Examples**

```
## Not run:
# standard usage (it builds a corpus from a set of text files):
stylo()
```

stylo.default.settings 37

```
# loading word frequencies from a tab-delimited file:
stylo(frequencies = "my_frequencies.txt")
# using an existing corpus (a list containing tokenized texts):
txt1 = c("to", "be", "or", "not", "to", "be")
txt2 = c("now", "i", "am", "alone", "o", "what", "a", "slave", "am", "i")
txt3 = c("though", "this", "be", "madness", "yet", "there", "is", "method")
custom.txt.collection = list(txt1, txt2, txt3)
 names(custom.txt.collection) = c("hamlet\_A", "hamlet\_B", "polonius\_A") \\
stylo(parsed.corpus = custom.txt.collection)
# using a custom set of features (words, n-grams) to be analyzed:
my.selection.of.function.words = c("the", "and", "of", "in", "if", "into",
                                   "within", "on", "upon", "since")
stylo(features = my.selection.of.function.words)
# loading a custom set of features (words, n-grams) from a file:
stylo(features = "wordlist.txt")
# batch mode, custom name of corpus directory:
my.test = stylo(gui = FALSE, corpus.dir = "ShakespeareCanon")
summary(my.test)
# batch mode, character 3-grams requested:
stylo(gui = FALSE, analyzed.features = "c", ngram.size = 3)
## End(Not run)
```

stylo.default.settings

Setting variables for the package stylo

### Description

Function that sets a series of global variables to be used by the package stylo and which can be modified by users via arguments passed to the function and/or via gui.stylo, gui.classify, or gui.oppose.

### Usage

```
stylo.default.settings(...)
```

### Arguments

... any variable as produced by this function can be set here to overwrite the default values.

38 stylo.pronouns

### **Details**

This function is typically called from inside stylo, classify, oppose, gui.stylo, gui.classify and gui.oppose.

### Value

The function returns a list of a few dozen variables, mostly options and parameters for different stylometric tests.

### Author(s)

Maciej Eder, Jan Rybicki, Mike Kestemont

#### See Also

```
stylo, gui.stylo
```

### **Examples**

```
stylo.default.settings()

# to see which variables have been set:
names(stylo.default.settings())

# to use the elements of the list as if they were independent variables:
my.variables = stylo.default.settings()
attach(my.variables)
```

 $\verb|stylo.pronouns||$ 

List of pronouns

### **Description**

This function returns a list of pronouns that can be used as a stop word list for different stylometric analyses. It has been shown that pronoun deletion improves, to some extent, attribution accuracy of stylometric procedures (e.g. in English novels: Hoover 2004a; 2004b).

### Usage

```
stylo.pronouns(language = "English")
```

### **Arguments**

language

an optional argument specifying the language of the texts analyzed: available languages are English, Latin, Polish, Dutch, French, German, Spanish, Italian, and Hungarian (default is English).

txt.to.features 39

#### Value

The function returns a vector of pronouns.

#### Author(s)

Jan Rybicki, Maciej Eder, Mike Kestemont

#### References

```
Hoover, D. (2004a). Testing Burrows's delta. "Literary and Linguistic Computing", 19(4): 453-75. Hoover, D. (2004b). Delta prime?. "Literary and Linguistic Computing", 19(4): 477-95.
```

#### See Also

```
stylo
```

### **Examples**

```
stylo.pronouns()
stylo.pronouns(language="Latin")
my.stop.words = stylo.pronouns(language="German")
```

txt.to.features

Split string of words or other countable features

### **Description**

Function that converts a vector of words into either words, or characters, and optionally parses them into n-grams.

### Usage

```
txt.to.features(tokenized.text, features = "w", ngram.size = 1)
```

### **Arguments**

tokenized.text a vector of tokinzed words

features an option for specifying the desired type of feature: w for words, c for characters

(default: w).

ngram.size an optional argument (integer) indicating the value of n, or the size of n-grams

to be created. If this argument is missing, the default value of 1 is used.

### **Details**

Function that carries out the preprocessing steps necessary for feature selection: converts an input text into the type of sequences needed (n-grams etc.) and returns a new vector of items. The function invokes make.ngrams to combine single units into pairs, triplets or longer n-grams. See help(make.ngrams) for details.

40 txt.to.words

#### Author(s)

Maciej Eder, Mike Kestemont

#### See Also

```
txt.to.words, txt.to.words.ext, make.ngrams
```

### **Examples**

```
# consider the string my.text:
my.text = "Quousque tandem abutere, Catilina, patientia nostra?"
# split it into a vector of consecutive words:
my.vector.of.words = txt.to.words(my.text)
# build a vector of word 2-grams:
txt.to.features(my.vector.of.words, ngram.size = 2)
# or produce character n-grams (in this case, character tetragrams):
txt.to.features(my.vector.of.words, features = "c", ngram.size = 4)
```

txt.to.words

Split text into words

### **Description**

Generic tokenization function for splitting a given input text into single words (chains of characters delimited by spaces or punctuation marks).

#### **Usage**

```
txt.to.words(input.text, splitting.rule = NULL, preserve.case = FALSE)
```

### **Arguments**

input.text a string of characters, usually a text.

splitting.rule an optional argument indicating an alternative splitting regexp. E.g., if your corpus contains no punctuation, you can use a very simple splitting sequence: "[\t\n]+" or "[[: space:]]+" (in this case, any whitespace is assumed to be a word delimiter). If you deal with non-latin scripts, especially with those that are not supported by the stylo package yet (e.g. Chinese, Japanese, Vietnamese, Georgian), you can indicate your letter characters explicitly: for most Cyrillic scripts try the following code "[^\u0400-\u0482\u048A\u04FF]+". Remember, however, that your texts need to be properly loaded into R (which is quite tricky on Windows; see below).

preserve.case

Whether or not to lowercase all characters in the corpus (default is FALSE).

txt.to.words 41

#### **Details**

The generic tokenization function for splitting a given input text into single words (chains of characters delimited with spaces or punctuation marks). In obsolete versions of the package stylo, the default splitting sequence of chars was "[^[:alpha:]]+" on Mac/Linux, and "\\W+\_" on Windows. Two different splitting rules were used, because regular expressions are not entirely platform-independent; type help(regexp) for more details. For the sake of compatibility, then, in the version >=0.5.6 a lengthy list of dozens of letters in a few alphabets (Latin, Cyrillic, Greek, Hebrew, Arabic so far) has been indicated explicitly:

```
paste("[^A-Za-z",
   # Latin supplement (Western):
    "\U00C0-\U00FF",
    # Latin supplement (Eastern):
    "\U0100-\U01BF",
    # Latin extended (phonetic):
    "\U01C4-\U02AF",
   # modern Greek:
    "\U0386\U0388-\U03FF",
   # Cyrillic:
    "\U0400-\U0481\U048A-\U0527",
   # Hebrew:
    "\U05D0-\U05EA\U05F0-\U05F4",
   # Arabic:
    "\U0620-\U065F\U066E-\U06D3\U06D5\U06DC",
   # extended Latin:
    "\U1E00-\U1EFF",
    # ancient Greek:
    "\U1F00-\U1FBC\U1FC2-\U1FCC\U1FD0-\U1FDB\U1FE0-\U1FEC\U1FF2-\U1FFC",
    "]+",
    sep="")
```

Alternatively, different tokenization rules can be applied through the option splitting.rule (see above). ATTENTION: this is the only piece of coding in the library stylo that might depend on the operating system used. While on Mac/Linux the native encoding is Unicode, on Windows you never know if your text will be loaded proprely. A considerable solution for Windows users is to convert your texts into Unicode (a variety of freeware converters are available on the internet), and to use an appropriate encoding option when loading the files: read.table("file.txt", encoding = "UTF-8" or scan("file.txt", what = "char", encoding = "UTF-8". If you use the functions provided by the library stylo, you should pass this option as an argument to your chosen function: stylo(encoding = "UTF-8"), classify(encoding = "UTF-8"), oppose(encoding = "UTF-8").

#### Value

The function returns a vector of tokenized words (or other units) as elements.

#### Author(s)

Maciej Eder, Jan Rybicki, Mike Kestemont

42 txt.to.words.ext

#### See Also

```
txt.to.words.ext, txt.to.features, make.ngrams, load.corpus
```

### **Examples**

txt.to.words.ext

Split text into words: extended version

### **Description**

Function for splitting a string of characters into single words, removing punctuation etc., and preserving some language-dependent idiosyncracies, such as common contractions in English.

### Usage

### Arguments

input.text a string of characters, usually a text.

language an optional argument specifying the language of the texts analyzed. Values that

will affect the function's output are: English.contr, English.all, Latin.corr

(their meaning is explained below). The default value is English.

splitting.rule if you are not satisfied with the default language settings (or your input string

of characters is not a regular text, but a sequence of, say, dance movements represented using symbolic signs), you can indicate your custom splitting regular expression here. This option will overwrite the above language settings. For

further details, refer to help(txt.to.words).

preserve.case Whether or not to lowercase all character in the corpus (default = F).

#### Details

Function for splitting a given input text into single words (chains of characters delimited with spaces or punctuation marks). It is build on top of the function txt.to.words and it is designed to manage some language-dependent text features during the tokenization process. In most languages, this is irrelevant. However, it might be important when with English or Latin texts: English.contr treats contractions as single, atomary words, i.e. strings such as "don't", "you've" etc. will not be split into two strings; English.all keeps the contractions (as above), and also prevents the function from splitting compound words (mother-in-law, double-decker, etc.). Latin.corr: since some editions

zeta.chisquare 43

do not distinguish the letters v/u, this setting provides a consistent conversion to "u" in the whole string. The option preserve case lets you specify whether you wish to lowercase all characters in the corpus.

#### Author(s)

Maciej Eder, Jan Rybicki, Mike Kestemont

### See Also

```
txt.to.words, txt.to.features, make.ngrams
```

### **Examples**

```
txt.to.words.ext("Nel mezzo del cammin di nostra vita / mi ritrovai per
    una selva oscura, che la diritta via era smarrita.")

# to see the difference between particular options for English,
# consider the following sentence from Joseph Conrad's "Nostromo":
sample.text = "That's how your money-making is justified here."
txt.to.words.ext(sample.text, language = "English")
txt.to.words.ext(sample.text, language = "English.contr")
txt.to.words.ext(sample.text, language = "English.all")
```

zeta.chisquare

Compare two subcorpora using a home-brew variant of Craig's Zeta

### **Description**

This is a function for comparing two sets of texts; unlike keywords analysis, it this method the goal is to split input texts into equal-sized slices, and to check the appearance of particular words over the slices. Number of slices in which a given word appeared in the subcorpus A and B is then compared using standard chisquare test (if p value exceeds 0.05, a difference is considered significant). This method is based on original Zeta as developed by Burrows and extended by Craig (Burrows 2007, Craig and Kinney 2009).

### Usage

```
zeta.chisquare(input.data)
```

#### **Arguments**

```
input.data a matrix of two columns.
```

#### Value

The function returns a list of two elements: the first contains words (or other units, like n-grams) statistically preferred by the authors of the primary subcorpus, while the second element contains avoided words. Since the applied measure is symmetrical, the preferred words are ipso facto avoided by the secondary authors, and vice versa.

44 zeta.craig

### Author(s)

Maciej Eder

### References

Burrows, J. F. (2007). All the way through: testing for authorship in different frequency strata. "Literary and Linguistic Computing", 22(1): 27-48.

Craig, H. and Kinney, A. F., eds. (2009). Shakespeare, Computers, and the Mystery of Authorship. Cambridge: Cambridge University Press.

#### See Also

```
oppose, zeta.eder, zeta.craig
```

### **Examples**

```
## Not run:
zeta.chisquare(input.data, filter.threshold)
## End(Not run)
```

zeta.craig

Compare two subcorpora using Craig's Zeta

### Description

This is a function for comparing two sets of texts; unlike keywords analysis, it this method the goal is to split input texts into equal-sized slices, and to check the appearance of particular words over the slices. Number of slices in which a given word appeared in the subcorpus A and B is then compared using Craig's formula, which is based on original Zeta as developed by Burrows (Craig and Kinney 2009, Burrows 2007).

### Usage

```
zeta.craig(input.data, filter.threshold)
```

### **Arguments**

```
input.data a matrix of two columns. filter.threshold
```

this parameter (default 0.1) gets rid of words of weak discrimination strength; the higher the number, the less words appear in the final wordlists. It does not normally exceed 0.5. In original Craig's Zeta, no threshold is used: instead, the results contain the fixed number of 500 top avoided and 500 top preferred words.

zeta.eder 45

#### Value

The function returns a list of two elements: the first contains words (or other units, like n-grams) statistically preferred by the authors of the primary subcorpus, while the second element contains avoided words. Since the applied measure is symmetrical, the preferred words are ipso facto avoided by the secondary authors, and vice versa.

### Author(s)

Maciej Eder

### References

Burrows, J. F. (2007). All the way through: testing for authorship in different frequency strata. "Literary and Linguistic Computing", 22(1): 27-48.

Craig, H. and Kinney, A. F., eds. (2009). Shakespeare, Computers, and the Mystery of Authorship. Cambridge: Cambridge University Press.

#### See Also

```
oppose, zeta.eder, zeta.chisquare
```

### **Examples**

```
## Not run:
zeta.craig(input.data, filter.threshold)
## End(Not run)
```

zeta.eder

Compare two subcorpora using Eder's Zeta

### Description

This is a function for comparing two sets of texts; unlike keywords analysis, it this method the goal is to split input texts into equal-sized slices, and to check the appearance of particular words over the slices. Number of slices in which a given word appeared in the subcorpus A and B is then compared using a distance derived from Canberra measure of similarity. Original Zeta was developed by Burrows and extended by Craig (Burrows 2007, Craig and Kinney 2009).

### Usage

```
zeta.eder(input.data, filter.threshold)
```

46 zeta.eder

### **Arguments**

```
input.data a matrix of two columns. filter.threshold
```

this parameter (default 0.1) gets rid of words of weak discrimination strength; the higher the number, the less words appear in the final wordlists. It does not normally exceed 0.5.

#### Value

The function returns a list of two elements: the first contains words (or other units, like n-grams) statistically preferred by the authors of the primary subcorpus, while the second element contains avoided words. Since the applied measure is symmetrical, the preferred words are ipso facto avoided by the secondary authors, and vice versa.

#### Author(s)

Maciej Eder

### References

Burrows, J. F. (2007). All the way through: testing for authorship in different frequency strata. "Literary and Linguistic Computing", 22(1): 27-48.

Craig, H. and Kinney, A. F., eds. (2009). Shakespeare, Computers, and the Mystery of Authorship. Cambridge: Cambridge University Press.

### See Also

```
oppose, zeta.eder, zeta.chisquare
```

### **Examples**

```
## Not run:
zeta.eder(input.data, filter.threshold)
## End(Not run)
```

## **Index**

```
assign.plot.colors, 2, 7
classify, 3, 10, 13, 23, 33, 34, 36
define.plot.area, 6
delete.markup, 7, 15
delete.stop.words, 8, 24
gui.classify, 10
gui.oppose, 11
gui.stylo, 10, 12, 38
load.corpus, 8, 13, 15, 20, 42
load.corpus.and.parse, 14, 20
make.ngrams, 16, 18, 40, 42, 43
make.samples, 15, 17
make.table.of.frequencies, 19
oppose, 5, 11, 13, 21, 34, 36, 44-46
perform.culling, 9, 23
perform.delta, 24, 27–30
perform.knn, 26, 26, 28-30
perform.naivebayes, 26, 27, 27, 29, 30
perform.nsc, 26-28, 28, 30
perform.svm, 26-29, 29
rolling.classify, 31
rolling.delta, 5, 13, 23, 33, 36
stylo, 5, 7, 12, 13, 23, 34, 35, 38, 39
stylo.default.settings, 11, 12, 37
stylo.pronouns, 9, 24, 38
txt.to.features, 8, 15, 17, 18, 39, 42, 43
txt.to.words, 8, 13, 15, 17, 18, 40, 40, 43
txt.to.words.ext, 8, 15, 17, 18, 40, 42, 42
zeta.chisquare, 43, 45, 46
zeta.craig, 44, 44
zeta.eder, 44, 45, 45, 46
```