Package ‘corpusTools’

May 8, 2023

Version  0.5.1
Date      2023-05-07
Title     Managing, Querying and Analyzing Tokenized Text
Description Provides text analysis in R, focusing on the use of a tokenized text format. In this format, the positions of tokens are maintained, and each token can be annotated (e.g., part-of-speech tags, dependency relations).
       Prominent features include advanced Lucene-like querying for specific tokens or contexts (e.g., documents, sentences), similarity statistics for words and documents, exporting to DTM for compatibility with many text analysis packages, and the possibility to reconstruct original text from tokens to facilitate interpretation.
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Depends   R (>= 3.5.0)
Imports   methods, wordcloud (>= 2.5), stringi, Rcpp (>= 0.12.12), R6, udpipe (>= 0.8.3), digest, data.table (>= 1.10.4), quanteda (>= 1.5.1), igraph, tokenbrowser (>= 0.1.5), RNewsflow (>= 1.2.1), Matrix (>= 1.2), parallel, pbapply (>= 1.4), rsyntax (>= 0.1.1)
Suggests  testthat, tm (>= 0.6), topicmodels, knitr, rmarkdown
LinkingTo Rcpp, RcppProgress
LazyData  true
Encoding  UTF-8
License   GPL-3
URL       https://github.com/kasperwelbers/corpusTools
RoxygenNote 7.2.1
VignetteBuilder knitr
NeedsCompilation yes
Repository CRAN
Date/Publication 2023-05-08 09:50:15 UTC
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add_multitoken_label

Choose and add multitoken strings based on multitoken categories

Description

Given a multitoken category (e.g., named entity ids), this function finds the most frequently occurring string in this category and adds it as a label for the category.

Usage

```r
add_multitoken_label(
  tc,  
  colloc_id,  
  feature = "token",  
  new_feature = sprintf("%s_l", colloc_id),  
  pref_subset = NULL
)
```

Arguments

tc a tcorpus object
colloc_id the data column containing the unique id for multitoken tokens
feature the name of the feature column
new_feature the name of the new feature column
pref_subset Optionally, a subset call, to specify a subset that has priority for finding the most frequently occurring string
**aggregate_rsyntax**

Aggregate rsyntax annotations

---

**Description**

A method for aggregating rsyntax annotations. The intended purpose is to compute aggregate values for a given label in an annotation column.

For example, you used `annotate_rsyntax` to add a column with subject-predicate labels, and now you want to concatenate the tokens with these labels. With `annotate_rsyntax` you would first aggregate the subject tokens, then aggregate the predicate tokens. By default (``txt = T``) the column with concatenated tokens are added.

You can specify any aggregation function using any column in `tc$tokens`. So say you want to perform a sentiment analysis on the quotes of politicians. You first used `annotate_rsyntax` to create an annotation column 'quote', that has the labels 'source', 'verb', and 'quote'. You also used `code_dictionary` to add a column with unique politician ID’s and a column with sentiment scores. Now you can aggregate the source tokens to get a single unique ID, and aggregate the quote tokens to get a single sentiment score.

**Usage**

```r
aggregate_rsyntax(
  tc,  
  annotation,  
  ...,  
  by_col = NULL,  
  txt = F,  
  labels = NULL,  
  rm_na = T  
)
```

**Arguments**

- **tc**
  - A `tCorpus`

- **annotation**
  - The name of the rsyntax annotation column

- **...**
  - To aggregate columns for specific

- **by_col**
  - A character vector with other column names in `tc$tokens` to aggregate by.

- **txt**
  - If `TRUE`, add columns with concatenated tokens for each label. Can also be a character vector specifying for which specific labels to create this column

- **labels**
  - Instead of using all labels, a character vector of labels can be given

- **rm_na**
  - If `TRUE`, remove rows with only NA values

**Value**

A `data.table`
Examples

```r
## Not run:
tc = tc_sotu_udpipe$copy()
tc$udpipe_clauses()

subject_verb_predicate = aggregate_rsyntax(tc, 'clause', txt=TRUE)
head(subject_verb_predicate)

## We can also add specific aggregation functions

## count number of tokens in predicate
aggregate_rsyntax(tc, 'clause',
                  agg_label('predicate', n = length(token_id)))

## same, but with txt for only the subject label
aggregate_rsyntax(tc, 'clause', txt='subject',
                  agg_label('predicate', n = length(token_id)))

## example application: sentiment scores for specific subjects

# first use queries to code subjects
tc$code_features(column = 'who',
                  query = c('I# I-s <this president>',
                            'we# we americans <american people>'))

# then use dictionary to get sentiment scores
dict = melt_quanteda_dict(quanteda::data_dictionary_LSD2015)
dict$sentiment = ifelse(dict$code %in% c('negative', 'neg_positive'), -1, 1)
tc$code_dictionary(dict)

sent = aggregate_rsyntax(tc, 'clause', txt='predicate',
                          agg_label('subject', subject = na.omit(who)[1]),
                          agg_label('predicate', sentiment = mean(sentiment, na.rm=TRUE)))
head(sent)
sent[,list(sentiment=mean(sentiment, na.rm=TRUE), n=.N), by='subject']

## End(Not run)
```

---

### agg_label

_Builder function for aggregate_rsyntax_

#### Description

This function is used within the `aggregate_rsyntax` function to facilitate aggregating by specific labels.

#### Usage

```r
agg_label(label, ...)
```
Args **label**  The rsyntax label. Needs to be an existing value in the annotation column (as specified when calling `aggregate_rsyntax`)

... Specify the new aggregated columns in name-value pairs. The name is the name of the new column, and the value should be a function over a column in $tokens. For example: subject = paste(token, collapse = ' ') would create the column 'subject', of which the values are the concatenated tokens. See examples for more.

**Value**

Not relevant. Should only be used within `aggregate_rsyntax`

**Examples**

tc = tc_sotu_udpipe$copy()
tc$udpipe_clauses()

## count number of tokens in predicate
aggregate_rsyntax(tc, 'clause', txt=FALSE,
                  agg_label('predicate', n = length(token_id)))

```r
agg_tcorpus
Aggregate the tokens data
```

**Description**

This is a wrapper for the data.table aggregate function, for easy aggregation of the tokens data grouped by columns in the tokens or meta data. The .id argument is an important addition, because token annotation often contain values that span multiple rows.

**Usage**

```r
agg_tcorpus(tc, ..., by = NULL, .id = NULL, wide = T)
```

**Arguments**

**tc** A tCorpus

... The name of the aggregated column and the function over an existing column are given as a name value pair. For example, count = length(token) will count the number of tokens in each group, and sentiment = mean(sentiment, na.rm=T) will calculate the mean score for a column with sentiment scores.

**by** A character vector with column names from the tokens and/or meta data.
.id  If an id column is given, only rows for which this id is not NA are used, and only one row for each id is used. This prevents double counting of values in annotations that span multiple rows. For example, a sentiment dictionary can match the tokens "not good", in which case the sentiment score (-1) will be assigned to both tokens. These annotations should have an _id column that indicates the unique matches.

wide  Should results be in wide or long format?

Value
A data table

Examples

tc = create_tcorpus(sotu_texts, doc_col='id')

library(quanteda)
dict = data_dictionary_LSD2015
dict = melt_quanteda_dict(dict)
dict$sentiment = ifelse(dict$code %in% c('positive','neg_negative'), 1, -1)
tc$code_dictionary(dict)

agg_tcorpus(tc, N = length(sentiment), sent = mean(sentiment), .id='code_id')
agg_tcorpus(tc, sent = mean(sentiment), .id='code_id', by='president')
agg_tcorpus(tc, sent = mean(sentiment), .id='code_id', by=c('president', 'token'))

as.tcorpus  Force an object to be a tCorpus class

Description
Force an object to be a tCorpus class

Usage
as.tcorpus(x, ...)

Arguments
x  the object to be forced
...  not used
**as.tcorpus.default**  
*Force an object to be a tCorpus class*

**Description**

Force an object to be a tCorpus class

**Usage**

```r
## Default S3 method:
as.tcorpus(x, ...)
```

**Arguments**

- `x`: the object to be forced
- `...`: not used

**Examples**

```r
## Not run:
x = c("First text", "Second text")
as.tcorpus(x) ## x is not a tCorpus object
## End(Not run)
```

---

**as.tcorpus.tCorpus**  
*Force an object to be a tCorpus class*

**Description**

Force an object to be a tCorpus class

**Usage**

```r
## S3 method for class 'tCorpus'
as.tcorpus(x, ...)
```

**Arguments**

- `x`: the object to be forced
- `...`: not used

**Examples**

```r
tc = create_tcorpus(c("First text", "Second text"))
as.tcorpus(tc)
```
backbone_filter

Extract the backbone of a network.

Description


Usage

backbone_filter(
    g,
    alpha = 0.05,
    direction = "none",
    delete_isolates = T,
    max_vertices = NULL,
    use_original_alpha = T,
    k_is_n = F
)

Arguments

g
A graph in the 'igraph' format.

alpha
The threshold for the alpha. Can be interpreted similar to a p value (see paper for clarification).

direction
direction = 'none' can be used for both directed and undirected networks, and is (supposed to be) the disparity filter proposed in Serrano et al. (2009) is used. By setting to 'in' or 'out', the alpha is only calculated for out or in edges. This is an experimental use of the backbone extraction (so beware!) but it seems a logical application.

delete_isolates
If TRUE, vertices with degree 0 (i.e. no edges) are deleted.

max_vertices
Optional. Set a maximum number of vertices for the network to be produced. The alpha is then automatically lowered to the point that only the given number of vertices remains connected (degree > 0). This can be usefull if the purpose is to make an interpretation friendly network. See e.g., http://jcom.sissa.it/archive/14/01/JCOM_1401_2015_A01.

use_original_alpha
if max_vertices is not NULL, this determines whether the lower alpha for selecting the top vertices is also used as a threshold for the edges, or whether the original value given in the alpha parameter is used.

k_is_n
the disparity filter method for backbone extraction uses the number of existing edges (k) for each node, which can be arbitrary if there are many very weak ties, which is often the case in a co-occurrence network. By setting k_is_n to TRUE, it is 'assumed' that all nodes are connected, which makes sense from a language model perspective (i.e. probability for co-occurrence is never zero).
Value

A graph in the Igraph format

Examples

```r
tc = create_tcorpus(sotu_texts, doc_column = 'id')
tc$preprocess('token', 'feature', remove_stopwords = TRUE, use_stemming = TRUE, min_docfreq = 10)

g = semnet_window(tc, 'feature', window.size = 10)
igraph::vcount(g)
igraph::ecount(g)
gb = backbone_filter(g, max_vertices = 100)
igraph::vcount(gb)
igraph::ecount(gb)
plot_semnet(gb)
```

---

### browse_hits

**View hits in a browser**

---

**Description**

Creates a static HTML file to view the query hits in the tcorpus in full text mode.

**Usage**

```r
browse_hits(
  tc, 
  hits, 
  token_col = "token", 
  n = 500, 
  select = c("first", "random"), 
  header = "", 
  subheader = NULL, 
  meta_cols = NULL, 
  seed = NA, 
  view = T, 
  filename = NULL
)
```

**Arguments**

- **tc**  
  a tCorpus

- **hits**  
  a featureHits object, as returned by `search_features`

- **token_col**  
  The name of the column in tc$tokens that contain the token text
n
If doc_ids is NULL, Only n of the results are printed (to prevent accidentally making huge browsers).

select
If n is smaller than the number of documents in tc, select determines how the n documents are selected

header
Optionally, a title presented at the top of the browser

subheader
Optionally, overwrite the subheader. By default the subheader reports the number of documents

meta_cols
A character vector with names of columns in tc$meta, used to only show the selected columns

seed
If select is "random", seed can be used to set a random seed

view
If TRUE (default), view the browser in the Viewer window (turn off if this is not supported)

filename
Optionally, save the browser at a specified location

Value
The url for the file location is returned (invisibly)

Examples

tc = create_tcorpus(sotu_texts, doc_column='id')
hits = search_features(tc, c("Terrorism# terroris*", "War# war*"))
browse_hits(tc, hits)

browse_texts
Create and view a full text browser

Description
Creates a static HTML file to view the texts in the tcorpus in full text mode.

Usage

browse_texts(
  tc,
  doc_ids = NULL,
  token_col = "token",
  n = 500,
  select = c("first", "random"),
  header = "",
  subheader = NULL,
  highlight = NULL,
  scale = NULL,
  category = NULL,
browse_texts

rsyntax = NULL,
value = NULL,
meta_cols = NULL,
seed = NA,
nav = NULL,
top_nav = NULL,
thsres_nav = 1,
view = T,
highlight_col = "yellow",
scale_col = c("red", "blue", "green"),
filename = NULL
)

Arguments

tc           a tCorpus

doc_ids      A vector with document ids to view

token_col    The name of the column in tc$tokens that contain the token text

n            Only n of the results are printed (to prevent accidentally making huge browsers).

select       If n is smaller than the number of documents in tc, select determines how the n
documents are selected

header       Optionally, a title presented at the top of the browser

subheader    Optionally, overwrite the subheader. By default the subheader reports the num-

ber of documents

highlight    Highlighte mode: provide the name of a numeric column in tc$tokens with values

between 0 and 1, used to highlight tokens. Can also be a character vector, in

which case al non-NA values are highlighted

scale        Scale mode: provide the name of a numeric column in tc$tokens with values

between -1 and 1, used to color tokens on a scale (set colors with scale_col)

category     Category mode: provide the name of a character or factor column in tc$tokens.

Each unique value will have its own color, and navigation for categories will be

added (nav cannot be used with this option)

rsyntax      rsyntax mode: provide the name of an rsyntax annotation column (see annotate_rsyntax)

value        rsyntax mode argument: if rsyntax mode is used, value can be a character vector

with values in the rsyntax annotation column. If used, only these values are fully

colored, and the other (non NA) values only have border colors.

meta_cols    A character vector with names of columns in tc$meta, used to only show the

selected columns

seed         If select is "random", seed can be used to set a random seed. After sampling the

seed is re-initialized with set.seed(NULL).

nav          Optionally, a column in tc$meta to add navigation (only supports simple filtering

on unique values). This is not possible if category is used.

top_nav      A number. If navigation based on token annotations is used, filters will only

apply to top x values with highest token occurrence in a document
calc_chi2

Vectorized computation of chi^2 statistic for a 2x2 crosstab containing the values [a, b] [c, d]

description

Vectorized computation of chi^2 statistic for a 2x2 crosstab containing the values [a, b] [c, d]

Usage

calc_chi2(a, b, c, d, correct = T, cochrans_criteria = F)

Arguments

a
topleft value of the table
b
topright value
c
bottomleft value
d
bottomright value
correct
if TRUE, use yates correction. Can be a vector of length a (i.e. the number of tables)
cochrans_criteria
if TRUE, check if cochrans_criteria indicate that a correction should be used. This overrides the correct parameter
**compare_corpus**

Compare tCorpus vocabulary to that of another (reference) tCorpus

**Description**

Compare tCorpus vocabulary to that of another (reference) tCorpus

**Usage**

```r
compare_corpus(
  tc, 
  tc_y, 
  feature, 
  smooth = 0.1, 
  min_ratio = NULL, 
  min_chi2 = NULL, 
  is_subset = F, 
  yates_cor = c("auto", "yes", "no"),
  what = c("freq", "docfreq", "cooccurrence")
)
```

**Arguments**

- `tc` a tCorpus
- `tc_y` the reference tCorpus
- `feature` the column name of the feature that is to be compared
- `smooth` Laplace smoothing is used for the calculation of the probabilities. Here you can set the added (pseudo-count) value.
- `min_ratio` threshold for the ratio value, which is the ratio of the relative frequency of a term in dtm.x and dtm.y
- `min_chi2` threshold for the chi^2 value
- `is_subset` Specify whether tc is a subset of tc_y. In this case, the term frequencies of tc will be subtracted from the term frequencies in tc_y
- `yates_cor` mode for using yates correction in the chi^2 calculation. Can be turned on ("yes") or off ("no"), or set to "auto", in which case cochrons rule is used to determine whether yates’ correction is used.
- `what` choose whether to compare the frequency ("freq") of terms, or the document frequency ("docfreq"). This also affects how chi^2 is calculated, comparing either freq relative to vocabulary size or docfreq relative to corpus size (N)

**Value**

A vocabularyComparison object
Examples

```r
tc = create_tcorpus(sotu_texts, doc_column = 'id')

tc$preprocess('token', 'feature', remove_stopwords = TRUE, use_stemming = TRUE)

obama = tc$subset_meta(president == 'Barack Obama', copy=TRUE)
bush = tc$subset_meta(president == 'George W. Bush', copy=TRUE)

comp = compare_corpus(tc, bush, 'feature')
comp = comp[order(-comp$chi),]
head(comp)
plot(comp)
```

### compare_documents

**Calculate the similarity of documents**

**Description**

Calculate the similarity of documents

**Usage**

```r
compare_documents(
  tc,  
  feature = "token",  
  date_col = NULL,  
  meta_cols = NULL,  
  hour_window = c(24),  
  measure = c("cosine", "overlap_pct"),  
  min_similarity = 0,  
  weight = c("norm_tfidf", "tfidf", "termfreq", "docfreq"),  
  ngrams = NA,  
  from_subset = NULL,  
  to_subset = NULL,  
  return_igraph = T,  
  verbose = T
)
```

**Arguments**

- `tc` A `tCorpus`
- `feature` the column name of the feature that is to be used for the comparison.
- `date_col` a date with time in POSIXct. If given together with `hour_window`, only documents within the given `hour_window` will be compared.
meta_cols a character vector with columns in the meta data/docvars. If given, only documents for which these values are identical are compared

hour_window A vector of length 1 or 2. If length is 1, the same value is used for the left and right side of the window. If length is 2, the first and second value determine the left and right side. For example, the value 12 will compare each document to all documents between the previous and next 12 hours, and c(-10, 36) will compare each document to all documents between the previous 10 and the next 36 hours.

measure the similarity measure. Currently supports cosine similarity (symmetric) and overlap_pct (asymmetric)

min_similarity A threshold for the similarity score

weight a weighting scheme for the document-term matrix. Default is term-frequency inverse document frequency with normalized rows (document length).

ngrams an integer. If given, ngrams of this length are used

from_subset An expression to select a subset. If given, only this subset will be compared to other documents

to_subset An expression to select a subset. If given, documents are only compared to this subset

return_igraph If TRUE, return as an igraph network. Otherwise, return as a list with the edge-list and meta data.

verbose If TRUE, report progress

Value
An igraph graph in which nodes are documents and edges represent similarity scores

Examples

d = data.frame(text = c('a b c d e',
  'e f g h i j k',
  'a b c'),
  date = as.POSIXct(c('2010-01-01','2010-01-01','2012-01-01')))
tc = create_tcorpus(d)

g = compare_documents(tc)
igraph::get.data.frame(g)

g = compare_documents(tc, measure = 'overlap_pct')
igraph::get.data.frame(g)

g = compare_documents(tc, date_col = 'date', hour_window = c(0,36))
igraph::get.data.frame(g)
compare_subset

Compare vocabulary of a subset of a tCorpus to the rest of the tCorpus

Description

Compare vocabulary of a subset of a tCorpus to the rest of the tCorpus

Usage

```r
compare_subset(
  tc,
  feature,
  subset_x = NULL,
  subset_meta_x = NULL,
  query_x = NULL,
  query_feature = "token",
  smooth = 0.1,
  min_ratio = NULL,
  min_chi2 = NULL,
  yates_cor = c("auto", "yes", "no"),
  what = c("freq", "docfreq", "cooccurrence")
)
```

Arguments

- `tc`: a `tCorpus`
- `feature`: the column name of the feature that is to be compared
- `subset_x`: an expression to subset the tCorpus. The vocabulary of the subset will be compared to the rest of the tCorpus
- `subset_meta_x`: like `subset_x`, but using using the meta data
- `query_x`: like `subset_x`, but using a query search to select documents (see `search_contexts`)
- `query_feature`: if `query_x` is used, the column name of the feature used in the query search.
- `smooth`: Laplace smoothing is used for the calculation of the probabilities. Here you can set the added (pseudocount) value.
- `min_ratio`: threshold for the ratio value, which is the ratio of the relative frequency of a term in `dtm.x` and `dtm.y`
- `min_chi2`: threshold for the chi^2 value
- `yates_cor`: mode for using yates correction in the chi^2 calculation. Can be turned on ("yes") or off ("no"), or set to "auto", in which case cochrans rule is used to determine whether yates' correction is used.
- `what`: choose whether to compare the frequency ("freq") of terms, or the document frequency ("docfreq"). This also affects how chi^2 is calculated, comparing either freq relative to vocabulary size or docfreq relative to corpus size (N)
Value

A vocabularyComparison object

Examples

tc = create_tcorpus(sotu_texts, doc_column = 'id')
tc$preprocess('token', 'feature', remove_stopwords = TRUE, use_stemming = TRUE)

comp = compare_subset(tc, 'feature', subset_meta_x = president == 'Barack Obama')
comp = comp[order(-comp$chi),]
head(comp)
plot(comp)

comp = compare_subset(tc, 'feature', query_x = 'terroris*')
comp = comp[order(-comp$chi),]
head(comp, 10)

corenlp_tokens  coreNLP example sentences

Description

coreNLP example sentences

Usage

data(corenlp_tokens)

Format

data.frame

count_tcorpus  Count results of search hits, or of a given feature in tokens

Description

Count results of search hits, or of a given feature in tokens
Usage

```r
count_tcorpus(
  tc,
  meta_cols = NULL,
  hits = NULL,
  feature = NULL,
  count = c("documents", "tokens", "hits"),
  wide = T
)
```

Arguments

- `tc`: A tCorpus
- `meta_cols`: The columns in the meta data by which the results should be grouped
- `hits`: featureHits or contextHits (output of `search_features`, `search_dictionary` or `search_contexts`)
- `feature`: Instead of hits, a specific feature column can be selected.
- `count`: How should the results be counted? Number of documents, tokens, or unique hits. The difference between tokens and hits is that hits can encompass multiple tokens (e.g., "Bob Smith" is 1 hit and 2 tokens).
- `wide`: Should results be in wide or long format?

Value

A data table

Examples

```r
tc = create_tcorpus(sotu_texts, doc_col='id')
hits = search_features(tc, c("US# <united states>", "Economy# econom*"))
count_tcorpus(tc, hits=hits)
count_tcorpus(tc, hits=hits, meta_cols='president')
count_tcorpus(tc, hits=hits, meta_cols='president', wide=FALSE)
```

---

**create_tcorpus**

Create a tCorpus

Description

Create a tCorpus from raw text input. Input can be a character (or factor) vector, data.frame or quanteda corpus. If a data.frame is given, all columns other than the document id and text columns are included as meta data. If a quanteda corpus is given, the ids and texts are already specified, and the docvars will be included in the tCorpus as meta data.
create_tcorpus

Usage

create_tcorpus(x, ...)

## S3 method for class 'character'
create_tcorpus(
  x,
  doc_id = 1:length(x),
  meta = NULL,
  udpipe_model = NULL,
  split_sentences = F,
  max_sentences = NULL,
  max_tokens = NULL,
  udpipe_model_path = getwd(),
  udpipe_cache = 3,
  udpipe_cores = NULL,
  udpipe_batchsize = 50,
  use_parser = F,
  remember_spaces = F,
  verbose = T,
  ...
)

## S3 method for class 'data.frame'
create_tcorpus(
  x,
  text_columns = "text",
  doc_column = "doc_id",
  udpipe_model = NULL,
  split_sentences = F,
  max_sentences = NULL,
  max_tokens = NULL,
  udpipe_model_path = getwd(),
  udpipe_cache = 3,
  udpipe_cores = NULL,
  udpipe_batchsize = 50,
  use_parser = F,
  remember_spaces = F,
  verbose = T,
  ...
)

## S3 method for class 'factor'
create_tcorpus(x, ...)

## S3 method for class 'corpus'
create_tcorpus(x, ...)
Arguments

- **x**: main input. can be a character (or factor) vector where each value is a full text, or a data.frame that has a column that contains full texts. If x (or a text_column in x) has leading or trailing whitespace, this is cut off (and you’ll get a warning about it).

- ...: Arguments passed to create_tcorpus.character

- **doc_id**: if x is a character/factor vector, doc_id can be used to specify document ids. This has to be a vector of the same length as x

- **meta**: A data.frame with document meta information (e.g., date, source). The rows of the data.frame need to match the values of x

- **udpipe_model**: Optionally, the name of a Universal Dependencies language model (e.g., "english-ewt", "dutch-alpino"), to use the udpipe package (udpipe_annotate) for natural language processing. You can use show_udpipe_models to get an overview of the available models. For more information about udpipe and performance benchmarks of the UD models, see the GitHub page of the udpipe package.

- **split_sentences**: Logical. If TRUE, the sentence number of tokens is also computed. (only if udpipe_model is not used)

- **max_sentences**: An integer. Limits the number of sentences per document to the specified number. If set when split_sentences == FALSE, split_sentences will be set to TRUE.

- **max_tokens**: An integer. Limits the number of tokens per document to the specified number

- **udpipe_model_path**: If udpipe_model is used, this path will be used to look for the model, and if the model doesn’t yet exist it will be downloaded to this location. Defaults to working directory

- **udpipe_cache**: The number of persistent caches to keep for inputs of udpipe. The caches store tokens in batches. This way, if a lot of data has to be parsed, or if R crashes, udpipe can continue from the latest batch instead of start over. The caches are stored in the corpustools_data folder (in udpipe_model_path). Only the most recent [udpipe_caches] caches will be stored.

- **udpipe_cores**: If udpipe_model is used, this sets the number of parallel cores. If not specified, will use the same number of cores as used by data.table (or limited to OMP_THREAD_LIMIT).

- **udpipe_batchsize**: In order to report progress and cache results, texts are parsed with udpipe in batches of 50. The price is that there will be some overhead for each batch, so for very large jobs it can be faster to increase the batchsize. If the number of texts divided by the number of parallel cores is lower than the batchsize, the texts are evenly distributed over cores.

- **use_parser**: If TRUE, use dependency parser (only if udpipe_model is used)

- **remember_spaces**: If TRUE, a column with spaces after each token and column with the start and end positions of tokens are included. Can turn it of for a bit more speed and less memory use, but some features won’t work.
create_tcorpus

**verbose**
If TRUE, report progress. Only if x is large enough to require multiple sequential batches

**text_columns**
if x is a data.frame, this specifies the column(s) that contains text. If multiple columns are used, they are pasted together separated by a double line break. If remember_spaces is true, a "field" column is also added that show the column name for each token, and the start/end positions are local within these fields.

**doc_column**
If x is a data.frame, this specifies the column with the document ids.

**Details**

By default, texts will only be tokenized, and basic preprocessing techniques (lowercasing, stemming) can be applied with the `preprocess` method. Alternatively, the udpipe package can be used to apply more advanced NLP preprocessing, by using the `udpipe_model` argument.

For certain advanced features you need to set `remember_spaces` to true. We are often used to forgetting all about spaces when we do bag-of-word type stuff, and that’s sad. With `remember_spaces`, the exact position of each token is remembered, including what type of space follows the token (like a space or a line break), and what text field the token came from (if multiple text_columns are specified in `create_tcorpus.data.frame`)

**Examples**

```r
## ...
tc = create_tcorpus(c('Text one first sentence. Text one second sentence', 'Text two'))
tc$tokens

tc = create_tcorpus(c('Text one first sentence. Text one second sentence', 'Text two'),
                   split_sentences = TRUE)
tc$tokens

## with meta (easier to S3 method for data.frame)
meta = data.frame(doc_id = c(1,2), source = c('a', 'b'))
tc = create_tcorpus(c('Text one first sentence. Text one second sentence', 'Text two'),
                   split_sentences = TRUE,
                   doc_id = c(1,2),
                   meta = meta)
tc
d = data.frame(text = c('Text one first sentence. Text one second sentence.',
                        'Text two', 'Text three'),
               date = c('2010-01-01', '2010-01-01', '2012-01-01'),
               source = c('A', 'B', 'B'))
tc = create_tcorpus(d, split_sentences = TRUE)
tc
tc$tokens

## use multiple text columns
d$headline = c('Head one', 'Head two', 'Head three')
## use custom doc_id
d$doc_id = c('#1', '#2', '#3')
tc = create_tcorpus(d, text_columns = c('headline', 'text'), doc_column = 'doc_id',
                   split_sentences = TRUE)
tc
tc$tokens
```
split_sentences = TRUE)

## It makes little sense to have full texts as factors, but it tends to happen.
## The create_tcorpus S3 method for factors is essentially identical to the
## method for a character vector.
text = factor(c('Text one first sentence', 'Text one second sentence'))
tc = create_tcorpus(text)
tc$tokens

library(quanteda)
create_tcorpus(data_corpus_inaugural)

docfreq_filter

**Support function for subset method**

**Description**

Support function to enable subsetting by document frequency stats of a given feature. Should only
be used within the tCorpus subset method, or any tCorpus method that supports a subset argument.

**Usage**

docfreq_filter(
  x,
  min = -Inf,
  max = Inf,
  top = NULL,
  bottom = NULL,
  doc_id = parent.frame()$doc_id
)

**Arguments**

- **x**
  the name of the feature column. Can be given as a call or a string.
- **min**
  A number, setting the minimum document frequency value
- **max**
  A number, setting the maximum document frequency value
- **top**
  A number. If given, only the top x features with the highest document frequency are TRUE
- **bottom**
  A number. If given, only the bottom x features with the highest document frequency are TRUE
- **doc_id**
  Added for reference, but should not be used. Automatically takes doc_id from tCorpus if the docfreq_filter function is used within the subset method.
**dtm_compare**

Comparison of two document term matrices.

### Examples

```r
tc = create_tcorpus(c('a a a b b', 'a a c c'))

tc$tokens

tc$subset(subset = docfreq_filter(token, min=2))
tc$tokens
```

### Description

Compare two document term matrices.

### Usage

```r
dtm_compare(  
dtm.x,  
dtm.y = NULL,  
smooth = 0.1,  
min_ratio = NULL,  
min_chi2 = NULL,  
select_rows = NULL,  
yates_cor = c("auto", "yes", "no"),  
x_is_subset = F,  
what = c("freq", "docfreq", "cooccurrence")
)
```

### Arguments

- **dtm.x**  
  the main document-term matrix

- **dtm.y**  
  the 'reference' document-term matrix

- **smooth**  
  Laplace smoothing is used for the calculation of the probabilities. Here you can set the added (pseudocount) value.

- **min_ratio**  
  threshold for the ratio value, which is the ratio of the relative frequency of a term in dtm.x and dtm.y

- **min_chi2**  
  threshold for the chi^2 value

- **select_rows**  
  Alternative to using dtm.y. Has to be a vector with rownames, by which

- **yates_cor**  
  mode for using yates correction in the chi^2 calculation. Can be turned on ("yes") or off ("no"), or set to "auto", in which case cochrans rule is used to determine whether yates' correction is used.

- **x_is_subset**  
  Specify whether dtm.x is a subset of dtm.y. In this case, the term frequencies of dtm.x will be subtracted from the term frequencies in dtm.y

- **what**  
  choose whether to compare the frequency ("freq") of terms, or the document frequency ("docfreq"). This also affects how chi^2 is calculated, comparing either freq relative to vocabulary size or docfreq relative to corpus size (N)
Value

A data frame with rows corresponding to the terms in dtm and the statistics in the columns

---

### dtm_wordcloud

Plot a word cloud from a dtm

---

Description

Compute the term frequencies for the dtm and plot a word cloud with the top n topics. You can either supply a document-term matrix or provide terms and freqs directly (in which case this is an alias for wordcloud::wordcloud with sensible defaults).

Usage

```r
R

dtm_wordcloud(
  dtm = NULL,
  nterms = 100,
  freq.fun = NULL,
  terms = NULL,
  freqs = NULL,
  scale = c(4, 0.5),
  min.freq = 1,
  rot.per = 0.15,
  ...
)
```

Arguments

- **dtm**: the document-term matrix
- **nterms**: the amount of words to plot (default 100)
- **freq.fun**: if given, will be applied to the frequenies (e.g. sqrt)
- **terms**: the terms to plot, ignored if dtm is given
- **freqs**: the frequencies to plot, ignored if dtm is given
- **scale**: the scale to plot (see wordcloud::wordcloud)
- **min.freq**: the minimum frequency to include (see wordcloud::wordcloud)
- **rot.per**: the percentage of vertical words (see wordcloud::wordcloud)
- **...**: other arguments passed to wordcloud::wordcloud
Examples

```r
## create DTM
tc = create_tcorpus(sotu_texts[1:100,], doc_column = 'id')
tc$preprocess('token', 'feature', remove_stopwords = TRUE)
dtm = get_dtm(tc, 'feature')

dtm_wordcloud(dtm, nterms = 20)

## or without a DTM
dtm_wordcloud(terms = c('in','the','cloud'), freqs = c(2,5,10))
```

---

**ego_semnet**

Create an ego network

**Description**

Create an ego network from an igraph object.

**Usage**

```r
go::ego_semnet(
  g,
  vertex_names,
  depth = 1,
  only_filter_vertices = T,
  weight_attr = "weight",
  min_weight = NULL,
  top_edges = NULL,
  max_edges_level = NULL,
  directed = c("out", "in")
)
```

**Arguments**

- `g` : an igraph object
- `vertex_names` : a character string with the names of the ego vertices/nodes
- `depth` : the number of degrees from the ego vertices/nodes that are included. 1 means that only the direct neighbours are included
- `only_filter_vertices` : if True, the algorithm will only filter out vertices/nodes that are not in the ego network. If False (default) then it also filters out the edges.
- `weight_attr` : the name of the edge attribute. if NA, no weight is used, and min_weight and top_edges are ignored
- `min_weight` : a number indicating the minimum weight
- `top_edges` : the number of top edges to select for each ego vertex/nodes
- `max_edges_level` : the maximum number of edges to select for each ego vertex/nodes
top_edges for each vertex within the given depth, only keep the top n edges with the strongest edge weight. Can also be a vector of the same length as the depth value, in which case a different value is used at each level: first value for level 1, second value for level 2, etc.

directed if the network is directed, specify whether 'out' degrees or 'in' degrees are used

Details

The function is similar to the ego function in igraph, but with some notable differences. Firstly, if multiple vertex_names are given, the ego network for both is given in 1 network (whereas igraph creates a list of networks). Secondly, the min_weight and top_edges parameters can be used to focus on the strongest edges.

Examples

tc = create_tcorpus(c('a b c', 'd e f', 'a d'))
g = semnet(tc, 'token')
igraph::get.data.frame(g)
plot_semnet(g)

## only keep nodes directly connected to given node
eg = ego_semnet(g, 'e')
igraph::get.data.frame(eg)
plot_semnet(eg)

## only keep edges directly connected to given node
eg = ego_semnet(g, 'e', only_filter_vertices = FALSE)
igraph::get.data.frame(eg)
plot_semnet(eg)

## only keep nodes connected to given node with a specified degree (i.e. distance)
eg = ego_semnet(g, 'e', depth = 2)
igraph::get.data.frame(eg)
plot_semnet(eg)

export_span_annotations

Export span annotations

Description

Export columns from a tCorpus as span annotations (annotations over a span of text). The annotations are returned as a data.frame where each row is an annotation, with columns: doc_id, variable, value, field, offset, length and text. The key purpose is that these span annotations are linked to exact character positions in the text. This also means that this function can only be used if position information is available (i.e. if remember_spaces=T was used when creating the tCorpus)
Usage

export_span_annotations(tc, variables)

Arguments

tc A tCorpus, created with `create_tcorpus`, where `remember_spaces` must have been set to TRUE
variables A character vector with variables (columns in `tc$tokens`) to export

Details

Note that if there are spans with gaps in them (e.g. based on proximity queries), they are split into different annotations. Thus some information can be lost.

Value

A data.table where each row is a span annotation, with columns: `doc_id`, `variable`, `value`, `field`, `offset`, `length`, `text`

Examples

tc = create_tcorpus(sotu_texts, c("president", "text"), doc_column="id", remember_spaces=TRUE)
tc$code_features(c("war# war peace", "us being# <(i we) (am are)>"))
export_span_annotations(tc, "code")
Arguments

- **tc**: a tCorpus
- **feature**: The name of the feature column in $tokens
- **query**: A character string that is a query. See search_features for documentation of the query language.
- **hits**: Alternatively, instead of giving a query, the results of search_features can be used.
- **query_feature**: If query is used, the column in $tokens on which the query is performed. By default uses 'token'
- **window**: The size of the word window (i.e. the number of words next to the feature)
- **n**: the top n of associated features
- **min_freq**: Optionally, ignore features that occur less than min_freq times
- **sort_by**: The value by which to sort the features
- **subset**: A call (or character string of a call) as one would normally pass to subset.tCorpus. If given, the keyword has to occur within the subset. This is for instance usefull to only look in named entity POS tags when searching for people or organization. Note that the condition does not have to occur within the subset.
- **subset_meta**: A call (or character string of a call) as one would normally pass to the subset_meta parameter of subset.tCorpus. If given, the keyword has to occur within the subset documents. This is for instance usefull to make queries date dependent. For example, in a longitudinal analysis of politicians, it is often required to take changing functions and/or party affiliations into account. This can be accomplished by using subset_meta = "date > xxx & date < xxx" (given that the appropriate date column exists in the meta data).
- **include_self**: If True, include the feature itself in the output

Value

- a data.frame

Examples

```r
# create tCorpus
tc = create_tcorpus(sotu_texts, doc_column = 'id')
tc$preprocess()

## directly from query
topf = feature_associations(tc, 'feature', 'war')
head(topf, 20) ## frequent words close to "war"

## adjust window size
topf = feature_associations(tc, 'feature', 'war', window = 5)
head(topf, 20) ## frequent words very close (five tokens) to "war"

## you can also first perform search_features, to get hits for (complex) queries
hits = search_features(tc, "war terror"~10)
```
topf = feature_associations(tc, 'feature', hits = hits)
head(topf, 20) ## frequent words close to the combination of "war" and "terror" within 10 words

---

**feature_stats**  
*Feature statistics*

**Description**

Compute a number of useful statistics for features: term frequency, idf, etc.

**Usage**

```r
feature_stats(tc, feature, context_level = c("document", "sentence"))
```

**Arguments**

- `tc` a tCorpus
- `feature` The name of the feature column
- `context_level` Should results be returned at document or sentence level

**Value**

a data.frame

**Examples**

```r
tc = create_tcorpus(c('Text one first sentence. Text one second sentence', 'Text two'),
                   split_sentences = TRUE)

fs = feature_stats(tc, 'token')
head(fs)
fs = feature_stats(tc, 'token', context_level = 'sentence')
head(fs)
```
fold_rsyntax

Fold rsyntax annotations

Description

If a tCorpus has rsyntax annotations (see annotate_rsyntax), it can be convenient to aggregate tokens that have a certain semantic label. For example, if you have a query for labeling "source" and "quote", you can add an aggregated value for the sources (such as a unique ID) as a column, and then remove the quote tokens.

Usage

fold_rsyntax(tc, annotation, by_label, ..., txt = F, rm_by = T)

Arguments

tc A tCorpus
annotation The name of an rsyntax annotation column
by_label The labels in this column for which you want to aggregate the tokens
... Specify the new aggregated columns in name-value pairs. The name is the name of the new column, and the value should be a function over a column in $tokens. For example: subject = paste(token, collapse = ' ') would create the column 'subject', of which the values are the concatenated tokens. See examples for more.
txt If TRUE, add _txt column with concatenated tokens for by_label.
rm_by If TRUE (default), remove the column(s) specified in by_label

Value

a transformed tCorpus

Examples

tc = tc_sotu_udpipe$copy()
tc$udpipe_clauses()

fold_rsyntax(tc, 'clause', by_label = 'subject', subject = paste(token, collapse=' '))
freq_filter

Support function for subset method

Description

Support function to enable subsetting by frequency stats of a given feature. Should only be used within the tCorpus subset method, or any tCorpus method that supports a subset argument.

Usage

freq_filter(x, min = -Inf, max = Inf, top = NULL, bottom = NULL)

Arguments

x
the name of the feature column. Can be given as a call or a string.

min
A number, setting the minimum frequency value

max
A number, setting the maximum frequency value

top
A number. If given, only the top x features with the highest frequency are TRUE

bottom
A number. If given, only the bottom x features with the highest frequency are TRUE

Examples

tc = create_tcorpus(c('a a a b b'))

tc$tokens
tc$subset(subset = freq_filter(token, min=3))
tc$tokens

get_dtm

Create a document term matrix.

Description

Create a document term matrix. The default output is a sparse matrix (Matrix, TsparseMatrix). Alternatively, the dtm style from the tm and quanteda package can be used.

The dfm function is shorthand for using quanteda’s dfm (document feature matrix) class. The meta data in the tcorpus is then automatically added as docvars in the dfm.
Usage

```r
get_dtm(
  tc,
  feature,
  context_level = c("document", "sentence"),
  weight = c("termfreq", "docfreq", "tfidf", "norm_tfidf"),
  drop_empty_terms = T,
  form = c("Matrix", "tm_dtm", "quanteda_dfm"),
  subset_tokens = NULL,
  subset_meta = NULL,
  context = NULL,
  context_labels = T,
  feature_labels = T,
  ngrams = NA,
  ngram_before_subset = F
)
```

```r
get_dfm(
  tc,
  feature,
  context_level = c("document", "sentence"),
  weight = c("termfreq", "docfreq", "tfidf", "norm_tfidf"),
  drop_empty_terms = T,
  subset_tokens = NULL,
  subset_meta = NULL,
  context = NULL,
  context_labels = T,
  feature_labels = T,
  ngrams = NA,
  ngram_before_subset = F
)
```

Arguments

tc a `tCorpus`

feature The name of the feature column

context_level Select whether the rows of the dtm should represent "documents" or "sentences".

weight Select the weighting scheme for the DTM. Currently supports term frequency (termfreq), document frequency (docfreq), term frequency inverse document frequency (tfidf) and tfidf with normalized document vectors.

drop_empty_terms If True, tokens that do not occur (i.e. column where sum is 0) are ignored.

form The output format. Default is a sparse matrix in the dgTMatrix class from the Matrix package. Alternatives are tm_dtm for a DocumentTermMatrix in the tm package format or quanteda_dfm for the document feature matrix from the quanteda package.

subset_tokens A subset call to select which rows to use in the DTM
subset_meta  A subset call for the meta data, to select which documents to use in the DTM context
Instead of using the document or sentence context, an custom context can be specified. Has to be a vector of the same length as the number of tokens, that serves as the index column. Each unique value will be a row in the DTM.

count_labels  If False, the DTM will not be given rownames
feature_labels  If False, the DTM will not be given column names

ngrams  Optionally, use ngrams instead of individual tokens. This is more memory efficient than first creating an ngram feature in the iCorpus.

ngram_before_subset  If a subset is used, ngrams can be made before the subset, in which case an ngram can contain tokens that have been filtered out after the subset. Alternatively, if ngrams are made after the subset, ngrams will span over the gaps of tokens that are filtered out.

Value
A document term matrix, in the format specified in the form argument

Examples

tc = create_tcorpus(c("First text first sentence. First text first sentence.",
"Second text first sentence"), doc_column = 'id', split_sentences = TRUE)

## Perform additional preprocessing on the 'token' column, and save as the 'feature' column
tc$preprocess('token', 'feature', remove_stopwords = TRUE, use_stemming = TRUE)
tc$tokens

## default: regular sparse matrix, using the Matrix package
m = get_dtm(tc, 'feature')
class(m)
m

## alternatively, create quanteda ('quanteda_dfm') or tm ('tm_dtm') class for DTM
m = get_dtm(tc, 'feature', form = 'quanteda_dfm')
class(m)
m

## create DTM with sentences as rows (instead of documents)
m = get_dtm(tc, 'feature', context_level = 'sentence')
nrow(m)

## use weighting
m = get_dtm(tc, 'feature', weight = 'norm_tfidf')
get_global_i  Compute global feature positions

Description

Features are given global ids, with an added distance (max_window_size) between contexts (e.g., documents, sentences). This way, the distance of features can be calculated across multiple contexts using a single vector.

Usage

get_global_i(
  tc,
  context_level = c("document", "sentence"),
  max_window_size = 200
)

Arguments

tc  tCorpus object
context_level  either 'document' or 'sentence'
max_window_size  Determines the size of the gap between documents. Called max_window_size because this gap determines what the maximum window size is for non-overlapping windows between documents.

Value

a tCorpus object

get_kwic  Get keyword-in-context (KWIC) strings

Description

Create a data.frame with keyword-in-context strings for given indices (i), search results (hits) or search strings (keyword).
get_kwic

Usage

going get_kwic(
  tc,
  hits = NULL,
  i = NULL,
  query = NULL,
  code = "",
  ntokens = 10,
  n = NA,
  nsample = NA,
  output_feature = "token",
  query_feature = "token",
  context_level = c("document", "sentence"),
  kw_tag = c("<", ">"),
  ...
)

Arguments

tc a tCorpus

hits results of feature search. see search_features.

i instead of the hits argument, you can give the indices of features directly.

query instead of using the hits or i arguments, a search string can be given directly. Note that this simply a convenient shorthand for first creating a hits object with search_features. If a query is given, then the ... argument is used to pass other arguments to search_features.

code if 'i' or 'query' is used, the code argument can be used to add a code label. Should be a vector of the same length that gives the code for each i or query, or a vector of length 1 for a single label.

ntokens an integers specifying the size of the context, i.e. the number of tokens left and right of the keyword.

n a number, specifying the total number of hits

nsample like n, but with a random sample of hits. If multiple codes are used, the sample is drawn for each code individually.

output_feature the feature column that is used to make the KWIC.

query_feature If query is used, the feature column that is used to perform the query

context_level Select the maxium context (document or sentence).

kw_tag a character vector of length 2, that gives the symbols before (first value) and after (second value) the keyword in the KWIC string. Can for instance be used to prepare KWIC with format tags for highlighting.

... See search_features for the query parameters
get_stopwords

Details

This is mainly for viewing results in the R console. If you want to create a subset corpus based on the context of query results, you can use subset_query with the window argument. Also, the browse_hits function is a good alternative for viewing query hits in full text.

Examples

tc = tokens_to_tcorpus(corenlp_tokens, sentence_col = 'sentence', token_id_col = 'id')

## look directly for a term (or complex query)
get_kwic(tc, query = 'love*')

## or, first perform a feature search, and then get the KWIC for the results
hits = search_features(tc, '(john OR mark) AND mary AND love*', context_level = 'sentence')
get_kwic(tc, hits=hits, context_level = 'sentence')

get_stopwords

Get a character vector of stopwords

Description

Get a character vector of stopwords

Usage

get_stopwords(lang)

Arguments


Value

A character vector containing stopwords

Examples

en_stop = get_stopwords('english')
nl_stop = get_stopwords('dutch')
ge_stop = get_stopwords('german')

head(en_stop)
head(nl_stop)
head(ge_stop)
**laplace**

Laplace (i.e. add constant) smoothing

**Description**

Laplace (i.e. add constant) smoothing

**Usage**

```r
laplace(freq, add = 0.5)
```

**Arguments**

- `freq`: A numeric vector of term frequencies (integers).
- `add`: The added value

**Value**

A numeric vector with the smoothed term proportions

**Examples**

```r
laplace(c(0,0,1,1,1,2,2,2,3,3,4,7,10))
```

---

**melt_quanteda_dict**

Convert a quanteda dictionary to a long data.table format

**Description**

This is used internally in the tCorpus dictionary search functions, but can be used manually for more control. For example, adding numeric scores for sentiment dictionaries, and specifying which label/code to use in `search_dictionary()`.

**Usage**

```r
melt_quanteda_dict(dict, column = "code", .index = NULL)
```

**Arguments**

- `dict`: The quanteda dictionary
- `column`: The name of the column with the label/code. If dictionary contains multiple levels, additional columns are added with the suffix `_l[i]`, where `[i]` is the level.
- `.index`: Do not use (used for recursive melting)
merge_tcorpora

Merge tCorpus objects

Description
Create one tcorpus based on multiple tcorpus objects

Usage
merge_tcorpora(
  ..., 
  keep_data = c("intersect", "all"),
  keep_meta = c("intersect", "all"),
  if_duplicate = c("stop", "rename", "drop"),
  duplicate_tag = "#D"
)

Arguments
... tCorpus objects, or a list with tcorpus objects
keep_data if 'intersect', then only the token data columns that occur in all tCorpus objects are kept
keep_meta if 'intersect', then only the document meta columns that occur in all tCorpus objects are kept
if_duplicate determine behaviour if there are duplicate doc_ids across tcorpora. By default, this yields an error, but you can set it to "rename" to change the names of duplicates (which makes sense of only the doc_ids are duplicate, but not the actual content), or "drop" to ignore duplicates, keeping only the first unique occurrence.
duplicate_tag a character string. if if_duplicates is "rename", this tag is added to the document id. (this is repeated till no duplicates remain)

Value
a tCorpus object

Examples

d = quanteda::data_dictionary_LSD2015
melt_quanteda_dict(d)
**plot.contextHits**  

**Description**  
S3 plot for contextHits class

**Usage**  
## S3 method for class 'contextHits'
plot(x, min_weight = 0, backbone_alpha = NA, ...)

**Arguments**  
x an object of class 'contextHits'
min_weight Optionally, the minimum weight for an edge in the network
backbone_alpha Optionally, the alpha threshold for backbone extraction (similar to a p-value, and lower is more strict)
... not used

**Examples**  
## Not run:
tc = create_tcorpus(sotu_texts, doc_column='id')
hits = search_contexts(tc, c('War# war* OR army OR bomb*','Terrorism# terroris*','Economy# econom* OR bank*','Education# educat* OR school*'))
plot(hits)
## End(Not run)

plot.featureAssociations

**Description**

visualize feature associations

**Usage**

```r
## S3 method for class 'featureAssociations'
plot(x, n = 25, size = c("chi2", "freq", "ratio"), ...)
```

**Arguments**

- `x`: a featureAssociations object, created with the `feature_associations` function
- `n`: the number of words in the plot
- `size`: use "freq", "chi2" or "ratio" for determining the size of words
- `...`: additional arguments passed to dtm_wordcloud

**Examples**

```r
## as example, compare SOTU paragraphs about taxes to rest
tc = create_tcorpus(sotu_texts[1:100,], doc_column = 'id')
comp = compare_subset(tc, 'token', query_x = 'tax*')

plot(comp, balance=TRUE)
plot(comp, mode = 'ratio_x')
plot(comp, mode = 'ratio_y')
```

plot.featureHits

**Description**

S3 plot for featureHits class

**Usage**

```r
## S3 method for class 'featureHits'
plot(x, min_weight = 0, backbone_alpha = NA, ...)
```
Arguments

- **x** a featureHits object, as returned by `search_features`
- **min_weight** Optionally, the minimum weight for an edge in the network
- **backbone_alpha** Optionally, the alpha threshold for backbone extraction (similar to a p-value, and lower is more strict)

Examples

```r
tc = create_tcorpus(sotu_texts, doc_column='id')
hits = search_features(tc, c('War# war* OR army OR bomb*','Terrorism# terroris*','Economy# econom* OR bank*','Education# educat* OR school*'))
plot(hits)
```

Description

visualize vocabularyComparison

Usage

```r
## S3 method for class 'vocabularyComparison'
plot(
  x,
  n = 25,
  mode = c("both", "ratio_x", "ratio_y"),
  balance = T,
  size = c("chi2", "freq", "ratio"),
  ...
)
```

Arguments

- **x** a vocabularyComparison object, created with the `compare_corpus` or `compare_subset` method
- **n** the number of words in the plot
- **mode** use "both" to plot both overrepresented and underrepresented words using the plot_words function. Whether a term is under- or overrepresented is indicated on the x-axis, which shows the log ratios (negative is underrepresented, positive is overrepresented). Use "ratio_x" or "ratio_y" to only plot overrepresented or underrepresented words using dtm_wordcloud
balance if TRUE, get an equal amount of terms on the left (underrepresented) and right (overrepresented) side. If FALSE, the top chi words are used, regardless of ratio.
size use "freq", "chi2" or "ratio" for determining the size of words
... additional arguments passed to plot_words ("both" mode) or dtm_wordcloud (ratio modes)

Examples

```r
## as example, compare SOTU paragraphs about taxes to rest
tc = create_tcorpus(sotu_texts[1:100], doc_column = 'id')
comp = compare_subset(tc, 'token', query_x = 'tax*')

plot(comp, balance=TRUE)
plot(comp, mode = 'ratio_x')
plot(comp, mode = 'ratio_y')
```

plot_semnet Visualize a semnet network

Description

plot_semnet is a wrapper for the plot.igraph() function optimized for plotting a semantic network of the "semnet" class.

Usage

```r
plot_semnet(
  g,
  weight_attr = "weight",
  min_weight = NA,
  delete_isolates = F,
  vertexsize_attr = "freq",
  vertexsize_coef = 1,
  vertexcolor_attr = NA,
  edgewidth_coef = 1,
  max_backbone_alpha = NA,
  labelsize_coef = 1,
  labelspace_coef = 1.1,
  reduce_labeloverlap = F,
  redo_layout = F,
  return_graph = T,
  vertex.label.dist = 0.25,
  layout_fun = igraph::layout_with_fr,
  ...
)
```
Arguments

- **g**: A network in the igraph format. Specifically designed for the output of coOccurrenceNetwork() and windowedCoOccurenceNetwork().
- **weight_attr**: The name of the weight attribute. Default is 'weight'.
- **min_weight**: The minimum weight. All edges with a lower weight are dropped.
- **delete_isolates**: If TRUE, isolate vertices (also after applying min_weight) are dropped.
- **vertexsize_attr**: A character string indicating a vertex attribute that represents size. Default is 'freq', which is created in the coOccurrenceNetwork functions to indicate the number of times a token occurred.
- **vertexsize_coef**: A coefficient for changing the vertex size.
- **vertexcolor_attr**: A character string indicating a vertex attribute that represents color. The attribute can also be a numeric value (e.g., a cluster membership) in which case colors are assigned to numbers. If no (valid) color attribute is given, vertex color are based on undirected fastgreedy.community() clustering.
- **edgewidth_coef**: A coefficient for changing the edge width.
- **max_backbone_alpha**: If g has an edge attribute named alpha (added if backbone extraction is used), this specifies the maximum alpha value.
- **labelsize_coef**: A coefficient for increasing or decreasing the size of the vertex label.
- **labelspace_coef**: A coefficient that roughly determines the minimal distance between vertex labels, based on the size of labels. Only used if reduce_labeloverlap is TRUE.
- **reduce_labeloverlap**: If TRUE, an algorithm is used to reduce overlap as best as possible.
- **redo_layout**: If TRUE, a new layout will be calculated using layout_with_fr(). If g does not have a layout attribute (g$layout), a new layout is automatically calculated.
- **return_graph**: If TRUE, plot_semnet() also returns the graph object with the attributes and layout as shown in the plot.
- **vertex.label.dist**: The distance of the label to the center of the vertex.
- **layout_fun**: The igraph layout function that is used.
- **...**: Additional arguments are passed on to plot.igraph().

Details

Before plotting the network, the set_network_attributes() function is used to set pretty defaults for plotting. Optionally, reduce_labeloverlap can be used to prevent label overlap (as much as possible).

Value

Plots a network, and returns the network object if return_graph is TRUE.
Examples

```r
tc = create_tcorpus(sotu_texts, doc_column = 'id')
tc$preprocess('token', 'feature', remove_stopwords = TRUE, use_stemming = TRUE, min_docfreq=10)

g = semnet_window(tc, 'feature', window.size = 10)
g = backbone_filter(g, max_vertices = 100)
plot_semnet(g)
```

---

**plot_words**

Plot a wordcloud with words ordered and coloured according to a dimension (x)

**Description**

Plot a wordcloud with words ordered and coloured according to a dimension (x)

**Usage**

```r
plot_words(
  x,
  y = NULL,
  words,
  wordfreq = rep(1, length(x)),
  xlab = '',
  ylab = '',
  yaxt = "n",
  scale = 1,
  random.y = T,
  xlim = NULL,
  ylim = NULL,
  col = c("darkred", "navyblue"),
  fixed_col = NULL,
  ...
)
```

**Arguments**

- `x` The (approximate) x positions of the words
- `y` The (approximate) y positions of the words
- `words` A character vector with the words to plot
- `wordfreq` The frequency of the words, defaulting to 1
- `xlab` Label of the x axis
- `ylab` Label of the y axis
- `yaxt` see `par` documentation
scale        Maximum size to scale the wordsize
random.y    if TRUE, the y position of words is random, otherwise it represents the word
            frequency.
xlim        Starting value of x axis
ylim        Starting value of y axis
col         A vector of colors that is passed to colorRamp to interpolate colors over x axis
fixed_col   Optionally, a vector of the exact colors given to words.
...         additional parameters passed to the plot function

Value

nothing

Examples

x = c(-10, -5, 3, 5)
y = c(0, 2, 5, 10)
words = c('words', 'where', 'you', 'like')

plot_words(x, y, words, c(1,2,3,4))

Description

Preprocess tokens in a character vector

Usage

preprocess_tokens(
  x,
  context = NULL,
  language = "english",
  use_stemming = F,
  lowercase = T,
  ngrams = 1,
  replace_whitespace = F,
  as_ascii = F,
  remove_punctuation = T,
  remove_stopwords = F,
  remove_numbers = F,
  min_freq = NULL,
  min_docfreq = NULL,
max_freq = NULL,
max_docfreq = NULL,
min_char = NULL,
max_char = NULL,
ngram_skip_empty = T)

Arguments

x       A character or factor vector in which each element is a token (i.e. a tokenized text)
context Optionally, a character vector of the same length as x, specifying the context of token (e.g., document, sentence). Has to be given if ngram > 1
language The language used for stemming and removing stopwords
use_stemming Logical, use stemming. (Make sure the specify the right language!)
lowercase Logical, make token lowercase
ngrams A number, specifying the number of tokens per ngram. Default is unigrams (1).
replace_whitespace Logical. If TRUE, all whitespace is replaced by underscores
as_ascii Logical. If TRUE, tokens will be forced to ascii
remove_punctuation Logical. if TRUE, punctuation is removed
remove_stopwords Logical. If TRUE, stopwords are removed. (Make sure to specify the right language!)
remove_numbers remove features that are only numbers
min_freq an integer, specifying minimum token frequency.
min_docfreq an integer, specifying minimum document frequency.
max_freq an integer, specifying minimum token frequency.
max_docfreq an integer, specifying minimum document frequency.
min_char an integer, specifying minimum number of characters in a term
max_char an integer, specifying maximum number of characters in a term
ngram_skip_empty if ngrams are used, determines whether empty (filtered out) terms are skipped (i.e. c("this", NA, "test"), becomes "this_test") or

Value

a factor vector
Examples

tokens = c('I', 'am', 'a', 'SHORT', 'example', 'sentence', '!

## default is lowercase without punctuation
preprocess_tokens(tokens)

## optionally, delete stopwords, perform stemming, and make ngrams
preprocess_tokens(tokens, remove_stopwords = TRUE, use_stemming = TRUE)
preprocess_tokens(tokens, context = NA, ngrams = 3)

print.contextHits  
S3 print for contextHits class

Description

S3 print for contextHits class

Usage

## S3 method for class 'contextHits'
print(x, ...)

Arguments

x  
a contextHits object, as returned by search_contexts

...  
not used

Examples

text = c('A B C', 'D E F. G H I', 'A D', 'GGG')
tc = create_tcorpus(text, doc_id = c('a','b','c','d'), split_sentences = TRUE)
hits = search_contexts(tc, c('query label# A AND B', 'second query# (A AND Q) OR ("D E") OR I'))
hits

print.featureHits  
S3 print for featureHits class

Description

S3 print for featureHits class

Usage

## S3 method for class 'featureHits'
print(x, ...)
refresh_tcorpus

Arguments

x  a featureHits object, as returned by search_features
...

Examples

text = c("Var A B C", "D E F. G H I", "A D", "GGG")
tc = create_tcorpus(text, doc_id = c("a","b","c","d"), split_sentences = TRUE)
hits = search_features(tc, c("query label# A AND B", "second query# (A AND Q) OR ("D E") OR I"))
hits

print.tCorpus  S3 print for tCorpus class

Description

S3 print for tCorpus class

Usage

## S3 method for class 'tCorpus'
print(x, ...)

Arguments

x  a tCorpus object
...

Examples

tc = create_tcorpus(c("First text", "Second text"))
print(tc)

refresh_tcorpus  Refresh a tCorpus object using the current version of corpustools

Description

As an R6 class, tCorpus contains its methods within the class object (i.e. itself). Therefore, if you use a new version of corpustools with an older tCorpus object (e.g., stored as a .rds. file), then the methods are not automatically updated. You can then use refresh_tcorpus() to reinitialize the tCorpus object with the current version of corpustools.
require_package

Usage

refresh_tcorpus(tc)

Arguments

tc a tCorpus object

Value

a tCorpus object

Examples

tc = create_tcorpus(c('First text', 'Second text'))
refresh_tcorpus(tc)

require_package

Check if package with given version exists

Description

Check if package with given version exists

Usage

require_package(package, min_version = NULL)

Arguments

package The name of the package
min_version The minimum version

Value

An error if package does not exist
search_contexts  

Search for documents or sentences using Boolean queries

Description

Search for documents or sentences using Boolean queries

Usage

search_contexts(
  tc,
  query, 
  code = NULL,  
  feature = "token", 
  context_level = c("document", "sentence"),  
  not = F,  
  verbose = F,  
  as_ascii = F
)

Arguments

tc            a tCorpus
query         A character string that is a query. See details for available query operators and modifiers. Can be multiple queries (as a vector), in which case it is recommended to also specify the code argument, to label results.
code          If given, used as a label for the results of the query. Especially useful if multiple queries are used.
feature       The name of the feature column
context_level Select whether the queries should occur within while "documents" or specific "sentences". Returns results at the specified level.
not           If TRUE, perform a NOT search. Return the articles/sentences for which the query is not found.
verbose       If TRUE, progress messages will be printed
as_ascii      if TRUE, perform search in ascii.

Details

Brief summary of the query language

The following operators and modifiers are supported:

- The standard Boolean operators: AND, OR and NOT. As a shorthand, an empty space can be used as an OR statement, so that "this that those" means "this OR that OR those". NOT statements strictly mean AND NOT, so should only be used between terms. If you want to find everything except certain terms, you can use * (wildcard for anything) like this: "* NOT (this that those)".
• For complex queries parentheses can (and should) be used. e.g. '(spam AND eggs) NOT (fish and (chips OR albatros))
• Wildcards ? and *. The questionmark can be used to match 1 unknown character or no character at all, e.g. "?at" would find "cat", "hat" and "at". The asterisk can be used to match any number of unknown characters. Both the asterisk and questionmark can be used at the start, end and within a term.
• Multitoken strings, or exact strings, can be specified using quotes. e.g. "united states"
• Tokens within a given token distance can be found using quotes plus tilde and a number specifying the token distance. e.g. "climate change"~10
• Alternatively, angle brackets (<>) can be used instead of quotes, which also enables nesting exact strings in proximity/window search
• Queries are not case sensitive, but can be made so by adding the ~s flag. e.g. COP~s only finds "COP" in uppercase. The ~s flag can also be used on quotes to make all terms within quotes case sensitive, and this can be combined with the token proximity flag. e.g. "Marco Polo"~s10

Value

A contextHits object, which is a list with $hits (data.frame with locations) and $queries (copy of queries for provenance)

Examples

text = c('A B C', 'D E F. G H I', 'A D', 'GGG')
tc = create_tcorpus(text, doc_id = c('a','b','c','d'), split_sentences = TRUE)
tc$tokens

hits = search_contexts(tc, c('query label# A AND B', 'second query# (A AND Q) OR ("D E") OR I\')
hits$hits ## print shows number of hits
summary(hits) ## summary gives hits per query

## sentence level

hits = search_contexts(tc, c('query label# A AND B', 'second query# (A AND Q) OR ("D E") OR I\'),
context_level = 'sentence')
hits$hits ## hits is a list, with hits$hits being a data.frame with specific contexts

## query language examples

## single term

search_contexts(tc, 'A')$hits

search_contexts(tc, 'G*')$hits ## wildcard *
search_contexts(tc, '*G')$hits ## wildcard *
search_contexts(tc, 'G*G')$hits ## wildcard *
search_contexts(tc, 'G?G')$hits ## wildcard ? (no hits)
## boolean

```
search_contexts(tc, 'A AND B')$hits
search_contexts(tc, 'A AND D')$hits
search_contexts(tc, 'A AND (B OR D)')$hits

search_contexts(tc, 'A NOT B')$hits
search_contexts(tc, 'A NOT (B OR D)')$hits
```

## sequence search (adjacent words)

```
search_contexts(tc, "A B")$hits
search_contexts(tc, "A C")$hits ## no hit, because not adjacent

search_contexts(tc, "A (B OR D)")$hits ## can contain nested OR
## cannot contain nested AND or NOT!!

search_contexts(tc, '<A B>')$hits ## can also use <> instead of "",
```

## proximity search (using ~ flag)

```
search_contexts(tc, "A <B>C">")$hits ## A AND C within a 5 word window
search_contexts(tc, "A <C>"-1")$hits ## no hit, because A and C more than 1 word apart

search_contexts(tc, "A (B OR D)"-5")$hits ## can contain nested OR
search_contexts(tc, "A <B>C">-5")$hits ## can contain nested sequence (must use <>)
## (<> is always OK, but cannot nest quotes in quotes)
## cannot contain nested AND or NOT!!
```

## case sensitive search

```
search_contexts(tc, 'g')$hits ## normally case insensitive
search_contexts(tc, 'g-s')$hits ## use ~s flag to make term case sensitive

search_contexts(tc, '(a OR g)-s')$hits ## use ~s flag on everything between parentheses
search_contexts(tc, '(a OR G)-s')$hits ## use ~s flag on everything between parentheses

search_contexts(tc, "a b"-s')$hits ## use ~s flag on everything between quotes
search_contexts(tc, "A B"-s')$hits ## use ~s flag on everything between quotes
```

---

**search_dictionary**

**Dictionary lookup**

**Description**

Similar to search_features, but for fast matching of large dictionaries.
**search_dictionary**

Usage

```r
search_dictionary(
  tc,
  dict,
  token_col = "token",
  string_col = "string",
  code_col = "code",
  sep = " ",
  mode = c("unique_hits", "features"),
  case_sensitive = F,
  use_wildcards = T,
  ascii = F,
  verbose = F
)
```

Arguments

- **tc**: A tCorpus
- **dict**: A dictionary. Can be either a data.frame or a quanteda dictionary. If a data.frame is given, it has to have a column named "string" (or use string_col argument) that contains the dictionary terms, and a column "code" (or use code_col argument) that contains the label/code represented by this string. Each row has a single string, that can be a single word or a sequence of words separated by a whitespace (e.g., "not bad"), and can have the common ? and * wildcards. If a quanteda dictionary is given, it is automatically converted to this type of data.frame with the `melt_quanteda_dict` function. This can be done manually for more control over labels.
- **token_col**: The feature in tc that contains the token text.
- **string_col**: If dict is a data.frame, the name of the column in dict with the dictionary lookup string. Default is "string".
- **code_col**: The name of the column in dict with the dictionary code/label. Default is "code". If dict is a quanteda dictionary with multiple levels, "code_l2", "code_l3", etc. can be used to select levels.
- **sep**: A regular expression for separating multi-word lookup strings (default is " ", which is what quanteda dictionaries use). For example, if the dictionary contains "Barack Obama", sep should be " " so that it matches the consecutive tokens "Barack" and "Obama". In some dictionaries, however, it might say "Barack+Obama", so in that case sep = '\+' should be used.
- **mode**: There are two modes: "unique_hits" and "features". The "unique_hits" mode prioritizes finding unique matches, which is recommended for counting how often a dictionary term occurs. If a term matches multiple dictionary terms (which should only happen for nested multi-word terms, such as "bad" and "not bad"), the longest term is always used. The features mode does not delete duplicates.
- **case_sensitive**: logical, should lookup be case sensitive?
- **use_wildcards**: Use the wildcards * (any number including none of any character) and ? (one or none of any character). If FALSE, exact string matching is used
search_features

If true, convert text to ascii before matching
If true, report progress

Value
A vector with the id value (taken from dict$id) for each row in tc$tokens

Examples

dict = data.frame(string = c('this is', 'for a', 'not big enough'), code=c('a','c','b'))
tc = create_tcorpus(c('this is a test','This town is not big enough for a test'))
search_dictionary(tc, dict)$hits

search_features  Find tokens using a Lucene-like search query

Description
Search tokens in a tokenlist using Lucene-like queries. For a detailed explanation of the query language, see the details below.

Usage
search_features(
  tc,
  query,
  code = NULL,
  feature = "token",
  mode = c("unique_hits", "features"),
  context_level = c("document", "sentence"),
  keep_longest = TRUE,
  as_ascii = F,
  verbose = F
)

Arguments

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>tc</td>
<td>a tCorpus</td>
</tr>
<tr>
<td>query</td>
<td>A character string that is a query. See details for available query operators and modifiers. Can be multiple queries (as a vector), in which case it is recommended to also specify the code argument, to label results.</td>
</tr>
<tr>
<td>code</td>
<td>The code given to the tokens that match the query (usefull when looking for multiple queries). Can also put code label in query with # (see details)</td>
</tr>
<tr>
<td>feature</td>
<td>The name of the feature column within which to search.</td>
</tr>
</tbody>
</table>
There are two modes: "unique_hits" and "features". The "unique_hits" mode prioritizes finding full and unique matches, which is recommended for counting how often a query occurs. However, this also means that some tokens for which the query is satisfied might not assigned a hit_id. The "features" mode, instead, prioritizes finding all tokens, which is recommended for coding coding features (the code_features and search_recode methods always use features mode).

Select whether the queries should occur within "documents" or specific "sentences".

If TRUE, then overlapping in case of overlapping queries strings in unique_hits mode, the query with the most separate terms is kept. For example, in the text "mr. Bob Smith", the query [smith OR "bob smith"] would match "Bob" and "Smith". If keep_longest is FALSE, the match that is used is determined by the order in the query itself. The same query would then match only "Smith".

If TRUE, perform search in ascii.

If TRUE, progress messages will be printed

Brief summary of the query language

The following operators and modifiers are supported:

- The standard Boolean operators: AND, OR and NOT. As a shorthand, an empty space can be used as an OR statement, so that "this that those" means "this OR that OR those". NOT statements strictly mean AND NOT, so should only be used between terms. If you want to find everything except certain terms, you can use * (wildcard for anything) like this: "* NOT (this that those)".

- For complex queries parentheses can (and should) be used. e.g. '(spam AND eggs) NOT (fish and (chips OR albatros))

- Wildcards ? and *. The questionmark can be used to match 1 unknown character or no character at all, e.g. "?at" would find "cat", "hat" and "at". The asterisk can be used to match any number of unknown characters. Both the asterisk and questionmark can be used at the start, end and within a term.

- Multitoken strings, or exact strings, can be specified using quotes. e.g. "united states"

- tokens within a given token distance can be found using quotes plus tilde and a number specifying the token distance. e.g. "climate chang*"~10

- Alternatively, angle brackets (<> can be used instead of quotes, which also enables nesting exact strings in proximity/window search

- Queries are not case sensitive, but can be made so by adding the ~s flag. e.g. COP~s only finds "COP" in uppercase. The ~s flag can also be used on parentheses or quotes to make all terms within case sensitive, and this can be combined with the token proximity flag. e.g. "Marco Polo"~s10

- The ~g (ghost) flag can be used to mark a term (or all terms within parentheses/quotes) as a ghost term. This has two effects. Firstly, features that match the query term will not be in the results. This is useful if a certain term is important for getting reliable search results, but not conceptually relevant. Secondly, ghost terms can be used multiple times, in different query
hits (only relevant in unique_hits mode). For example, in the text "A B C", the query 'A~g AND (B C)' will return both B and C as separate hit, whereas 'A AND (B C)' will return A and B as a single hit.

- A code label can be included at the beginning of a query, followed by a # to start the query (label# query). Note that to search for a hashtag symbol, you need to escape it with \ in R character vector

- Aside from the feature column (specified with the feature argument) a query can include any column in the token data. To manually select a column, use 'columnname: ' at the start of a query or nested query (i.e. between parentheses or quotes). See examples for clarification.

Value

A featureHits object, which is a list with $hits (data.frame with locations) and $queries (copy of queries for provenance)

Examples

text = c('A B C', 'D E F. G H I', 'A D', 'GGG')
tc = create_tcorpus(text, doc_id = c('a','b','c','d'), split_sentences = TRUE)
tc$tokens ## (example uses letters instead of words for simple query examples)

hits = search_features(tc, c('query label# A AND B', 'second query# (A AND Q) OR ("D E") OR I'))
ahits$hits ## print shows number of hits

summary(hits) ## summary gives hits per query

## sentence level
hits = search_features(tc, c('query label# A AND B', 'second query# (A AND Q) OR ("D E") OR I'),
 context_level = 'sentence')
hits$hits ## hits is a list, with hits$hits being a data.frame with specific features

## query language examples

## single term
search_features(tc, 'A')$hits

search_features(tc, 'G*')$hits ## wildcard *

search_features(tc, '*G')$hits ## wildcard *

search_features(tc, 'G*G')$hits ## wildcard *

search_features(tc, 'G?G')$hits ## wildcard ?

search_features(tc, 'G?')$hits ## wildcard ? (no hits)

## boolean
search_features(tc, 'A AND B')$hits
search_features(tc, 'A AND D')$hits
search_features(tc, 'A AND (B OR D)')$hits
search_features(tc, 'A NOT B')$hits
search_features(tc, 'A NOT (B OR D)')$hits

## sequence search (adjacent words)
search_features(tc, '"A B"')$hits
search_features(tc, '"A C"')$hits ## no hit, because not adjacent

search_features(tc, '"A (B OR D)"')$hits ## can contain nested OR
## cannot contain nested AND or NOT!!

search_features(tc, '<A B>')$hits ## can also use <> instead of "."

## proximity search (using ~ flag)
search_features(tc, '"A C"~5')$hits ## A AND C within a 5 word window
search_features(tc, '"A C"~1')$hits ## no hit, because A and C more than 1 word apart

search_features(tc, '"A (B OR D)"~5')$hits ## can contain nested OR
search_features(tc, '"A <B C>"~5')$hits ## can contain nested sequence (must use <>)
search_features(tc, '<A <B C>~5')$hits ## <> is always OK, but cannot nest "" in ""
## cannot contain nested AND or NOT!!

## case sensitive search (~s flag)
search_features(tc, 'g')$hits ## normally case insensitive
search_features(tc, 'g~s')$hits ## use ~s flag to make term case sensitive

search_features(tc, '(a OR g)~s')$hits ## use ~s flag on everything between parentheses
search_features(tc, '(a OR G)~s')$hits

search_features(tc, '"a b"~s')$hits ## use ~s flag on everything between quotes
search_features(tc, '"A B"~s')$hits ## use ~s flag on everything between quotes

## ghost terms (~g flag)
search_features(tc, 'A AND B~g')$hits ## ghost term (~g) has to occur, but is not returned
search_features(tc, 'A AND Q~g')$hits ## no hi

# (can also be used on parentheses/quotes/anglebrackets for all nested terms)

## "unique_hits" versus "features" mode
tc = create_tcorpus('A A B')

search_features(tc, 'A AND B')$hits ## in "unique_hits" (default), only match full queries
# (B is not repeated to find a second match of A AND B)

search_features(tc, 'A AND B', mode = 'features')$hits ## in "features", match any match
# (note that hit_id in features mode is irrelevant)

# ghost terms (used for conditions) can be repeated
search_features(tc, 'A AND B~g')$hits
semnet

Create a semantic network based on the co-occurrence of tokens in documents

Description

This function calculates the co-occurrence of features and returns a network/graph in the igraph format, where nodes are tokens and edges represent the similarity/adjacency of tokens. Co-occurrence is calculated based on how often two tokens occurred within the same document (e.g., news article, chapter, paragraph, sentence). The semnet_window() function can be used to calculate co-occurrence of tokens within a given token distance.

Usage

```r
semnet(
  tc,
  feature = "token",
  measure = c("con_prob", "con_prob_weighted", "cosine", "count_directed",
               "count_undirected", "chi2"),
  context_level = c("document", "sentence"),
  backbone = F,
  n.batches = NA
)
```

Arguments

- `tc`: a tCorpus or a featureHits object (i.e. the result of search_features)
- `feature`: The name of the feature column
- `measure`: The similarity measure. Currently supports: "con_prob" (conditional probability), "con_prob_weighted", "cosine" similarity, "count_directed" (i.e number of cooccurrences) and "count_undirected" (same as count_directed, but returned as an undirected network, chi2 (chi-square score))
- `context_level`: Determine whether features need to co-occurn within "documents" or "sentences"
- `backbone`: If True, add an edge attribute for the backbone alpha
- `n.batches`: If a number, perform the calculation in batches

Value

an Igraph graph in which nodes are features and edges are similarity scores

Examples

```r
text = c('A B C', 'D E F G H I', 'A D', 'GGG')
tc = create_tcorpus(text, doc_id = c('a', 'b', 'c', 'd'), split_sentences = TRUE)
g = semnet(tc, 'token')
```
semnet_window

Create a semantic network based on the co-occurrence of tokens in token windows

Description

This function calculates the co-occurrence of features and returns a network/graph in the igraph format, where nodes are tokens and edges represent the similarity/adjacency of tokens. Co-occurrence is calculated based on how often two tokens co-occur within a given token distance.

If a featureHits object is given as input, then for for query hits that have multiple positions (i.e. terms connected with AND statements or word proximity) the raw count score is biased. For the count_* measures therefore only the first position of the query hit is used.

Usage

semnet_window(
  tc,
  feature = "token",
  measure = c("con_prob", "cosine", "count_directed", "count_undirected", "chi2"),
  context_level = c("document", "sentence"),
  window.size = 10,
  direction = "<>",
  backbone = F,
  n.batches = 5,
  matrix_mode = c("positionXwindow", "windowXwindow")
)

Arguments

- **tc**: a tCorpus or a featureHits object (i.e. the result of search_features)
- **feature**: The name of the feature column
- **measure**: The similarity measure. Currently supports: "con_prob" (conditional probability), "cosine" similarity, "count_directed" (i.e. number of cooccurrences) and "count_undirected" (same as count_directed, but returned as an undirected network, chi2 (chi-square score))
- **context_level**: Determine whether features need to co-occur within "documents" or "sentences"
- **window.size**: The token distance within which features are considered to co-occur
- **direction**: Determine whether co-occurrence is asymmetric ("<>") or takes the order of tokens into account. If direction is '<', then the from/x feature needs to occur before the to/y feature. If direction is '>', then after.
- **backbone**: If True, add an edge attribute for the backbone alpha
n.batches
To limit memory use the calculation is divided into batches. This parameter controls the number of batches.

matrix_mode
There are two approaches for calculating window co-occurrence (see details). By default we use positionXmatrix, but matrixXmatrix is optional because it might be favourable for some uses, and might make more sense for cosine similarity.

Details
There are two approaches for calculating window co-occurrence. One is to measure how often a feature occurs within a given token window, which can be calculating by calculating the inner product of a matrix that contains the exact position of features and a matrix that contains the occurrence window. We refer to this as the "positionXwindow" mode. Alternatively, we can measure how much the windows of features overlap, for which take the inner product of two window matrices, which we call the "windowXwindow" mode. The positionXwindow approach has the advantage of being easy to interpret (e.g. how likely is feature "Y" to occur within 10 tokens from feature "X"?). The windowXwindow mode, on the other hand, has the interesting feature that similarity is stronger if tokens co-occur more closely together (since then their windows overlap more), but this only works well for similarity measures that normalize the similarity (e.g., cosine). Currently, we only use the positionXwindow mode, but windowXwindow could be interesting to use as well, and for cosine it might actually make more sense.

Value
an Igraph graph in which nodes are features and edges are similarity scores

Examples
```r
text = c("A B C", "D E F. G H I", "A D", "GGG")
tc = create_tcorpus(text, doc_id = c("a","b","c","d"), split_sentences = TRUE)
g = semnet_window(tc, 'token', window.size = 1)
g
igraph::get.data.frame(g)
plot_semnet(g)
```

---

set_network_attributes

Set some default network attributes for pretty plotting

Description
The purpose of this function is to create some default network attribute options to plot networks in a nice and insightfull way.
Usage

```
set_network_attributes(
  g,
  size_attribute = "freq",
  color_attribute = NA,
  redo_layout = F,
  edgewidth_coef = 1,
  layout_fun = igraph::layout_with_fr
)
```

Arguments

- **g**: A graph in the Igraph format.
- **size_attribute**: the name of the vertex attribute to be used to set the size of nodes
- **color_attribute**: the name of the attribute that is used to select the color
- **redo_layout**: if TRUE, force new layout if layout already exists as a graph attribute
- **edgewidth_coef**: A coefficient for changing the edge width
- **layout_fun**: THe igraph layout function used

Value

a network in the Igraph format

Examples

```r
tc = create_tcorpus(c('A B C', 'B C', 'B D'))
g = semnet(tc, 'token')

igraph::get.edge.attribute(g)
igraph::get.vertex.attribute(g)
plot(g)
g = set_network_attributes(g, size_attribute = 'freq')
igraph::get.edge.attribute(g)
igraph::get.vertex.attribute(g)
plot(g)
```

---

**sgt**

*Simple Good Turing smoothing*

Description

Usage
sgt(freq)

Arguments
freq A numeric vector of frequencies (integers).

Value
A numeric vector with the smoothed term proportions

show_udpipe_models  Show the names of udpipe models

Description
Returns a data.table with the language, treebank and udpipe_model name. Uses the default model repository provided by the udpipe package (udpipe_download_model). For more information about udpipe and performance benchmarks of the UD models, see the GitHub page of the udpipe package.

Usage
show_udpipe_models()

Value
a data.frame

Examples
show_udpipe_models()

sotu_texts  State of the Union addresses

Description
State of the Union addresses

Usage
data(sotu_texts)

Format
data.frame
stopwords_list

---

**stopwords_list**  
*Basic stopword lists*

---

**Description**

Basic stopword lists

**Usage**

data(stopwords_list)

**Format**

A named list, with names matching the languages used by SnowballC

---

**subset.tCorpus**  
*S3 subset for tCorpus class*

---

**Description**

S3 subset for tCorpus class

**Usage**

```r
## S3 method for class 'tCorpus'
subset(x, subset = NULL, subset_meta = NULL, window = NULL, ...)
```

**Arguments**

- `x`  
a tCorpus object

- `subset`  
logical expression indicating rows to keep in the tokens data.

- `subset_meta`  
logical expression indicating rows to keep in the document meta data.

- `window`  
If not NULL, an integer specifying the window to be used to return the subset. For instance, if the subset contains token 10 in a document and window is 5, the subset will contain token 5 to 15. Naturally, this does not apply to subset_meta.

- `...`  
not used
## Examples

```r
## create tcorpus of 5 bush and obama docs
tc = create_tcorpus(sotu_texts[c(1:5,801:805),], doc_col='id')

## subset to keep only tokens where token_id <= 20 (i.e.first 20 tokens)
tcs1 = subset(tc, token_id < 20)
tcs1

## subset to keep only documents where president is Barack Obama
tcs2 = subset(tc, subset_meta = president == 'Barack Obama')
tcs2
```

### subset_query

**Subset tCorpus token data using a query**

**Description**

A convenience function that searches for contexts (documents, sentences), and uses the results to subset the tCorpus token data.

**Usage**

```r
subset_query(
  tc,  
  query,  
  feature = "token",  
  context_level = c("document", "sentence"),  
  not = F,  
  as_ascii = F,  
  window = NA
)
```

**Arguments**

- **tc**  
  A tCorpus

- **query**  
  A character string that is a query. See `search_contexts` for query syntax.

- **feature**  
  The name of the feature columns on which the query is used.

- **context_level**  
  Select whether the query and subset are performed at the document or sentence level.

- **not**  
  If TRUE, perform a NOT search. Return the articles/sentences for which the query is not found.

- **as_ascii**  
  If TRUE, perform search in ascii.

- **window**  
  If used, uses a word distance as the context (overrides context_level)

**Details**

See the documentation for `search_contexts` for an explanation of the query language.
```r
Examples
text = c('A B C', 'D E F. G H I', 'A D', 'GGG')
tc = create_tcorpus(text, doc_id = c('a','b','c','d'), split_sentences = TRUE)

## subset by reference
tc2 = subset_query(tc, 'A')
tc2$meta

summary.contextHits  S3 summary for contextHits class

Description
S3 summary for contextHits class

Usage
## S3 method for class 'contextHits'
summary(object, ...)

Arguments
object a contextHits object, as returned by search_contexts
...
not used

Examples
text = c('A B C', 'D E F. G H I', 'A D', 'GGG')
tc = create_tcorpus(text, doc_id = c('a','b','c','d'), split_sentences = TRUE)
hits = search_contexts(tc, c('query label# A AND B', 'second query# (A AND Q) OR ("D E") OR I'))

summary(hits)

summary.featureHits  S3 summary for featureHits class

Description
S3 summary for featureHits class

Usage
## S3 method for class 'featureHits'
summary(object, ...)
```
Arguments

object  a featureHits object, as returned by search_features
...  not used

Examples

text = c('A B C', 'D E F G H I', 'A D', 'GGG')
tc = create_tcorpus(text, doc_id = c('a','b','c','d'), split_sentences = TRUE)
hits = search_features(tc, c('query label# A AND B', 'second query# (A AND Q) OR ("D E") OR I'))

summary(hits)

summary.tCorpus  Summary of a tCorpus object

Description

Summary of a tCorpus object

Usage

## S3 method for class 'tCorpus'
summary(object, ...)

Arguments

object  A tCorpus object
...  not used

Examples

tc = create_tcorpus(c('First text', 'Second text'))
summary(tc)

---

tCorpus  tCorpus: a corpus class for tokenized texts

Description

The tCorpus is a class for managing tokenized texts, stored as a data.frame in which each row represents a token, and columns contain the positions and features of these tokens.
Methods and Functions

The corpustools package uses both functions and methods for working with the tCorpus. Methods are used for all operations that modify the tCorpus itself, such as subsetting or adding columns. This allows the data to be modified by reference. Methods are accessed using the dollar sign after the tCorpus object. For example, if the tCorpus is named tc, the subset method can be called as tc$subset(...). Functions are used for all operations that return a certain output, such as search results or a semantic network. These are used in the common R style that you know and love. For example, if the tCorpus is named tc, a semantic network can be created with semnet(tc, ...).

Overview of methods and functions

The primary goal of the tCorpus is to facilitate various corpus analysis techniques. The documentation for currently implemented techniques can be reached through the following links.

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---

tCorpus$annotate_rsyntax

Annotation tokens based on rsyntax queries

Description

Apply queries to extract syntax patterns, and add the results as three columns to a tokenlist. The first column contains the ids for each hit. The second column contains the annotation label. The third column contains the fill level (which you probably won’t use, but is important for some features). Only nodes that are given a name in the tquery (using the label parameter) will be added as annotation.

Note that while queries only find 1 node for each labeled component of a pattern (e.g., quote queries have 1 node for "source" and 1 node for "quote"), all children of these nodes can be annotated by setting fill to TRUE. If a child has multiple ancestors, only the most direct ancestors are used (see documentation for the fill argument).

Usage:

```r
annotate_rsyntax(column, ..., block = NULL, fill = TRUE,
                   overwrite = FALSE, block_fill = FALSE, copy = TRUE,
                   verbose = FALSE)
```
Arguments

- **column**: The name of the column in which the annotations are added. The unique ids are added as column_id.
- **block**: Optionally, specify ids (doc_id - sentence - token_id triples) that are blocked from querying and filling (ignoring the id and recursive searches through the id).
- **fill**: Logical. If TRUE (default) also assign the fill nodes (as specified in the tquery). Otherwise these are ignored.
- **overwrite**: Applies if column already exists. If TRUE, existing column will be overwritten. If FALSE, the existing annotations in the column will be blocked, and new annotations will be added. This is identical to using multiple queries.
- **block_fill**: If TRUE (and overwrite is FALSE), the existing fill nodes will also be blocked. In other words, the new annotations will only be added if the
- **verbose**: If TRUE, report progress (only useful if multiple queries are given).

Examples

```r
library(rsyntax)

## spacy tokens for: Mary loves John, and Mary was loved by John
tokens = tokens_spacy[tokens_spacy$doc_id == 'text3',]
tc = tokens_to_tcorpus(tokens)

## two simple example tqueries
passive = tquery(pos = "VERB*", label = "predicate",
    children(relation = c("agent"), label = "subject"))
active = tquery(pos = "VERB*", label = "predicate",
    children(relation = c("nsubj", "nsubjpass"), label = "subject"))

tc$annotate_rsyntax("clause", pas=passive, act=active)
tc$tokens

if (interactive()) {
    plot_tree(tc$tokens, annotation='clause')
}
if (interactive()) {
    syntax_reader(tc$tokens, annotation = 'clause', value='subject')
}
```

Dictionary lookup
Description

Add a column to the token data that contains a code (the query label) for tokens that match the dictionary

Usage:

```
## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

code_dictionary(...)  
```

Arguments

dict A dictionary. Can be either a data.frame or a quanteda dictionary. If a data.frame is given, it has to have a column named "string" (or use string_col argument) that contains the dictionary terms. All other columns are added to the tCorpus $tokens data. Each row has a single string, that can be a single word or a sequence of words separated by a whitespace (e.g., "not bad"), and can have the common ? and * wildcards. If a quanteda dictionary is given, it is automatically converted to this type of data.frame with the `melt_quanteda_dict` function. This can be done manually for more control over labels.

token_col The feature in tc that contains the token text.

string_col If dict is a data.frame, the name of the column in dict that contains the dictionary lookup string

sep A regular expression for separating multi-word lookup strings (default is " ", which is what quanteda dictionaries use). For example, if the dictionary contains "Barack Obama", sep should be " " so that it matches the consecutive tokens "Barack" and "Obama". In some dictionaries, however, it might say "Barack+Obama", so in that case sep = '\+' should be used.

case_sensitive logical, should lookup be case sensitive?

column The name of the column added to $tokens. [column]_id contains the unique id of the match. If a quanteda dictionary is given, the label for the match is in the column named [column]. If a dictionary has multiple levels, these are added as [column]_l[level].

use_wildcards Use the wildcards * (any number including none of any character) and ? (one or none of any character). If FALSE, exact string matching is used. (":-)") versus ":: "")). This is only behind the scenes for the dictionary lookup, and will not affect tokenization in the corpus.

ascii If true, convert text to ascii before matching

verbose If true, report progress

Value

the tCorpus
Examples

dict = data.frame(string = c('good','bad','ugly','nice','not pretty', ':)'), ':('), sentiment=c(1,-1,-1,1,-1,1,-1))
tc = create_tcorpus(c('The good, the bad and the ugly, is nice :) but not pretty :('))
tc$code_dictionary(dict)
tc$tokens

tCorpus$code_features  Code features in a tCorpus based on a search string

Description

like search_features, but instead of return hits only adds a column to the token data that contains a code (the query label) for tokens that match the query. Note that only one code can be assigned to each token, so if there are overlapping results for different queries, the code for the last query will be used. This means that the order of queries (in the query argument) matters.

Usage:

## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

code_features(query, code=NULL, feature='token', column='code', ...)

Arguments

query  A character string that is a query. See search_features for documentation of the query language.

code  The code given to the tokens that match the query (usefull when looking for multiple queries). Can also put code label in query with # (see details)

feature  The name of the feature column within which to search.

column  The name of the column that is added to the data

add_column  list of name-value pairs, used to add additional columns. The name will become the column name, and the value should be a vector of the same length as the query vector.

context_level  Select whether the queries should occur within "documents" or specific "sentences".

as_ascii  if TRUE, perform search in ascii.

verbose  If TRUE, progress messages will be printed

overwrite  If TRUE (default) and column already exists, overwrite previous results.

...  alternative way to specify name-value pairs for adding additional columns

Examples

tc = create_tcorpus('Anna and Bob are secretive')
tc$code_features(c("actors# anna bob", "associations# secretive"))
tc$tokens
**tCorpus$context**  
*Get a context vector*

**Description**
Depending on the purpose, the context of an analysis can be the document level or sentence level. The `tCorpus$context()` method offers a convenient way to get the context id of tokens for different settings.

**Arguments**
- `context_level` Select whether the context is document or sentence level
- `with_labels` Return context as only ids (numeric, starting at 1) or with labels (factor)

**Details**

**Usage:**
```r
## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

data(context_level = c('document','sentence'), with_labels = T)
```

**Examples**
```r
tc <- create_tcorpus(c('Text one first sentence. Text one second sentence', 'Text two'),
                        split_sentences = TRUE)

doc <- tc$context() ## default context is doc_id (document level)
doc

sent <- tc$context('sentence') ## can specify sentence level
sent
```

**tCorpus$deduplicate**  
*Deduplicate documents*

**Description**
Deduplicate documents based on similarity scores. Can be used to filter out identical documents, but also similar documents.

Note that deduplication occurs by reference (`tCorpus_modify_by_reference`) unless `copy` is set to `TRUE`.

**Usage:**
```r
## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).
deduplicate(feature='token', date_col=NULL, meta_cols=NULL, hour_window=NULL, min_docfreq=2, max_docfreq_pct=0.5, measure=c('cosine','overlap_pct'), similarity=1, keep=c('first','last','random'), weight=c('norm_tfidf','tfidf','termfreq','docfreq'), ngrams=NA, print_duplicates=F, copy=F)
```
Arguments

- **feature**
  the column name of the feature that is to be used for the comparison.

- **date_col**
  The column name for a column with a date vector (in POSIXct). If given together with hour_window, only documents within the given hour_window will be compared.

- **meta_cols**
  a vector with names for columns in the meta data. If given, documents are only considered duplicates if the values of these columns are identical (in addition to having a high similarity score)

- **hour_window**
  A vector of length 1 or 2. If length is 1, the same value is used for the left and right side of the window. If length is 2, the first and second value determine the left and right side. For example, the value 12 will compare each document to all documents between the previous and next 12 hours, and c(-10, 36) will compare each document to all documents between the previous 10 and the next 36 hours.

- **min_docfreq**
  a minimum document frequency for features. This is mostly to lighten computational load. Default is 2, because terms that occur once cannot overlap across documents

- **max_docfreq_pct**
  a maximum document frequency percentage for features. High frequency terms contain little information for identifying duplicates. Default is 0.5 (i.e. terms that occur in more than 50 percent of documents are ignored),

- **lowercase**
  If True, make feature lowercase

- **measure**
  the similarity measure. Currently supports cosine similarity (symmetric) and overlap_pct (asymmetric)

- **similarity**
  the similarity threshold used to determine whether two documents are duplicates. Default is 1, meaning 100 percent identical.

- **keep**
  select either 'first', 'last' or 'random'. Determines which document of duplicates to delete. If a date is given, 'first' and 'last' specify whether the earliest or latest document is kept.

- **weight**
  a weighting scheme for the document-term matrix. Default is term-frequency inverse document frequency with normalized rows (document length).

- **ngrams**
  an integer. If given, ngrams of this length are used

- **print_deduplicates**
  if TRUE, print ids of duplicates that are deleted

- **verbose**
  if TRUE, report progress

- **copy**
  If TRUE, the method returns a new tCorpus object instead of deduplicating the current one by reference.

Examples

```r
d = data.frame(text = c('a b c d e',
  'e f g h i j k',
  'a b c'),
  date = as.POSIXct(c('2010-01-01','2010-01-01','2012-01-01'))) tc = create_tcorpus(d)
```
tCorpus$delete_columns

```r

tc$meta
dedup = tc$deduplicate(feature = 'token', date_col = 'date', similarity = 0.8, copy = TRUE)
dedup$meta

dedup = tc$deduplicate(feature = 'token', date_col = 'date', similarity = 0.8, keep = 'last',
                        copy = TRUE)
dedup$meta
```

tCorpus$delete_columns

*Delete column from the data and meta data*

Description

Usage:

Arguments

- `cnames`: The names of the columns to delete

Details

### R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

```r

delete_columns(cnames)
delete_meta_columns(cnames)
```

Examples

```r

d = data.frame(text = c('Text one', 'Text two', 'Text three'),
               date = c('2010-01-01', '2010-01-01', '2012-01-01'))
tc = create_tcorpus(d)
tc$tokens
tc$delete_columns('token')
tc$tokens
tc$meta
tc$delete_meta_columns('date')
tc$meta
```
tCorpus$feats_to_columns

Cast the "feats" column in UDpipe tokens to columns

Description

If the UDpipe parser is used in create_tcorpus, the 'feats' column contains strings with features (e.g., Number=Sing|PronType=Dem). To work with these nested features it is more convenient to cast them to columns.

Arguments

- keep: Optionally, the names of features to keep
- drop: Optionally, the names of features to drop
- rm_column: If TRUE (default), remove the original column

Details

Usage:

```r
## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

to_columns(keep=NULL, drop=NULL, rm_column=TRUE)
```

Examples

```r
if (interactive()) {
    tc = create_tcorpus('This is a test Bobby.', udpipe_model='english-ewt')
tc$feats_to_columns()
tc$tokens

tc = create_tcorpus('This is a test Bobby.', udpipe_model='english-ewt')
tc$feats_to_columns(keep = c('Gender', 'Tense', 'Person'))
tc$tokens
}
```

---

### Filter features

tCorpus$feature_subset

Filter features
**Description**

Similar to using `tCorpus$subset`, but instead of deleting rows it only sets rows for a specified feature to NA. This can be very convenient, because it enables only a selection of features to be used in an analysis (e.g. a topic model) but maintaining the context of the full article, so that results can be viewed in this context (e.g. a topic browser).

Just as in subset, it is easy to use objects and functions in the filter, including the special functions for using term frequency statistics (see documentation for `tCorpus$subset`).

**Usage:**

```r
## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

feature_subset(column, new_column, subset)
```

**Arguments**

- `column`: the column containing the feature to be used as the input
- `subset`: logical expression indicating rows to keep in the tokens data. i.e. rows for which the logical expression is FALSE will be set to NA.
- `new_column`: the column to save the filtered feature. Can be a new column or overwrite an existing one.
- `min_freq`: an integer, specifying minimum token frequency.
- `min_docfreq`: an integer, specifying minimum document frequency.
- `max_freq`: an integer, specifying minimum token frequency.
- `max_docfreq`: an integer, specifying minimum document frequency.
- `min_char`: an integer, specifying minimum characters in a token
- `max_char`: an integer, specifying maximum characters in a token

**Examples**

```r
tc = create_tcorpus('a a a a b b b c c')

tc$feature_subset('token', 'tokens_subset1', subset = token_id < 5)

tc$feature_subset('token', 'tokens_subset2', subset = freq_filter(token, min = 3))

tc$tokens
```

---

**Fold rsyntax annotations**

If a tCorpus has rsyntax annotations (see `annotate_rsyntax`), it can be convenient to aggregate tokens that have a certain semantic label. For example, if you have a query for labeling "source" and "quote", you can add an aggregated value for the sources (such as a unique ID) as a column, and then remove the quote tokens.
Arguments

- **annotation**: The name of an rsyntax annotation column
- **by_label**: The labels in this column for which you want to aggregate the tokens

... Specify the new aggregated columns in name-value pairs. The name is the name of the new column, and the value should be a function over a column in $tokens. For example: `subject = paste(token, collapse = ' ')` would create the column 'subject', of which the values are the concatenated tokens. See examples for more.

- **txt**: If TRUE, add _txt column with concatenated tokens for by_label
- **rm_by**: If TRUE (default), remove the column(s) specified in by_label
- **copy**: If TRUE, return a copy of the transformed tCorpus, instead of transforming the tCorpus by reference

Details

Usage:

```r
## R6 method for class tCorpus. Use as tc-method (where tc is a tCorpus object).

tc$fold_rsyntax(annotation, by_label, ...,
    to_label=NULL, rm_by=T, copy=F)
```

Examples

```r
tc = tc_sotu_udpipe$copy()

tc$udpipe_clauses()

tc$fold_rsyntax('clause', by_label = 'subject', subject = paste(token, collapse=' '))

tc$tokens
```

---

**tCorpus$get**

**Access the data from a tCorpus**

Description

Get (a copy of) the token and meta data. For quick access recommend using `tc$tokens` and `tc$meta` to get the tokens and meta data.tables, which does not copy the data. However, you should then make sure to not change the data.tables by reference, or you might break the tCorpus.

Usage:

```r
## R6 active method for class tCorpus. Use as tc-method (where tc is a tCorpus object).

tc$get(columns=NULL, keep_df=F, as.df=F, subset=NULL, doc_id=NULL, token_id=NULL, safe_copy=T)

tc$get_meta(columns=NULL, keep_df=F, as.df=F, subset=NULL, doc_id=NULL, safe_copy=T)
```
Arguments
- **columns**: character vector with the names of the columns
- **keep_df**: if True, the output will be a data.table (or data.frame) even if it only contains 1 columns
- **as_df**: if True, the output will be a regular data.frame instead of a data.table
- **subset**: Optionally, only get a subset of rows (see `tCorpus$subset` method)
- **doc_id**: A vector with document ids to select rows. Faster than subset, because it uses binary search. Cannot be used in combination with subset. If duplicate doc_ids are given, duplicate rows are returned.
- **token_id**: A vector with token indices. Can only be used in pairs with doc_id. For example, if `doc_id = c(1,1,1,2,2)` and `token_id = c(1,2,3,1,2)`, then the first three tokens of doc 1 and the first 2 tokens of doc 2 are returned. This is mainly usefull for fast (binary search) retrieval of specific tokens.
- **safe_copy**: for advanced use. The get methods always return a copy of the data, even if the full data is returned (i.e. use get without parameters). This is to prevent accidental changes within tCorpus data (which can break it) if the returned data is modified by reference (see data.table documentation). If safe_copy is set to FALSE and get is called without parameters—`tc$get(safe_copy=F)”—then no copy is made, which is much faster and more memory efficient. Use this if you need speed and efficiency, but make sure not to change the output data.table by reference.

Examples
```
d = data.frame(text = c('Text one first sentence. Text one second sentence', 'Text two'),
               medium = c('A','B'),
               date = c('2010-01-01','2010-02-01'),
               doc_id = c('D1','D2'))
tc = create_tcorpus(d, split_sentences = TRUE)
## get token data
  tc$tokens  # full data.table
  tc$get(c('doc_id','token'))  # data.table with selected columns
  head(tc$get('doc_id'))  # single column as vector
  head(tc$get(as.df = TRUE))  # return as regular data.frame

## get subset
  tc$get(subset = token_id %in% 1:2)

## subset on keys using (fast) binary search
  tc$get(doc_id = 'D1')  # for doc_id
  tc$get(doc_id = 'D1', token_id = 5)  # for doc_id / token pairs

##### use get for meta data with get_meta
  tc$meta

## option to repeat meta data to match tokens
```

tCorpus$lda_fit

Estimate a LDA topic model

Description

Estimate an LDA topic model using the LDA function from the topicmodels package. The parameters other than dtm are simply passed to the sampler but provide a workable default. See the description of that function for more information.

Usage:

## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

tda_fit(feature, create_feature=NULL, K=50, num.iterations=500, alpha=50/K, eta=.01, burnin=250, context_level=c('document','sentence'), ...)

Arguments

- feature: the name of the feature columns
- create_feature: optionally, add a feature column that indicates the topic to which a feature was assigned (in the last iteration). Has to be a character string, that will be the name of the new feature column
- K: the number of clusters
- num.iterations: the number of iterations
- method: set method. see documentation for LDA function of the topicmodels package
- alpha: the alpha parameter
- eta: the eta parameter
- burnin: The number of burnin iterations

Value

A fitted LDA model, and optionally a new column in the tcorpus (added by reference)

Examples

if (interactive()) {
  tc = create_tcorpus(sotu_texts, doc_column = 'id')
tc$preprocess('token', 'feature', remove_stopwords = TRUE, use_stemming = TRUE, min_freq=10)
set.seed(1)
m = tc$lda_fit('feature', create_feature = 'lda', K = 5, alpha = 0.1)
m
topicmodels::terms(m, 10)
  tc$tokens
}
Merge the token and meta data.tables of a tCorpus with another data.frame

Description

Add columns to token/meta by merging with a data.frame df. Only possible for unique matches (i.e. the columns specified in by are unique in df)

Arguments

df          A data.frame (can be regular, data.table or tibble)
by          The columns to match on. Must exist in both tokens/meta and df. If the columns in tokens/meta and df have different names, use by.x and by.y
by.x        The names of the columns used in tokens/meta
by.y        The names of the columns used in df
columns     Optionally, specify which specific columns from df to merge to tokens

Details

Usage:

## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

merge(df, by, by.x, by.y)
merge_meta(df, by, by.x, by.y)

Examples

d = data.frame(text = c('This is an example. Best example ever.','oh my god','so good'),
id = c('a','b','c'),
source =c('aa','bb','cc'))
tc = create_tcorpus(d, doc_col='id', split_sentences = TRUE)

df = data.frame(doc_id=c('a','b'), test=c('A','B'))
tc$merge(df, by='doc_id')
tc$tokens

df = data.frame(doc_id=c('a','b'), sentence=1, test2=c('A','B'))
tc$merge(df, by=c('doc_id', 'sentence'))
tc$tokens

df = data.frame(doc_id=c('a','b'), sentence=1, token_id=c(3,4), test3=c('A','B'))
tc$merge(df, by=c('doc_id', 'sentence', 'token_id'))
tc$tokens

meta = data.frame(doc_id=c('a','b'), test=c('A','B'))
Description

Usage:

Arguments

column the column containing the feature to be used as the input
new_column the column to save the preprocessed feature. Can be a new column or overwrite an existing one.
lowercase make feature lowercase
ngrams create ngrams. The ngrams match the rows in the token data, with the feature in the row being the last token of the ngram. For example, given the features "this is an example", the third feature ("an") will have the trigram "this_is_an". Ngrams at the beginning of a context will have empty spaces. Thus, in the previous example, the second feature ("is") will have the trigram ".is_an".
ngram_context Ngrams will not be created across contexts, which can be documents or sentences. For example, if the context_level is sentences, then the last token of sentence 1 will not form an ngram with the first token of sentence 2.
as_ascii convert characters to ascii. This is particularly useful for dealing with special characters.
remove_punctuation remove (i.e. make NA) any features that are only punctuation (e.g., dots, comma's)
remove_stopwords remove (i.e. make NA) stopwords. (!) Make sure to set the language argument correctly.
remove_numbers remove features that are only numbers
use_stemming reduce features (tokens) to their stem
language The language used for stopwords and stemming
min_freq an integer, specifying minimum token frequency.
min_docfreq an integer, specifying minimum document frequency.
max_freq an integer, specifying minimum token frequency.
max_docfreq an integer, specifying minimum document frequency.
min_char an integer, specifying minimum number of characters in a term
max_char an integer, specifying maximum number of characters in a term
Details

## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

```
preprocess(column='token', new_column='feature', lowercase=T, ngrams=1, ngram_context=c('document', 'sentence'), as_ascii=F, remove_punctuation=T, remove_stopwords=F, remove_numbers=F, use_stemming=F, language='english', min_freq=NULL, min_docfreq=NULL, max_freq=NULL, max_docfreq=NULL, min_char=NULL, max_char=NULL)
```

Examples

tc = create_tcorpus('I am a SHORT example sentence! That I am!')

## default is lowercase without punctuation
tc$preprocess('token', 'preprocessed_1')

## delete stopwords and perform stemming
tc$preprocess('token', 'preprocessed_2', remove_stopwords = TRUE, use_stemming = TRUE)

## filter on minimum frequency
tc$preprocess('token', 'preprocessed_3', min_freq=2)

## make ngrams
tc$preprocess('token', 'preprocessed_4', ngrams = 3)

Description

Uses `search_dictionary`, and replaces tokens that match the dictionary lookup term with the dictionary code. Multi-token matches (e.g., "Barack Obama") will become single tokens. Multiple lookup terms per code can be used to deal with alternatives such as "Barack Obama", "president Obama" and "Obama".

This method can also be used to concatenate ASCII symbols into emoticons, given a dictionary of emoticons.

Usage:

## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

`replace_dictionary(...)`
Arguments

**dict**
A dictionary. Can be either a data.frame or a quanteda dictionary. If a data.frame is given, it has to have a column named "string" (or use string_col argument) that contains the dictionary terms, and a column "code" (or use code_col argument) that contains the label/code represented by this string. Each row has a single string, that can be a single word or a sequence of words separated by whitespace (e.g., "not bad"), and can have the common ? and * wildcards. If a quanteda dictionary is given, it is automatically converted to this type of data.frame with the `melt_quanteda_dict` function. This can be done manually for more control over labels. Finally, you can also just pass a character vector. All multi word strings (like emoticons) will then be collapsed into single tokens.

token_col
The feature in tc that contains the token text.

string_col
If dict is a data.frame, the name of the column in dict with the dictionary lookup string. Default is "string".

code_col
The name of the column in dict with the dictionary code/label. Default is "code". If dict is a quanteda dictionary with multiple levels, "code_l2", "code_l3", etc. can be used to select levels.

replace_cols
The names of the columns in tc$tokens that will be replaced by the dictionary code. Default is the column on which the dictionary is applied, but in some cases it might make sense to replace multiple columns (like token and lemma)

sep
A regular expression for separating multi-word lookup strings (default is " ", which is what quanteda dictionaries use). For example, if the dictionary contains "Barack Obama", sep should be " " so that it matches the consecutive tokens "Barack" and "Obama". In some dictionaries, however, it might say "Barack+Obama", so in that case sep = "+" should be used.

code_from_features
If TRUE, instead of replacing features with the matched code columnn, use the most frequent occurring string in the features.

code_sep
If code_from_features is TRUE, the separator for pasting features together. Default is an underscore, which is recommended because it has special features in corpustools. Most importantly, if a query or dictionary search is performed, multi-word tokens concatenated with an underscore are treated as separate consecutive words. So, "Bob_Smith" would still match a lookup for the two consecutive words "bob smith"

decrement_ids
If TRUE (default), decrement token ids after concatenating multi-token matches. So, if the tokens c(";", ")", "yay") have token_id c(1,2,3), then after concatenating ASCII emoticons, the tokens will be c(";", "yay") with token_id c(1,2)

case_sensitive
logical, should lookup be case sensitive?

use_wildcards
Use the wildcards * (any number including none of any character) and ? (one or none of any character). If FALSE, exact string matching is used

ascii
If true, convert text to ascii before matching

verbose
If true, report progress

Value
A vector with the id value (taken from dict$id) for each row in tc$tokens
Examples

tc = create_tcorpus('happy :) sad :( happy 8-)

tc$tokens  ## tokenization has broken up emoticons (as it should)

# corpustools dictionary lookup automatically normalizes tokenization of
# tokens and dictionary strings. The dictionary string ";)") would match both
# the single token ";)") and two consecutive tokens c(";", ")"). This
# makes it easy and foolproof to look for emoticons like this:
emoticon_dict = data.frame(
    code = c('happy_emo', 'happy_emo', 'sad_emo'),
    string = c(';', '8-'), ':(')

tc$replace_dictionary(emoticon_dict)
tc$tokens

# If a string is passed to replace dictionary, it will collapse multi-word
# strings...

tc = create_tcorpus('happy :) sad :( Barack Obama')
tc$tokens
tc$replace_dictionary(c(';', '8-'), 'Barack Obama'))
tc$tokens

---

`tCorpus$search_recode  Recode features in a tCorpus based on a search string`

Description

Search features (see `search_features`) and replace features with a new value

Usage:

```r
search_recode(feature, new_value, keyword, condition = NA, condition_once = FALSE)
```

Arguments

- `feature`  The feature in which to search
- `new_value`  the character string with which all features that are found are replaced
- `query`  See `search_features` for the query parameters
- `...`  Additional search_features parameters. See `search_features`
tCorpus$set

Modify the token and meta data.tables of a tCorpus

Description

Modify the token/meta data.table by setting the values of one (existing or new) column. The subset argument can be used to modify only subsets of columns, and can be a logical vector (select TRUE rows), numeric vector (indices of TRUE rows) or logical expression (e.g. pos == 'noun'). If a new column is made while using a subset, then the rows outside of the selection are set to NA.

Arguments

column     Name of a new column (to create) or existing column (to transform)
value      An expression to be evaluated within the token/meta data, or a vector of the same length as the number of rows in the data. Note that if a subset is used, the length of value should be the same as the length of the subset (the TRUE cases of the subset expression) or a single value.
subset     logical expression indicating rows to keep in the tokens data or meta data
subset_value If subset is used, should value also be subsetted? Default is TRUE, which is what you want if the value has the same length as the full data.table (which is the case if a column in tokens is used). However, if the vector of values is already of the length of the subset, subset_value should be FALSE

Details

Usage:
## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

set(column, value, subset)

Examples

tc = create_tcorpus(sotu_texts[1:5,], doc_column = 'id')

tc$tokens   ## show original

## create new column
i <- 1:tc$n
tc$set(column = 'i', i)

## create new column based on existing column(s)
tc$set(column = 'token_upper', toupper(token))

## use subset to modify existing column

tc$set('token', paste0('**', token, '**'), subset = token_id == 1)

## use subset to create new column with NA's

tc$set('second_token', token, subset = token_id == 2)
tCorpus$set_levels

### Change levels of factor columns

**Description**

For factor columns, the levels can be changed directly (and by reference). This is particularly useful for fast preprocessing (e.g., making tokens lowercase).

**Arguments**

- `column` the name of the column
- `levels` The new levels

**Details**

**Usage:**

```r
## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

set_levels(column, levels)

set_meta_levels(column, levels)
```

**Examples**

```r
tc = create_tcorpus(c('Text one first sentence. Text one second sentence', 'Text two'))

## change factor levels of a column in the token data
unique_tokens <- tc$get_levels('token')
tc$set_levels('token', toupper(unique_tokens))
tc$tokens
```
tCorpus$set_name  Change column names of data and meta data

Description

Usage:

Arguments

oldname       the current/old column name
newname       the new column name

Details

## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

set_name(oldname, newname)

set_meta_name(oldname, newname)

Examples

tc = create_tcorpus(sotu_texts[1:5,], doc_column = 'id')

## change column name in token data
tc$names ## original column names
tc$set_name(oldname = 'token', newname = 'word')

tc$tokens

## change column name in meta data
tc$meta_names ## original column names
tc$set_meta_name(oldname = 'party', newname = 'clan')
tc$set_meta_name(oldname = 'president', newname = 'clan leader')

tc$meta
Description

Returns the subset of a tCorpus. The selection can be made separately (and simultaneously) for the token data (using subset) and the meta data (using subset_meta). The subset arguments work according to the subset.data.table function.

There are two flavours. You can either use subset(tc, ...) or tc$subset(...). The difference is that the second approach changes the tCorpus by reference. In other words, tc$subset() will delete the rows from the tCorpus, instead of creating a new tCorpus. Modifying the tCorpus by reference is more efficient (which becomes important if the tCorpus is large), but the more classic subset(tc, ...) approach is often more obvious.

Subset can also be used to select rows based on token/feature frequencies. This is a common step in corpus analysis, where it often makes sense to ignore very rare and/or very frequent tokens. To do so, there are several special functions that can be used within a subset call. The freq_filter() and docfreq_filter() can be used to filter terms based on term frequency and document frequency, respectively. (see examples)

The subset_meta() method is an alternative for using subset(subset_meta = ...), that is added for consistency with the other _meta methods.

Note that you can also use the tCorpus$feature_subset method if you want to filter out low/high frequency tokens, but do not want to delete the rows in the tCorpus.

Usage:

```r
## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

subset(tc, subset = NULL, subset_meta = NULL,
       window = NULL)
tc$subset(subset = NULL, subset_meta = NULL,
          window = NULL, copy = F)
tc$subset_meta(subset = NULL, copy = F)
```

Arguments

- **subset**: logical expression indicating rows to keep in the tokens data.
- **subset_meta**: logical expression indicating rows to keep in the document meta data.
- **window**: If not NULL, an integer specifying the window to be used to return the subset. For instance, if the subset contains token 10 in a document and window is 5, the subset will contain token 5 to 15. Naturally, this does not apply to subset_meta.
- **copy**: If TRUE, the method returns a new tCorpus object instead of subsetting the current one. This is added for convenience when analyzing a subset of the data. e.g., tc_nyt = tc$subset_meta(medium == "New_York_Times", copy=T)

Examples

```r
tc = create_tcorpus(sotu_texts[1:5,], doc_column = 'id')
tc$n # original number of tokens

tc2 = subset(tc, token_id < 20)
tc2$n
```

## Description

Returns the subset of a tCorpus. The selection can be made separately (and simultaneously) for the token data (using `subset`) and the meta data (using `subset_meta`). The subset arguments work according to the `subset.data.table` function.

There are two flavours. You can either use `subset(tc, ...)` or `tc$subset(...)`. The difference is that the second approach changes the tCorpus by reference. In other words, `tc$subset()` will delete the rows from the tCorpus, instead of creating a new tCorpus. Modifying the tCorpus by reference is more efficient (which becomes important if the tCorpus is large), but the more classic `subset(tc, ...)` approach is often more obvious.

Subset can also be used to select rows based on token/feature frequencies. This is a common step in corpus analysis, where it often makes sense to ignore very rare and/or very frequent tokens. To do so, there are several special functions that can be used within a subset call. The `freq_filter()` and `docfreq_filter()` can be used to filter terms based on term frequency and document frequency, respectively. (see examples)

The `subset_meta()` method is an alternative for using `subset(subset_meta = ...)`, that is added for consistency with the other _meta methods.

Note that you can also use the `tCorpus$feature_subset` method if you want to filter out low/high frequency tokens, but do not want to delete the rows in the tCorpus.

### Usage:

```r
## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

subset(tc, subset = NULL, subset_meta = NULL,
       window = NULL)
tc$subset(subset = NULL, subset_meta = NULL,
          window = NULL, copy = F)
tc$subset_meta(subset = NULL, copy = F)
```

### Arguments

- **subset**: logical expression indicating rows to keep in the tokens data.
- **subset_meta**: logical expression indicating rows to keep in the document meta data.
- **window**: If not NULL, an integer specifying the window to be used to return the subset. For instance, if the subset contains token 10 in a document and window is 5, the subset will contain token 5 to 15. Naturally, this does not apply to `subset_meta`.
- **copy**: If TRUE, the method returns a new tCorpus object instead of subsetting the current one. This is added for convenience when analyzing a subset of the data. e.g., `tc_nyt = tc$subset_meta(medium == "New_York_Times", copy=T)`

### Examples

```r
tc = create_tcorpus(sotu_texts[1:5,], doc_column = 'id')
tc$n # original number of tokens

tc2 = subset(tc, token_id < 20)
tc2$n
```
## Note that the original is untouched
tc$n

## Now we subset by reference. This doesn't make a copy, but changes tc itself
tc$subset(token_id < 20)
tc$n

## you can filter on term frequency and document frequency with the freq_filter() and
docfreq_filter() functions
tc = create_tcorpus(sotu_texts[c(1:5,800:805),], doc_column = 'id')
tc$subset( freq_filter(token, min = 2, max = 4) )
tc$tokens

#### subset can be used for meta data by using the subset_meta argument, or the subset_meta method
tc$n_meta
tc$meta
tc$subset(subset_meta = president == 'Barack Obama')
tc$n_meta

---

### tCorpus$subset_query

**Subset tCorpus token data using a query**

---

**Description**

A convenience function that searches for contexts (documents, sentences), and uses the results to subset the tCorpus token data.

See the documentation for search_contexts for an explanation of the query language.

**Usage:**

## R6 method for class tCorpus. Use as tc$method (where tc is a tCorpus object).

subset_query(query, feature = 'token', context_level = c('document','sentence','window'))

**Arguments**

- **query**: A character string that is a query. See search_contexts for query syntax.
- **feature**: The name of the feature columns on which the query is used.
- **context_level**: Select whether the query and subset are performed at the document or sentence level.
- **window**: If used, uses a word distance as the context (overrides context_level)
- **as_ascii**: if TRUE, perform search in ascii.
- **not**: If TRUE, perform a NOT search. Return the articles/sentences for which the query is not found.
- **copy**: If TRUE, return modified copy of data instead of subsetting the input tcorpus by reference.
Examples

text = c('A B C', 'D E F. G H I', 'A D', 'GGG')
tc = create_tcorpus(text, doc_id = c('a','b','c','d'), split_sentences = TRUE)

## subset by reference
tc$subset_query('A')
tc$meta

## using copy mechanic
class(tc$tokens$doc_id)
tc2 = tc$subset_query('A AND D', copy=TRUE)
tc2$get_meta()
tc$meta ## (unchanged)

---

**Add columns indicating who did what**

Description

An off-the-shelf application of rsyntax for extracting subject-verb clauses. Designed for working with a tCorpus created with `udpipe_tcorpus`.

Arguments

column The name of the column in $Tokens to store the results.
tqueries A list of tQueries. By default uses the off-the-shelf tqueries in `udpipe_clause_tqueries`.

Value

a tCorpus

Examples

tc = tc_sotu_udpipe$copy()
tc$udpipe_clauses()
if (interactive()) {
tc_plot_tree(tc, token, lemma, POS, annotation='clause')
browse_texts(tc, rsyntax='clause', value='subject')
}
tCorpus$udpipe_quotes  Add columns indicating who said what

Description

An off-the-shelf application of rsyntax for extracting quotes. Designed for working with a tCorpus created with udpipe_tcorpus.

Arguments

tqueries  A list of tqueries. By default uses the off-the-shelf tqueries in udpipe_quote_tqueries.
span_tqueries  Additional tqueries for finding candidates for 'span quotes' (i.e. quotes that span multiple sentences, indicated by quotation marks). By default uses the off-the-shelf tqueries in udpipe_spanquote_tqueries.

Details

Default tqueries are provided for detecting source-quote relations within sentences (udpipe_quote_tqueries), and for detecting source candidates for text between quotation marks that can span across multiple sentences (udpipe_spanquote_tqueries). These have mainly been developed and tested for the english-ewt udpipe model.

There are two ways to customize this function. One is to specify a custom character vector of verb lemma. This vector should then be passed as an argument to the two functions for generating the default tqueries. The second (more advanced) way is to provide a custom list of tqueries. (Note that the udpipe_quote_tqueries and udpipe_spanquote_tqueries functions simply create lists of queries. You can create new lists, or add tqueries to these lists). !! If you create custom tqueries, make sure that the labels for the quote and source tokens are 'source' and 'quote'. For the spanquote_tqueries, the label for the source candidate should be 'source'.

Value

the columns 'quote', 'quote_id', and 'quote_verbatim' are added to tokens

Examples

## Not run:
txt = 'Bob said that he likes Mary. John did not like that: "how dare he!". "It is I, John, who likes Mary!!"
tc = udpipe_tcorpus(txt, model = 'english-ewt')
tc$udpipe_quotes()

if (interactive()) {
    tc_plot_tree(tc, token, lemma, POS, annotation='quote')
browse_texts(tc, rsyntax='quote', value='source')
}

## you can provide your own lists of tqueries, or use the two
## query generating functions to customize the specific 'verb lemma'
 custom_verb_lemma = c('say','state')  ## this should be longer
quote_tqueries = udpipe_quote_tqueries(custom_verb_lemma)
span_quote_tqueries = udpipe_spanquote_tqueries(custom_verb_lemma)

## note that these use simply lists with tqueries, so you can also
## create your own list or customize these lists

quote_tqueries
span_quote_tqueries

if (interactive()) {
  tc$udpipe_quotes(tqueries = quote_tqueries, span_tqueries = span_quote_tqueries)
  tc_plot_tree(tc, token, lemma, POS, annotation='quote')
  browse_texts(tc, rsyntax='quote', value='source')
}

## End(Not run)
**tCorpus_data**  
Methods and functions for viewing, modifying and subsetting tCorpus data

---

**Description**

(back to overview)

**Details**

**Get data**

$get()  
Get (by default deep copy) token data, with the possibility to select columns and subset. Instead of copying you can also access the token data with tc$tokens

$get_meta()  
Get meta data, with the possibility to select columns and subset. Like tokens, you can also access meta data with tc$meta

get_dtm()  
Create a document term matrix

get_dfm()  
Create a document term matrix, using the Quanteda dfm format

$context()  
Get a context vector. Currently supports documents or globally unique sentences.

**Modify**

The token and meta data can be modified with the set* and delete* methods. All modifications are performed by reference.

$set()  
Modify the token data by setting the values of one (existing or new) column.

$set_meta()  
The set method for the document meta data

$set_levels()  
Change the levels of factor columns.

$set_meta_levels()  
Change the levels of factor columns in the meta data

$set_name()  
Modify column names of token data.

$set_meta_name()  
Delete columns in the meta data

$delete_columns()  
Delete columns.

$delete_meta_columns()  
Delete columns in the meta data

Modifying is restricted in certain ways to ensure that the data always meets the assumptions required for tCorpus methods. tCorpus automatically tests whether assumptions are violated, so you don’t have to think about this yourself. The most important limitations are that you cannot subset or append the data. For subsetting, you can use the tCorpus$subset method, and to add data to a tcorpus you can use the merge_tcorpora function.

**Subsetting, merging/adding**

subset()  
Modify the token and/or meta data using the subset function. A subset expression can be specified for both

subset_query()  
Subset the tCorpus based on a query, as used in search_contexts

$subset()  
Like subset, but as an R6 method that changes the tCorpus by reference

$subset_query()  
Like subset_query, but as an R6 method that changes the tCorpus by reference

**Fields**

For the sake of convenience, the number of rows and column names of the data and meta data.tables
can be accessed directly.

\begin{itemize}
  \item \$n\quad \text{The number of tokens (i.e. rows in the data)}$
  \item \$n\_\text{meta}\quad \text{The number of documents (i.e. rows in the document meta data)}$
  \item \$\text{names}\quad \text{The names of the token data columns}$
  \item \$\text{names}\_\text{meta}\quad \text{The names of the document meta data columns}$
\end{itemize}

\section*{tCorpus_docsim \quad Document similarity}

\subsection*{Description}
\textit{(back to overview)}

\subsection*{Details}
\textbf{Compare documents, and perform similarity based deduplication}

\begin{itemize}
  \item \texttt{compare\_documents()}\quad \text{Compare documents}$
  \item \texttt{$\text{deduplicate}()$}\quad \text{Remove duplicate documents}$
\end{itemize}

\section*{tCorpus_features \quad Preprocessing, subsetting and analyzing features}

\subsection*{Description}
\textit{(back to overview)}

\subsection*{Details}
\textbf{Pre-process features}

\begin{itemize}
  \item \texttt{$\text{preprocess}()$}\quad \text{Create or modify a feature by preprocessing an existing feature}$
  \item \texttt{$\text{feature\_subset}()$}\quad \text{Similar to using subset, but instead of deleting rows it only sets rows for a specified feature to NA}.$
\end{itemize}

\textbf{Inspect features}

\begin{itemize}
  \item \texttt{$\text{feature\_stats}()$}\quad \text{Create a data.frame with feature statistics}$
  \item \texttt{$\text{top\_features}()$}\quad \text{Show top features, optionally grouped by a given factor}$
\end{itemize}
tCorpus_modify_by_reference

Modify tCorpus by reference

Description

(Back to overview)

Details

If any tCorpus method is used that changes the corpus (e.g., set, subset), the change is made by reference. This is convenient when working with a large corpus, because it means that the corpus does not have to be copied when changes are made, which is slower and less memory efficient.

To illustrate, for a tCorpus object named ‘tc’, the subset method can be called like this:

```
tc$subset(doc_id %in% selection)
```

The ‘tc’ object itself is now modified, and does not have to be assigned to a name, as would be the more common R philosophy. Like this:

```
tc = tc$subset(doc_id %in% selection)
```

The results of both lines of code are the same. The assignment in the second approach is not necessary, but doesn’t harm either because tc$subset returns the modified corpus invisibly (see ?invisible if that sounds spooky).

Be aware, however, that the following does not work!!

```
tc2 = tc$subset(doc_id %in% selection)
```

In this case, tc2 does contain the subsetted corpus, but tc itself will also be subsetted!!

Using the R6 method for subset forces this approach on you, because it is faster and more memory efficient. If you do want to make a copy, there are several solutions.

Firstly, for some methods we provide identical functions. For example, instead of the $subset() R6 method, we can use the subset() function.

```
tc2 = subset(tc, doc_id %in% selection)
```

We promise that only the R6 methods (called as tc$method()) will change the data by reference.

A second option is that R6 methods where copying is often usefull have copy parameter Modifying by reference only happens in the R6 methods

```
tc2 = tc$subset(doc_id %in% selection, copy=TRUE)
```

Finally, you can always make a deep copy of the entire tCorpus before modifying it, using the $copy() method.

```
tc2 = tc$copy()
```
Use Boolean queries to analyze the tCorpus

**Description**

(back to overview)

**Details**

**Feature-level queries**

- `search_features()` Search for features based on keywords and conditions
- `$code_features()` Add a column to the token data based on feature search results
- `$search_recode()` Use the search_features query syntax to recode features
- `feature_associations()` Given a query, get words that often co-occur nearby
- `kwic()` Get keyword-in-context (kwic) strings
- `browse_hits()` Create full-text browsers with highlighted search hits

**Context-level queries**

- `search_contexts()` Search for documents or sentences using Lucene-like queries
- `$subset_query()` use the search_contexts query syntax to subset the tCorpus

---

**tCorpus_semnet**

*Feature co-occurrence based semantic network analysis*

**Description**

(back to overview)

**Details**

**Create networks**

- `semnet()` Feature co-occurrence within contexts (documents, sentences)
- `semnet_window()` Feature co-occurrence within a specified token distance

**Support functions for analyzing and visualizing the semantic network**

- `ego_semnet()` Create an ego network from an Igraph network
- `plot_semnet()` Convenience function for visualizing an Igraph network, specialized for semantic networks
**tCorpus_topmod**  
*Topic modeling*

**Description**

(back to overview)

**Details**

**Train a topic model**

$Lda_fit()$  
Latent Dirichlet Allocation

---

**tc_plot_tree**  
*Visualize a dependency tree*

**Description**

A wrapper for the `plot_tree` function, that can be used directly on a tCorpus.

**Usage**

```
tc_plot_tree(
  tc,
  ..., annotation = NULL,
  sentence_i = 1,
  doc_id = NULL,
  pdf_file = NULL
)
```

**Arguments**

- `tc`  
a tCorpus

- `...`  
Arguments passed to `plot_tree`. Most importantly, this is used to select which specific columns to display on the bottom rows. For instance, `tc_plot_tree(tc, token, lemma, POS)` shows only these three columns.

- `annotation`  
Optionally, the name of a column with an rsyntax annotation.

- `sentence_i`  
By default, `plot_tree` uses the first sentence (`sentence_i = 1`) in the data. `sentence_i` can be changed to select other sentences by position (the i-th unique sentence in the data). Note that `sentence_i` does not refer to the values in the sentence column (for this use the sentence argument together with `doc_id`).

- `doc_id`  
Optionally, the document id can be specified. If so, `sentence_i` refers to the i-th sentence within the given document.

- `pdf_file`  
Directly save the plot as a pdf file
Value

plots a dependency tree.

Examples

```r
if (interactive())
  tc_plot_tree(tc_sotu_udpipe, token, lemma, POS)
```

---

**tc_sotu_udpipe**

A tCorpus with a small sample of sotu paragraphs parsed with udpipe

---

**Description**

A tCorpus with a small sample of sotu paragraphs parsed with udpipe

**Usage**

```r
data(tc_sotu_udpipe)
```

---

**Format**

data.frame

---

**tokens_to_tcorpus**

Create a tcorpus based on tokens (i.e. preprocessed texts)

---

**Description**

Create a tcorpus based on tokens (i.e. preprocessed texts)

**Usage**

```r
tokens_to_tcorpus(
  tokens,
  doc_col = "doc_id",
  token_id_col = "token_id",
  token_col = NULL,
  sentence_col = NULL,
  parent_col = NULL,
  meta = NULL,
  meta_cols = NULL,
  feature_cols = NULL,
  sent_is_local = T,
  token_is_local = T,
  ...
)
```
tokens_to_tcorpus

Arguments

- **tokens**: A data.frame in which rows represent tokens, and columns indicate (at least) the document in which the token occurred (doc_col) and the position of the token in that document or globally (token_id_col)
- **doc_col**: The name of the column that contains the document ids/names
- **token_id_col**: The name of the column that contains the positions of tokens. If NULL, it is assumed that the data.frame is ordered by the order of tokens and does not contain gaps (e.g., filtered out tokens)
- **token_col**: Optionally, the name of the column that contains the token text. This column will then be renamed to "token" in the tcorpus, which is the default name for many functions (e.g., querying, printing text)
- **sentence_col**: Optionally, the name of the column that indicates the sentences in which tokens occurred. This can be necessary if tokens are not local at the document level (see token_is_local argument), and sentence information can be used in several tcorpus functions.
- **parent_col**: Optionally, the name of the column that contains the id of the parent (if a dependency parser was used). If token_is_local = FALSE, then the token_ids will be transformed, so parent ids need to be changed as well. Default is 'parent', but if this column is not present the parent is ignored.
- **meta**: Optionally, a data.frame with document meta data. Needs to contain a column with the document ids (with the same name)
- **meta_cols**: Alternatively, if there are document meta columns in the tokens data.table, meta_cols can be used to recognize them. Note that these values have to be unique within documents.
- **feature_cols**: Optionally, specify which columns to include in the tcorpus. If NULL, all columns are included (except the specified columns for documents, sentences and positions)
- **sent_is_local**: Sentences in the tCorpus are assumed to be locally unique within documents. If sent_is_local is FALSE, then sentences are transformed to be locally unique. However, it is then assumed that the first sentence in a document is sentence 1, which might not be the case if tokens (input) is a subset.
- **token_is_local**: Same as sent_is_local, but for token_id. !! if the data has a parent column, make sure to specify parent_col, so that the parent ids are also transformed

Examples

```r
head(corenlp_tokens)
tc = tokens_to_tcorpus(corenlp_tokens, doc_col = 'doc_id',
                      sentence_col = 'sentence', token_id_col = 'id')
tc

meta = data.frame(doc_id = 1, medium = 'A', date = '2010-01-01')
tc = tokens_to_tcorpus(corenlp_tokens, doc_col = 'doc_id',
                      sentence_col = 'sentence', token_id_col = 'id', meta=meta)
tc
```
tokenWindowOccurence

Gives the window in which a term occurred in a matrix.

Description

This function returns the occurrence of tokens (position.matrix) and the window of occurrence (window.matrix). This format enables the co-occurrence of tokens within sliding windows (i.e., token distance) to be calculated by multiplying position.matrix with window.matrix.

Usage

```r
tokenWindowOccurence(
  tc,
  feature,
  context_level = c("document", "sentence"),
  window.size = 10,
  direction = "<>",
  distance_as_value = F,
  batch_rows = NULL,
  drop_empty_terms = T
)
```

Arguments

- `tc`: a tCorpus object
- `feature`: The name of the feature column
- `context_level`: Select whether to use "document" or "sentence" as context boundaries
- `window.size`: The distance within which tokens should occur from each other to be counted as a co-occurrence.
- `direction`: a string indicating whether only the left ('<') or right ('>') side of the window, or both ('<>'), should be used.
- `distance_as_value`: If True, the values of the matrix will represent the shortest distance to the occurrence of a feature
- `batch_rows`: Used in functions that call this function in batches
- `drop_empty_terms`: If TRUE, empty terms (with zero occurrence) will be dropped

Value

A list with two matrices. position.mat gives the specific position of a term, and window.mat gives the window in which each token occurred. The rows represent the position of a term, and matches the input of this function (position, term and context). The columns represent terms.
top_features

Description
Show top features

Usage
top_features(
tc,
feature,
n = 10,
group_by = NULL,
group_by_meta = NULL,
rank_by = c("freq", "chi2"),
dropNA = T,
return_long = F
)

Arguments
tc a tCorpus
feature The name of the feature
n Return the top n features
group_by A column in the token data to group the top features by. For example, if token data contains part-of-speech tags (pos), then grouping by pos will show the top n feature per part-of-speech tag.
group_by_meta A column in the meta data to group the top features by.
rank_by The method for ranking the terms. Currently supports frequency (default) and the 'Chi2' value for the relative frequency of a term in a topic compared to the overall corpus. If return_long is used, the Chi2 score is also returned, but note that there are negative Chi2 scores. This is used to indicate that the relative frequency of a feature in a group was lower than the relative frequency in the corpus (i.e. under-represented).
dropNA if TRUE, drop NA features
return_long if TRUE, results will be returned in a long format that contains more information.

Value
a data.frame
Examples

tc = tokens_to_tcorpus(corenlp_tokens, token_id_col = 'id')

top_features(tc, 'lemma')
top_features(tc, 'lemma', group_by = 'NER', group_by_meta='doc_id')

transform_rsyntax  Apply rsyntax transformations

Description

This is an experimental function for applying rsyntax transformations directly on a tcorpus, to create a new tcorpus with the transformed tokens. The argument f should be self defined function that wraps rsyntax transformations. Or more generally, a function that takes a tokens data.frame (or data.table) as input, and returns a tokens data.frame (or data.table). For examples, see corpus-tools::ud_relcl, or corpus-tools::udpipe_simplify for a function that wraps multiple transformations.

Usage

transform_rsyntax(tc, f, ...)

Arguments

tc  a tCorpus
f  functions that perform rsyntax tree transformations
...  arguments passed to f

Value

a tCorpus after applying the transformations

Examples

if (interactive()) {
tc = tc_sotu_udpipe$copy()
tc2 = transform_rsyntax(tc, udpipe_simplify)

browse_texts(tc2)
  rsyntax::plot_tree(tc$tokens, token, lemma, POS, sentence_i=20)
  rsyntax::plot_tree(tc2$tokens, token, lemma, POS, sentence_i=20)
}

udpipe_clause_tqueries

Get a list of tqueries for extracting who did what

Description
An off-the-shelf list of tqueries for extracting subject-verb clauses. Designed for working with a tCorpus created with udpipe_tcorpus.

Usage
udpipe_clause_tqueries(verbs = NULL, exclude_verbs = verb_lemma("quote"))

Arguments
verbs A character vector for specific verbs to use. By default uses all verbs (except for those specified in exclude_verbs)
exclude_verbs A character vector for specific verbs NOT to use. By default uses the verbs that indicate speech (that are used for extracting who said what, in udpipe_quote_tqueries)

Examples
udpipe_clause_tqueries()

udpipe_quote_tqueries

Get a list of tqueries for extracting quotes

Description
An off-the-shelf list of tqueries for extracting quotes. Designed for working with a tCorpus created with udpipe_tcorpus.

Usage
udpipe_quote_tqueries(say_verbs = verb_lemma("quote"))

Arguments
say_verbs A character vector of verb lemma that indicate speech (e.g., say, state). A default list is included in verb_lemma('quote'), but certain lemma might be more accurate/appropriate depending on the corpus.

Examples
udpipe_quote_tqueries()
udpipe_simplify

Simplify tokenIndex created with the udpipe parser

Description

This is an off-the-shelf implementation of several rsyntax transformation for simplifying text.

Usage

    udupipe_simplify(
        tokens,
        split_conj = T,
        rm_punct = F,
        new_sentences = F,
        rm_mark = F
    )

Arguments

    tokens       A tokenIndex, based on output from the ud parser.
    split_conj   If TRUE, split conjunctions into separate sentences
    rm_punct     If TRUE, remove punctuation afterwards
    new_sentences If TRUE, assign new sentence and token_id after splitting
    rm_mark      If TRUE, remove children with a mark relation if this is used in the simplification.

Value

    a tokenIndex

Examples

    if (interactive()) {
      tc = tc_sotu_udpipe$copy()
      tc2 = transform_rsyntax(tc, udupipe_simplify)

      browse_texts(tc2)
        rsyntax::plot_tree(tc_sotu_udpipe$tokens, token, lemma, POS, sentence_i=20)
        rsyntax::plot_tree(tc2$tokens, token, lemma, POS, sentence_i=20)
    }
udpipe_spanquote_tqueries

Get a list of tqueries for finding candidates for span quotes.

Description

Quote extraction with tqueries is limited to quotes within sentences. When (verbatim) quotes span multiple sentences (which we call span quotes here), they are often indicated with quotation marks. While it is relatively easy to identify these quotes, it is less straightforward to identify the sources of these quotes. A good approach is to first apply tqueries for finding quotes within sentences, because a source mentioned just before (we use 2 sentences) a span quote is often also the source of this span quote. For cases where there is no previous source, we can apply simple queries for finding source candidates. That's what the tqueries created with the current function are for.

Usage

udpipe_spanquote_tqueries(say_verbs = verb_lemma("quote"))

Arguments

say_verbs A character vector of verb lemma that indicate speech (e.g., say, state). A default list is included in verb_lemma('quote'), but certain lemma might be more accurate/appropriate depending on the corpus.

Details

This procedure is supported in rsyntax with the add_span_quotes function. In corpustools this function is implemented within the udpipe_quotes method. The current function provides the default tqueries for the span quotes.

Examples

udpipe_spanquote_tqueries()

udpipe_tcorpus

Create a tCorpus using udpipe

Description

This is simply shorthand for using create_tcorpus with the udpipe_ arguments and certain specific settings. This is the way to create a tCorpus if you want to use the syntax analysis functionalities.
Usage

udpipe_tcorpus(x, ...)

## S3 method for class 'character'
udpipe_tcorpus(
  x,
  model = "english-ewt",
  doc_id = 1:length(x),
  meta = NULL,
  max_sentences = NULL,
  model_path = getwd(),
  cache = 3,
  cores = NULL,
  batchsize = 50,
  use_parser = T,
  start_end = F,
  verbose = T,
  ...
)

## S3 method for class 'data.frame'
udpipe_tcorpus(
  x,
  model = "english-ewt",
  text_columns = "text",
  doc_column = "doc_id",
  max_sentences = NULL,
  model_path = getwd(),
  cache = 3,
  cores = 1,
  batchsize = 50,
  use_parser = T,
  start_end = F,
  verbose = T,
  ...
)

## S3 method for class 'factor'
udpipe_tcorpus(x, ...)

## S3 method for class 'corpus'
udpipe_tcorpus(x, ...)

Arguments

x main input. can be a character (or factor) vector where each value is a full text, or a data.frame that has a column that contains full texts.

... Arguments passed to create_tcorpus.character
model
The name of a Universal Dependencies language model (e.g., "english-ewt", "dutch-alpino"), to use the udpipe package (udpipe_annotate). If you don’t know the model name, just type the language and you’ll get a suggestion. Otherwise, use show_udpipe_models to get an overview of the available models. For more information about udpipe and performance benchmarks of the UD models, see the GitHub page of the udpipe package.

doc_id
if x is a character/factor vector, doc_id can be used to specify document ids. This has to be a vector of the same length as x

meta
A data.frame with document meta information (e.g., date, source). The rows of the data.frame need to match the values of x

max_sentences
An integer. Limits the number of sentences per document to the specified number.

model_path
If udpipe_model is used, this path will be used to look for the model, and if the model doesn’t yet exist it will be downloaded to this location. Defaults to working directory

cache
The number of persistent caches to keep for inputs of udpipe. The caches store tokens in batches. This way, if a lot of data has to be parsed, or if R crashes, udpipe can continue from the latest batch instead of start over. The caches are stored in the corpustools_data folder (in udpipe_model_path). Only the most recent [udpipe_caches] caches will be stored.

cores
If udpipe_model is used, this sets the number of parallel cores. If not specified, will use the same number of cores as used by data.table (or limited to OMP_THREAD_LIMIT)

batchsize
In order to report progress and cache results, texts are parsed with udpipe in batches of 50. The price is that there will be some overhead for each batch, so for very large jobs it can be faster to increase the batchsize. If the number of texts divided by the number of parallel cores is lower than the batchsize, the texts are evenly distributed over cores.

use_parser
If TRUE, use dependency parser (only if udpipe_model is used)

start_end
If TRUE, include start and end positions of tokens

verbose
If TRUE, report progress. Only if x is large enough to require multiple sequential batches

text_columns
if x is a data.frame, this specifies the column(s) that contains text. The texts are paste together in the order specified here.

doc_column
If x is a data.frame, this specifies the column with the document ids.

Examples

```r
### ...
if (interactive()) {
  tc = udpipe_tcorpus(c('Text one first sentence. Text one second sentence', 'Text two'),
                     model = 'english-ewt')
  tc$tokens
}
if (interactive()) {
  tc = udpipe_tcorpus(sotu_texts[1:5,], doc_column='id', model = 'english-ewt')
```
It makes little sense to have full texts as factors, but it tends to happen. The create_tcorpus S3 method for factors is essentially identical to the method for a character vector.

```r
text = factor(c('Text one first sentence', 'Text one second sentence'))
if (interactive()) {
  tc = udpipe_tcorpus(text, 'english-ewt-')
tc$tokens
}
```

---

**untokenize**  
Reconstruct original texts

**Description**

If the tCorpus was created with remember_spaces = T, you can rebuild the original texts.

**Usage**

```r
untokenize(tc)
```

**Arguments**

- `tc` A tCorpus, created with `create_tcorpus`, with remember_spaces = TRUE

**Value**

A data.table with the text fields and meta fields as columns.

**Examples**

```r
tc = create_tcorpus(sotu_texts, doc_column='id')
untokenize(tc)
```
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