

Package ‘kgrams’

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Title Classical k-gram Language Models

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Description Training and evaluating k-gram language models in R, supporting several probability smoothing techniques, perplexity computations, random text generation and more.

License GPL (>= 3)

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Contents

dictionary	2
EOS	4
kgram_freqs	5
language_model	10
midsummer	11
much_ado	12
parameters	13
perplexity	14
preprocess	17
probability	18
query	19
sample_sentences	21
smoothers	22
tknz_sent	24
word_context	25
%+%	26
Index	28

dictionary	<i>Word dictionaries</i>
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Description

Construct or coerce to and from a dictionary.

Usage

```
dictionary(object, ...)
```

```
## S3 method for class 'kgram_freqs'
dictionary(object, size = NULL, cov = NULL, thresh = NULL, ...)
```

```
## S3 method for class 'character'
dictionary(
  object,
  .preprocess = identity,
  size = NULL,
  cov = NULL,
  thresh = NULL,
  ...
)
```

```
## S3 method for class 'connection'
dictionary(
  object,
```

```

    .preprocess = identity,
    size = NULL,
    cov = NULL,
    thresh = NULL,
    max_lines = Inf,
    batch_size = max_lines,
    ...
)

as_dictionary(object)

## S3 method for class 'kgrams_dictionary'
as_dictionary(object)

## S3 method for class 'character'
as_dictionary(object)

## S3 method for class 'kgrams_dictionary'
as.character(x, ...)
```

Arguments

object	object from which to extract a dictionary, or to be coerced to dictionary.
...	further arguments passed to or from other methods.
size	either NULL or a length one positive integer. Predefined size of the required dictionary (the top size most frequent words are retained).
cov	either NULL or a length one numeric between 0 and 1. Predefined text coverage fraction of the dictionary (the most frequent words providing the required coverage are retained).
thresh	either NULL or length one a positive integer. Minimum word count threshold to include a word in the dictionary.
.preprocess	a function taking a character vector as input and returning a character vector as output. Optional preprocessing transformation applied to text before creating the dictionary.
max_lines	a length one positive integer or Inf. Maximum number of lines to be read from the connection. If Inf, keeps reading until the End-Of-File.
batch_size	a length one positive integer less than or equal to max_lines. Size of text batches when reading text from connection.
x	a dictionary.

Details

These generic functions are used to build dictionary objects, or to coerce from other formats to dictionary, and from a dictionary to a character vector. By now, the only non-trivial type coercible to dictionary is character, in which case each entry of the input vector is considered as a single word. Coercion from dictionary to character returns the list of words included in the dictionary as a regular character vector.

Dictionaries can be extracted from `kgram_freqs` objects, or *built* from text coming either directly from a character vector or a connection.

A single preprocessing transformation can be applied before processing the text for unique words. After preprocessing, *anything delimited by one or more white space characters* in the transformed text input *is counted as a word* and may be added to the dictionary modulo additional constraints.

The possible constraints for including a word in the dictionary can be of three types: (i) fixed size of dictionary, implemented by the `size` argument; (ii) fixed text covering fraction, as specified by the `cov` argument; or (iii) minimum word count threshold, `thresh` argument. *Only one of these constraints can be applied at a time*, so that specifying more than one of `size`, `cov` or `thresh` results in an error.

Value

A dictionary for `dictionary()` and `as_dictionary()`, a character vector for the `as.character()` method.

Author(s)

Valerio Gherardi

Examples

```
# Building a dictionary from Shakespeare's "Much Ado About Nothing"

dict <- dictionary(much_ado)
length(dict)
query(dict, "leonato") # TRUE
query(dict, c("thy", "thou")) # c(TRUE, TRUE)
query(dict, "smartphones") # FALSE

# Getting list of words as regular character vector
words <- as.character(dict)
head(words)

# Building a dictionary from a list of words
dict <- as_dictionary(c("i", "the", "a"))
```

EOS

Special Tokens

Description

Return Begin-Of-Sentence, End-Of-Sentence and Unknown-Word special tokens.

Usage`EOS()``BOS()``UNK()`**Details**

These functions return the internal representation of BOS, EOS and UNK tokens respectively. Their actual returned values are irrelevant and their only purpose is to simplify queries of k-gram counts and probabilities involving the special tokens, as shown in the examples.

Value

a string representing the appropriate special token.

Author(s)

Valerio Gherardi

Examples

```
f <- kgram_freqs("a b b a b", 2)
query(f, c(BOS(), EOS(), UNK()))

m <- language_model(f, "add_k", k = 1)
probability(c("a", "b") %|% BOS(), m)
probability("a b b a" %+% EOS(), m)

# The actual values of BOS(), EOS() and UNK() are irrelevant
c(BOS(), EOS(), UNK())
```

`kgram_freqs`*k-gram Frequency Tables*

Description

Extract k-gram frequency counts from a text or a connection.

Principal methods supported by objects of class `kgram_freqs`:

- `query()`: query k-gram counts from the table. See [query](#)
- `probability()`: compute word continuation and sentence probabilities using Maximum Likelihood estimates. See [probability](#).
- `language_model()`: build a k-gram language model using various probability smoothing techniques. See [language_model](#).

Usage

```
kgram_freqs(object, ...)  
  
## S3 method for class 'numeric'  
kgram_freqs(  
  object,  
  .preprocess = identity,  
  .tknz_sent = identity,  
  dict = NULL,  
  ...  
)  
  
## S3 method for class 'kgram_freqs'  
kgram_freqs(object, ...)  
  
## S3 method for class 'character'  
kgram_freqs(  
  object,  
  N,  
  .preprocess = identity,  
  .tknz_sent = identity,  
  dict = NULL,  
  open_dict = is.null(dict),  
  verbose = FALSE,  
  ...  
)  
  
## S3 method for class 'connection'  
kgram_freqs(  
  object,  
  N,  
  .preprocess = identity,  
  .tknz_sent = identity,  
  dict = NULL,  
  open_dict = is.null(dict),  
  verbose = FALSE,  
  max_lines = Inf,  
  batch_size = max_lines,  
  ...  
)  
  
process_sentences(  
  text,  
  freqs,  
  .preprocess = attr(freqs, ".preprocess"),  
  .tknz_sent = attr(freqs, ".tknz_sent"),  
  open_dict = TRUE,  
  in_place = TRUE,
```

```

    verbose = FALSE,
    ...
)

## S3 method for class 'character'
process_sentences(
  text,
  freqs,
  .preprocess = attr(freqs, ".preprocess"),
  .tknz_sent = attr(freqs, ".tknz_sent"),
  open_dict = TRUE,
  in_place = TRUE,
  verbose = FALSE,
  ...
)

## S3 method for class 'connection'
process_sentences(
  text,
  freqs,
  .preprocess = attr(freqs, ".preprocess"),
  .tknz_sent = attr(freqs, ".tknz_sent"),
  open_dict = TRUE,
  in_place = TRUE,
  verbose = FALSE,
  max_lines = Inf,
  batch_size = max_lines,
  ...
)

```

Arguments

object	any type allowed by the available methods. The type defines the behaviour of <code>kgram_freqs()</code> as a default constructor, a copy constructor or a constructor of a non-trivial object. See ‘Details’.
...	further arguments passed to or from other methods.
.preprocess	a function taking a character vector as input and returning a character vector as output. Optional preprocessing transformation applied to text before k-gram tokenization. See ‘Details’.
.tknz_sent	a function taking a character vector as input and returning a character vector as output. Optional sentence tokenization step applied to text after preprocessing and before k-gram tokenization. See ‘Details’.
dict	anything coercible to class dictionary . Optional pre-specified word dictionary.
N	a length one integer. Maximum order of k-grams to be considered.
open_dict	TRUE or FALSE. If TRUE, any new word encountered during processing not appearing in the original dictionary is included into the dictionary. Otherwise, new words are replaced by an unknown word token. It is by default TRUE if dict is specified, FALSE otherwise.

verbose	Print current progress to the console.
max_lines	a length one positive integer or Inf. Maximum number of lines to be read from the connection. If Inf, keeps reading until the End-Of-File.
batch_size	a length one positive integer less than or equal to max_lines. Size of text batches when reading text from connection.
text	a character vector or a connection. Source of text from which k-gram frequencies are to be extracted.
freqs	a kgram_freqs object, to which new k-gram counts from text are to be added.
in_place	TRUE or FALSE. Should the initial kgram_freqs object be modified in place?

Details

The function `kgram_freqs()` is a generic constructor for objects of class `kgram_freqs`, i.e. k-gram frequency tables. The constructor from `integer` returns an empty 'kgram_freqs' of fixed order, with an optional predefined dictionary (which can be empty) and `.preprocess` and `.tknz_sent` functions to be used as defaults in other `kgram_freqs` methods. The constructor from `kgram_freqs` returns a copy of an existing object, and it is provided because, in general, `kgram_freqs` objects have reference semantics, as discussed below.

The following discussion focuses on `process_sentences()` generic, as well as on the character and connection methods of the constructor `kgram_freqs()`. These functions extract k-gram frequency counts from a text source, which may be either a character vector or a connection. The second option is useful if one wants to avoid loading the full text corpus in physical memory, allowing to process text from different sources such as files, compressed files or URLs.

The returned object is of class `kgram_freqs` (a thin wrapper around the internal C++ class where all k-gram computations take place). `kgram_freqs` objects have methods for querying bare k-gram frequencies ([query](#)) and maximum likelihood estimates of sentence probabilities or word continuation probabilities (see [probability](#)). More importantly `kgram_freqs` objects are used to create [language_model](#) objects, which support various probability smoothing techniques.

The function `kgram_freqs()` is used to *construct* a new `kgram_freqs` object, initializing it with the k-gram counts from the `text` input, whereas `process_sentences()` is used to add k-gram counts from a new `text` to an *existing* `kgram_freqs` object, `freqs`. In this second case, the initial object `freqs` can either be modified in place (for `in_place == TRUE`, the default) or by making a copy (`in_place == FALSE`), see the examples below. The final object is returned invisibly when modifying in place, visibly in the second case. It is worth to mention that modifying in place a `kgram_freqs` object `freqs` will also affect `language_model` objects created from `freqs` with `language_model()`, which will also be updated with the new information. If one wants to avoid this behaviour, one can make copies using either the `kgram_freqs()` copy constructor, or the `in_place = FALSE` argument.

The `dict` argument allows to provide an initial set of known words. Subsequently, one can either work with such a closed dictionary (`open_dict == FALSE`), or extended the dictionary with all new words encountered during k-gram processing (`open_dict == TRUE`).

The `.preprocess` and `.tknz_sent` functions are applied *before* k-gram counting takes place, and are in principle arbitrary transformations of the original text. *After* preprocessing and sentence tokenization, each line of the transformed input is presented to the k-gram counting algorithm as a separate sentence (these sentences are implicitly padded with $N - 1$ Begin-Of-Sentence (BOS) and one End-Of-Sentence (EOS) tokens, respectively). This is illustrated in the examples). For basic

usage, this package offers the utilities [preprocess](#) and [tknz_sent](#). Notice that, strictly speaking, there is some redundancy in these two arguments, as the processed input to the k-gram counting algorithm is `.tknz_sent(.preprocess(text))`. They appear explicitly as separate arguments for two main reasons:

- The presence of `.tknz_sent` is a reminder of the fact that sentences have to be explicitly separated in different entries of the processed input, in order for `kgram_freqs()` to append the correct Begin-Of-Sentence and End-Of-Sentence paddings to each sentence.
- At prediction time (e.g. with [probability](#)), by default only `.preprocess` is applied when computing conditional probabilities, whereas both `.preprocess()` and `.tknz_sent()` are applied when computing sentence absolute probabilities.

Value

A `kgram_freqs` class object: k-gram frequency table storing k-gram counts from text. For `process_sentences()`, the updated `kgram_freqs` object is returned invisibly if `in_place` is `TRUE`, visibly otherwise.

Author(s)

Valerio Gherardi

See Also

[query](#), [probability language_model](#), [dictionary](#)

Examples

```
# Build a k-gram frequency table from a character vector

f <- kgram_freqs("a b b a a", 3)
f
summary(f)
query(f, c("a", "b")) # c(3, 2)
query(f, c("a b", "a" %>% EOS(), BOS() %>% "a b")) # c(1, 1, 1)
query(f, "a b b a") # NA (counts for k-grams of order k > 3 are not known)

process_sentences("b", f)
query(f, c("a", "b")) # c(3, 3): 'f' is updated in place

f1 <- process_sentences("b", f, in_place = FALSE)
query(f, c("a", "b")) # c(3, 3): 'f' is copied
query(f1, c("a", "b")) # c(3, 4): the new 'f1' stores the updated counts

# Build a k-gram frequency table from a file connection

## Not run:
f <- kgram_freqs(file("my_text_file.txt"), 3)
```

```
## End(Not run)

# Build a k-gram frequency table from an URL connection
## Not run:
f <- kgram_freqs(url("http://my.website/my_text_file.txt"), 3)

## End(Not run)
```

language_model

k-gram Language Models

Description

Build a k-gram language model.

Principal methods supported by objects of class `language_model`:

- `probability()`: compute word continuation and sentence probabilities. See [probability](#).
- `sample_sentences()`: generate random text by sampling from the language model probability distribution at arbitrary temperature. See [sample_sentences](#).
- `perplexity()`: Compute the language model perplexity on a test corpus. See [perplexity](#).

Usage

```
language_model(object, ...)

## S3 method for class 'language_model'
language_model(object, ...)

## S3 method for class 'kgram_freqs'
language_model(object, smoother = "m1", N = param(object, "N"), ...)
```

Arguments

<code>object</code>	an object which stores the information required to build the k-gram model. At present, necessarily a <code>kgram_freqs</code> object, or a <code>language_model</code> object of which a copy is desired (see Details).
<code>...</code>	possible additional parameters required by the smoother.
<code>smoother</code>	a length one character vector. Indicates the smoothing technique to be applied to compute k-gram continuation probabilities. A list of available smoothers can be obtained with <code>smoothers()</code> , and further information on a particular smoother through <code>info()</code> .
<code>N</code>	a length one integer. Maximum order of k-grams to use in the language model. This must be less than or equal to the order of the underlying <code>kgram_freqs</code> object.

Details

These generics are used to construct objects of class `language_model`. The `language_model` method is only needed to create copies of `language_model` objects (that is to say, new copies which are not altered by methods which modify the original object in place, see e.g. [parameters](#)). The discussion below focuses on language models and the `kgram_freqs` method.

`kgrams` supports several k-gram language models, including Interpolated Kneser-Ney, Stupid Back-off and others (see [smoothers](#)). The objects created by `language_models()` have methods for computing word continuation and sentence probabilities (see [probability](#)), random text generation (see [sample_sentences](#)) and other type of language modeling tasks such as computing perplexities and word prediction accuracies.

Smoothers have often tuning parameters, which need to be specified by (exact) name through the `...` arguments; otherwise, `language_model()` will use default values and, once per session, throw a warning. `info(smooth)` lists all parameters needed by a specific smoother, together with their allowed parameter space.

The run-time of `language_model()` may vary substantially for different smoothing methods, depending on whether or not a method requires the computation of additional quantities (that is to say, beyond k-gram counts) for its operativity (this is, for instance, the case for the Kneser-Ney smoother).

Value

A `language_model` object.

Author(s)

Valerio Gherardi

Examples

```
# Create an interpolated Kneser-Ney 2-gram language model

freqs <- kgram_freqs("a a b a a b a b a b a b", 2)
model <- language_model(freqs, "kn", D = 0.5)
model
summary(model)
probability("a" %|% "b", model)

# For more examples, see ?probability, ?sample_sentences and ?perplexity.
```

midsummer

A Midsummer Night's Dream

Description

The entire play "A Midsummer Night's Dream" from William Shakespeare.

Usage

midsummer

Format

A length one character vector, containing the entire text of "A Midsummer Night's Dream" from William Shakespeare. The script used for generating this file is available [here](#)

Source

<https://www.folger.edu/>

See Also

[much_ado](#)

Examples

midsummer[840]

much_ado

Much Ado About Nothing

Description

The entire play "Much Ado About Nothing" from William Shakespeare.

Usage

much_ado

Format

A length one character vector, containing the entire text of "Much Ado About Nothing" from William Shakespeare. The script used for generating this file is available [here](#)

Source

<https://www.folger.edu/>

See Also

[midsummer](#)

Examples

much_ado[840]

parameters	<i>Language Model Parameters</i>
------------	----------------------------------

Description

Get and set parameters of a language model.

Usage

```
param(object, which)

## S3 method for class 'kgram_freqs'
param(object, which)

param(object, which) <- value

parameters(object)
```

Arguments

object	a language_model or kgram_freqs class object.
which	a string. Name of the parameter to get or set.
value	new value for the parameter specified by which. Typically a length one numeric.

Details

These functions are used to retrieve or modify the parameters of a language_model or a kgram_freqs object. Any object of, or inheriting from, any of these two classes has at least two parameters:

- N: higher order of k-grams considered in the model for language_model, or stored in memory for kgram_freqs.
- V: size of the dictionary (excluding the special tokens BOS(), EOS(), UNK()).

For an object of class kgram_freqs, these are the only parameters, and they are read-only. language_models allow to set N less than or equal to the order of the underlying kgram_freqs object.

In addition to these, language_models can have additional parameters, e.g. discount values or interpolation constants, depending on the particular smoother employed by the model. A list of parameters available for a given smoother can be obtained through info() (see [smoothers](#)).

It is important to mention that setting a parameter is an in-place operation. This implies that if, say, object m is a language_model object, the code m1 <- m ; param(m1, which) <- value will set the parameter which to value both for m1 *and* m. The reason for this is that, behind the scenes, both m and m1 are pointers to the same C++ object. In order to create a true copy, one can use the copy constructor language_model(), see [language_model](#).

Value

a list for parameters(), a single value, typically numeric, for param() and param()<- (the new value, in this last case).

Examples

```
# Get and set k-gram model parameters

f <- kgram_freqs("a a b a b", 3)
param(f, "N")
parameters(f)

m <- language_model(f, "sbo", lambda = 0.2)
param(m, "V")
param(m, "lambda")
param(m, "N") <- 2
param(m, "lambda") <- 0.4

if (FALSE) {
  param(m, "V") <- 5 # Error: dictionary size cannot be set.
}

if (FALSE) {
  param(f, "N") <- 4 # Error: parameters of 'kgram_freqs' cannot be set
}

m1 <- m
param(m1, "lambda") <- 0.5
param(m, "lambda") # 0.5 ; param() modifies 'm' by reference!

m2 <- language_model(m) # This creates a true copy
param(m2, "lambda") <- 0.6
param(m, "lambda") # 0.5
```

perplexity

Language Model Perplexities

Description

Compute language model perplexities on a test corpus.

Usage

```
perplexity(
  text,
  model,
  .preprocess = attr(model, ".preprocess"),
  .tknz_sent = attr(model, ".tknz_sent"),
  exp = TRUE,
  ...
)

## S3 method for class 'character'
```

```

perplexity(
  text,
  model,
  .preprocess = attr(model, ".preprocess"),
  .tknz_sent = attr(model, ".tknz_sent"),
  exp = TRUE,
  detailed = FALSE,
  ...
)

## S3 method for class 'connection'
perplexity(
  text,
  model,
  .preprocess = attr(model, ".preprocess"),
  .tknz_sent = attr(model, ".tknz_sent"),
  exp = TRUE,
  batch_size = Inf,
  ...
)

```

Arguments

<code>text</code>	a character vector or connection. Test corpus from which language model perplexity is computed.
<code>model</code>	an object of class <code>language_model</code> .
<code>.preprocess</code>	a function taking a character vector as input and returning a character vector as output. Preprocessing transformation applied to input before computing perplexity.
<code>.tknz_sent</code>	a function taking a character vector as input and returning a character vector as output. Optional sentence tokenization step applied before computing perplexity.
<code>exp</code>	TRUE or FALSE. If TRUE, returns the actual perplexity - exponential of cross-entropy per token - otherwise returns its natural logarithm.
<code>...</code>	further arguments passed to or from other methods.
<code>detailed</code>	TRUE or FALSE. If TRUE, the output has a "details" attribute, which is a data-frame containing the cross-entropy of each individual sentence tokenized from text.
<code>batch_size</code>	a length one positive integer or <code>Inf</code> . Size of text batches when reading text from a connection. If <code>Inf</code> , all input text is processed in a single batch.

Details

These generic functions are used to compute a `language_model` perplexity on a test corpus, which may be either a plain character vector of text, or a connection from which text can be read in batches. The second option is useful if one wants to avoid loading the full text in physical memory, and allows to process text from different sources such as files, compressed files or URLs.

"Perplexity" is defined here, following Ref. (Chen and Goodman 1999), as the exponential of the normalized language model cross-entropy with the test corpus. Cross-entropy is normalized by the total number of words in the corpus, where we include the End-Of-Sentence tokens, but not the Begin-Of-Sentence tokens, in the word count.

The custom `.preprocess` and `.tknz_sent` arguments allow to apply transformations to the text corpus before the perplexity computation takes place. By default, the same functions used during model building are employed, c.f. [kgram_freqs](#) and [language_model](#).

A note of caution is in order. Perplexity is not defined for all language models available in [kgrams](#). For instance, smoother "sbo" (i.e. Stupid Backoff (Brants et al. 2007)) does not produce normalized probabilities, and this is signaled by a warning (shown once per session) if the user attempts to compute the perplexity for such a model. In these cases, when possible, perplexity computations are performed anyway case, as the results might still be useful (e.g. to tune the model's parameters), even if their probabilistic interpretation does no longer hold.

Value

a number. Perplexity of the language model on the test corpus.

Author(s)

Valerio Gherardi

References

Brants T, Popat AC, Xu P, Och FJ, Dean J (2007). "Large Language Models in Machine Translation." In *Proceedings of the 2007 Joint Conference on Empirical Methods in Natural Language Processing and Computational Natural Language Learning (EMNLP-CoNLL)*, 858–867. <https://aclanthology.org/D07-1090/>.

Chen SF, Goodman J (1999). "An empirical study of smoothing techniques for language modeling." *Computer Speech & Language*, **13**(4), 359–394.

Examples

```
# Train 4-, 6-, and 8-gram models on Shakespeare's "Much Ado About Nothing",
# compute their perplexities on the training and test corpora.
# We use Shakespeare's "A Midsummer Night's Dream" as test.
```

```
train <- much_ado
test <- midsummer
```

```
tknz <- function(text) tknz_sent(text, keep_first = TRUE)
f <- kgram_freqs(train, 8, .tknz_sent = tknz)
m <- language_model(f, "kn", D = 0.75)
```

```
# Compute perplexities for 4-, 6-, and 8-gram models
FUN <- function(N) {
  param(m, "N") <- N
  c(train = perplexity(train, m), test = perplexity(test, m))
}
```



```
    }
  sapply(c("N = 4" = 4, "N = 6" = 6, "N = 8" = 8), FUN)
```

preprocess

*Text preprocessing***Description**

A minimal text preprocessing utility.

Usage

```
preprocess(input, erase = "[^.?!;:'[:alnum:][:space:]]", lower_case = TRUE)
```

Arguments

input	a character vector.
erase	a length one character vector. Regular expression matching parts of text to be <i>erased</i> from input. The default removes anything not alphanumeric (<code>[A-z0-9]</code>), space (white space, tab, vertical tab, newline, form feed, carriage return), apostrophes or punctuation characters (<code>"[.?!:;]"</code>).
lower_case	a length one logical vector. If TRUE, puts everything to lower case.

Details

The expressions `preprocess(x, erase = pattern, lower_case = TRUE)` and `preprocess(x, erase = pattern, lower_case = FALSE)` are roughly equivalent to `tolower(gsub(pattern, "", x))` and `gsub(pattern, "", x)`, respectively, provided that the regular expression 'pattern' is correctly recognized by R.

Note. This function, as well as `tknz_sent`, are included in the library for illustrative purposes only, and are not optimized for performance. Furthermore (for performance reasons) the function has a separate implementation for Windows and UNIX OS types, respectively, so that results obtained in the two cases may differ slightly. In contexts that require full reproducibility, users are encouraged to define their own preprocessing and tokenization custom functions - or to work with externally processed data.

Value

a character vector containing the processed output.

Author(s)

Valerio Gherardi

Examples

```
preprocess("#This Is An Example@-@!#")
```

probability *Language Model Probabilities*

Description

Compute sentence probabilities and word continuation conditional probabilities from a language model

Usage

```
probability(object, model, .preprocess = attr(model, ".preprocess"), ...)

## S3 method for class 'kgrams_word_context'
probability(object, model, .preprocess = attr(model, ".preprocess"), ...)

## S3 method for class 'character'
probability(
  object,
  model,
  .preprocess = attr(model, ".preprocess"),
  .tknz_sent = attr(model, ".tknz_sent"),
  ...
)
```

Arguments

object	a character vector for sentence probabilities, a word-context conditional expression created with the conditional operator <code>% %</code> (see word_context). for word continuation probabilities.
model	an object of class <code>language_model</code> .
.preprocess	a function taking a character vector as input and returning a character vector as output. Preprocessing transformation applied to input before computing probabilities
...	further arguments passed to or from other methods.
.tknz_sent	a function taking a character vector as input and returning a character vector as output. Optional sentence tokenization step applied before computing sentence probabilities.

Details

The generic function `probability()` is used to obtain both sentence unconditional probabilities (such as `Prob("I was starting to feel drunk")`) and word continuation conditional probabilities (such as `Prob("you" | "i love")`). In plain words, these probabilities answer the following related but conceptually different questions:

- Sentence probability `Prob(s)`: what is the probability that extracting a single sentence (from a corpus of text, say) we will obtain exactly 's'?

- Continuation probability $\text{Prob}(w|c)$: what is the probability that a given context 'c' will be followed exactly by the word 'w'?

In order to compute continuation probabilities (i.e. $\text{Prob}(w|c)$), one must create conditional expressions with the infix operator `%|%`, as shown in the examples below. Both probability and `%|%` are vectorized with respect to words (left hand side of `%|%`), but the context must be a length one character (right hand side of `%|%`).

The input is treated as in [query](#) for what concerns word tokenization: anything delimited by (one or more) white space(s) is tokenized as a word. For sentence probabilities, Begin-Of-Sentence and End-Of-Sentence paddings are implicitly added to the input, but specifying them explicitly does not produce wrong results as BOS and EOS tokens are ignored by `probability()` (see the examples below). For continuation probabilities, any context of more than $N - 1$ words (where N is the k -gram order the language model) is truncated to the last $N - 1$ words.

By default, the same `.preprocess()` and `.tknz_sent()` functions used during model building are applied to the input, but this can be overridden with arbitrary functions. Notice that the `.tknz_sent` can be useful (for sentence probabilities) if e.g. the input is a length one unprocessed character vector.

Value

a numeric vector. Probabilities of the sentences or word continuations.

Author(s)

Valerio Gherardi

Examples

```
# Usage of probability()

f <- kgram_freqs("a b b a b a b", 2)

m <- language_model(f, "add_k", k = 1)
probability(c("a", "b", EOS(), UNK()) %|% BOS(), m) # c(0.4, 0.2, 0.2, 0.2)
probability("a" %|% UNK(), m) # not NA
```

query

Query k-gram frequency tables or dictionaries

Description

Return the frequency count of k -grams in a k -gram frequency table, or whether words are contained in a dictionary.

Usage

```

query(object, x)

## S3 method for class 'kgram_freqs'
query(object, x)

## S3 method for class 'kgrams_dictionary'
query(object, x)

```

Arguments

<code>object</code>	a <code>kgram_freqs</code> or dictionary class object.
<code>x</code>	a character vector. A list of k-grams if <code>object</code> is of class <code>kgram_freqs</code> , a list of words if <code>object</code> is a dictionary.

Details

This generic has slightly different behaviors when querying for the presence of words in a dictionary and for k-gram counts in a frequency table respectively. For words, `query()` looks for exact matches between the input and the dictionary entries. Queries of Begin-Of-Sentence (`BOS()`) and End-Of-Sentence (`EOS()`) tokens always return `TRUE`, and queries of the Unknown-Word token return `FALSE` (see [special_tokens](#)).

On the other hand, queries of k-gram counts first perform a word level tokenization, so that anything separated by one or more space characters in the input is considered as a single word (thus, for instance queries of strings such as "i love you", " i love you"), or "i love you ") all produce the same outcome). Moreover, querying for any word outside the underlying dictionary returns the counts corresponding to the Unknown-Word token (`UNK()`) (e.g., if the word "prcsrnr" is outside the dictionary, querying "i love prcsrnr" is the same as querying `paste("i love", UNK())`). Queries from k-grams of order $k > N$ will return `NA`.

A subsetting equivalent of `query`, with syntax `object[x]` is available (see the examples). `query(object, x)`. The query of the empty string "" returns the total count of words, including the `EOS` and `UNK` tokens, but not the `BOS` token.

See also the examples below.

Value

an integer vector, containing k-gram counts of `x`, if `object` is a `kgram_freqs` class object, a logical vector if `object` is a dictionary. Vectorized over `x`.

Author(s)

Valerio Gherardi

Examples

```

# Querying a k-gram frequency table
f <- kgram_freqs("a a b a b b a b", N = 2)
query(f, c("a", "b")) # query single words

```

```

query(f, c("a b")) # query a 2-gram
identical(query(f, "c"), query(f, "d")) # TRUE, both "c" and "d" are <UNK>
identical(query(f, UNK()), query(f, "c")) # TRUE
query(f, EOS()) # 1, since text is a single sentence
f[c("b b", "b")] # query with subsetting syntax
f[""] # 9 (includes the EOS token)

# Querying a dictionary
d <- as_dictionary(c("a", "b"))
query(d, c("a", "b", "c")) # query some words
query(d, c(BOS(), EOS(), UNK())) # c(TRUE, TRUE, FALSE)
d["a"] # query with subsetting syntax

```

sample_sentences

Random Text Generation

Description

Sample sentences from a language model's probability distribution.

Usage

```
sample_sentences(model, n, max_length, t = 1)
```

Arguments

model	an object of class <code>language_model</code> .
n	an integer. Number of sentences to sample.
max_length	an integer. Maximum length of sampled sentences.
t	a positive number. Sampling temperature (optional); see Details.

Details

This function samples sentences according the prescribed language model's probability distribution, with an optional temperature parameter. The temperature transform of a probability distribution is defined by $p(t) = \exp(\log(p) / t) / Z(t)$ where $Z(t)$ is the partition function, fixed by the normalization condition $\sum(p(t)) = 1$.

Sampling is performed word by word, using the already sampled string as context, starting from the Begin-Of-Sentence context (i.e. $N - 1$ BOS tokens). Sampling stops either when an End-Of-Sentence token is encountered, or when the string exceeds `max_length`, in which case a truncated output is returned.

Some language models may give a non-zero probability to the the Unknown word token, but this is never produced in text generated by `sample_sentences()`: when randomly sampled, it is simply ignored.

Finally, a word of caution on some special smoothers: "sbo" smoother (Stupid Backoff), does not produce normalized continuation probabilities, but rather continuation *scores*. Sampling is here

performed by assuming that Stupid Backoff scores are *proportional* to actual probabilities. 'ml' smoother (Maximum Likelihood) does not assign probabilities when the k-gram count of the context is zero. When this happens, the next word is chosen uniformly at random from the model's dictionary.

Value

a character vector of length n. Random sentences generated from the language model's distribution.

Author(s)

Valerio Gherardi

Examples

```
# Sample sentences from 8-gram Kneser-Ney model trained on Shakespeare's
# "Much Ado About Nothing"
```

```
### Prepare the model and set seed
freqs <- kgram_freqs(much_ado, 8, .tknz_sent = tknz_sent)
model <- language_model(freqs, "kn", D = 0.75)
set.seed(840)
```

```
sample_sentences(model, n = 3, max_length = 10)
```

```
### Sampling at high temperature
sample_sentences(model, n = 3, max_length = 10, t = 100)
```

```
### Sampling at low temperature
sample_sentences(model, n = 3, max_length = 10, t = 0.01)
```

smoothers

k-gram Probability Smoothers

Description

Information on available k-gram continuation probability smoothers.

List of smoothers currently supported by kgrams:

- "ml": Maximum Likelihood estimate (Markov 1913).
- "add_k": Add-k smoothing (Dale and Laplace 1995; Lidstone 1920; Johnson 1932; Jeffreys 1998).
- "abs": Absolute discounting (Ney and Essen 1991).
- "wb": Witten-Bell smoothing (Bell et al. 1990; Witten and Bell 1991)

- "kn": Interpolated Kneser-Ney. (Kneser and Ney 1995; Chen and Goodman 1999).
- "mkn": Interpolated modified Kneser-Ney. (Chen and Goodman 1999).
- "sbo": Stupid Backoff (Brants et al. 2007).

Usage

```
smoothers()
```

```
info(smooth)
```

Arguments

smoother a string. Code name of probability smoother.

Value

smoothers() returns a character vector, the list of code names of probability smoothers available in `kgrams`. `info(smooth)` returns NULL (invisibly) and prints some information on the selected smoothing technique.

Author(s)

Valerio Gherardi

References

Bell TC, Cleary JG, Witten IH (1990). *Text compression*. Prentice-Hall, Inc.

Brants T, Popat AC, Xu P, Och FJ, Dean J (2007). "Large Language Models in Machine Translation." In *Proceedings of the 2007 Joint Conference on Empirical Methods in Natural Language Processing and Computational Natural Language Learning (EMNLP-CoNLL)*, 858–867. <https://aclanthology.org/D07-1090/>.

Chen SF, Goodman J (1999). "An empirical study of smoothing techniques for language modeling." *Computer Speech & Language*, **13**(4), 359–394.

Dale AI, Laplace P (1995). *Philosophical essay on probabilities*. Springer.

Jeffreys H (1998). *The theory of probability*. OUP Oxford.

Johnson WE (1932). "Probability: The deductive and inductive problems." *Mind*, **41**(164), 409–423.

Kneser R, Ney H (1995). "Improved backing-off for M-gram language modeling." *1995 International Conference on Acoustics, Speech, and Signal Processing*, **1**, 181-184 vol.1.

Lidstone GJ (1920). "Note on the general case of the Bayes-Laplace formula for inductive or a posteriori probabilities." *Transactions of the Faculty of Actuaries*, **8**(182-192), 13.

Markov AA (1913). “Essai d’une Recherche Statistique Sur le Texte du Roman Eugene Oneguine.” *Bull. Acad. Imper. Sci. St. Petersburg*, **7**.

Ney H, Essen U (1991). “On smoothing techniques for bigram-based natural language modelling.” In *Acoustics, Speech, and Signal Processing, IEEE International Conference on*, 825–828. IEEE Computer Society.

Witten IH, Bell TC (1991). “The zero-frequency problem: Estimating the probabilities of novel events in adaptive text compression.” *Ieee transactions on information theory*, **37**(4), 1085–1094.

Examples

```
# List available smoothers
smoothers()

# Get information on smoother "kn", i.e. Interpolated Kneser-Ney
info("kn")
```

tknz_sent	<i>Sentence tokenizer</i>
-----------	---------------------------

Description

Extract sentences from a batch of text lines.

Usage

```
tknz_sent(input, EOS = "[.?!:;]+", keep_first = FALSE)
```

Arguments

input	a character vector.
EOS	a regular expression matching an End-Of-Sentence delimiter.
keep_first	TRUE or FALSE? Should the first character of the matches be appended to the returned sentences (with a space)?

Details

tknz_sent() splits text into sentences, where sentence delimiters are specified by a regular expression through the EOS argument. Specifically, when an EOS token is found, the next sentence begins at the first position in the input string not containing any of the EOS tokens *or white space* (so that entries like "Hi there!!!" or "Hello . . ." are both recognized as a single sentence).

If keep_first is FALSE, the delimiters are stripped off from the returned sequences. Otherwise, the first character of the substrings matching the EOS regular expressions are appended to the corresponding sentences, preceded by a white space.

In the absence of any EOS delimiter, `tknz_sent()` returns the input as is, since parts of text corresponding to different entries of the input vector `x` are understood as parts of separate sentences.

Note. This function, as well as [preprocess](#), are included in the library for illustrative purposes only, and are not optimized for performance. Furthermore (for performance reasons) the function has a separate implementation for Windows and UNIX OS types, respectively, so that results obtained in the two cases may differ slightly. In contexts that require full reproducibility, users are encouraged to define their own preprocessing and tokenization custom functions - or to work with externally processed data.

Value

a character vector, each entry of which corresponds to a single sentence.

Author(s)

Valerio Gherardi

Examples

```
tknz_sent("Hi there! I'm using kgrams.")
```

word_context

Word-context conditional expression

Description

Create word-context conditional expression with the `%|%` operator.

Usage

```
word %|% context
```

Arguments

word	a character vector. Word or words to include as the variable part of the conditional expression.
context	a character vector of length one. The fixed (or "given") part of the conditional expression.

Details

The intuitive meaning of the operator `%|%` is that of the mathematical symbol `|` (given). This operator is used to create conditional expressions representing the occurrence of some word after a given context (for instance, the expression `"you" %|% "i love"` would represent the occurrence of the word `"you"` after the string `"i love"`). The purpose of `%|%` is to create objects which can be given as input to `probability()` (see [probability](#) for further examples).

Value

a `word_context` class object.

Author(s)

Valerio Gherardi

Examples

```
f <- kgram_freqs(much_ado, 2, .tknz_sent = tknz_sent)
m <- language_model(f, "kn", D = 0.5)
probability("leonato" %|% "enter", m)
```

%+%

String concatenation

Description

String concatenation

Usage

```
lhs %+% rhs
```

Arguments

lhs	a string or vector of strings.
rhs	a string or vector of strings.

Details

The expression `lhs %+% rhs` is equivalent to `paste(lhs, rhs, sep = " ", collapse = NULL, recycle0 = FALSE)`. See [paste](#) for more details.

Value

a string or vector of strings.

Author(s)

Valerio Gherardi
Brief synthax for string concatenation.

See Also

[paste](#)

`%+%`

27

Examples

```
"i love" %+% c("cats", "jazz", "you")
```

Index

- * **datasets**
 - midsummer, [11](#)
 - much_ado, [12](#)
- %+%, [26](#)
- as.character.kgrams_dictionary
(dictionary), [2](#)
- as_dictionary (dictionary), [2](#)
- BOS (EOS), [4](#)
- dictionary, [2](#), [7](#), [9](#)
- EOS, [4](#)
- info (smoothers), [22](#)
- kgram_freqs, [5](#), [16](#)
- kgrams, [11](#), [16](#), [23](#)
- language_model, [5](#), [8](#), [9](#), [10](#), [13](#), [16](#)
- midsummer, [11](#), [12](#)
- much_ado, [12](#), [12](#)
- param (parameters), [13](#)
- param<- (parameters), [13](#)
- parameters, [11](#), [13](#)
- paste, [26](#)
- perplexity, [10](#), [14](#)
- preprocess, [9](#), [17](#), [25](#)
- probability, [5](#), [8–11](#), [18](#), [25](#)
- process_sentences (kgram_freqs), [5](#)
- query, [5](#), [8](#), [9](#), [19](#), [19](#)
- sample_sentences, [10](#), [11](#), [21](#)
- smoothers, [11](#), [13](#), [22](#)
- special_tokens, [20](#)
- special_tokens (EOS), [4](#)
- tknz_sent, [9](#), [17](#), [24](#)
- UNK (EOS), [4](#)
- word_context, [18](#), [25](#)