# Package 'LSX'

December 9, 2025

```
Description A word embeddings-based semi-supervised model for document scaling Watan-
     abe (2020) <doi:10.1080/19312458.2020.1832976>.
     LSS allows users to analyze large and complex corpora on arbitrary dimen-
     sions with seed words exploiting efficiency of word embeddings (SVD, Glove).
     It can generate word vectors on a users-provided corpus or incorporate a pre-trained word vectors.
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as.seedwords

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as.s	eedwords Convert a list or a dictionary to seed words	

# Description

Convert a list or a dictionary to seed words

# Usage

```
as.seedwords(x, upper = 1, lower = 2, concatenator = "_")
```

# Arguments

a list of characters vectors or a dictionary object.
 upper
 numeric index or key for seed words for higher scores.
 lower
 numeric index or key for seed words for lower scores.
 concatenator
 character to replace separators of multi-word seed words.

# Value

named numeric vector for seed words with polarity scores

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as.textmodel\_lss

Create a Latent Semantic Scaling model from various objects

# **Description**

Create a new textmodel\_lss object from an existing or foreign objects.

## Usage

```
as.textmodel_lss(x, ...)
## S3 method for class 'matrix'
as.textmodel_lss(
 х,
  seeds,
  terms = NULL,
  slice = NULL,
  simil_method = "cosine",
  auto_weight = FALSE,
  verbose = FALSE,
)
## S3 method for class 'numeric'
as.textmodel_lss(x, ...)
## S3 method for class 'textmodel_lss'
as.textmodel_lss(x, ...)
## S3 method for class 'textmodel_wordvector'
as.textmodel_lss(x, seeds, terms = NULL, verbose = FALSE, spatial = TRUE, \dots)
```

# Arguments

X	an object from which a new textmodel_lss object is created. See details.
	arguments used to create a new object. seeds must be given when x is a dense matrix or a fitted textmodel_lss.
seeds	a character vector or named numeric vector that contains seed words. If seed words contain "*", they are interpreted as glob patterns. See quanteda::valuetype.
terms	a character vector or named numeric vector that specify words for which polarity scores will be computed; if a numeric vector, words' polarity scores will be weighted accordingly; if NULL, all the features in x except those less frequent than min_count will be used.
slice	a number or indices of the components of word vectors used to compute similarity; slice < k to further truncate word vectors; useful for diagnosys and simulation.

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simil_method	specifies method to compute similarity between features. The value is passed to quanteda.textstats::textstat_simil(), "cosine" is used otherwise.
auto_weight	automatically determine weights to approximate the polarity of terms to seed words. Deprecated.
verbose	show messages if TRUE.
spatial	if TRUE, return a spatial model. Otherwise, a probabilistic model.

## **Details**

If x is a textmodel\_lss, original word vectors are reused to compute polarity scores with new seed words. It is also possible to subset word vectors via slice if it was trained originally using SVD.

If x is a dense matrix, it is treated as a column-oriented word vectors with which polarity of words are computed. If x is a named numeric vector, the values are treated as polarity scores of the words in the names.

#### Value

a dummy textmodel\_lss object

bootstrap\_lss [experimental] Compute polarity scores with different hyperparameters

# **Description**

A function to compute polarity scores of words and documents by resampling hyper-parameters from a fitted LSS model.

```
bootstrap_lss(
    x,
    what = c("seeds", "k"),
    mode = c("terms", "coef", "predict"),
    remove = FALSE,
    from = 100,
    to = NULL,
    by = 50,
    verbose = FALSE,
    ...
)
```

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#### **Arguments**

x a fitted textmodel\_lss object.

what choose the hyper-parameter to resample in bootstrapping.

mode choose the type of the result of bootstrapping. If coef, returns the polarity scores

of words; if terms, returns words sorted by the polarity scores in descending

order; if predict, returns the polarity scores of documents.

remove if TRUE, remove each seed word when what = "seeds".

from, to, by passed to seq() to generate values for k; only used when what = "k".

verbose show messages if TRUE.

... additional arguments passed to as.textmodel\_lss() and predict().

#### **Details**

bootstrap\_lss() creates LSS fitted textmodel\_lss objects internally by resampling hyper-parameters and computes polarity of words or documents. The resulting matrix can be used to asses the validity and the reliability of seeds or k.

Note that the objects created by as.textmodel\_lss() does not contain data, users must pass newdata via ... when mode = "predict".

 $coef.textmodel\_lss$ 

Extract model coefficients from a fitted textmodel\_lss object

# Description

coef() extract model coefficients from a fitted textmodel\_lss object. coefficients() is an alias.

#### Usage

```
## $3 method for class 'textmodel_lss'
coef(object, ...)
coefficients.textmodel_lss(object, ...)
```

# **Arguments**

object a fitted textmodel\_lss object.

... not used.

data\_dictionary\_ideology

Seed words for analysis of left-right political ideology

#### **Description**

Seed words for analysis of left-right political ideology

# **Examples**

as.seedwords(data\_dictionary\_ideology)

data\_dictionary\_sentiment

Seed words for analysis of positive-negative sentiment

#### **Description**

Seed words for analysis of positive-negative sentiment

#### References

Turney, P. D., & Littman, M. L. (2003). Measuring Praise and Criticism: Inference of Semantic Orientation from Association. ACM Trans. Inf. Syst., 21(4), 315–346. doi:10.1145/944012.944013

#### **Examples**

as.seedwords(data\_dictionary\_sentiment)

data\_textmodel\_lss\_russianprotests

A fitted LSS model on street protest in Russia

## Description

This model was trained on a Russian media corpus (newspapers, TV transcripts and newswires) to analyze framing of street protests. The scale is protests as "freedom of expression" (high) vs "social disorder" (low). Although some slots are missing in this object (because the model was imported from the original Python implementation), it allows you to scale texts using predict.

## References

Lankina, Tomila, and Kohei Watanabe. "'Russian Spring' or 'Spring Betrayal'? The Media as a Mirror of Putin's Evolving Strategy in Ukraine." Europe-Asia Studies 69, no. 10 (2017): 1526–56. doi:10.1080/09668136.2017.1397603.

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optimize\_lss [experimental] Compute variance ratios with different hyperparameters

# Description

[experimental] Compute variance ratios with different hyper-parameters

# Usage

```
optimize_lss(x, ...)
```

# **Arguments**

x a fitted textmodel\_lss object.

... additional arguments passed to bootstrap\_lss.

#### **Details**

optimize\_lss() computes variance ratios with different values of hyper-parameters using bootstrap\_lss. The variance ration v is defined as

$$v = \sigma_{documents}^2/\sigma_{words}^2.$$

It maximizes when the model best distinguishes between the documents on the latent scale.

# **Examples**

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```
predict.textmodel_lss
Prediction method for textmodel_lss
```

#### **Description**

Prediction method for textmodel\_lss

# Usage

```
## S3 method for class 'textmodel_lss'
predict(
  object,
  newdata = NULL,
  se_fit = FALSE,
  density = FALSE,
  rescale = TRUE,
  cut = NULL,
  min_n = 0L,
  ...
)
```

#### **Arguments**

```
a fitted LSS textmodel.
object
                  a dfm on which prediction should be made.
newdata
se_fit
                  if TRUE, returns standard error of document scores.
density
                  if TRUE, returns frequency of polarity words in documents.
rescale
                  if TRUE, normalizes polarity scores using scale().
cut
                   a vector of one or two percentile values to dichotomized polarty scores of words.
                   When two values are given, words between them receive zero polarity.
                   set the minimum number of polarity words in documents.
min_n
                   not used
```

#### Details

Polarity scores of documents are the means of polarity scores of words weighted by their frequency. When  $se_fit = TRUE$ , this function returns the weighted means, their standard errors, and the number of polarity words in the documents. When rescale = TRUE, it converts the raw polarity scores to z sores for easier interpretation. When rescale = FALSE and cut is used, polarity scores of documents are bounded by [-1.0, 1.0].

Documents tend to receive extreme polarity scores when they have only few polarity words. This is problematic when LSS is applied to short documents (e.g. social media posts) or individual sentences, but users can alleviate this problem by adding zero polarity words to short documents using min\_n. This setting does not affect empty documents.

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seedwords

Seed words for Latent Semantic Analysis

# **Description**

Seed words for Latent Semantic Analysis

# Usage

```
seedwords(type)
```

# **Arguments**

type

type of seed words currently only for sentiment (sentiment) or political ideology (ideology).

## References

Turney, P. D., & Littman, M. L. (2003). Measuring Praise and Criticism: Inference of Semantic Orientation from Association. ACM Trans. Inf. Syst., 21(4), 315–346. doi:10.1145/944012.944013

# **Examples**

```
seedwords('sentiment')
```

smooth\_lss

Smooth predicted polarity scores

# **Description**

Smooth predicted polarity scores by local polynomial regression.

```
smooth_lss(
    x,
    lss_var = "fit",
    date_var = "date",
    span = 0.1,
    groups = NULL,
    from = NULL,
    to = NULL,
    by = "day",
    engine = c("loess", "locfit"),
    ...
)
```

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## **Arguments**

Х a data.frame containing polarity scores and dates. lss\_var the name of the column in x for polarity scores. the name of the column in x for dates. date\_var the level of smoothing. span specify the columns in x to smooth separately by the group; the columns must groups be factor, character or logical. from, to, by the the range and the internal of the smoothed scores; passed to seq.Date. specifies the function to be used for smoothing. engine additional arguments passed to the smoothing function. . . .

#### **Details**

Smoothing is performed using stats::loess() or locfit::locfit(). When the x has more than 10000 rows, it is usually better to choose the latter by setting engine = "locfit". In this case, span is passed to locfit::lp(nn = span).

textmodel\_lss

Fit a Latent Semantic Scaling model

#### **Description**

Latent Semantic Scaling (LSS) is a semi-supervised algorithm for document scaling based on word embedding.

```
textmodel_lss(x, ...)
## S3 method for class 'dfm'
textmodel_lss(
 Х,
  seeds,
  terms = NULL,
  k = 300,
  slice = NULL,
 weight = "count",
  cache = FALSE,
  simil_method = "cosine",
  engine = c("RSpectra", "irlba", "rsvd"),
  auto_weight = FALSE,
  include_data = FALSE,
  group_data = FALSE,
  verbose = FALSE,
```

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```
)
## S3 method for class 'fcm'
textmodel_lss(
 х,
 seeds,
  terms = NULL,
 k = 50,
 max\_count = 10,
 weight = "count",
  cache = FALSE,
  simil_method = "cosine",
  engine = "rsparse",
  auto_weight = FALSE,
  verbose = FALSE,
)
## S3 method for class 'tokens'
textmodel_lss(
 Х,
  seeds,
  terms = NULL,
 k = 200,
 min_count = 5,
 engine = "wordvector",
  tolower = TRUE,
  include_data = FALSE,
  group_data = FALSE,
  spatial = TRUE,
  verbose = FALSE,
)
```

is a dfm.

# **Arguments**

х	a dfm or fcm created by quanteda::dfm(), quanteda::fcm(), quanteda::tokens or quanteda::tokens_xptr object.
	additional arguments passed to the underlying engine.
seeds	a character vector or named numeric vector that contains seed words. If seed words contain "*", they are interpreted as glob patterns. See quanteda::valuetype.
terms	a character vector or named numeric vector that specify words for which polarity scores will be computed; if a numeric vector, words' polarity scores will be weighted accordingly; if NULL, all the features in x except those less frequent than min_count will be used.
k	the number of singular values requested to the SVD engine. Only used when x

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slice a number or indices of the components of word vectors used to compute similarity; slice < k to further truncate word vectors; useful for diagnosys and simweight weighting scheme passed to quanteda::dfm\_weight(). Ignored when engine = "rsparse". if TRUE, save the result of SVD for next execution with identical x and settings. cache Use the base::options(lss\_cache\_dir) to change the location cache files to be save. simil\_method specifies method to compute similarity between features. The value is passed to quanteda.textstats::textstat\_simil(), "cosine" is used otherwise. select the engine to factorize x to generate word vectors. If x is a dfm, RSpectra::svds(), engine irlba::irlba() or rsvd::rsvd(). If x is a fcm, rsparse::GloVe(). If x is a tokens (or tokens\_xptr), wordvector::textmodel\_word2vec(). auto\_weight automatically determine weights to approximate the polarity of terms to seed words. Deprecated. include\_data if TRUE, the fitted model includes the dfm supplied as x. group\_data if TRUE, apply dfm\_group(x) before saving the dfm. verbose show messages if TRUE. passed to x\_max in rsparse::GloVe\$new() where cooccurrence counts are max\_count ceiled to this threshold. It should be changed according to the size of the corpus. Used only when x is a fcm. min\_count the minimum frequency of the words. Words less frequent than this in x are removed before training.

# Details

tolower

spatial

Latent Semantic Scaling (LSS) is a semisupervised document scaling method. textmodel\_lss() constructs word vectors from use-provided documents (x) and weights words (terms) based on their semantic proximity to seed words (seeds). Seed words are any known polarity words (e.g. sentiment words) that users should manually choose. The required number of seed words are usually 5 to 10 for each end of the scale.

[experimental] if FALSE, return a probabilistic model. See the details.

if TRUE, lower-case all the words in the model.

If seeds is a named numeric vector with positive and negative values, a bipolar model is construct; if seeds is a character vector, a unipolar model. Usually bipolar models perform better in document scaling because both ends of the scale are defined by the user.

A seed word's polarity score computed by textmodel\_lss() tends to diverge from its original score given by the user because it's score is affected not only by its original score but also by the original scores of all other seed words. If auto\_weight = TRUE, the original scores are weighted automatically using stats::optim() to minimize the squared difference between seed words' computed and original scores. Weighted scores are saved in seed\_weighted in the object.

When x is a tokens or tokens\_xptr object, wordvector::textmodel\_word2vec is called internally with type = "skip-gram" and other arguments passed via .... If spatial = TRUE, it return a spatial model; otherwise a probabilistic model. While the polarity scores of words are their cosine

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similarity to seed words in spatial models, they are predicted probability that the seed words to occur in their contexts. The probabilistic models are still experimental, so use them with caution.

Please visit the package website for examples.

#### References

Watanabe, Kohei. 2020. "Latent Semantic Scaling: A Semisupervised Text Analysis Technique for New Domains and Languages", Communication Methods and Measures. doi:10.1080/19312458.2020.1832976.

Watanabe, Kohei. 2017. "Measuring News Bias: Russia's Official News Agency ITAR-TASS' Coverage of the Ukraine Crisis" European Journal of Communication. doi:10.1177/0267323117695735.

textplot\_simil

Plot similarity between seed words

#### **Description**

Plot similarity between seed words

# Usage

```
textplot_simil(x)
```

#### **Arguments**

Х

fitted textmodel\_lss object.

textplot\_terms

Plot polarity scores of words

# Description

Plot polarity scores of words

```
textplot_terms(
    x,
    highlighted = NULL,
    max_highlighted = 50,
    max_words = 1000,
    sampling = c("absolute", "relative"),
    ...
)
```

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#### **Arguments**

a fitted textmodel\_lss object. highlighted quanteda::pattern to select words to highlight. If a quanteda::dictionary is passed, words in the top-level categories are highlighted in different colors. max\_highlighted the maximum number of words to highlight. When highlighted = NULL, words are randomly sampled proportionally to beta ^ 2 \* log(frequency) for highlighting. max\_words the maximum number of words to plot. Words are randomly sampled to keep the number below the limit. if "relative", words are sampled based on their squared deviation from the mean sampling for highlighting; if "absolute", they are sampled based on the squared distance from zero. passed to underlying functions. See the Details.

# **Details**

Users can customize the plots through ..., which is passed to ggplot2::geom\_text() and ggrepel::geom\_text\_repel(). The colors are specified internally but users can override the settings by appending ggplot2::scale\_colour\_manual() or ggplot2::scale\_colour\_brewer(). The legend title can also be modified using ggplot2::labs().

textstat\_context

Identify context words

# **Description**

Identify context words using user-provided patterns.

```
textstat_context(
    x,
    pattern,
    valuetype = c("glob", "regex", "fixed"),
    case_insensitive = TRUE,
    window = 10,
    min_count = 10,
    remove_pattern = TRUE,
    n = 1,
    skip = 0,
    ...
)
char_context(
    x,
```

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```
pattern,
valuetype = c("glob", "regex", "fixed"),
case_insensitive = TRUE,
window = 10,
min_count = 10,
remove_pattern = TRUE,
p = 0.001,
n = 1,
skip = 0
)
```

#### **Arguments**

x a tokens object created by quanteda::tokens().

pattern quanteda::pattern() to specify target words.

valuetype the type of pattern matching: "glob" for "glob"-style wildcard expressions;

"regex" for regular expressions; or "fixed" for exact matching. See quanteda::valuetype()

for details.

case\_insensitive

if TRUE, ignore case when matching.

window size of window for collocation analysis.

min\_count minimum frequency of words within the window to be considered as colloca-

tions.

remove\_pattern if TRUE, keywords do not contain target words.

n integer vector specifying the number of elements to be concatenated in each

n-gram. Each element of this vector will define a n in the n-gram(s) that are

produced.

skip integer vector specifying the adjacency skip size for tokens forming the n-grams,

default is 0 for only immediately neighbouring words. For skipgrams, skip can be a vector of integers, as the "classic" approach to forming skip-grams is to set skip = k where k is the distance for which k or fewer skips are used to construct the n-gram. Thus a "4-skip-n-gram" defined as skip = 0:4 produces results that include 4 skips, 3 skips, 2 skips, 1 skip, and 0 skips (where 0 skips are typical

n-grams formed from adjacent words). See Guthrie et al (2006).

... additional arguments passed to quanteda.textstats::textstat\_keyness().

p threshold for statistical significance of collocations.

#### See Also

```
quanteda.textstats::textstat_keyness()
```

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