## Package 'topicmodels.etm'

October 14, 2022

Type Package

Title Topic Modelling in Embedding Spaces

Version 0.1.0

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#### Description

Find topics in texts which are semantically embedded using techniques like word2vec or Glove. This topic modelling technique models each word with a categorical distribution whose natural parameter is the inner product between a word embedding and an embedding of its assigned topic. The techniques are explained in detail in the paper 'Topic Modeling in Embedding Spaces' by Adji B. Dieng, Francisco J. R. Ruiz, David M. Blei (2019), available at <arXiv:1907.04907>.

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**Encoding** UTF-8

SystemRequirements LibTorch (https://pytorch.org/)

**Depends** R (>= 2.10)

**Imports** graphics, stats, Matrix, torch (>= 0.5.0)

Suggests udpipe (>= 0.8.4), word2vec, uwot, tinytest, textplot (>= 0.2.0), ggrepel, ggalt

#### RoxygenNote 7.1.2

#### NeedsCompilation no

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#### **Repository** CRAN

Date/Publication 2021-11-08 08:40:02 UTC

### **R** topics documented:

as.matrix.ETM	 		 2
ETM	 		 3
ng20	 		 6
plot.ETM	 		 6
predict.ETM	 	•••	 8
summary.ETM	 		 9
			11

#### Index

as.matrix.ETM Get matrices out of an ETM object

#### Description

Convenience function to extract

- embeddings of the topic centers
- embeddings of the words used in the model
- words emmitted by each topic (beta), which is the softmax-transformed inner product of word embedding and topic embeddings

#### Usage

```
## S3 method for class 'ETM'
as.matrix(x, type = c("embedding", "beta"), which = c("topics", "words"), ...)
```

#### Arguments

х	an object of class ETM
type	character string with the type of information to extract: either 'beta' (words emttied by each topic) or 'embedding' (embeddings of words or topic centers). Defaults to 'embedding'.
which	a character string with either 'words' or 'topics' to get either the embeddings of the words used in the model or the embedding of the topic centers. Defaults to 'topics'. Only used if type = 'embedding'.
	not used

#### Value

a numeric matrix containing, depending on the value supplied in type either the embeddings of the topic centers, the embeddings of the words or the words emitted by each topic

#### See Also

ETM

#### ETM

#### Examples

```
library(torch)
library(topicmodels.etm)
path <- system.file(package = "topicmodels.etm", "example", "example_etm.ckpt")</pre>
model <- torch_load(path)</pre>
                   <- as.matrix(model, type = "embedding", which = "topics")
topic.centers
word.embeddings <- as.matrix(model, type = "embedding", which = "words")</pre>
topic.terminology <- as.matrix(model, type = "beta")</pre>
```

ETM

Topic Modelling in Semantic Embedding Spaces

#### Description

ETM is a generative topic model combining traditional topic models (LDA) with word embeddings (word2vec).

- It models each word with a categorical distribution whose natural parameter is the inner product between a word embedding and an embedding of its assigned topic.
- The model is fitted using an amortized variational inference algorithm on top of libtorch.

#### Usage

```
ETM(
  k = 20,
  embeddings,
 dim = 800,
 activation = c("relu", "tanh", "softplus", "rrelu", "leakyrelu", "elu", "selu",
    "glu"),
  dropout = 0.5,
  vocab = rownames(embeddings)
```

#### Arguments

)

k	the number of topics to extract
embeddings	either a matrix with pretrained word embeddings or an integer with the dimen- sion of the word embeddings. Defaults to 50 if not provided.
dim	dimension of the variational inference hyperparameter theta (passed on to nn_linear). Defaults to 800.
activation	character string with the activation function of theta. Either one of 'relu', 'tanh', 'softplus', 'rrelu', 'leakyrelu', 'elu', 'selu', 'glu'. Defaults to 'relu'.

ETM

dropout	dropout percentage on the variational distribution for theta (passed on to nn_dropout) Defaults to 0.5.
vocab	a character vector with the words from the vocabulary. Defaults to the rownames of the embeddings argument.

#### Value

an object of class ETM which is a torch nn\_module containing o.a.

- num\_topics: the number of topics
- vocab: character vector with the terminology used in the model
- vocab\_size: the number of words in vocab
- rho: The word embeddings
- alphas: The topic embeddings

#### Methods

fit(data, optimizer, epoch, batch\_size, normalize = TRUE, clip = 0, lr\_anneal\_factor = 4, lr\_anneal\_nonmor Fit the model on a document term matrix by splitting the data in 70/30 training/test set and updating the model weights.

#### Arguments

data bag of words document term matrix in dgCMatrix format

optimizer object of class torch\_Optimizer

epoch integer with the number of iterations to train

batch\_size integer with the size of the batch

normalize logical indicating to normalize the bag of words data

- clip number between 0 and 1 indicating to do gradient clipping passed on to nn\_utils\_clip\_grad\_norm\_
- **lr\_anneal\_factor** divide the learning rate by this factor when the loss on the test set is monotonic for at least lr\_anneal\_nonmono training iterations
- **lr\_anneal\_nonmono** number of iterations after which learning rate annealing is executed if the loss does not decreases

#### References

https://arxiv.org/pdf/1907.04907.pdf

#### Examples

```
library(torch)
library(topicmodels.etm)
library(word2vec)
library(udpipe)
data(brussels_reviews_anno, package = "udpipe")
##
## Toy example with pretrained embeddings
```

##

```
## a. build word2vec model
           <- subset(brussels_reviews_anno, language %in% "nl")
х
           <- paste.data.frame(x, term = "lemma", group = "doc_id")
Х
set.seed(4321)
w2v
           <- word2vec(x = x$lemma, dim = 15, iter = 20, type = "cbow", min_count = 5)
embeddings <- as.matrix(w2v)</pre>
## b. build document term matrix on nouns + adjectives, align with the embedding terms
dtm <- subset(brussels_reviews_anno, language %in% "nl" & upos %in% c("NOUN", "ADJ"))
dtm <- document_term_frequencies(dtm, document = "doc_id", term = "lemma")</pre>
dtm <- document_term_matrix(dtm)</pre>
dtm <- dtm_conform(dtm, columns = rownames(embeddings))</pre>
dtm <- dtm[dtm_rowsums(dtm) > 0, ]
## create and fit an embedding topic model - 8 topics, theta 100-dimensional
if (torch::torch_is_installed()) {
set.seed(4321)
torch_manual_seed(4321)
            <- ETM(k = 8, dim = 100, embeddings = embeddings, dropout = 0.5)
model
optimizer <- optim_adam(params = model$parameters, lr = 0.005, weight_decay = 0.0000012)
           <- model$fit(data = dtm, optimizer = optimizer, epoch = 40, batch_size = 1000)
overview
            <- predict(model, dtm, type = "topics")
scores
lastbatch <- subset(overview$loss, overview$loss$batch_is_last == TRUE)</pre>
plot(lastbatch$epoch, lastbatch$loss)
plot(overview$loss_test)
## show top words in each topic
terminology <- predict(model, type = "terms", top_n = 7)</pre>
terminology
##
## Toy example without pretrained word embeddings
##
set.seed(4321)
torch_manual_seed(4321)
           <- ETM(k = 8, dim = 100, embeddings = 15, dropout = 0.5, vocab = colnames(dtm))
model
optimizer <- optim_adam(params = model$parameters, lr = 0.005, weight_decay = 0.000012)
          <- model$fit(data = dtm, optimizer = optimizer, epoch = 40, batch_size = 1000)</pre>
overview
terminology <- predict(model, type = "terms", top_n = 7)</pre>
terminology
```

ng20

#### Description

Data available at https://github.com/bnosac-dev/ETM/tree/master/data/20ng

#### Examples

```
data(ng20)
str(ng20$vocab)
str(ng20$bow_tr$tokens)
str(ng20$bow_tr$counts)
```

plot.ETM

Plot functionality for an ETM object

#### Description

Convenience function allowing to plot

- the evolution of the loss on the training / test set in order to inspect training convergence
- the ETM model in 2D dimensional space using a umap projection. This plot uses function textplot\_embedding\_2d from the textplot R package and plots the top\_n most emitted words of each topic and the topic centers in 2 dimensions

#### Usage

```
## S3 method for class 'ETM'
plot(
    x,
    type = c("loss", "topics"),
    which,
    top_n = 4,
    title = "ETM topics",
    subtitle = "",
    encircle = FALSE,
    points = FALSE,
    ...
)
```

#### plot.ETM

#### Arguments

х	an object of class ETM
type	character string with the type of plot to generate: either 'loss' or 'topics'
which	an integer vector of topics to plot, used in case type = 'topics'. Defaults to all topics. See the example below.
top_n	passed on to summary.ETM in order to visualise the top_n most relevant words for each topic. Defaults to 4.
title	passed on to textplot_embedding_2d, used in case type = 'topics'
subtitle	passed on to textplot_embedding_2d, used in case type = 'topics'
encircle	passed on to textplot_embedding_2d, used in case type = 'topics'
points	passed on to textplot_embedding_2d, used in case type = 'topics'
	arguments passed on to summary.ETM

#### Value

In case type is set to 'topics', maps the topic centers and most emitted words for each topic to 2D using summary.ETM and returns a ggplot object by calling textplot\_embedding\_2d. For type 'loss', makes a base graphics plot and returns invisibly nothing.

#### See Also

ETM, summary.ETM, textplot\_embedding\_2d

#### Examples

```
library(torch)
library(topicmodels.etm)
path <- system.file(package = "topicmodels.etm", "example", "example_etm.ckpt")
model <- torch_load(path)
plot(model, type = "loss")</pre>
```

predict.ETM

#### Description

Predict to which ETM topic a text belongs or extract which words are emitted for each topic.

#### Usage

```
## S3 method for class 'ETM'
predict(
   object,
   newdata,
   type = c("topics", "terms"),
   batch_size = nrow(newdata),
   normalize = TRUE,
   top_n = 10,
   ...
)
```

#### Arguments

object	an object of class ETM
newdata	bag of words document term matrix in dgCMatrix format. Only used in case type = 'topics'.
type	a character string with either 'topics' or 'terms' indicating to either predict to which topic a document encoded as a set of bag of words belongs to or to extract the most emitted terms for each topic
batch_size	integer with the size of the batch in order to do chunkwise predictions in chunks of batch_size rows. Defaults to the whole dataset provided in newdata. Only used in case type = 'topics'.
normalize	logical indicating to normalize the bag of words data. Defaults to TRUE similar as the default when building the ETM model. Only used in case type = 'topics'.
top_n	integer with the number of most relevant words for each topic to extract. Only used in case type = 'terms'.
	not used

#### Value

Returns for

- type 'topics': a matrix with topic probabilities of dimension nrow(newdata) x the number of topics
- type 'terms': a list of data.frame's where each data.frame has columns term, beta and rank indicating the top\_n most emitted terms for that topic. List element 1 corresponds to the top terms emitted by topic 1, element 2 to topic 2 ...

#### summary.ETM

#### See Also

ETM

#### Examples

```
library(torch)
library(topicmodels.etm)
path <- system.file(package = "topicmodels.etm", "example", "example_etm.ckpt")
model <- torch_load(path)
# Get most emitted words for each topic
terminology <- predict(model, type = "terms", top_n = 5)
terminology
# Get topics probabilities for each document
path <- system.file(package = "topicmodels.etm", "example", "example_dtm.rds")
dtm <- neadRDS(path)
dtm <- head(dtm, n = 5)
scores <- predict(model, newdata = dtm, type = "topics")
scores
```

summary.ETM

#### Project ETM embeddings using UMAP

#### Description

Uses the uwot package to map the word embeddings and the center of the topic embeddings to a 2-dimensional space

#### Usage

```
## S3 method for class 'ETM'
summary(object, type = c("umap"), n_components = 2, top_n = 20, ...)
```

#### Arguments

object	object of class ETM
type	character string with the type of summary to extract. Defaults to 'umap', no other summary information currently implemented.
n_components	the dimension of the space to embed into. Passed on to umap. Defaults to 2.
top_n	passed on to predict.ETM to get the top_n most relevant words for each topic in the 2-dimensional space
	further arguments passed onto umap

#### Value

a list with elements

- center: a matrix with the embeddings of the topic centers
- words: a matrix with the embeddings of the words
- embed\_2d: a data.frame which contains a lower dimensional presentation in 2D of the topics and the top\_n words associated with the topic, containing columns type, term, cluster (the topic number), rank, beta, x, y, weight; where type is either 'words' or 'centers', x/y contain the lower dimensional positions in 2D of the word and weight is the emitted beta scaled to the highest beta within a topic where the topic center always gets weight 0.8

#### See Also

umap, ETM

#### Examples

# Index

as.matrix.ETM, 2

ETM, 2, 3, 7, 9, 10

ng20, 6
nn\_dropout, 4
nn\_linear, 3
nn\_utils\_clip\_grad\_norm\_, 4

plot.ETM,6 predict.ETM,8,9

summary.ETM, 7,9

textplot\_embedding\_2d, 6, 7

umap, *9*, *10*