Package ‘deepMOU’

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Abstracts

Abstracts dataset

Description

Dataset composed by titles and abstracts of articles published in 2015 in five statistical journals.

Usage

data(Abstracts)

Format

A matrix for the bag-of-words representation of the Abstracts dataset; terms are sorted in alphabetical order.

Details

These are the titles and abstracts of all the articles published in 2015 by the following journals:
- Journal of American Statistical Association (JASA)
- Journal of the Royal Statistical Society - Series B
- Annals of Statistics
- Biometrika
- Statistical Science

The dataset comprises 379 articles with a vocabulary of 606 words already pre-processed (stemmed, lemmatized, stopwords removal etc.); terms with entropy less than 0.3 were discarded (rule-of-thumb threshold).

Examples

```r
x = data(Abstracts)
print(head(x))
```
bubble_clust  Bubble plot

Description

Bi-dimensional representation (via Multi-Dimensional Scaling) of the clusters, where each bubble corresponds to a cluster, its size is proportional to the relative frequency of documents and color saturation reflects cluster cohesion.

Usage

```r
bubble_clust(
  x,
  clusters,
  bubble_size = 1,
  bubble_col = c("red", "white"),
  cex_text = 1,
  main = NULL
)
```

Arguments

- **x**: Document-term matrix describing the frequency of terms that occur in a collection of documents. Rows correspond to documents in the collection and columns correspond to terms.
- **clusters**: Integer vector of length of the number of cases, which indicates a clustering. The clusters have to be numbered from 1 to the number of clusters.
- **bubble_size**: Graphical parameter for bubbles size.
- **bubble_col**: Choose palette with two colors (default "red" and "white"). The first (darker) color will denote homogeneous clusters, while the latter (lighter) more heterogeneous ones.
- **cex_text**: Size of texts inside bubbles.
- **main**: A title for the plot.

Value

A graphical aid to visualize and to describe the obtained clusters.

Examples

```r
# Load the CNAE2 dataset
data("CNAE2")

# Perform parameter estimation and clustering
mou_CNAE2 = mou_EM(x = CNAE2, k = 2)
```
# Usage of the function

```r
bubble_clust(mou_CNAE2$x, mou_CNAE2$clusters, bubble_size = 5)
```

---

### cl_CNAE

**Classification labels of the CNAE2 data set**

**Description**

True classification labels of documents from the CNAE2 data set.

**Usage**

```r
data(CNAE2)
```

**Format**

A vector containing the true classification labels of documents of CNAE2 data set.

**Source**

See CNAE2 for further details.

---

### CNAE2

**CNAE dataset on classes 4 and 9**

**Description**

The data set CNAE2 is a subset of the original CNAE-9 data, that comprises 1080 documents categorized into 9 topics of free text business descriptions of Brazilian companies. Specifically, CNAE2 contains only the documents belonging to topics "4" and "9". The data set is already pre-processed and provides the bag-of-words representation of the documents; the columns with null counts are removed leading to a matrix with 240 documents on a vocabulary with cardinality 357. This data set is highly sparse (98 Class labels are stored in cl_CNAE

**Usage**

```r
data(CNAE2)
```

**Format**

A matrix for the bag-of-words representation of the CNAE2 dataset.

**Source**

Original CNAE9 dataset
### Examples

```r
x = data(CNAE2)
print(head(x))
```

### Description

Performs parameter estimation by means of Gibbs sampling and cluster allocation for the Deep Mixture of Unigrams.

### Usage

```r
dep_mou_gibbs(x, k, g, n_it = 500, seed_choice = 1, burn_in = 200)
```

### Arguments

- **x**: Document-term matrix describing the frequency of terms that occur in a collection of documents. Rows correspond to documents in the collection and columns correspond to terms.
- **k**: Number of clusters/groups at the top layer.
- **g**: Number of clusters at the bottom layer.
- **n_it**: Number of Gibbs steps.
- **seed_choice**: Set seed for reproducible results.
- **burn_in**: Number of initial Gibbs samples to be discarded and not included in the computation of final estimates.

### Details

Starting from the data matrix `x`, the Deep Mixture of Unigrams is fitted and `k` clusters are obtained. The algorithm for the estimation of the parameters is the Gibbs sampling. In particular, the function assigns initial values to all the parameters to be estimated. Then `n_it` samples for the parameters are obtained using conditional distributions on all the other parameters. The final estimates are obtained by averaging the samples given that initial `burn_in` samples are discarded. Clustering is eventually performed by maximizing the posterior distribution of the latent variables. For further details see the references.

### Value

A list containing the following elements:

- **x**: The data matrix.
- **clusters**: the clustering labels.
- **k**: the number of clusters at the top layer.
\[ g \] the number of clusters at the bottom layer.
\[ \text{numobs} \] the sample size.
\[ p \] the vocabulary size.
\[ z1 \] the allocation variables at the top layer.
\[ z2 \] the allocation variables at the bottom layer.
\[ \text{Alpha} \] the estimates of Alpha parameters.
\[ \text{Beta} \] the estimates of the Beta parameters.
\[ \text{pi\_hat} \] estimated probabilities of belonging to the \( k \) clusters at the top layer conditional to the \( g \) clusters at the bottom layer.
\[ \text{pi\_hat\_2} \] estimated probabilities of belonging to the \( g \) clusters at the bottom layer.

References


Examples

```r
# Load the CNAE2 dataset
data("CNAE2")

# Perform parameter estimation and clustering, very few iterations used for this example
depth_CNAE2 = deep_mou_gibbs(x = CNAE2, k = 2, g = 2, n_it = 5, burn_in = 2)

# Shows cluster labels to documents
depth_CNAE2$clusters
```

```
dir_mult_GD

Dirichlet-Multinomial mixture model by Gradient Descend algorithm

Description


Usage

```r
dir_mult_GD(  
  x,  
  k,  
  numobs = 100,  
  eps = 1e-05,  
  seed_choice = 1,  
  KK = 20,  
  min_iter = 2,  
  init = NULL  
)
```
Arguments

- **x**: Document-term matrix describing the frequency of terms that occur in a collection of documents. Rows correspond to documents in the collection and columns correspond to terms.
- **k**: Number of clusters/groups.
- **n_it**: Number of Gradient Descend steps.
- **eps**: Tolerance level for the convergence of the algorithm. Default is $1e^{-05}$.
- **seed_choice**: Set seed for reproducible results.
- **KK**: Maximum number of iterations allowed for the nlminb function (see below).
- **min_iter**: Minimum number of Gradient Descend steps.
- **init**: Vector containing the initial document allocations for the initialization of the algorithm. If NULL (default) initialization is carried out via spherical k-means (skmeans).

Details

Starting from the data given by x the Dirichlet-Multinomial mixture model is fitted and k clusters are obtained. The algorithm for the parameter estimation is the Gradient Descend. In particular, the function assigns initial values to weights of the Dirichlet-Multinomial distribution for each cluster and initial weights for the elements of the mixture. The estimates are obtained with maximum n_it steps of the Descent Algorithm algorithm or until a tolerance level eps is reached; by using the posterior distribution of the latent variable z, the documents are allocated to the cluster which maximizes the posterior distribution. For further details see the references.

Value

A list containing the following elements:

- **x**: The data matrix.
- **clusters**: the clustering labels.
- **k**: the number of clusters.
- **numobs**: the sample size.
- **p**: the vocabulary size.
- **likelihood**: vector containing the likelihood values at each iteration.
- **pi_hat**: estimated probabilities of belonging to the k clusters.
- **Theta**: matrix containing the estimates of the Theta parameters for each cluster.
- **f_z_x**: matrix containing the posterior probabilities of belonging to each cluster.
- **AIC**: Akaike Information Criterion (AIC) value of the fitted model.
- **BIC**: Bayesian Information Criterion (BIC) value of the fitted model.

References

Examples

# Load the CNAE2 dataset
data("CNAE2")

# Perform parameter estimation and clustering, very
# few iterations are used for this example
dir_CNAE2 = dir_mult_GD(x = CNAE2, k = 2, n_it = 2)

# Shows cluster labels to documents
dir_CNAE2$clusters

heatmap_words

Heatmap of word frequencies by cluster

Description

Displays the heatmap of the cluster frequency distributions of the most frequent terms sorted by the most informative ones.

Usage

heatmap_words(
  x,
  clusters,
  n_words = 50,
  legend_position = "bottom",
  font_size = 12,
  low_color = "grey92",
  top_color = "red",
  main = "Row frequencies of terms distribution",
  xlabel = NULL,
  ylabel = NULL,
  legend_title = "Entropy"
)

Arguments

x Document-term matrix describing the frequency of terms that occur in a collection of documents. Rows correspond to documents in the collection and columns correspond to terms.

clusters Integer vector of length of the number of cases, which indicates a clustering. The clusters have to be numbered from 1 to the number of clusters.

n_words Number of words to include in the heatmap (default is 50).

legend_position Position of the legend ("none", "left", "right", "bottom", "top", or two-element numeric vector as in theme). Default is "bottom".
font_size  
low_color  
top_color  
main  
xlabel  
ylabel  
legend_title

Details

Takes as input the bag-of-words matrix and returns a heatmap displaying the row frequency distribution of terms according to the clusters. Words are sorted by entropy.

Value

A graphical aid to describe the clusters according to the most informative words.

Examples

# Load the CNAE2 dataset
data("CNAE2")

# Get document labels by clustering using mou_EM
mou_CNAE2 = mou_EM(x = CNAE2, k = 2)

# Usage of the function
heatmap_words(x = mou_CNAE2$x, clusters = mou_CNAE2$clusters)

mou_EM

Description

Performs parameter estimation by means of the Expectation-Maximization (EM) algorithm and cluster allocation for the Mixture of Unigrams.

Usage

mou_EM(x, k, n_it = 100, eps = 1e-07, seed_choice = 1, init = NULL)
Arguments

- **x**: Document-term matrix describing the frequency of terms that occur in a collection of documents. Rows correspond to documents in the collection and columns correspond to terms.
- **k**: Number of clusters/groups.
- **n_it**: Number of iterations for the Expectation-Maximization algorithm.
- **eps**: Tolerance level for the convergence of the algorithm. Default is $1e^{-07}$.
- **seed_choice**: Set seed for reproducible results.
- **init**: Vector containing the initial document allocations for the initialization of the algorithm. If NULL (default) initialization is carried out via spherical k-means (skmeans).

Details

Starting from the data given by **x** the Mixture of Unigrams is fitted and **k** clusters are obtained. The algorithm for the parameter estimation is the Expectation-Maximization (EM). In particular, the function assigns initial values to weights of the Multinomial distribution for each cluster and initial weights for the components of the mixture. The estimates are obtained with maximum **n_it** steps of the EM algorithm or until the tolerance level **eps** is reached; by using the posterior distribution of the latent variable **z**, the documents are allocated to the cluster which maximizes the posterior distribution. For further details see the references.

Value

A list containing the following elements:

- **x**: The data matrix.
- **clusters**: the clustering labels.
- **k**: the number of clusters.
- **numobs**: the sample size.
- **p**: the vocabulary size.
- **likelihood**: vector containing the likelihood values at each iteration.
- **pi_hat**: estimated probabilities of belonging to the **k** clusters.
- **omega**: matrix containing the estimates of the omega parameters for each cluster.
- **f_z_x**: matrix containing the posterior probabilities of belonging to each cluster.
- **AIC**: Akaike Information Criterion (AIC) value of the fitted model.
- **BIC**: Bayesian Information Criterion (BIC) value of the fitted model.

References

Examples

```r
# Load the CNAE2 dataset
data("CNAE2")

# Perform parameter estimation and clustering
mou_CNAE2 = mou_EM(x = CNAE2, k = 2)

# Shows cluster labels to documents
mou_CNAE2$clusters
```

```
plot.deepMOU

Plotting method for "shallow" and deep mixtures of Unigrams and mixtures of Dirichlet-Multinomials

Description

Bi-dimensional representation (via Multi-Dimensional Scaling) of the clusters, where each bubble corresponds to a cluster, its size is proportional to the relative frequency of documents and color saturation reflects cluster cohesion.

Usage

```r
## S3 method for class 'deepMOU'
plot(
  x,
  y,
  bubble_size = 1,
  bubble_col = c("red", "white"),
  cex_text = 1,
  main = NULL,
  ...
)
```

Arguments

- `x`: Output from `mou_EM`, `dir_mult_GD` or `deep_mou_gibbs`.
- `y`: Parameter not used
- `bubble_size`: Graphical parameter for bubbles size.
- `bubble_col`: Choose palette with two colors (default "red" and "white"). The first (darker) color will denote homogeneous clusters, while the latter (lighter) more heterogeneous ones.
- `cex_text`: Size of texts inside bubbles.
- `main`: A main title for the plot.
- `...`: Parameter not used
Details

The default graphical representation of mou_EM, dir_mult_GD and deep_mou_gibbs is the bubble plot. Namely, a bi-dimensional representation (via Multi-Dimensional Scaling) of the clusters, each bubble corresponds to a cluster, its size is proportional to the relative frequency of documents and color saturation reflects cluster cohesion.

Value

A graphical aid to visualize and to describe the obtained clusters.

Examples

```r
# Load the CNAE2 dataset
data("CNAE2")

# Perform parameter estimation and clustering
mou_CNAE2 = mou_EM(x = CNAE2, k = 2)

# Usage of the function
plot(mou_CNAE2, bubble_size = 5 )
```

words_freq_plot

Graph of most frequent words of each cluster

Description

Graphical plot of the most frequent words of each cluster

Usage

```r
words_freq_plot(
  x,
  clusters,
  clust_label = NULL,
  n_words = 5,
  words_size = 2,
  axis_size = 1,
  set_colors = NA,
  main = "Most frequent words for each cluster",
  xlabel = "",
  ylabel = ""
)
```
words_freq_plot

Arguments

- **x**: Document-term matrix describing the frequency of terms that occur in a collection of documents. Rows correspond to documents in the collection and columns correspond to terms.

- **clusters**: Integer vector of length of the number of cases, which indicates a clustering. The clusters have to be numbered from 1 to the number of clusters.

- **clust_label**: Vector of length of the number of cluster containing the cluster names to be displayed (by default "Cluster_1", "Cluster_2", ...).

- **n_words**: Number of words to display.

- **words_size**: A numerical value giving the amount by which plotting words should be magnified with respect to the default setting.

- **axis_size**: Magnification to be used for axis annotation with respect to the default setting.

- **set_colors**: Choose palette for word colors.

- **main**: A title for the plot. Default is "Most frequent words for each cluster".

- **xlabel**: A title for the x-axis. Default is empty.

- **ylabel**: A title for the y-axis. Default is empty.

Details

The number of most frequent words to be shown can be set by n_words and also clusters names can be passed beforehand as a character vector to clust_label.

Value

A graphical aid for visualizing the most frequent terms for each cluster.

Examples

```r
# Load the CNAE2 dataset
data("CNAE2")

# Perform parameter estimation and clustering
mou_CNAE2 = mou_EM(x = CNAE2, k = 2)

# Usage of the function
words_freq_plot(mou_CNAE2$x, mou_CNAE2$clusters,n_words = 4, words_size = 2, main = "Example")
```
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