

Package: rrecsys (via r-universe)

September 8, 2024

Type Package

Title Environment for Evaluating Recommender Systems

Version 0.9.7.2

Date 2018-02-10

URL <https://rrecsys.inf.unibz.it/>

BugReports <https://github.com/ludovikcoba/rrecsys/issues>

Description Processes standard recommendation datasets (e.g., a user-item rating matrix) as input and generates rating predictions and lists of recommended items. Standard algorithm implementations which are included in this package are the following: Global/Item/User-Average baselines, Weighted Slope One, Item-Based KNN, User-Based KNN, FunkSVD, BPR and weighted ALS. They can be assessed according to the standard offline evaluation methodology (Shani, et al. (2011) <[doi:10.1007/978-0-387-85820-3_8](https://doi.org/10.1007/978-0-387-85820-3_8)>) for recommender systems using measures such as MAE, RMSE, Precision, Recall, F1, AUC, NDCG, RankScore and coverage measures. The package (Coba, et al.(2017) <[doi:10.1007/978-3-319-60042-0_36](https://doi.org/10.1007/978-3-319-60042-0_36)>) is intended for rapid prototyping of recommendation algorithms and education purposes.

Imports methods, Rcpp

Depends R (>= 3.1.2), registry, MASS, stats, knitr, ggplot2

License GPL-3

VignetteBuilder knitr

Encoding UTF-8

LinkingTo Rcpp

Repository <https://ludovikcoba.r-universe.dev>

RemoteUrl <https://github.com/ludovikcoba/rrecsys>

RemoteRef HEAD

RemoteSha 8511f2770c10b30cabb630621c121b44afe4f5bd

Contents

algAverageClass	2
BPRclass	3
dataChart	3
dataSet-class	4
defineData	5
evalChart	6
evalModel	7
evalModel-class	8
evalPred	9
evalRec	10
evalRecResults	11
eval_nDCG	12
getAUC	13
histogram	14
IBclass	14
ml100k	15
mlLatest100k	15
PPLclass	16
predict	16
rankScore	17
recommend	18
rrecsys	19
setStoppingCriteria	22
slopeOneClass	23
sparseDataSet-class	24
SVDclass	25
UBclass	25
wALSclass	26
_ds-class	26
Index	27

algAverageClass	<i>Baseline algorithms exploiting global/item and user averages.</i>
-----------------	--

Description

Container for the model learned using any average(global, user or item) based model.

Slots

alg: The algorithm denominator, of class "character".

data: the dataset used for training the model, class "matrix".

average: average calculated either globally, on user or item, class "matrix".

Methods

```
show signature(object = "algAverageClass")
```

See Also

[rrecsys.](#)

BPRclass

Bayesian Personalized Ranking based model.

Description

Container for the model learned using any Bayesian Personalized Ranking based model.

Slots

alg: The algorithm denominator, of class "character".

data: the dataset used for training the model, class "matrix".

factors: user(U) and items(V) factors, class "list".

parameters: the parameters(such as number of factors k, learning rate lambda, user regularization term regU, positive rated item regularization term regI, negative rated item regularization term regJ and the Boolean updateJ to decide whatever negative updates are required) used in the model, class "list".

Methods

```
show signature(object = "BPRclass")
```

See Also

[rrecsys.](#)

dataChart

Visualization of data characteristics.

Description

This method visualizes data characteristics on a two dimensional graph, where "x" axes shows either items ordered by descending popularity, or users based on the number of ratings they have submitted. Moreover the "y" axes shows the number of ratings.

Usage

```
dataChart(data, x = "items", y = "num_of_ratings")
```

Arguments

data	the dataset, class "_ds".
x	class "character", is the variable that will be shown on the "x" axis. Possible values are: "items", "users".
y	class "character", is the variable that will be shown on the "y" axis. Possible values are: "num_of_ratings", "%_of_ratings".

Value

Plot results.

See Also

See Also as [_ds-class](#).

Examples

```
data(mLlatest100k)
a <- defineData(mLlatest100k)
dataChart(a, x = "items", y = "num_of_ratings")
```

dataSet-class

Dataset class.

Description

Container for a dense dataset that distinguishes between binary and non-binary feedback datasets. Extends [_ds](#).

Slots

data: the dataset, class "matrix".

binary: class "logical", determines if the item dataset contains binary (i.e. 1/0) or non-binary ratings.

minimum: class "numeric", defines the minimal value present in the dataset.

maximum: class "numeric", defines the maximal value present in the dataset.

intScale: object of class "logical", if **TRUE** the range of ratings in the dataset contains as well half star values.

Methods

nrow signature(object = "dataSet"): number of rows of the dataset.

ncol signature(object = "dataSet"): number of columns of the dataset.

dim signature(object = "dataSet"): returns the dimensions of the dataset.

rowRatings signature(object = "dataSet"): returns the number of ratings on each row.

colRatings signature(object = "dataSet"): returns the number of ratings on each column.

numRatings signature(object = "dataSet"): returns the total number of ratings.

[signature(x = "dataSet", i = "ANY", j = "ANY", drop = "ANY")]: returns a subset of the dataset.

coerce signature(from = "dataSet", to = "matrix")

rowAverages signature(object = "dataSet"): returns the average rating on each row.

colAverages signature(object = "dataSet"): returns the average rating on each column.

Examples

```
x <- matrix(sample(c(0:5), size = 100, replace = TRUE,
  prob = c(.6,.08,.08,.08,.08,.08)), nrow = 20, byrow = TRUE)

x <- defineData(x)

colRatings(x)

rowRatings(x)

numRatings(x)

sparsity(x)

a <- x[1:10,2:3]
```

defineData

Define dataset.

Description

Defines your dataset, if either it is implicit or explicit.

Arguments

data	the dataset, class "matrix".
sparseMatrix	class "logical". If FALSE implies that the input is a dense two dimensional matrix. If TRUE implies that the input is arranged as coordinate list where entries are stored as list of (row, column, value) tuples.

binary	class "logical", defines if the item dataset consists of binary (i.e. NA/1) or non-binary ratings. Default value FALSE.
minimum	class "numeric", defines the minimal value present in the dataset. Default value 0.5.
maximum	class "numeric", defines the maximal value present in the dataset. Default value 5.
intScale	object of class "logical", if TRUE the range of ratings in the dataset contains as well half star values. Default value FALSE.
positiveThreshold	class "numeric", in case binary is TRUE, positiveThreshold defines the threshold value for binarizing the dataset (i.e. any rating value \geq positiveThreshold will be transformed to 1 and all other values to NA(corresponding to a not rated item)). Default value 0.5.

Value

Returns an object of class "dataSet".

See Also

See Also as [dataSet-class](#).

Examples

```
data(mLLatest100k)
a <- defineData(mLLatest100k)
b <- defineData(mLLatest100k,binary = TRUE ,positiveThreshold = 3)
```

evalChart

Visualization of data characteristics.

Description

This method visualizes data characteristics on a two dimensional graph, where "x" axes shows either items ordered by descending popularity, or users based on the number of ratings they have submitted. Moreover the "y" axes shows the number of ratings.

Usage

```
evalChart(res, x = "items", y = "TP", x_label, y_label, y_lim)
```

Arguments

res	evaluation results, class "evalRecResults".
x	class "character", is the variable that will be shown on the "x" axis. Possible values are: "items", "users".
y	class "character", is the variable that will be shown on the "y" axis. Possible values are: "num_of_ratings", "%_of_ratings".
x_label	class "character", the label to be printed on the "x" axes.
y_label	class "character", the label to be printed on the "y" axes.
y_lim	class "numeric", scale of the "y" axes.

Value

Plot results.

See Also

See Also as [evalRecResults-class](#).

evalModel

Creating the evaluation model.

Description

Creates the dataset split for evaluation where ratings of each user are uniformly distributed over k random folds. The function returns the list of items that are assigned to each fold, such that algorithms can be compared on the same train/test splits.

Usage

```
evalModel(data, folds)
```

Arguments

data	dataset, of class <code>_ds</code> .
folds	The number of folds to use in the k-fold cross validation, of class <code>numeric</code> , default value set to 5.

Value

An object of class [evalModel-class](#).

See Also

[evalModel-class](#), [evalRec](#), [_ds](#).

Examples

```
x <- matrix(sample(c(0:5), size = 200, replace = TRUE,
  prob = c(.6, .08, .08, .08, .08)), nrow = 20, byrow = TRUE)

d <- defineData(x)

my_2_folds <- evalModel(d, 2)          #output class evalModel.

my_2_folds
# 2 - fold cross validation model on the dataset with 20 users and 10 items.

my_2_folds@data           #the dataset.
my_2_folds@folds         #the number of folds in the model.
my_2_folds@fold_indices  #the index of each item in the fold.
```

evalModel-class	<i>Evaluation model.</i>
-----------------	--------------------------

Description

Class that contains the data and a distribution of the uniform distribution of ratings onto k-folds.

Details

The `fold_indices` list contains the indexes to access the dataset on one dimension. A matrix can be addressed as a one dimensional array, considered as an extension of each column after another. E.g: in a matrix `M` with 10 rows and 20 columns, `M[10] == M[10, 1]`; `M[12] == M[2,2]`.

Slots

data: the dataset, class "matrix".

folds: number of k - folds, class "numeric".

fold_indices: a list with k slots, each slot represents a fold and contains the index of items assigned to that fold, class "list".

fold_indices_x_user: a list that specifies specifically for each user the distribution of the items in the folds, class "list".

Methods

```
show signature(object = "evalModel")
```

evalPred	<i>Evaluates the requested prediction algorithm.</i>
----------	--

Description

Evaluates the prediction task of an algorithm with a given configuration and based on the given evaluation model. RMSE and MAE are both calculated individually for each user and then averaged over all users (in this case they will be referred as RMSE and MAE) as well as determined as the average error over all predictions (in this case they are named globalRMSE and globalMAE).

Usage

```
evalPred(model, ...)  
## S4 method for signature 'evalModel'  
evalPred(model, alg, ... )
```

Arguments

model	Object of type evalModel. See evalModel-class .
alg	The algorithm to be used in the evaluation. Of type character.
...	other attributes specific to the algorithm to be deployed. Refer to rrecsys .

Value

Returns a data frame with the RMSE, MAE, globalRMSE and globalMAE for each of the k-folds defined in the evaluation model and an average over all folds.

References

F. Ricci, L. Rokach, B. Shapira, and P. B. Kantor, editors. *Recommender Systems Handbook*. Springer, 2011. ISBN 978-0-387-85819-7. URL <http://www.springerlink.com/content/978-0-387-85819-7>.

See Also

[evalModel-class](#), [rrecsys](#).

Examples

```
x <- matrix(sample(c(0:5), size = 200, replace = TRUE,  
  prob = c(.6, .8, .8, .8, .8, .8)), nrow = 20, byrow = TRUE)  
  
x <- defineData(x)  
  
e <- evalModel(x, 2)  
  
SVDEvaluation <- evalPred(e, "FunkSVD", k = 4)  
  
SVDEvaluation
```

```
IBEvaluation <- evalPred(e, "IBKNN", simFunct = "cos", neigh = 5, coRatedThreshold = 2)

IBEvaluation
```

evalRec

Evaluates the requested recommendation algorithm.

Description

Evaluates the recommendation task of an algorithm with a given configuration and based on the given evaluation model.

Arguments

model	Object of type <code>evalModel</code> . See evalModel-class .
alg	The algorithm to be used in the evaluation. Of class character.
topN	Object of class <code>numeric</code> , specifying the number of items to be recommended per user.
topNGen	Object of class character, specifying the function used to produce the recommendations. Values: "hpr" and "mf" (currently available only for IB and UB methods).
positiveThreshold	Object of class <code>numeric</code> , indicating the threshold of the ratings to be considered a good. This attribute is not used when evaluating implicit feedback.
alpha	Object of class <code>numeric</code> , is the half-life parameter for the rankscore metric.
...	other attributes specific to the algorithm to be deployed. Refer to rrecsys .

Value

Returns an object of class `evalRecResults` with the precision, recall, F1, nDCG, RankScore, true positives(TP), false positives(FP), true negatives(TN), false negatives(FN) for each of the k-folds defined in the evaluation model and the overall average.

References

F. Ricci, L. Rokach, B. Shapira, and P. B. Kantor, editors. *Recommender Systems Handbook*. Springer, 2011. ISBN 978-0-387-85819-7. URL <http://www.springerlink.com/content/978-0-387-85819-7>.

See Also

[evalModel-class](#), [rrecsys](#), [evalRecResults-class](#).

Examples

```
x <- matrix(sample(c(0:5), size = 200, replace = TRUE,
  prob = c(.6, .8, .8, .8, .8, .8)), nrow = 20, byrow = TRUE)

x <- defineData(x)

e <- evalModel(x, 2)

SVDEvaluation <- evalRec(e, "FunkSVD", positiveThreshold = 4, k = 4)

SVDEvaluation
```

evalRecResults	<i>Evaluation results.</i>
----------------	----------------------------

Description

Defines a structure for the results obtained by evaluating an algorithm

Slots

data: class `"_ds"`, the dataset.

alg: class `"character"`, the name of the used algorithm.

topN: class `"numeric"`, the number N of Top-N items recommended to each user.

topNGen: class `"character"`, the name of the recommendation algorithm.

positiveThreshold: class `"numeric"`, indicating the threshold of the ratings to be considered a good. This attribute is not used when evaluating implicit feedback.

alpha: class `numeric`, is the half-life parameter for the rankscore metric.

parameters: class `"list"`, parameters used in the configuration of the algorithm.

TP: class `"numeric"`, True Positives count on each fold.

FP: class `"numeric"`, False Positives count on each fold.

TN: class `"numeric"`, True Negatives count on each fold.

FN: class `"numeric"`, False Negatives count on each fold.

precision: class `"numeric"`, precision measured on each fold.

recall: class `"numeric"`, recall measured on each fold.

F1: class `"numeric"`, F1 measured on each fold.

nDCG: class `"numeric"`, nDCG measured on each fold.

rankscore: class `"numeric"`, rankscore measured on each fold.

item_coverage: class `"numeric"`, item coverage.

user_coverage: class "numeric", user coverage.
 ex.time: class "numeric", the execution time.
 TP_count: class "numeric", True positives count on each item.
 rec_counts: class "numeric", counts how many times an item was recommended.
 rec_popularity: class "numeric", popularity of recommendations.

Methods

show signature(object = "evalRecResults")
 results signature(object = "evalRecResults", metrics = "character"): returns a subset of the results based on the required metric.

eval_nDCG

Normalized Discounted Cumulative Gain

Description

Metric for information retrieval where positions are discounted logarithmically.

Usage

eval_nDCG(recommendedIDX, testSetIDX)

Arguments

recommendedIDX indices of the recommended items. Object of class numeric.
 testSetIDX indices of the items in the test set. Object of class numeric

Details

nDCG is computed as the ratio between Discounted Cumulative Gain(DCG) and idealized Discounted Cumulative Gain(IDCG):

$$DCG_{pos} = rel_1 + \sum_{i=2}^{pos} \frac{rel_i}{\log_2 i}$$

$$IDCG_{pos} = rel_1 + \sum_{i=2}^{|h|-1} \frac{rel_i}{\log_2 i}$$

$$nDCG_{pos} = \frac{DCG}{IDCG}$$

References

Asela Gunawardana, Guy Shani, Evaluating Recommender Systems.

getAUC	Returns the Area under the ROC curve.
--------	---------------------------------------

Description

Computes the Area Under the ROC curve for a recommendation task of an algorithm with its given configuration and based on the given evaluation model.

Usage

```
getAUC(model, ...)  
## S4 method for signature 'evalModel'  
getAUC(model, alg, ... )
```

Arguments

model	Object of type evalModel. See evalModel-class .
alg	The algorithm to be used in the evaluation. Of class character.
...	other attributes specific to the algorithm to be deployed. Refer to rrecsys .

Value

Returns a data frame with the AUC for each of the k-folds defined in the evaluation model and the overall average.

References

T. Fawcett, “ROC Graphs: Notes and Practical Considerations for Data Mining Researchers ROC Graphs : Notes and Practical Considerations for Data Mining Researchers,” HP Inven., p. 27, 2003.

See Also

[evalModel-class](#), [rrecsys](#).

Examples

```
x <- matrix(sample(c(NA, 1:5), size = 200, replace = TRUE,  
  prob = c(.6, .8, .8, .8, .8)), nrow = 20, byrow = TRUE)  
  
x <- defineData(x)  
  
e <- evalModel(x, 5)  
  
auc <- getAUC(e, "FunkSVD", k = 4)  
  
auc
```

histogram	<i>Ratings histogram.</i>
-----------	---------------------------

Description

Histogram of the ratings grouped by value.

Usage

```
histogram(data, title = "", x = "Rating values", y = "# of ratings")
```

Arguments

data	class "_ds", the dataset.
title	class "character", eventual caption of for the chart.
x	class "character", label for the x-axis.
y	class "character", label for the y-axis.

IBclass	<i>Item based model.</i>
---------	--------------------------

Description

Container for the model learned using any k-nearest neighbor item-based collaborative filtering algorithm.

Slots

alg: The algorithm denominator, of class "character".
data: the dataset used for training the model, class "matrix".
sim: The item - item similarity matrix, class "matrix".
sim_index_kNN: The index of the k nearest neighbors for each item, class "matrix".
parameters: the parameters used in the model, class "list".

Methods

```
show signature(object = "IBclass")
```

See Also

[rrecsys](#).

ml100k

MovieLens 100K Dataset

Description

MovieLens data sets were collected by the GroupLens Research Project at the University of Minnesota.

This data set consists of:

1. 100,000 ratings (1-5) from 943 users on 1682 movies.
2. Each user has rated at least 20 movies.

The data was collected through the MovieLens web site (movielens.umn.edu) during the seven-month period from September 19th, 1997 through April 22nd, 1998. This data has been cleaned up - users who had less than 20 ratings or did not have complete demographic information were removed from this data set. Detailed descriptions of the data file can be found at the end of this file.

Source

<http://grouplens.org/datasets/movielens/100k/>

mlLatest100k

MovieLens Latest

Description

This dataset (ml-latest-small) is a 5-star rating dataset from [MovieLens](<http://movielens.org>), a movie recommendation service of the GroupLens research group at the University of Minnesota. It contains 100234 ratings across 8927 movies. The data was created by 718 users between March 26, 1996 and August 05, 2015. This dataset was generated on August 06, 2015. Users were selected at random for inclusion. All selected users had rated at least 20 movies. The data is edited and structured as a matrix and distributed as such. Below the usage license of this redistributed data is cited below.

Usage

```
data("mlLatest100k")
```

Format

The format is: num [1:718, 1:8915] 5 3 0 0 4 4 0 3 0 0 ... - attr(*, "dimnames")=List of 2 ..\$: chr [1:718] "1" "2" "3" "4"\$: chr [1:8915] "Toy Story (1995)" "Jumanji (1995)" "GoldenEye (1995)" "Twelve Monkeys (a.k.a. 12 Monkeys) (1995)" ...

Source

<http://grouplens.org/datasets/movielens/latest/>

PPLclass	<i>Popularity based model.</i>
----------	--------------------------------

Description

Container for the model learned by an unpersonalized popularity-based algorithm.

Slots

alg: The algorithm denominator, of class "character".

data: the dataset used for training the model, class "matrix".

indices: the indices of items ordered by popularity, class "integer".

parameters: the parameters used in the model, class "list".

Methods

show signature(object = "PPLclass")

See Also

[rrecsys](#).

predict	<i>Generate predictions.</i>
---------	------------------------------

Description

Generate predictions on any of the previously trained models.

Arguments

model A previously trained model, see [rrecsys](#)

Round object of class "logical", if **TRUE** all the predictions are rounded to integer values, else values are returned as calculated.

Value

All unrated items are predicted and the entire matrix is returned with the new ratings.

See Also

[rrecsys](#), [IBclass](#), [SVDclass](#).

Examples

```

data("mlLatest100k")

smallM1 <- mlLatest100k[1:50, 1:100]

exExp1 <- defineData(smallM1)

model1exp <- rrecsys(exExp1, alg = "funk", k = 10, learningRate = 0.01, regCoef = 0.001)

pre1 <- predict(model1exp, Round = TRUE)

```

rankScore

*Rank Score***Description**

Rank Score extends the recall metric to take the positions of correct items in a ranked list into account.

Usage

```
rankScore(recommendedIDX, testSetIDX, alpha)
```

Arguments

recommendedIDX indices of the recommended items. Object of class `numeric`.
testSetIDX indices of the items in the test set. Object of class `numeric`.
alpha is the ranking half life. Object of class `numeric`.

Details

Rank Score is defined as the ratio of the Rank Score of the correct items to best theoretical Rank Score achievable for the user:

$$rankscore_p = \sum_{i \in h} 2^{-\frac{rank(i)-1}{\alpha}}$$

$$rankscore_{max} = \sum_{i=1}^{|T|} 2^{-\frac{i-1}{\alpha}}$$

$$rankscore = \frac{rankscore_p}{rankscore_{max}}$$

recommend	<i>Generate recommendation.</i>
-----------	---------------------------------

Description

This method generates top-n recommendations based on a model that has been trained before. Two main methods: `recommendHPR`, `recommendMF`. The first method recommends the highest predicted ratings on a user. Instead `recommendMF` (currently available only for IBKNN and UBKNN), recommends the most frequent item in the user's neighborhood.

Usage

```
recommendHPR(model, topN = 3)
recommendMF(model, topN = 3, pt)
```

Arguments

<code>model</code>	the trained model of any algorithm.
<code>topN</code>	number of items to be recommended per user, class <code>numeric</code> .
<code>pt</code>	positive threshold, class <code>numeric</code> .

Value

Returns a list with suggested items for each user.

See Also

[rrecsys](#).

Examples

```
myratings <- matrix(sample(c(0:5), size = 200, replace = TRUE,
  prob = c(.6, .08, .08, .08, .08, .08)), nrow = 20, byrow = TRUE)

myratings <- defineData(myratings)

r <- rrecsys(myratings, alg = "FunkSVD", k = 2)

rec <- recommendHPR(r)
```

rrecsys *Create a recommender system.*

Description

Based on the specific given algorithm a recommendation model will be trained.

Usage

```
rrecsys(data, alg, ...)
```

Arguments

data	Training set of class "matrix". The columns correspond to items and the rows correspond to users.
alg	A "character" string specifying the recommender algorithm to apply on the data.
...	other attributes, see details .

Details

Based on the value of *alg* the attributes will have different names and values. Possible configuration of *alg* and it's meaning:

1. **itemAverage**. When *alg* = "itemAverage" the average rating of an item is used to make predictions and recommendations.
2. **userAverage**. When *alg* = "userAverage" the average rating of a user is used to make predictions and recommendations.
3. **globalAverage**. When *alg* = "globalAverage" the overall average of all ratings is used to make predictions and recommendations.
4. **Mostpopular**. The most popular algorithm (*alg* = "mostpopular") is the most simple algorithm for recommendations. Item will be ordered based on the number of times that they were rated. Recommendations for a particular user will be the most popular items from the data set which are not contained in the user's training set.
5. **IBKNN**. As *alg* = "IBKNN" a k-nearest neighbor item-based collaborative filtering algorithm. Given two items *a* and *b*, we consider them as rating vectors \vec{a} and \vec{b} . If the argument *simFunct* is set to "cos" the method computes the cosine similarity as:

$$sim(\vec{a}, \vec{b}) = cos(\vec{a}, \vec{b}) = \frac{\vec{a} \cdot \vec{b}}{|\vec{a}| * |\vec{b}|}$$

If the argument *simFunct* is set to "adjCos" the method determines the "adjusted cosine" distance among the items as:

$$sim(\vec{a}, \vec{b}) = \frac{\sum_{u \in U} (r_{u,a} - \bar{r}_u) * (r_{u,b} - \bar{r}_u)}{\sqrt{(r_{u,a} - \bar{r}_u)^2} * \sqrt{(r_{u,b} - \bar{r}_u)^2}}$$

It extracts, based on the value of the *neigh* attribute, the number of closest neighbors for each item.

6. **UBKNN**. As `alg = "UBKNN"` a k-nearest neighbor user-based collaborative filtering algorithm. Given two users u and v , we consider them as rating vectors \vec{u} and \vec{v} . If the argument `simFunct` is set to "cos" the method computes the cosine similarity as:

$$\text{sim}(\vec{u}, \vec{v}) = \text{cos}(\vec{u}, \vec{v}) = \frac{\vec{u} \cdot \vec{v}}{|\vec{u}| * |\vec{v}|}$$

If the argument `simFunct` is set to "Pearson" the method determines the "Pearson correlation" among the users as:

$$\text{sim}(\vec{u}, \vec{v}) = \text{Pearson}(\vec{u}, \vec{v}) = \frac{\sum_{i \in I_u \cap I_v} (R_{ui} - \bar{R}_u) * (R_{vi} - \bar{R}_v)}{\sqrt{\sum_{i \in I_u \cap I_v} (R_{ui} - \bar{R}_u)^2 * \sum_{i \in I_u \cap I_v} (R_{vi} - \bar{R}_v)^2}}$$

It extracts, based on the value of the *neigh* attribute, the number of closest neighbors for each item.

7. **FunkSVD**. It implements `alg = "funkSVD"` a stochastic gradient descent optimization technique. The U(user) and V(item) factor matrices are initialized at small values and cropped to k features. Each feature is trained until *convergence* (the convergence value has to be specified by the user, by configuring the `steps` argument). On each loop the algorithm predicts r'_{ui} and calculates the error as:

$$r'_{ui} = u_u * v_i^T$$

$$e_{ui} = r_{ui} - r'_{ui}$$

The factors are updated:

$$v_{ik} \leftarrow v_{ik} + \text{learningRate} * (e_{ui} * u_{uk} - \text{regCoef} * v_{ik})$$

$$u_{uk} \leftarrow u_{uk} + \text{lambda} * (e_{ui} * v_{ik} - \text{gamma} * u_{uk})$$

. The attribute `learningRate` represents the learning rate, while `regCoef` corresponds to the weight of the regularization term. If the argument `biases` is TRUE, the biases will be computed to update the features and generate predictions.

8. **wALS**. The `alg = "wALS"` weighted Alternated Least squares method. For a given non-negative weight matrix W the algorithm will perform updates on the item V and user U feature matrix as follows:

$$U_i = R_i * \widetilde{W}_i * V * (V^T * \widetilde{W}_i * V + \text{lambda} * (\sum_j W_{ij}) I)^{-1}$$

$$V_j = R_j^T * \widetilde{W}_j * U * (V^T * \widetilde{W}_j * U + \text{lambda} * (\sum_i W_{ij}) I)^{-1}$$

Initially the V matrix is initialized with Gaussian random numbers with mean zero and small standard deviation. Than U and V are updated until convergence. The attribute scheme must specify the scheme(`uni`, `uo`, `io`, `co`) to use.

9. **BPR**. In this implementation of BPR (alg = "BPR") is applied a stochastic gradient descent approach that randomly choose triples from D_R and trains the model Θ . In this implementation the BPR optimization criterion is applied on matrix factorization. If $R = U \times V^T$, where U and V are the usual feature matrix cropped to k features, the parameter vector of the model is $\Theta = \langle U, V \rangle$. The Boolean randomInit parameter determines whatever the feature matrix are initialized to a random value or at a static 0.1 value. The algorithm will use three regularization terms, RegU for the user features U , RegI for positive updates and RegJ for negative updates of the item features V , lambda is the learning rate, autoConvergence is a toggle to the auto convergence validation, convergence upper limit to the convergence, and updateJ if true updates negative item features.
10. **SlopeOne** The Weighted Slope One (alg = "slopeOne") performs prediction for a missing rating \hat{r}_{ui} for user u on item i as the following average:

$$\hat{r}_{ui} = \frac{\sum_{\forall r_{uj}} (dev_{ij} + r_{uj}) c_{ij}}{\sum_{\forall r_{uj}} c_{ij}}.$$

The average deviation rating dev_{ij} between co-rated items is defined by:

$$dev_{ij} = \sum_{\forall u \in users} \frac{r_{ui} - r_{uj}}{c_{ij}}.$$

Where c_{ij} is the number of co-ratings between items i and j and r_{ui} is an existing rating for user u on item i . The Weighted Slope One takes into account both, information from users who rated the same item and the number of observed ratings.

To view a full list of available algorithms and their default configuration execute `rrecsysRegistry`.

Value

Depending on the alg value it will be either an object of type `SVDclass` or `IBclass`.

References

- D. Jannach, M. Zanker, A. Felfernig, and G. Friedrich. *Recommender Systems: An Introduction*. Cambridge University Press, New York, NY, USA, 1st edition, 2010. ISBN 978-0-521-49336-9.
- Funk, S., 2006, *Netflix Update: Try This at Home*, <http://sifter.org/~simon/journal/20061211.html>.
- Y. Koren, R. Bell, and C. Volinsky. *Matrix Factorization Techniques for Recommender Systems*. *Computer*, 42(8):30–37, Aug. 2009. ISSN 0018-9162. doi: 10.1109/MC.2009.263. <http://dx.doi.org/10.1109/MC.2009.263>.
- R. Pan, Y. Zhou, B. Cao, N. Liu, R. Lukose, M. Scholz, and Q. Yang. *One-Class Collaborative Filtering*. In *Data Mining, 2008. ICDM '08. Eighth IEEE International Conference on*, pages 502–511, Dec 2008. doi: 10.1109/ICDM.2008.16.
- S. Rendle, C. Freudenthaler, Z. Gantner, and L. Schmidt-Thieme. *BPR: Bayesian Personalized Ranking from Implicit Feedback*. In *Proceedings of the Twenty-Fifth Conference on Uncertainty in Artificial Intelligence, UAI '09*, pages 452–461, Arlington, Virginia, United States, 2009. AUAI Press. ISBN 978-0-9749039-5-8. URL <http://dl.acm.org/citation.cfm?id=1795114.1795167>.

Examples

```
myratings <- matrix(sample(c(0:5), size = 200, replace = TRUE,
  prob = c(.6, .08, .08, .08, .08, .08)), nrow = 20, byrow = TRUE)

myratings <- defineData(myratings)

r <- rrecsys(myratings, alg = "funktSVD", k = 2)

r2 <- rrecsys(myratings, alg = "IBKNN", simFunct = "cos", neigh = 5)

rrecsysRegistry$get_entries()
```

setStoppingCriteria *Set stopping criteria.*

Description

Define stopping criteria for functions that need a convergence check.

Usage

```
setStoppingCriteria(autoConverge = FALSE,
  deltaErrorThreshold = 1e-05, nrLoops = NULL, minNrLoops = 10)
showStoppingCriteria()
showDeltaError()
```

Arguments

autoConverge	class "logical", turns on the auto-convergence algorithm.
deltaErrorThreshold	class "numeric", is the threshold for the auto-convergence algorithm.
nrLoops	class "numeric", number of loops that will be performed in case autoConvergence is FALSE
minNrLoops	class "numeric", the minimum number of loops to consider before verifying the deltaErrorThreshold.

Details

If autoConvergence = TRUE tells the package to monitor the difference of global RMSE on two consecutive iterations, and to see if it drops below a threshold value. Whenever it drops under the specified value the iteration is considered converged. If FALSE the limit of iterations is delimited by nrLoops

Methods

`showStoppingCriteria` Print on console the current configuration of the convergence algorithm.

`showDeltaError` Report the delta error on each iteration of the algorithm that requires an auto-convergence algorithm.

References

M. D. Ekstrand, M. Ludwig, J. Kolb, and J. T. Riedl, “*LensKit: a modular recommender framework*”, Proc. fifth ACM Conf. Recomm. Syst. - RecSys '11, p. 349, 2011.

See Also

See Also as [rrecsys](#), [SVDclass](#), [wALSclass](#), [BPRclass](#).

Examples

```
setStoppingCriteria(autoConverge = TRUE)
```

```
setStoppingCriteria(nrLoops = 30)
```

slopeOneClass	<i>Slope One model.</i>
---------------	-------------------------

Description

Container for the model learned using Slope One algorithm.

Slots

`alg`: The algorithm denominator, of class "character".

`data`: the dataset used for training the model, class "matrix".

`devcard`: Deviation and Cardinality between columns, class "list".

Methods

`show` signature(object = "SVDclass")

See Also

[rrecsys](#).

sparseDataSet-class *Dataset class for tuples (user, item, rating).*

Description

Container for a sparse dataset that distinguishes between binary and non-binary feedback datasets. Data are stored as tuples (user, item, rating). Extends `_ds`.

Slots

data: the dataset, class "matrix".

binary: class "logical", determines if the item dataset contains binary (i.e. 1/0) or non-binary ratings.

minimum: class "numeric", defines the minimal value present in the dataset.

maximum: class "numeric", defines the maximal value present in the dataset.

intScale: object of class "logical", if **TRUE** the range of ratings in the dataset contains as well half star values.

userID: class "numeric", array containing all user IDs.

itemID: class "numeric", array containing all item IDs.

userPointers: class "list", pointer to all users position in the dataset.

itemPointers: class "list", pointer to all items position in the dataset.

Methods

nrow signature(object = "sparseDataSet"): number of rows of the dataset.

ncol signature(object = "sparseDataSet"): number of columns of the dataset.

dim signature(object = "sparseDataSet"): returns the dimensions of the dataset.

rowRatings signature(object = "sparseDataSet"): returns the number of ratings on each row.

colRatings signature(object = "sparseDataSet"): returns the number of ratings on each column.

numRatings signature(object = "sparseDataSet"): returns the total number of ratings.

[signature(x = "sparseDataSet", i = "ANY", j = "ANY", drop = "ANY"): returns a subset of the dataset.

coerce signature(from = "sparseDataSet", to = "matrix")

rowAverages signature(object = "sparseDataSet"): returns the average rating on each row.

colAverages signature(object = "sparseDataSet"): returns the average rating on each column.

SVDclass

SVD model.

Description

Container for the model learned using any matrix factorization algorithm.

Slots

alg: The algorithm denominator, of class "character".

data: the dataset used for training the model, class "matrix".

factors: user(U) and items(V) factors, class "list".

parameters: the parameters used in the model, class "list".

baselines: Global, user and item baselines, class "list".

Methods

show signature(object = "SVDclass")

See Also

[rrecsys.](#)

UBclass

Item based model.

Description

Container for the model learned using any k-nearest neighbor item-based collaborative filtering algorithm.

Slots

alg: The algorithm denominator, of class "character".

data: the dataset used for training the model, class "matrix".

sim: The item - item similarity matrix, class "matrix".

sim_index_kNN: The index of the k nearest neighbors for each item, class "matrix".

parameters: the parameters used in the model, class "list".

Methods

show signature(object = "UBclass")

See Also

[rrecsys.](#)

`wALSclass`*Weighted Alternating Least Squares based model.*

Description

Container for the model learned using any weighted Alternating Least Squares based algorithm.

Slots

alg: The algorithm denominator, of class "character".

data: the dataset used for training the model, class "matrix".

factors: user(U) and items(V) factors, class "list".

weightScheme: The weighting scheme used in updating the factors, class "matrix".

parameters: the parameters(such as number of factors k, learning rate lambda, number of iterations until convergence and the weighting scheme) used in the model, class "list".

Methods

`show signature(object = "wALSclass")`

See Also

[rrecsys](#).

`_ds-class`*Dataset class.*

Description

Defines a structure for a dataset that distinguishes between binary and non-binary feedback datasets.

Slots

binary: class "logical", determines if the item dataset contains binary (i.e. 1/0) or non-binary ratings.

minimum: class "numeric", defines the minimal value present in the dataset.

maximum: class "numeric", defines the maximal value present in the dataset.

intScale: object of class "logical", if **TRUE** the range of ratings in the dataset contains as well half star values.

Methods

`show signature(object = "_ds")`

sparsity `signature(object = "_ds")`: returns the sparsity of the dataset.

summary `signature(object = "_ds")`: summary of the characteristics of the dataset.

Index

- * **datasets**
 - ml100k, [15](#)
 - mlLatest100k, [15](#)
- [, dataSet, ANY, ANY, missing-method (dataSet-class), [4](#)
- [, sparseDataSet, ANY, ANY, missing-method (sparseDataSet-class), [24](#)
- _ds, [4](#), [7](#), [24](#)
- _ds (_ds-class), [26](#)
- _ds-class, [26](#)
- algAverageClass, [2](#)
- algAverageClass-class (algAverageClass), [2](#)
- BPRclass, [3](#), [23](#)
- BPRclass-class (BPRclass), [3](#)
- coerce, dataSet, matrix-method (dataSet-class), [4](#)
- coerce, sparseDataSet, matrix-method (sparseDataSet-class), [24](#)
- colAverages (_ds-class), [26](#)
- colAverages, dataSet-method (dataSet-class), [4](#)
- colAverages, sparseDataSet-method (sparseDataSet-class), [24](#)
- colRatings (_ds-class), [26](#)
- colRatings, dataSet-method (dataSet-class), [4](#)
- colRatings, sparseDataSet-method (sparseDataSet-class), [24](#)
- dataChart, [3](#)
- dataSet (dataSet-class), [4](#)
- dataSet-class, [4](#)
- defineData, [5](#)
- defineData, matrix-method (defineData), [5](#)
- details, [19](#)
- dim, dataSet-method (dataSet-class), [4](#)
- dim, sparseDataSet-method (sparseDataSet-class), [24](#)
- eval_nDCG, [12](#)
- evalChart, [6](#)
- evalModel, [7](#)
- evalModel, _ds-method (evalModel), [7](#)
- evalModel, dataSet-method (evalModel), [7](#)
- evalModel, sparseDataSet-method (evalModel), [7](#)
- evalModel-class, [8](#)
- evalPred, [9](#)
- evalPred, evalModel, list-method (evalPred), [9](#)
- evalPred, evalModel-method (evalPred), [9](#)
- evalRec, [7](#), [10](#)
- evalRec, evalModel, list-method (evalRec), [10](#)
- evalRec, evalModel-method (evalRec), [10](#)
- evalRecResults, [11](#)
- evalRecResults-class (evalRecResults), [11](#)
- getAUC, [13](#)
- getAUC, evalModel (getAUC), [13](#)
- getAUC, evalModel-method (getAUC), [13](#)
- histogram, [14](#)
- IBclass, [14](#), [16](#), [21](#)
- IBclass-class (IBclass), [14](#)
- ml100k, [15](#)
- ml100k_array (ml100k), [15](#)
- mlLatest100k, [15](#)
- ncol, dataSet-method (dataSet-class), [4](#)
- ncol, sparseDataSet-method (sparseDataSet-class), [24](#)
- nrow, dataSet-method (dataSet-class), [4](#)

- nrow, sparseDataSet-method
(sparseDataSet-class), 24
- numRatings (_ds-class), 26
- numRatings, dataSet-method
(dataSet-class), 4
- numRatings, sparseDataSet-method
(sparseDataSet-class), 24

- PPLclass, 16
- PPLclass-class (PPLclass), 16
- predict, 16
- predict, slopeOneClass (predict), 16
- predict, algAverageClass-method
(predict), 16
- predict, BPRclass-method (predict), 16
- predict, IBclass-method (predict), 16
- predict, slopeOneClass-method
(slopeOneClass), 23
- predict, SVDclass-method (predict), 16
- predict, UBclass-method (predict), 16
- predict, wALSclass-method (predict), 16

- rankScore, 17
- recommend, 18
- recommendHPR (recommend), 18
- recommendMF (recommend), 18
- results (evalRecResults), 11
- results, evalRecResults-method
(evalRecResults), 11
- rowAverages (_ds-class), 26
- rowAverages, dataSet-method
(dataSet-class), 4
- rowAverages, sparseDataSet-method
(sparseDataSet-class), 24
- rowRatings (_ds-class), 26
- rowRatings, dataSet-method
(dataSet-class), 4
- rowRatings, sparseDataSet-method
(sparseDataSet-class), 24
- rrecsys, 3, 9, 10, 13, 14, 16, 18, 19, 23, 25, 26
- rrecsys, _ds-method (rrecsys), 19
- rrecsysRegistry (rrecsys), 19

- setStoppingCriteria, 22
- show, _ds-method (_ds-class), 26
- show, algAverageClass-method
(algAverageClass), 2
- show, BPRclass-method (BPRclass), 3
- show, evalModel-method
(evalModel-class), 8
- show, evalRecResults-method
(evalRecResults), 11
- show, IBclass-method (IBclass), 14
- show, PPLclass-method (PPLclass), 16
- show, slopeOneClass-method
(slopeOneClass), 23
- show, SVDclass-method (SVDclass), 25
- show, UBclass-method (UBclass), 25
- show, wALSclass-method (wALSclass), 26
- showDeltaError (setStoppingCriteria), 22
- showStoppingCriteria
(setStoppingCriteria), 22
- slopeOneClass, 23
- slopeOneClass-class (slopeOneClass), 23
- sparseDataSet (sparseDataSet-class), 24
- sparseDataSet-class, 24
- sparsity (_ds-class), 26
- sparsity, _ds-method (_ds-class), 26
- summary, _ds-method (_ds-class), 26
- SVDclass, 16, 21, 23, 25
- SVDclass-class (SVDclass), 25

- UBclass, 25
- UBclass-class (UBclass), 25

- wALSclass, 23, 26
- wALSclass-class (wALSclass), 26