Package: fastadi (via r-universe)

October 11, 2024

```
Type Package
Title Self-Tuning Data Adaptive Matrix Imputation
Version 0.1.1
Description Implements the AdaptiveImpute matrix completion algorithm
     of 'Intelligent Initialization and Adaptive Thresholding for
     Iterative Matrix Completion',
     <https://amstat.tandfonline.com/doi/abs/10.1080/10618600.2018.1518238>.
     AdaptiveImpute is useful for embedding sparsely observed
     matrices, often out performs competing matrix completion
     algorithms, and self-tunes its hyperparameter, making usage
     easy.
License MIT + file LICENSE
URL https://github.com/RoheLab/fastadi
BugReports https://github.com/RoheLab/fastadi/issues
Depends LRMF3, Matrix, R (>= 3.1)
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adaptive_imputation

Create an Adaptive Imputation object

Description

adaptive_imputation objects are a subclass of LRMF3::svd_like(), with an additional field alpha.

Usage

```
adaptive_imputation(u, d, v, alpha, ...)
```

Arguments

u	A matrix "left singular-ish" vectors.
d	A vector of "singular-ish" values.
v	A matrix of "right singular-ish" vectors.
alpha	Value of alpha after final iteration.
	Optional additional items to pass to the constructor.

Value

An adaptive_imputation object.

adaptive_impute	<i>AdaptiveImpute</i>
adaptive_impute	Adaptıveli

Description

An implementation of the AdaptiveImpute algorithm for matrix completion for sparse matrices.

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Usage

```
adaptive_impute(
  Χ,
  rank,
  initialization = c("svd", "adaptive-initialize", "approximate"),
 max_iter = 200L,
  check_interval = 1L,
  epsilon = 1e-07,
  additional = NULL
)
## S3 method for class 'sparseMatrix'
adaptive_impute(
 Χ,
 rank,
  initialization = c("svd", "adaptive-initialize", "approximate"),
  additional = NULL
)
## S3 method for class 'LRMF'
adaptive_impute(
 Χ,
  rank,
  epsilon = 1e-07,
 max_iter = 200L,
  check_interval = 1L
)
```

Arguments

X A sparse matrix of Matrix::sparseMatrix() class.

rank

Desired rank (integer) to use in the low rank approximation. Must be at least 2L and at most the rank of X. Note that the rank of X is typically unobserved and computations may be unstable or even fail when rank is near or exceeds this threshold.

... Unused additional arguments.

initialization How to initialize the low rank approximation. Options are:

- "svd" (default). In the initialization step, this treats unobserved values as zeroes.
- "adaptive-initialize". In the initialization step, this treats unobserved values as actually unobserved. However, the current AdaptiveInitialize implementation relies on dense matrix computations that are only suitable for relatively small matrices.

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> • "approximate". An approximate variant of AdaptiveInitialize that is less computationally expensive. See adaptive_initialize for details.

Note that initialization matters as AdaptiveImpute optimizes a non-convex objective. The current theory shows that initializing with AdaptiveInitialize leads to a consistent estimator, but it isn't know if this is the case for SVD initialization. Empirically we have found that SVD initialization works well nonetheless.

max_iter

Maximum number of iterations to perform (integer). Defaults to 200L. In practice 10 or so iterations will get you a decent approximation to use in exploratory analysis, and and 50-100 will get you most of the way to convergence. Must be at least 1L.

check_interval Integer specifying how often to perform convergence checks. Defaults to 1L. In practice, check for convergence requires a norm calculation that is expensive for large matrices and decreasing the frequency of convergence checks will reduce computation time. Can also be set to NULL, which case max_iter iterations of the algorithm will occur with no possibility of stopping due to small relative change in the imputed matrix. In this case delta will be reported as Inf.

epsilon

Convergence criteria, measured in terms of relative change in Frobenius norm of the full imputed matrix. Defaults to 1e-7.

additional

Ignored except when alpha_method = "approximate" in which case it controls the precise of the approximation to alpha. The approximate computation of alpha will always understand alpha, but the approximation will be better for larger values of additional. We recommend making additional as large as computationally tolerable.

Value

A low rank matrix factorization represented by an adaptive_imputation() object.

References

- 1. Cho, Juhee, Donggyu Kim, and Karl Rohe. "Asymptotic Theory for Estimating the Singular Vectors and Values of a Partially-Observed Low Rank Matrix with Noise." Statistica Sinica, 2018. https://doi.org/10.5705/ss.202016.0205.
- —. "Intelligent Initialization and Adaptive Thresholding for Iterative Matrix Completion: Some Statistical and Algorithmic Theory for Adaptive-Impute." Journal of Computational and Graphical Statistics 28, no. 2 (April 3, 2019): 323-33. https://doi.org/10.1080/10618600.2018.1518238.

Examples

```
mf <- adaptive_impute(ml100k, rank = 3L, max_iter = 5L, check_interval = NULL)</pre>
```

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Description

An implementation of the AdaptiveInitialize algorithm for matrix imputation for sparse matrices. At the moment the implementation is only suitable for small matrices with on the order of thousands of rows and columns at most.

Usage

```
adaptive_initialize(
   X,
   rank,
   ...,
   p_hat = NULL,
   alpha_method = c("exact", "approximate"),
   additional = NULL
)

## S3 method for class 'sparseMatrix'
adaptive_initialize(
   X,
   rank,
   ...,
   p_hat = NULL,
   alpha_method = c("exact", "approximate"),
   additional = NULL
)
```

Arguments

X A sparse matrix of sparseMatrix class. Explicit (observed) zeroes in X can be

dropped for

rank Desired rank (integer) to use in the low rank approximation. Must be at least 2L

and at most the rank of X.

... Ignored.

p_hat The portion of X that is observed. Defaults to NULL, in which case p_hat is set to

the number of observed elements of X. Primarily for internal use in citation_impute()

or advanced users.

alpha_method Either "exact" or "approximate", defaulting to "exact". "exact" is com-

putationally expensive and requires taking a complete SVD of matrix of size nrow(X) x nrow(X), and matches the AdaptiveInitialize algorithm exactly. "approximate" departs from the AdaptiveInitialization algorithm to compute a truncated SVD of rank rank + additional instead of a complete SVD. This reduces computational burden, but the resulting estimates of singular-ish values will not be penalized as much as in the AdaptiveInitialize algorithm.

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additional

Ignored except when alpha_method = "approximate" in which case it controls the precise of the approximation to alpha. The approximate computation of alpha will always understand alpha, but the approximation will be better for larger values of additional. We recommend making additional as large as computationally tolerable.

Value

A low rank matrix factorization represented by an adaptive_imputation() object.

Examples

```
mf <- adaptive_initialize(
  ml100k,
  rank = 3,
  alpha_method = "approximate",
  additional = 2
)
mf</pre>
```

citation_impute

CitationImpute

Description

An implementation of the AdaptiveImpute algorithm using efficient sparse matrix computations, specialized for the case when missing values in the upper triangle are taken to be *explicitly observed* zeros, as opposed to missing values. This is primarily useful for spectral decompositions of adjacency matrices of graphs with (near) tree structure, such as citation networks.

Usage

```
citation_impute(
   X,
   rank,
   ...,
   initialization = c("svd", "adaptive-initialize", "approximate"),
   max_iter = 200L,
   check_interval = 1L,
   epsilon = 1e-07,
   additional = NULL
)

## S3 method for class 'sparseMatrix'
citation_impute(
   X,
```

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```
rank,
  initialization = c("svd", "adaptive-initialize", "approximate"),
  additional = NULL
)
## S3 method for class 'LRMF'
citation_impute(
  Χ,
  rank,
  . . . ,
  epsilon = 1e-07,
 max_iter = 200L,
  check_interval = 1L
)
```

Arguments

Χ

A square sparse matrix of Matrix::sparseMatrix() class. Implicit zeros in the upper triangle of this matrix are considered observed and predictions on these elements contribute to the objective function minimized by AdaptiveImpute.

rank

Desired rank (integer) to use in the low rank approximation. Must be at least 2L and at most the rank of X. Note that the rank of X is typically unobserved and computations may be unstable or even fail when rank is near or exceeds this threshold.

Unused additional arguments.

initialization How to initialize the low rank approximation. Options are:

- "svd" (default). In the initialization step, this treats unobserved values as
- "adaptive-initialize". In the initialization step, this treats unobserved values as actually unobserved. However, the current AdaptiveInitialize implementation relies on dense matrix computations that are only suitable for relatively small matrices.
- "approximate". An approximate variant of AdaptiveInitialize that is less computationally expensive. See adaptive_initialize for details.

Note that initialization matters as AdaptiveImpute optimizes a non-convex objective. The current theory shows that initializing with AdaptiveInitialize leads to a consistent estimator, but it isn't know if this is the case for SVD initialization. Empirically we have found that SVD initialization works well nonetheless.

max_iter

Maximum number of iterations to perform (integer). Defaults to 200L. In practice 10 or so iterations will get you a decent approximation to use in exploratory analysis, and and 50-100 will get you most of the way to convergence. Must be at least 1L.

check_interval Integer specifying how often to perform convergence checks. Defaults to 1L. In practice, check for convergence requires a norm calculation that is expensive for

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large matrices and decreasing the frequency of convergence checks will reduce computation time. Can also be set to NULL, which case max_iter iterations of the algorithm will occur with no possibility of stopping due to small relative change in the imputed matrix. In this case delta will be reported as Inf.

epsilon

Convergence criteria, measured in terms of relative change in Frobenius norm of the full imputed matrix. Defaults to 1e-7.

additional

Ignored except when alpha_method = "approximate" in which case it controls the precise of the approximation to alpha. The approximate computation of alpha will always understand alpha, but the approximation will be better for larger values of additional. We recommend making additional as large as computationally tolerable.

Details

If OpenMP is available, citation_impute will automatically use getOption("Ncpus", 1L) OpenMP threads to parallelize some key computations. Note that some computations are performed with the Armadillo C++ linear algebra library and may also be parallelized dependent on your BLAS and LAPACK installations and configurations.

Value

A low rank matrix factorization represented by an adaptive_imputation() object.

Examples

```
# create a (binary) square sparse matrix to demonstrate on
set.seed(887)

n <- 10
A <- rsparsematrix(n, n, 0.1, rand.x = NULL)

mf <- citation_impute(A, rank = 3L, max_iter = 1L, check_interval = NULL)
mf</pre>
```

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