# Package 'BiocNeighbors'

April 6, 2025

Version 2.1.3

Date 2025-03-04

Title Nearest Neighbor Detection for Bioconductor Packages

Imports Rcpp, methods

Suggests BiocParallel, testthat, BiocStyle, knitr, rmarkdown

biocViews Clustering, Classification

Description Implements exact and approximate methods for nearest neighbor detection, in a framework that allows them to be easily switched within Bioconductor packages or workflows. Exact searches can be performed using the k-means for k-nearest neighbors algorithm or with vantage point trees. Approximate searches can be performed using the Annoy or HNSW libraries. Searching on either Euclidean or Manhattan distances is supported. Parallelization is achieved for all methods by using BiocParallel. Functions are also provided to search for all neighbors within a given distance.

License GPL-3

LinkingTo Rcpp, assorthead

VignetteBuilder knitr

**SystemRequirements** C++17

RoxygenNote 7.3.2

**Encoding UTF-8** 

git\_url https://git.bioconductor.org/packages/BiocNeighbors

git\_branch devel

git\_last\_commit 25e7813

git\_last\_commit\_date 2025-03-05

**Repository** Bioconductor 3.21

Date/Publication 2025-04-06

Author Aaron Lun [aut, cre, cph]

Maintainer Aaron Lun <infinite.monkeys.with.keyboards@gmail.com>

# **Contents**

Bioc	Neighbors-package	BiocNei ages	ghbors	s: Nea	rest N	leighba	or Dete	ction for	Bioconduc	tor Pack-	-
Index											26
	VptreeParam										24
	queryNeighbors										22
	queryKNN										
	queryDistance										18
	KmknnParam										17
	HnswParam										
	findNeighbors										
	findMutualNN										
	findKNN										
	findDistance										_
	ExhaustiveParam .										8
	defineBuilder										
	BiocNeighborParam buildIndex										5 6
	BiocNeighborIndex										
	AnnoyParam										3
	BiocNeighbors-pack										2

# **Description**

Implements exact and approximate methods for nearest neighbor detection, in a framework that allows them to be easily switched within Bioconductor packages or workflows. Exact searches can be performed using the k-means for k-nearest neighbors algorithm or with vantage point trees. Approximate searches can be performed using the Annoy or HNSW libraries. Searching on either Euclidean or Manhattan distances is supported. Parallelization is achieved for all methods by using BiocParallel. Functions are also provided to search for all neighbors within a given distance.

# Author(s)

Maintainer: Aaron Lun <infinite.monkeys.with.keyboards@gmail.com> [copyright holder]

AnnoyParam 3

AnnoyParam	The AnnoyParam class	

## **Description**

A class to hold parameters for the Annoy algorithm for approximate nearest neighbor identification.

## Usage

```
AnnoyParam(
  ntrees = 50,
  search.mult = ntrees,
  distance = c("Euclidean", "Manhattan", "Cosine")
)
## S4 method for signature 'AnnoyParam'
defineBuilder(BNPARAM)
```

## **Arguments**

ntrees Integer scalar, number of trees to use for index generation.

search.mult Numeric scalar, multiplier for the number of points to search.

distance String specifying the distance metric to use. Cosine distances are implemented

as Euclidean distances on L2-normalized coordinates.

BNPARAM An AnnoyParam instance.

# Details

The Approximate nearest neighbors Oh Yeah (Annoy) algorithm is based on recursive hyperplane partitions. Briefly, a tree is constructed where a random hyperplane splits the points into two subsets at each internal node. Leaf nodes are defined when the number of points in a subset falls below a threshold (close to twice the number of dimensions for the settings used here). Multiple trees are constructed in this manner, each of which is different due to the random choice of hyperplanes. For a given query point, each tree is searched to identify the subset of all points in the same leaf node as the query point. The union of these subsets across all trees is exhaustively searched to identify the actual nearest neighbors to the query.

The ntrees parameter controls the trade-off between accuracy and computational work. More trees provide greater accuracy at the cost of more computational work (both in terms of the indexing time and search speed in downstream functions).

The search.mult controls the parameter known as search\_k in the original Annoy documentation. Specifically, search\_k is defined as k \* search.mult where k is the number of nearest neighbors to identify in downstream functions. This represents the number of points to search exhaustively and determines the run-time balance between speed and accuracy. The default search.mult=ntrees is based on the Annoy library defaults. Note that this parameter is not actually used in the index construction itself, and is only included here so that the output index fully parametrizes the search.

4 BiocNeighborIndex

Technically, the index construction algorithm is stochastic but, for various logistical reasons, the seed is hard-coded into the C++ code. This means that the results of the Annoy neighbor searches will be fully deterministic for the same inputs, even though the theory provides no such guarantees.

## Value

The AnnoyParam constructor returns an instance of the AnnoyParam class.

The defineBuilder method returns a list that can be used in buildIndex to construct an Annoy index

# Author(s)

Aaron Lun

## See Also

BiocNeighborParam, for the parent class and its available methods.

https://github.com/spotify/annoy, for details on the underlying algorithm.

# **Examples**

```
(out <- AnnoyParam())
out[['ntrees']]
out[['ntrees']] <- 20L
out</pre>
```

BiocNeighborIndex

The BiocNeighborIndex class

## **Description**

A virtual class for indexing structures of different nearest-neighbor search algorithms. Developers should define subclasses for their own buildIndex and/or defineBuilder methods.

## **Details**

In general, the internal structure of a BiocNeighborIndex class is arbitrary and left to the discretion of the developer. If an arbitrary structure is used, the associated methods should be written for all downstream generics like findKNN, etc.

Alternatively, developers may choose to derive from the BiocNeighborGenericIndex class. This expects:

- A ptr slot containing an external pointer that refers to a BiocNeighbors::Prebuilt object (see definition in system. file("include", "BiocNeighbors.h", package="BiocNeighbors")).
- A names slot containing a character vector with the names of the observations, or NULL if no names are available. This is used by subset= in the various find\* generics.

In this case, no additional methods are required for the downstream generics.

BiocNeighborParam 5

## Author(s)

Aaron Lun

BiocNeighborParam

The BiocNeighborParam class

# **Description**

A virtual class for specifying the type of nearest-neighbor search algorithm and associated parameters.

## **Details**

The BiocNeighborParam class is a virtual base class on which other parameter objects are built. There are currently 5 concrete subclasses in **BiocNeighbors**:

KmknnParam: Exact nearest-neighbor search with the KMKNN algorithm.

VptreeParam: Exact nearest-neighbor search with the tree algorithm.

ExhaustiveParam: Exact nearest-neighbor search via brute-force.

AnnoyParam: Approximate nearest-neighbor search with the Annoy algorithm.

HnswParam: Approximate nearest-neighbor search with the HNSW algorithm.

These objects hold parameters specifying how each algorithm should be run on an arbitrary data set. See the associated documentation pages for more details.

## Methods

In the following code snippets, x and object are BiocNeighborParam objects.

show(object): Display the class and arguments of object.

bndistance(object): Return a string specifying the distance metric to be used for searching. This should be one of "Euclidean", "Manhattan" or "Cosine".

x[[i]]: Return the value of slot i, as used in the constructor for x.

x[[i]] <- value: Set slot i to the specified value.

# Author(s)

Aaron Lun

## See Also

 ${\it KmknnParam, VptreeParam, AnnoyParam, and HnswParam for constructors.}$ 

buildIndex, findKNN and queryKNN for dispatch.

6 buildIndex

buildIndex	Build a nearest-neighbor index	

## **Description**

Build indices for nearest-neighbor searching with different algorithms.

# Usage

```
buildIndex(X, transposed = FALSE, ..., BNPARAM = NULL)
```

# **Arguments**

_	
X	A numeric matrix where rows correspond to data points and columns correspond to variables (i.e., dimensions).
transposed	Logical scalar indicating whether X is transposed, i.e., rows are variables and columns are data points.
	Further arguments to be passed to individual methods.
BNPARAM	A BiocNeighborParam object specifying the type of index to be constructed. If NULL, this defaults to a KmknnParam object.

Alternatively, this may be a list returned by defineBuilder.

## **Details**

Each buildIndex method is expected to return an instance of a BiocNeighborIndex subclass. The structure of this subclass is arbitrary and left to the discretion of the method developer. Developers are also responsible for defining methods for their subclass in each of the relevant functions (e.g., findKNN, queryKNN). The exception is if the method returns an instance of a BiocNeighborGenericIndex subclass, which can be used with the existing methods for findKNN, etc. without further effort.

# Value

A BiocNeighborIndex object can be used in findKNN and related functions as the X= argument. Users should assume that the index is not serializable, i.e., cannot be saved or transferred between processes.

## Author(s)

Aaron Lun

# **Examples**

```
Y <- matrix(rnorm(100000), ncol=20)
(k.out <- buildIndex(Y))
(a.out <- buildIndex(Y, BNPARAM=AnnoyParam()))</pre>
```

defineBuilder 7

defineBuilder

Define an index builder

## **Description**

Define a builder object that can construct C++ indices for neighbor searches.

## **Usage**

defineBuilder(BNPARAM)

## **Arguments**

**BNPARAM** 

A BiocNeighborParam object specifying the type of index to be constructed. If NULL, this defaults to a KmknnParam object.

## **Details**

The external pointer returned in builder should refer to a BiocNeighbors::Builder object, see the definition in system.file("include", "BiocNeighbors.h", package="BiocNeighbors") for details. If a developer defines a defineBuilder method for a search algorithm, they do not have to define a new buildIndex method. The existing buildIndex methods will automatically create an instance of the appropriate BiocNeighborGenericIndex subclass based on class, which can be immediately used in all generics (e.g., findKNN, queryNeighbors) without further effort.

Note that the pointer returned by defineBuilder should *not* be used as the ptr in the BiocNeighborIndex subclasses. The ptr slot is expected to contain a pointer referring to a BiocNeighbors::Prebuilt object, as returned by the default buildIndex. Using the pointer from builder will probably crash the R session.

Needless to say, users should not attempt to serialize the external pointer returned by this generic. Attempting to use a deserialized pointer in buildIndex will cause the R session to crash.

#### Value

List containing:

- builder, a pointer to a builder instance that can be used to construct a prebuilt index in buildIndex.
- class, the constructor for a BiocNeighborGenericIndex subclass that accepts ptr and names arguments.

## Author(s)

Aaron Lun

#### See Also

defineBuilder, KmknnParam-method, defineBuilder, VptreeParam-method, defineBuilder, AnnoyParam-method and defineBuilder, HnswParam-method for specific methods.

8 ExhaustiveParam

## **Examples**

```
(out <- defineBuilder())
(out2 <- defineBuilder(AnnoyParam()))</pre>
```

ExhaustiveParam

The ExhaustiveParam class

# **Description**

A class to hold parameters for the exhaustive algorithm for exact nearest neighbor identification.

# Usage

```
ExhaustiveParam(distance = c("Euclidean", "Manhattan", "Cosine"))
## S4 method for signature 'ExhaustiveParam'
defineBuilder(BNPARAM)
```

## **Arguments**

distance String specifying the distance metric to use. Cosine distances are implemented

as Euclidean distances on L2-normalized coordinates.

BNPARAM An ExhaustiveParam instance.

## **Details**

The exhaustive search computes all pairwise distances between data and query points to identify nearest neighbors of the latter. It has quadratic complexity and is theoretically the worst-performing method; however, it has effectively no overhead from constructing or querying indexing structures, making it faster for in situations where indexing provides little benefit. This includes queries against datasets with few data points or very high dimensionality.

All that said, this algorithm is largely provided as a baseline for comparing against the other algorithms.

# Value

The ExhaustiveParam constructor returns an instance of the ExhaustiveParam class.

The defineBuilder method returns an external pointer that can be used in buildIndex to construct an exhaustive index.

## Author(s)

Allison Vuong

## See Also

BiocNeighborParam, for the parent class and its available methods.

findDistance 9

## **Examples**

```
(out <- ExhaustiveParam())</pre>
```

findDistance

Distance to the k-th nearest neighbor

## **Description**

Find the distance to the k-th nearest neighbor for each point in a dataset.

## Usage

```
findDistance(X, k, num.threads = 1, subset = NULL, ..., BNPARAM = NULL)
```

## **Arguments**

Χ	A numeric matrix where rows correspond to data points and columns correspond
	to variables (i.e., dimensions). Alternatively, a prebuilt BiocNeighborIndex ob-

ject from buildIndex.

k A positive integer scalar specifying the number of nearest neighbors to retrieve.

Alternatively, an integer vector of length equal to the number of points in X, specifying the number of neighbors to identify for each point. If subset is provided, this should have length equal to the length of subset. Users should wrap this vector in an AsIs class to distinguish length-1 vectors from integer scalars.

All k should be less than or equal to the number of points in X minus 1, otherwise

the former will be capped at the latter with a warning.

num. threads Integer scalar specifying the number of threads to use for the search.

subset An integer, logical or character vector specifying the indices of points in X for

which the nearest neighbors should be identified. This yields the same result as (but is more efficient than) subsetting the output matrices after computing

neighbors for all points.

Further arguments to pass to buildIndex when X is not an external pointer.

BNPARAM A BiocNeighborParam object specifying how the index should be constructed.

If NULL, this defaults to a KmknnParam. Ignored if x contains a prebuilt index.

## **Details**

If multiple queries are to be performed to the same X, it may be beneficial to build the index from X with buildIndex. The resulting pointer object can be supplied as X to multiple findDistance calls, avoiding the need to repeat index construction in each call.

10 findKNN

# Value

Numeric vector of length equal to the number of points in X (or subset, if provided), containing the distance from each point to its k-th nearest neighbor. This is equivalent to but more memory efficient than using findKNN and subsetting to the last distance.

# Author(s)

Aaron Lun

## See Also

buildIndex, to build an index ahead of time.

# **Examples**

```
Y <- matrix(rnorm(100000), ncol=20)
out <- findDistance(Y, k=8)
summary(out)</pre>
```

findKNN

Find k-nearest neighbors

# Description

Find the k-nearest neighbors of each point in a dataset.

# Usage

```
findKNN(
   X,
   k,
   get.index = TRUE,
   get.distance = TRUE,
   num.threads = 1,
   subset = NULL,
   ...,
   BNPARAM = NULL
)
```

## **Arguments**

Χ

A numeric matrix where rows correspond to data points and columns correspond to variables (i.e., dimensions). Alternatively, a prebuilt BiocNeighborIndex object from buildIndex.

findKNN 11

k A positive integer scalar specifying the number of nearest neighbors to retrieve.

Alternatively, an integer vector of length equal to the number of points in X, specifying the number of neighbors to identify for each point. If subset is provided, this should have length equal to the length of subset. Users should wrap this vector in an AsIs class to distinguish length-1 vectors from integer scalars.

All k should be less than or equal to the number of points in X minus 1, otherwise the former will be capped at the latter with a warning.

get.index A logical scalar indicating whether the indices of the nearest neighbors should

be recorded. Setting this to FALSE improves efficiency if the indices are not of

interest.

Alternatively, if k is an integer scalar, this may be a string containing "normal" or "transposed". The former is the same as TRUE, while the latter returns the

index matrix in transposed format.

get.distance A logical scalar indicating whether distances to the nearest neighbors should be

recorded. Setting this to FALSE improves efficiency if the distances are not of

interest.

Alternatively, if k is an integer scalar, this may be a string containing "normal" or "transposed". The former is the same as TRUE, while the latter returns the

distance matrix in transposed format.

num. threads Integer scalar specifying the number of threads to use for the search.

subset An integer, logical or character vector specifying the indices of points in X for

which the nearest neighbors should be identified. This yields the same result as (but is more efficient than) subsetting the output matrices after computing

neighbors for all points.

... Further arguments to pass to buildIndex when X is not an external pointer.

BNPARAM A BiocNeighborParam object specifying how the index should be constructed.

If NULL, this defaults to a KmknnParam. Ignored if x contains a prebuilt index.

#### **Details**

If multiple queries are to be performed to the same X, it may be beneficial to build the index from X with buildIndex. The resulting pointer object can be supplied as X to multiple findKNN calls, avoiding the need to repeat index construction in each call.

## Value

List containing index (if get.index is not FALSE) and distance (if get.distance is not FALSE).

• If get.index=TRUE or "normal" and k is an integer scalar, index is an integer matrix with k columns where each row corresponds to a point (denoted here as i) in X. The i-th row contains the indices of points in X that are the nearest neighbors to point i, sorted by increasing distance from i. i will not be included in its own set of nearest neighbors.

If get.index=FALSE or "transposed" and k is an integer scalar, index is as described above but transposed, i.e., the i-th column contains the indices of neighboring points in X.

12 findMutualNN

• If get.distance=TRUE or "normal" and k is an integer scalar, distance is a numeric matrix of the same dimensions as index. The *i*-th row contains the distances of neighboring points in X to the point *i*, sorted in increasing order.

If get.distance=FALSE or "transposed" and k is an integer scalar, distance is as described above but transposed, i.e., the i-th column contains the distances to neighboring points in X.

- If get.index is not FALSE and k is an integer vector, index is a list of integer vectors where each vector corresponds to a point (denoted here as i) in X. The i-th vector has length k[i] and contains the indices of points in X that are the nearest neighbors to point i, sorted by increasing distance from i.
- If get.distance is not FALSE and k is an integer vector, distance is a list of numeric vectors of the same lengths as those in index. The *i*-th vector contains the distances of neighboring points in X to the point *i*, sorted in increasing order.

## Author(s)

Aaron Lun

## See Also

buildIndex, to build an index ahead of time.

findDistance, to efficiently obtain the distance to the k-th nearest neighbor.

## **Examples**

```
Y <- matrix(rnorm(100000), ncol=20)
out <- findKNN(Y, k=8)
head(out$index)
head(out$distance)</pre>
```

findMutualNN

Find mutual nearest neighbors

## **Description**

Find mutual nearest neighbors (MNN) across two data sets.

## **Usage**

```
findMutualNN(data1, data2, k1, k2 = k1, BNINDEX1 = NULL, BNINDEX2 = NULL, ...)
```

findMutualNN 13

## **Arguments**

data1	A numeric matrix containing points in the rows and variables/dimensions in the columns.
data2	A numeric matrix like data1 for another dataset with the same variables/dimensions.
k1	Integer scalar specifying the number of neighbors to search for in data1.
k2	Integer scalar specifying the number of neighbors to search for in data2.
BNINDEX1	A pre-built index for data1. If NULL, this is constructed from data1 within the internal queryKNN call.
BNINDEX2	A pre-built index for data2. If NULL, this is constructed from data2 within the internal queryKNN call.
	Other arguments to be passed to the underlying queryKNN calls, e.g., BNPARAM, .

## **Details**

For each point in dataset 1, the set of k2 nearest points in dataset 2 is identified. For each point in dataset 2, the set of k1 nearest points in dataset 1 is similarly identified. Two points in different datasets are considered to be part of an MNN pair if each point lies in the other's set of neighbors. This concept allows us to identify matching points across datasets, which is useful for, e.g., batch correction.

Any values for the BNINDEX1 and BNINDEX2 arguments should be equal to the output of buildIndex for the respective matrices, using the algorithm specified with BNPARAM. These arguments are only provided to improve efficiency during repeated searches on the same datasets (e.g., for comparisons between all pairs). The specification of these arguments should not, generally speaking, alter the output of the function.

## Value

A list containing the integer vectors first and second, containing row indices from data1 and data2 respectively. Corresponding entries in first and second specify a MNN pair consisting of the specified rows from each matrix.

## Author(s)

Aaron Lun

## See Also

queryKNN for the underlying neighbor search code.

fastMNN and related functions from the **batchelor** package, from which this code was originally derived.

# Examples

```
B1 <- matrix(rnorm(10000), ncol=50) # Batch 1
B2 <- matrix(rnorm(10000), ncol=50) # Batch 2
out <- findMutualNN(B1, B2, k1=20)
head(out$first)
```

14 findNeighbors

head(out\$second)

findNeighbors

Find neighbors within a threshold distance

# Description

Find all neighbors within a threshold distance of each point of a dataset.

# Usage

```
findNeighbors(
   X,
   threshold,
   get.index = TRUE,
   get.distance = TRUE,
   num.threads = 1,
   subset = NULL,
   ...,
   BNPARAM = NULL
)
```

# Arguments

X	A numeric matrix where rows correspond to data points and columns correspond to variables (i.e., dimensions). Alternatively, a prebuilt BiocNeighborIndex object from buildIndex.
threshold	A positive numeric scalar specifying the maximum distance at which a point is considered a neighbor. Alternatively, a vector containing a different distance threshold for each point.
get.index	A logical scalar indicating whether the indices of the neighbors should be recorded.
get.distance	A logical scalar indicating whether distances to the neighbors should be recorded.
num.threads	Integer scalar specifying the number of threads to use for the search.
subset	An integer, logical or character vector specifying the indices of points in X for which the nearest neighbors should be identified. This yields the same result as (but is more efficient than) subsetting the output matrices after computing neighbors for all points.
	Further arguments to pass to buildIndex when X is not an external pointer.
BNPARAM	A BiocNeighborParam object specifying how the index should be constructed. If NULL, this defaults to a KmknnParam. Ignored if x contains a prebuilt index.

findNeighbors 15

## **Details**

This function identifies all points in X that within threshold of each point in X. For Euclidean distances, this is equivalent to identifying all points in a hypersphere centered around the point of interest. Not all implementations support this search mode, but we can use KmknnParam and VptreeParam.

If threshold is a vector, each entry is assumed to specify a (possibly different) threshold for each point in X. If subset is also specified, each entry is assumed to specify a threshold for each point in subset. An error will be raised if threshold is a vector of incorrect length.

If multiple queries are to be performed to the same X, it may be beneficial to build the index from X with buildIndex. The resulting pointer object can be supplied as X to multiple findNeighbors calls, avoiding the need to repeat index construction in each call.

## Value

A list is returned containing:

- index, if get.index=TRUE. This is a list of integer vectors where each entry corresponds to a point (denoted here as *i*) in X. The vector for *i* contains the set of row indices of all points in X that lie within threshold of point *i*. Neighbors for *i* are sorted by increasing distance.
- distance, if get.distance=TRUE. This is a list of numeric vectors where each entry corresponds to a point (as above) and contains the distances of the neighbors from *i*. Elements of each vector in distance match to elements of the corresponding vector in index.

If both get.index=FALSE and get.distance=FALSE, an integer vector is returned of length equal to the number of observations. The i-th entry contains the number of neighbors of i within threshold.

If subset is not NULL, each entry of the above vector/lists corresponds to a point in the subset, in the same order as supplied in subset.

# Author(s)

Aaron Lun

## See Also

buildIndex, to build an index ahead of time.

## **Examples**

```
Y <- matrix(runif(100000), ncol=20)
out <- findNeighbors(Y, threshold=1)
summary(lengths(out$index))</pre>
```

16 HnswParam

HnswParam

The HnswParam class

## **Description**

A class to hold parameters for the HNSW algorithm for approximate nearest neighbor identification.

# Usage

```
HnswParam(
  nlinks = 16,
  ef.construction = 200,
  ef.search = 10,
  distance = c("Euclidean", "Manhattan", "Cosine")
)
## S4 method for signature 'HnswParam'
defineBuilder(BNPARAM)
```

## Arguments

nlinks Integer scalar, number of bi-directional links per element for index generation.

ef.construction

Integer scalar, size of the dynamic list for index generation.

ef. search Integer scalar, size of the dynamic list for neighbor searching.

distance String specifying the distance metric to use. Cosine distances are implemented

as Euclidean distances on L2-normalized coordinates.

BNPARAM A HsnwParam instance.

## **Details**

In the HNSW algorithm (Malkov and Yashunin, 2016), each point is a node in a "nagivable small world" graph. The nearest neighbor search proceeds by starting at a node and walking through the graph to obtain closer neighbors to a given query point. Nagivable small world graphs are used to maintain connectivity across the data set by creating links between distant points. This speeds up the search by ensuring that the algorithm does not need to take many small steps to move from one cluster to another. The HNSW algorithm extends this idea by using a hierarchy of such graphs containing links of different lengths, which avoids wasting time on small steps in the early stages of the search where the current node position is far from the query.

Larger values of nlinks improve accuracy at the expense of speed and memory usage. Larger values of ef.construction improve index quality at the expense of indexing time. The value of ef.search controls the accuracy of the neighbor search at run time, where larger values improve accuracy at the expense of a slower search.

Technically, the index construction algorithm is stochastic but, for various logistical reasons, the seed is hard-coded into the C++ code. This means that the results of the HNSW neighbor searches will be fully deterministic for the same inputs, even though the theory provides no such guarantees.

KmknnParam 17

# Value

The HnswParam constructor returns an instance of the HnswParam class.

The defineBuilder method returns an external pointer that can be used in buildIndex to construct a HNSW index.

## Author(s)

Aaron Lun

# See Also

BiocNeighborParam, for the parent class and its available methods.

https://github.com/nmslib/hnswlib, for details on the underlying algorithm.

# **Examples**

```
(out <- HnswParam())
out[['nlinks']]
out[['nlinks']] <- 20L
out</pre>
```

KmknnParam

The KmknnParam class

# Description

A class to hold parameters for the k-means k-nearest-neighbors (KMKNN) algorithm for exact nearest neighbor identification.

# Usage

```
KmknnParam(..., distance = c("Euclidean", "Manhattan", "Cosine"))
## S4 method for signature 'KmknnParam'
defineBuilder(BNPARAM)
```

# **Arguments**

Further arguments, ignored.

distance String specifying the distance metric to use. Cosine distances are implemented

as Euclidean distances on L2-normalized coordinates.

BNPARAM A KmknnParam instance.

18 queryDistance

## **Details**

In the KMKNN algorithm (Wang, 2012), k-means clustering is first applied to the data points using the square root of the number of points as the number of cluster centers. The cluster assignment and distance to the assigned cluster center for each point represent the KMKNN indexing information. This speeds up the nearest neighbor search by exploiting the triangle inequality between cluster centers, the query point and each point in the cluster to narrow the search space. The advantage of the KMKNN approach is its simplicity and minimal overhead, resulting in performance improvements over conventional tree-based methods for high-dimensional data where most points need to be searched anyway. It is also trivially extended to find all neighbors within a threshold distance from a query point.

Note that KMKNN operates much more naturally with Euclidean distances. Computational efficiency may not be optimal when using it with other choices of distance, though the results will still be exact.

#### Value

The KmknnParam constructor returns an instance of the KmknnParam class.

The defineBuilder method returns a list that can be used in buildIndex to construct a KMKNN index.

## Author(s)

Aaron Lun, using code from the cydar package.

## References

Wang X (2012). A fast exact k-nearest neighbors algorithm for high dimensional search using k-means clustering and triangle inequality. *Proc Int Jt Conf Neural Netw*, 43, 6:2351-2358.

## See Also

BiocNeighborParam, for the parent class and its available methods.

## **Examples**

```
(out <- KmknnParam(iter.max=100))</pre>
```

queryDistance

Distance to the k-th nearest neighbor to query points

## **Description**

Query a reference dataset to determine the distance to the k-th nearest neighbor of each point in a query dataset.

queryDistance 19

## Usage

```
queryDistance(
   X,
   query,
   k,
   num.threads = 1,
   subset = NULL,
   transposed = FALSE,
   ...,
   BNPARAM = NULL
)
```

## **Arguments**

X The reference dataset to be queried. This should be a numeric matrix where rows correspond to reference points and columns correspond to variables (i.e., dimensions). Alternatively, a prebuilt BiocNeighborIndex object from buildIndex.

A numeric matrix of query points, containing the same number of columns as X.

A positive integer scalar specifying the number of nearest neighbors to retrieve.

Alternatively, an integer vector of length equal to the number of points in query, specifying the number of neighbors to identify for each point. If subset is provided, this should have length equal to the length of subset. Users should wrap this vector in an AsIs class to distinguish length-1 vectors from integer

All k should be less than or equal to the number of points in X, otherwise the former will be capped at the latter with a warning.

num. threads

Integer scalar specifying the number of threads to use for the search.

subset An integer, logical or character vector indicating the rows of query (or columns,

if transposed=TRUE) for which the nearest neighbors should be identified.

transposed A logical scalar indicating whether X and query are transposed, in which case

both matrices are assumed to contain dimensions in the rows and data points in

the columns.

scalars.

... Further arguments to pass to buildIndex when X is not an external pointer.

BNPARAM A BiocNeighborParam object specifying how the index should be constructed.

If NULL, this defaults to a KmknnParam. Ignored if x contains a prebuilt index.

## **Details**

If multiple queries are to be performed to the same X, it may be beneficial to build the index from X with buildIndex. The resulting pointer object can be supplied as X to multiple queryKNN calls, avoiding the need to repeat index construction in each call.

## Value

Numeric vector of length equal to the number of points in query (or subset, if provided), containing the distance from each point to its k-th nearest neighbor. This is equivalent to but more memory efficient than using queryKNN and subsetting to the last distance.

20 queryKNN

## Author(s)

Aaron Lun

## See Also

buildIndex, to build an index ahead of time.

## **Examples**

```
Y <- matrix(rnorm(100000), ncol=20)
Z <- matrix(rnorm(20000), ncol=20)
out <- queryDistance(Y, query=Z, k=5)
head(out)</pre>
```

queryKNN

Query k-nearest neighbors

## **Description**

Query a reference dataset for the k-nearest neighbors of each point in a query dataset.

## Usage

```
queryKNN(
   X,
   query,
   k,
   get.index = TRUE,
   get.distance = TRUE,
   num.threads = 1,
   subset = NULL,
   transposed = FALSE,
   ...,
   BNPARAM = NULL
)
```

## **Arguments**

Χ

The reference dataset to be queried. This should be a numeric matrix where rows correspond to reference points and columns correspond to variables (i.e., dimensions). Alternatively, a prebuilt BiocNeighborIndex object from buildIndex.

query

A numeric matrix of query points, containing the same number of columns as X.

k

A positive integer scalar specifying the number of nearest neighbors to retrieve. Alternatively, an integer vector of length equal to the number of points in query, specifying the number of neighbors to identify for each point. If subset is provided, this should have length equal to the length of subset. Users should

queryKNN 21

wrap this vector in an AsIs class to distinguish length-1 vectors from integer scalars.

All k should be less than or equal to the number of points in X, otherwise the former will be capped at the latter with a warning.

get.index A logical scalar indicating whether the indices of the nearest neighbors should

be recorded. Setting this to FALSE improves efficiency if the indices are not of

interest.

Alternatively, if k is an integer scalar, this may be a string containing "normal" or "transposed". The former is the same as TRUE, while the latter returns the

index matrix in transposed format.

get.distance A logical scalar indicating whether distances to the nearest neighbors should be

recorded. Setting this to FALSE improves efficiency if the distances are not of

interest.

Alternatively, if k is an integer scalar, this may be a string containing "normal" or "transposed". The former is the same as TRUE, while the latter returns the

distance matrix in transposed format.

num. threads Integer scalar specifying the number of threads to use for the search.

subset An integer, logical or character vector indicating the rows of query (or columns,

if transposed=TRUE) for which the nearest neighbors should be identified.

transposed A logical scalar indicating whether X and query are transposed, in which case

both matrices are assumed to contain dimensions in the rows and data points in

the columns.

... Further arguments to pass to buildIndex when X is not an external pointer.

BNPARAM A BiocNeighborParam object specifying how the index should be constructed.

If NULL, this defaults to a KmknnParam. Ignored if x contains a prebuilt index.

# Details

If multiple queries are to be performed to the same X, it may be beneficial to build the index from X with buildIndex. The resulting pointer object can be supplied as X to multiple queryKNN calls, avoiding the need to repeat index construction in each call.

# Value

List containing index (if get.index is not FALSE) and distance (if get.distance is not FALSE).

• If get.index=TRUE or "normal" and k is an integer scalar, index is an integer matrix with k columns where each row corresponds to a point (denoted here as i) in query. The i-th row contains the indices of points in X that are the nearest neighbors to point i, sorted by increasing distance from i.

If get.index=FALSE or "transposed" and k is an integer scalar, index is as described above but transposed, i.e., the i-th column contains the indices of neighboring points in X.

• If get.distance=TRUE or "normal" and k is an integer scalar, distance is a numeric matrix of the same dimensions as index. The *i*-th row contains the distances of neighboring points in X to the point *i*, sorted in increasing order.

If get.distance=FALSE or "transposed" and k is an integer scalar, distance is as described above but transposed, i.e., the i-th column contains the distances to neighboring points in X.

22 queryNeighbors

• If get.index is not FALSE and k is an integer vector, index is a list of integer vectors where each vector corresponds to a point (denoted here as i) in X. The i-th vector has length k[i] and contains the indices of points in X that are the nearest neighbors to point i, sorted by increasing distance from i.

• If get.distance is not FALSE and k is an integer vector, distance is a list of numeric vectors of the same lengths as those in index. The *i*-th vector contains the distances of neighboring points in X to the point *i*, sorted in increasing order.

## Author(s)

Aaron Lun

## See Also

buildIndex, to build an index ahead of time. queryDistance, to obtain the distance from each query point to its k-th nearest neighbor.

# **Examples**

```
Y <- matrix(rnorm(100000), ncol=20)
Z <- matrix(rnorm(20000), ncol=20)
out <- queryKNN(Y, query=Z, k=5)
head(out$index)
head(out$distance)</pre>
```

queryNeighbors

Query neighbors within a threshold distance

## **Description**

Find all points in a reference dataset that lie within a threshold distance of each point in a query dataset.

# Usage

```
queryNeighbors(
   X,
   query,
   threshold,
   get.index = TRUE,
   get.distance = TRUE,
   num.threads = 1,
   subset = NULL,
   transposed = FALSE,
   ...,
   BNPARAM = NULL
)
```

queryNeighbors 23

## **Arguments**

X	The reference dataset to be queried. This should be a numeric matrix where rows correspond to reference points and columns correspond to variables (i.e., dimensions). Alternatively, a prebuilt BiocNeighborIndex object from buildIndex.
query	A numeric matrix of query points, containing the same number of columns as X.
threshold	A positive numeric scalar specifying the maximum distance at which a point is considered a neighbor. Alternatively, a vector containing a different distance threshold for each query point.
get.index	A logical scalar indicating whether the indices of the neighbors should be recorded.
get.distance	A logical scalar indicating whether distances to the neighbors should be recorded.
num.threads	Integer scalar specifying the number of threads to use for the search.
subset	An integer, logical or character vector indicating the rows of query (or columns, if transposed=TRUE) for which the nearest neighbors should be identified.
transposed	A logical scalar indicating whether X and query are transposed, in which case both matrices are assumed to contain dimensions in the rows and data points in the columns.
	Further arguments to pass to buildIndex when X is not an external pointer.
BNPARAM	A BiocNeighborParam object specifying how the index should be constructed. If NULL, this defaults to a KmknnParam. Ignored if x contains a prebuilt index.

## **Details**

This function identifies all points in X that within threshold of each point in query. For Euclidean distances, this is equivalent to identifying all points in a hypersphere centered around the point of interest. Not all implementations support this search mode, but we can use KmknnParam and VptreeParam.

If threshold is a vector, each entry is assumed to specify a (possibly different) threshold for each point in query. If subset is also specified, each entry is assumed to specify a threshold for each point in subset. An error will be raised if threshold is a vector of incorrect length.

If multiple queries are to be performed to the same X, it may be beneficial to build the index from X with buildIndex. The resulting pointer object can be supplied as X to multiple queryKNN calls, avoiding the need to repeat index construction in each call.

## Value

A list is returned containing:

- index, if get.index=TRUE. This is a list of integer vectors where each entry corresponds to a point (denoted here as i) in query. The vector for i contains the set of row indices of all points in X that lie within threshold of point i. Neighbors for i are sorted by increasing distance from i.
- distance, if get.distance=TRUE. This is a list of numeric vectors where each entry corresponds to a point (as above) and contains the distances of the neighbors from *i*. Elements of each vector in distance match to elements of the corresponding vector in index.

24 VptreeParam

If both get.index=FALSE and get.distance=FALSE, an integer vector is returned of length equal to the number of observations. The i-th entry contains the number of neighbors of i within threshold.

If subset is not NULL, each entry of the above vector/lists refers to a point in the subset, in the same order as supplied in subset.

## Author(s)

Aaron Lun

## See Also

buildIndex, to build an index ahead of time.

# **Examples**

```
Y <- matrix(rnorm(100000), ncol=20)
Z <- matrix(rnorm(20000), ncol=20)
out <- queryNeighbors(Y, query=Z, threshold=3)
summary(lengths(out$index))</pre>
```

**VptreeParam** 

The VptreeParam class

# **Description**

A class to hold parameters for the vantage point (VP) tree algorithm for exact nearest neighbor identification.

# Usage

```
VptreeParam(distance = c("Euclidean", "Manhattan", "Cosine"))
## S4 method for signature 'VptreeParam'
defineBuilder(BNPARAM)
```

# **Arguments**

distance String specifying the distance metric to use. Cosine distances are implemented

as Euclidean distances on L2-normalized coordinates.

BNPARAM A VptreeParam instance.

VptreeParam 25

## **Details**

In a VP tree (Yianilos, 1993), each node contains a subset of points that is split into two further partitions. The split is determined by picking an arbitrary point inside that subset as the node center, computing the distance to all other points from the center, and taking the median as the "radius". The left child of this node contains all points within the median distance from the radius, while the right child contains the remaining points. This is applied recursively until all points resolve to individual nodes. The nearest neighbor search traverses the tree and exploits the triangle inequality between query points, node centers and thresholds to narrow the search space.

VP trees are often faster than more conventional KD-trees or ball trees as the former uses the points themselves as the nodes of the tree, avoiding the need to create many intermediate nodes and reducing the total number of distance calculations. Like KMKNN, it is also trivially extended to find all neighbors within a threshold distance from a query point.

## Value

The VptreeParam constructor returns an instance of the VptreeParam class.

The defineBuilder method returns an external pointer that can be used in buildIndex to construct a VP tree index.

# Author(s)

Aaron Lun

## References

Yianilos PN (1993). Data structures and algorithms for nearest neighbor search in general metric spaces. *Proceedings of the Fourth Annual ACM-SIAM Symposium on Discrete Algorithms*, 311-321.

#### See Also

BiocNeighborParam, for the parent class and its available methods.

https://stevehanov.ca/blog/index.php?id=130, for a description of the algorithm.

# **Examples**

```
(out <- VptreeParam())</pre>
```

# **Index**

[[,BiocNeighborParam-method	defineBuilder, HnswParam-method
(BiocNeighborParam), 5	(HnswParam), 16
<pre>[[&lt;-,BiocNeighborParam-method</pre>	<pre>defineBuilder,KmknnParam-method     (KmknnParam), 17</pre>
(Blocheighbor Far alli), 5	defineBuilder, missing-method
AnnoyIndex (AnnoyParam), 3	(defineBuilder), 7
AnnoyIndex (AnnoyParam), 3	defineBuilder,NULL-method
AnnoyParam, 3, 5	(defineBuilder), 7
AnnoyParam-class (AnnoyParam), 3	defineBuilder,VptreeParam-method
AsIs, 9, 11, 19, 21	(VptreeParam), 24
BiocNeighborGenericIndex, 6, 7	ExhaustiveIndex (ExhaustiveParam), 8
BiocNeighborGenericIndex-class	ExhaustiveIndex-class
(BiocNeighborIndex),4	(ExhaustiveParam), 8
BiocNeighborIndex, 4, 6, 7, 9, 10, 14, 19, 20,	ExhaustiveParam, 5, 8
23	ExhaustiveParam-class
BiocNeighborIndex-class	(ExhaustiveParam), 8
(BiocNeighborIndex), 4	
BiocNeighborParam, 4, 5, 6–9, 11, 14, 17–19,	findDistance, 9, 12
21, 23, 25	find Distance, Bioc Neighbor Generic Index, ANY-method
BiocNeighborParam-class	(findDistance), 9
(BiocNeighborParam), 5	<pre>findDistance,BiocNeighborGenericIndex-method</pre>
BiocNeighbors (BiocNeighbors-package), 2	(findDistance), 9
BiocNeighbors-package, 2	findDistance,matrix,ANY-method
<pre>bndistance (BiocNeighborParam), 5</pre>	(findDistance), 9
buildIndex, 4, 5, 6, 7–15, 17–25	findDistance,matrix-method
<pre>buildIndex,matrix,BiocNeighborParam-method</pre>	(findDistance), 9
(buildIndex), 6	findKNN, 4–7, 10, 10
<pre>buildIndex,matrix,list-method</pre>	findKNN,BiocNeighborGenericIndex,ANY-method
(buildIndex), 6	(findKNN), 10
<pre>buildIndex,matrix,missing-method</pre>	findKNN,BiocNeighborGenericIndex-method
(buildIndex), 6	(findKNN), 10
buildIndex, matrix, NULL-method	findKNN, matrix, ANY-method (findKNN), 10
(buildIndex), 6	findKNN, matrix-method (findKNN), 10
	findKNN, missing, ANY-method (findKNN), 10
defineBuilder, 4, 6, 7, 8, 17, 18, 25	findKNN, missing-method (findKNN), 10
defineBuilder, AnnoyParam-method	findMutualNN, 12
(AnnoyParam), 3	findNeighbors, 14
defineBuilder, ExhaustiveParam-method	findNeighbors, BiocNeighborGenericIndex, ANY-method
(ExhaustiveParam).8	(findNeighbors), 14

INDEX 27

```
findNeighbors, BiocNeighborGenericIndex-methodqueryNeighbors, matrix-method
        (findNeighbors), 14
                                                         (queryNeighbors), 22
findNeighbors, matrix, ANY-method
                                                 queryNeighbors, missing, ANY-method
        (findNeighbors), 14
                                                         (queryNeighbors), 22
findNeighbors,matrix-method
                                                 queryNeighbors, missing-method
                                                         (queryNeighbors), 22
        (findNeighbors), 14
findNeighbors,missing,ANY-method
                                                 show, AnnoyParam-method (AnnoyParam), 3
        (findNeighbors), 14
                                                 show, BiocNeighborIndex-method
findNeighbors,missing-method
                                                         (BiocNeighborIndex), 4
        (findNeighbors), 14
                                                 show, BiocNeighborParam-method
                                                         (BiocNeighborParam), 5
HnswIndex (HnswParam), 16
                                                 show, HnswParam-method (HnswParam), 16
HnswIndex-class (HnswParam), 16
HnswParam, 5, 16
                                                 VptreeIndex (VptreeParam), 24
HnswParam-class (HnswParam), 16
                                                 VptreeIndex-class (VptreeParam), 24
                                                 VptreeParam, 5, 15, 23, 24
KmknnIndex (KmknnParam), 17
                                                 VptreeParam-class (VptreeParam), 24
KmknnIndex-class (KmknnParam), 17
KmknnParam, 5-7, 9, 11, 14, 15, 17, 19, 21, 23
KmknnParam-class (KmknnParam), 17
queryDistance, 18, 22
queryDistance,BiocNeighborGenericIndex,ANY-method
        (queryDistance), 18
queryDistance,BiocNeighborGenericIndex-method
        (queryDistance), 18
queryDistance, matrix, ANY-method
        (queryDistance), 18
queryDistance, matrix-method
        (queryDistance), 18
queryKNN, 5, 6, 13, 19, 20
{\tt queryKNN}, {\tt BiocNeighborGenericIndex}, {\tt ANY-method}
        (queryKNN), 20
queryKNN,BiocNeighborGenericIndex-method
        (queryKNN), 20
queryKNN, matrix, ANY-method (queryKNN),
queryKNN, matrix-method (queryKNN), 20
queryKNN, missing, ANY-method (queryKNN),
queryKNN, missing-method (queryKNN), 20
queryNeighbors, 7, 22
queryNeighbors,BiocNeighborGenericIndex,ANY-method
        (queryNeighbors), 22
queryNeighbors,BiocNeighborGenericIndex-method
        (queryNeighbors), 22
queryNeighbors, matrix, ANY-method
        (queryNeighbors), 22
```