

# Package ‘stream’

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**Title** Infrastructure for Data Streams

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**Description** A framework for data stream modelling and associated data mining tasks such as clustering and classification.

**Depends** R (>= 2.10.0), clusterGeneration, proxy

**Imports** rJava (>= 0.6-3), MASS

**SystemRequirements** Java (>= 5.0)

**License** GPL-2

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cluster

*Clustering data based with input algorithm and data stream*


---

### Description

Clusters a number of input points from a data stream into a clustering object.

### Usage

```
cluster(dsc, dsd, n = 1000)
```

### Arguments

dsc	a DSC object
dsd	a DSD object
n	number of points to cluster

### Details

Clusters input data. The underlying clustering in the DSC object is implicitly updated as if the object was 'passed by reference' in a traditional programming language.

### Value

The updated DSC object is returned invisibly for reassignment.

### Examples

```
dsc <- DSC_DenStream()
dsd <- DSD_MOA()
cluster(dsc, dsd, 2000)
plot(dsc)
```

---

DSClusterer

*DSClusterer*


---

### Description

The abstract base class for all DSC classes.

### Details

The DSClusterer class cannot be instantiated, but it serves as a base class from which all DSC objects inherit from.

DSClusterer provides several generic functions that can operate on all DSC subclasses: `get_centers()`, which returns the centers of the micro-clusters generated; `nclusters()`, which returns the number of micro-clusters generated; `print()`, which prints general data about the DSC; and `plot()`, which will plot the centers of the micro-clusters.

**See Also**

[DSCclusterer](#), [DSC\\_CobWeb](#), [DSC\\_ClusTree](#), [DSC\\_StreamKM](#), [DSC\\_DenStream](#), [DSC\\_Clustream](#), [DSDData](#)

---

DSC\_Clustream

*DataStreamClusterer: Clustream*


---

**Description**

Creates a new DataStreamClusterer with the Clustream algorithm

**Usage**

```
DSC_Clustream(timeWindow = 1000, maxNumKernels = 100)
```

**Arguments**

timeWindow	Defines the time window to be used in Clustream
maxNumKernels	Defines the maximum number of kernels used in Clustream

**Details**

Clustream was originally introduced in Aggarwal's paper cited below. Like most data stream clustering algorithms, Clustream summarizes the incoming data points into micro-clusters. At any given time, it can be assumed there are  $q$  micro-clusters maintained by Clustream. This means that when a new micro-cluster is created, another one must be deleted, and the opposite is true for when two micro-clusters merge with one another. A significant amount of thought in Clustream is given to the temporal structure of data streams. Snapshots of calculated micro-clusters are stored in a "pyramidal time frame" pattern. This allows the user to turn back to a specific location in the data stream to mine information from a particular time period.

The `timeWindow` parameter changes the granularity at which the snapshots are taken for the micro-cluster summary maintenance. `maxNumKernels` is analogous to the  $q$  micro-clusters that are maintained within the algorithm.

**Value**

description	The name of the algorithm in the DSC object.
options	The CLI params defined when creating the DSC object.
javaObj	The underlying Java object associated with DSC_MOA objects.

**References**

Aggarwal CC, Han J, Wang J, Yu PS (2003). "A Framework for Clustering Evolving Data Streams." In "Proceedings of the International Conference on Very Large Data Bases (VLDB '03)," pp. 81-92.

Bifet A, Holmes G, Pfahringer B, Kranen P, Kremer H, Jansen T, Seidl T (2010). "MOA: Massive Online Analysis, a Framework for Stream Classification and Clustering." In "Journal of Machine Learning Research (JMLR) Workshop and Conference Proceedings,"

**See Also**

[DSClusterer](#), [DSC\\_MOA](#), [DSC\\_DenStream](#), [DSC\\_CobWeb](#), [DSC\\_StreamKM](#), [DSC\\_ClusTree](#)

**Examples**

```
dsc <- DSC_Clustream()
dsd <- DSD_Gaussian_Static()
cluster(dsc, dsd, 5000)
plot(dsc)
```

---

DSC\_ClusTree

*DataStreamClusterer: ClusTree*


---

**Description**

Creates a new DataStreamClusterer with the ClusTree algorithm

**Usage**

```
DSC_ClusTree(timeWindow = 1000, maxHeight=8)
```

**Arguments**

timeWindow	The timeWindow parameter used in MOA.
maxHeight	The maximum height of the tree.

**Details**

The authors of ClusTree define it as a "parameter free algorithm". Their innovation automatically adjusts to the speed of the incoming points generated by a data stream. ClusTree is able to detect concept drift and outliers in the stream being operated upon.

**Value**

A list of class DSC, DSC\_MOA, and DSC\_ClusTree. The list contains the following items:

description	The name of the algorithm in the DSC object.
options	The CLI params defined when creating the DSC object.
javaObj	The underlying Java object associated with DSC_MOA objects.

**References**

Philipp Kranen, Ira Assent, Corinna Baldauf, and Thomas Seidl. 2009. Self-Adaptive Anytime Stream Clustering. In Proceedings of the 2009 Ninth IEEE International Conference on Data Mining (ICDM '09). IEEE Computer Society, Washington, DC, USA, 249-258. DOI=10.1109/ICDM.2009.47 <http://dx.doi.org/10.1109/ICDM.2009.47>

**See Also**

[DSClusterer](#), [DSC\\_MOA](#), [DSC\\_DenStream](#), [DSC\\_CobWeb](#), [DSC\\_StreamKM](#), [DSC\\_Clustream](#)

**Examples**

```
dsc <- DSC_ClusTree()
dsd <- DSD_Gaussian_Static()
cluster(dsc, dsd, 5000)
plot(dsc)
```

DSC\_CobWeb

*DataStreamClusterer: CobWeb***Description**

Creates a new DataStreamClusterer with the CobWeb algorithm

**Usage**

```
DSC_CobWeb(acuity = 1, cutoff = 0.002, randomSeed = 1)
```

**Arguments**

acuity	The acuity used in the underlying implementation.
cutoff	The cut off.
randomSeed	The random seed used in the MOA implementation.

**Value**

A list of class DSC, DSC\_MOA, and DSC\_CobWeb. The list contains the following items:

description	The name of the algorithm in the DSC object.
options	The CLI params defined when creating the DSC object.
javaObj	The underlying Java object associated with DSC_MOA objects.

**References**

Douglas H. Fisher. 1987. Knowledge Acquisition Via Incremental Conceptual Clustering. Mach. Learn. 2, 2 (September 1987), 139-172.

**See Also**

[DSCclusterer](#), [DSC\\_MOA](#), [DSC\\_DenStream](#), [DSC\\_Clustream](#), [DSC\\_StreamKM](#), [DSC\\_ClusTree](#)

**Examples**

```
#dsc <- DSC_CobWeb()
#dsd <- DSD_Gaussian_Static()
#cluster(dsc, dsd, 1000)
#plot(dsc)
```

---

DSC\_DenStream

*DataStreamClusterer: DenStream*


---

## Description

Creates a new DataStreamClusterer with the DenStream algorithm

## Usage

```
DSC_DenStream(epsilon = 0.1, minPoints = 10, lambda = 0.006, beta = 0.001, mu =
```

## Arguments

epsilon	-e (defines the epsilon neighborhood, range: 0 to 1)
minPoints	-p (min. num. points a cluster must have)
lambda	-l (range: 0 to 1)
beta	-b (range: 0 to 1)
mu	-m (range: 0 to max(double))
initPoints	-i (number of points to use for initialization)

## Details

DSC\_DenStream is based on the DenStream algorithm. DenStream is a density based clustering algorithm that summarizes the incoming data points into micro-clusters based upon the frequency of data points in a particular location within a defined boundary (i.e., their density). DenStream includes a fading function that causes the weight of data points to decrease exponentially as t increases. This means that temporal structures within the data stream are also visible using this algorithm.

The parameters are defined as follows: epsilon: refers to the neighborhood in which the density of each micro-cluster is calculated (in other words, the radius).

minPoints: is the minimum number of points a micro-cluster must have.

lambda: alters the fading function: the larger lambda, the lower the weight of the historic data.

beta: is the outlier threshold, and is used in conjunction with mu

mu: the minimum weight a micro-cluster must have within the epsilon neighborhood

epsilon: the neighborhood to be considered a core-micro-cluster (as opposed to an outlier or a potential-micro-cluster).

initPoints: is the number of data points to initialize the algorithm with.

## Value

A list of class DSC, DSC\_MOA, and DSC\_DenStream. The list contains the following items:

description	The name of the algorithm in the DSC object.
options	The CLI params defined when creating the DSC object.
javaObj	The underlying Java object associated with DSC_MOA objects.

## References

Cao F, Ester M, Qian W, Zhou A (2006). "Density-Based Clustering over an Evolving Data Stream with Noise." In "Proceedings of the 2006 SIAM International Conference on Data Mining," pp 326-337. SIAM.

Bifet A, Holmes G, Pfahringer B, Kranen P, Kremer H, Jansen T, Seidl T (2010). "MOA: Massive Online Analysis, a Framework for Stream Classification and Clustering." In "Journal of Machine Learning Research (JMLR) Workshop and Conference Proceedings,"

## See Also

[DSClusterer](#), [DSC\\_MOA](#), [DSC\\_Clustream](#), [DSC\\_CobWeb](#), [DSC\\_StreamKM](#), [DSC\\_ClusTree](#)

## Examples

```
dsc <- DSC_DenStream()
dsd <- DSD_Gaussian_Static()
cluster(dsc, dsd, 5000)
plot(dsc)
```

---

DSC\_MOA

*DSC\_MOA*


---

## Description

An abstract class that inherits from the base DSC class, DSClusterer. All DSC objects that use underlying MOA code will inherit this class because they have a specific way of operating.

## Details

DSC\_MOA classes operate in a different way in that the centers of the micro-clusters have to be extracted from the underlying Java object. This is done by performing method calls directly in the JRI and unpackaging the multi-dimensional Java array into a local R data type.

## See Also

[DSClusterer](#), [DSC\\_CobWeb](#), [DSC\\_ClusTree](#), [DSC\\_StreamKM](#), [DSC\\_DenStream](#), [DSC\\_Clustream](#), [DSDData](#)

---

DSC\_StreamKM

*DataStreamClusterer: StreamKM*


---

## Description

Creates a new DataStreamClusterer with the StreamKM++ algorithm

## Usage

```
DSC_StreamKM(sizeCoreset=100, k=5, width=1000, randomSeed=1)
```

**Arguments**

sizeCoreset	The size of the coreset tree.
k	The number of clusters.
width	
randomSeed	The random seed used in the k-means algorithm.

**Value**

A list of class DSC, DSC\_MOA, and DSC\_StreamKM. The list contains the following items:

description	The name of the algorithm in the DSC object.
options	The CLI params defined when creating the DSC object.
javaObj	The underlying Java object associated with DSC_MOA objects.

**References**

Marcel R. Ackermann, Christiane Lammersen, Marcus M"artens, Christoph Raupach, Christian Sohler, Kamil Swierkot. StreamKM++: A Clustering Algorithm for Data Streams. In Proceedings of the 12th Workshop on Algorithm Engineering and Experiments (ALENEX '10), pp. 173-187, Society for Industrial and Applied Mathematics, 2010.

**See Also**

[DSCclusterer](#), [DSC\\_MOA](#), [DSC\\_DenStream](#), [DSC\\_CobWeb](#), [DSC\\_Clustream](#), [DSC\\_ClusTree](#)

**Examples**

```
#dsc <- DSC_StreamKM()
#dsd <- DSD_Gaussian_Static()
#cluster(dsc, dsd, 5000)
#plot(dsc)
```

---

DSC\_tNN

*DataStreamClusterer: tNN*


---

**Description**

Creates a new DataStreamClusterer with the tNN (threshold Nearest Neighbor) algorithm

**Usage**

```
DSC_tNN(threshold = 0.2, measure = "euclidean",
centroids = identical(tolower(measure), "euclidean"), lambda=0)
```

**Arguments**

threshold	The threshold in the nearest neighborhood algorithm.
measure	The measure used to calculate cluster proximity.
centroids	The measure used to calculate centroid proximity.
lambda	The lambda used in the fading function.



**Value**

<code>description</code>	The name of the algorithm in the DSC object.
<code>clusterFun</code>	The cluster function specific to DSC_tNN
<code>RObj</code>	The underlying R object associated with DSC_R objects.

**References**

M.H. Dunham, Y. Meng, J. Huang (2004): Extensible Markov Model, In: ICDM '04: Proceedings of the Fourth IEEE International Conference on Data Mining, pp. 371-374.

M. Hahsler, M. H. Dunham (2010): rEMM: Extensible Markov Model for Data Stream Clustering in R, Journal of Statistical Software, 35(5), 1-31, URL <http://www.jstatsoft.org/v35/i05/>

**See Also**

[DSCClusterer](#), [DSC\\_MOA](#), [DSC\\_DenStream](#), [DSC\\_CobWeb](#), [DSC\\_StreamKM](#), [DSC\\_Clustream](#), [DSC\\_ClusTree](#)

**Examples**

```
dsc <- DSC_tNN()
dsd <- DSD_Gaussian_Static()
cluster(dsc, dsd, 5000)
plot(dsc)
```

---

DSData

*DSData*


---

**Description**

The abstract base class for all `DataStreamData` classes.

**Details**

The `DSData` class cannot be instantiated, but it serves as a base class from which all DSD objects inherit from. There is only one function that needs to be implemented: `get_points()`. `DSData` also provides a generic `print()` function that displays basic information about the class.

In addition to this function, each DSD class also needs a constructor specific to that class. The links below contain various DSD classes that have been implemented.

**See Also**

[DSCClusterer](#), [DSD\\_Gaussian\\_Static](#),  
[DSD\\_ReadStream](#), [DSD\\_DataFrame](#),

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DSD_DataFrame	<i>DataStreamData: DataFrame (a wrapper for data frames and matrices)</i>
---------------	---

---

## Description

A DataStreamData class that wraps either a data frame or matrix that was created in R. The data can either be looped or replayed manually to give the exact same data to 2 different DataStreamTask objects.

## Usage

```
DSD_DataFrame(df, k, loop=FALSE)
```

## Arguments

df	A data frame or matrix with the data to be used in the stream.
k	Optional: The number of clusters
loop	A flag that tells the stream to loop or not to loop over the data frame.

## Details

The DSD\_DataFrame class is designed to be a wrapper class for data that is generated within R in either a data frame or matrix form. It removes the step of having to write the data to a file, then read in by a connection through DSD\_ReadStream.

It works like the other DSD classes—by calling `get_points()` to retrieve data from the stream. The function `reset_stream()` can be used to move the counter back to the beginning of the stream. This is an important feature to be able to replay stream data for multiple clusters as shown in the example below.

The value returned is a list of class DSD, and DSD\_DataFrame. The important items with the list are `strm`, the data frame being wrapped, `state` the environment which contains the counter, and `d`, the number of dimensions in the stream.

## Value

description	The name of the class of the DSD object.
strm	The data frame or matrix that the stream is wrapping.
state	The environment variable which holds the counter for the data frame (accessed through <code>state\$counter</code> ).
d	The number of dimensions ( <code>ncol(strm)</code> ).
k	The number of clusters (may not be defined).
loop	The flag that determines if looping is or is not enabled.

## See Also

[DSD\\_MOA](#), [DSD\\_ReadStream](#),  
[DSD\\_Gaussian\\_Static](#), [write\\_stream](#), [reset\\_stream](#)

**Examples**

```
# creating the DSD_DataFrame from other stream data
dsd <- DSD_Gaussian_Static(k=3, d=2)
replayer <- DSD_DataFrame(get_points(dsd, 100), k=3)

# creating 2 clusterers of different algorithms
dsc1 <- DSC_Clustream()
dsc2 <- DSC_CobWeb()

# clustering the same data in 2 DSC objects
cluster(dsc1, replayer, 100)
reset_stream(replayer) # resetting the dsd to its original state
cluster(dsc2, replayer, 100)
```

---

DSD\_Gaussian\_Static

*DataStreamData: Static data stream*


---

**Description**

A DataStreamData that generates random data based upon either a defined covariance matrix or a randomly generated covariance matrix.

**Usage**

```
DSD_Gaussian_Static(k=2, d=2, mu, sigma, p, noise=0)
```

**Arguments**

k	Determines the number of clusters.
d	Determines the number of dimensions.
mu	A matrix of means for each dimension of each cluster.
sigma	The covariance matrix.
p	A vector of probabilities that determines the likelihood of generated a data point from a particular cluster.
noise	Noise is generated in the stream based on this parameter.

**Details**

DSD\_Gaussian\_Static is a general purpose DSD generator for stream data. It has been implemented entirely in R, so there is no computational overhead with communicating to the Java Runtime Interface (JRI) or native C code. An important characteristic of DSD\_Gaussian\_Static is that once it has been initialized according to the input parameters defined, the defined clusters will not move (i.e., they are static). This means that DSD\_Gaussian\_Static is not an ideal DSD for examining the temporal structure of data streams for drastic changes such as the splitting or merging of clusters.

Its initialization function accepts 5 main parameters: the number of clusters *k*, the number of dimensions *d*, a matrix of means *mu*, a covariance matrix *sigma*, and a probability vector *p*.

By default, DSD\_Gaussian\_Static generates 2 dimensionality data in 2 different clusters, but the user is able to define any number of clusters with any number of dimensions. Additionally, the

user may define `mu`, `sigma`, and `p`, but if left undefined the constructor will generate these values automatically. When `get_points()` is called on `DSD_Gaussian_Static`, the data points are generated using the `mvnrm()` function, making it important to seed the random number generator to reproduce the experiment. This can be done with `set.seed()`.

### Value

Returns a `DSD_Gaussian_Static` object which is a list of the defined params. The params are either passed in from the function or created internally. They include:

<code>description</code>	A brief description of the DSD object.
<code>k</code>	The number of clusters.
<code>d</code>	The number of dimensions.
<code>mu</code>	The matrix of means of the dimensions in each cluster.
<code>sigma</code>	The covariance matrix.
<code>p</code>	The probability vector for the cluters.
<code>noise</code>	A flag that determines if or if not noise is generated.

### See Also

[DSD\\_MOA](#), [DSD\\_ReadStream](#),  
[DSD\\_DataFrame](#), [write\\_stream](#)

### Examples

```
# other params will be generated internally
dsd <- DSD_Gaussian_Static(k=2, d=2)
dsc <- DSC_DenStream()

# obtaining some data
sample <- get_points(dsd, 500)

# clustering some data, this will take some time
cluster(dsc, dsd, 5000)

# plotting the data
plot(sample, xlab='x', ylab='y', pch=4, cex=.5)
points(get_centers(dsc), col='red', cex=3, lwd=2)
```

---

DSD\_MOA

*DataStreamData - MOA*


---

### Description

A `DataStreamData` that generates random data based upon the `RandomRBFGenerator` implemented in MOA.

### Usage

```
DSD_MOA(k=3, d=2, avgRadius=0, modelSeed=1, instanceSeed=1)
```

**Arguments**

<code>k</code>	The number of clusters.
<code>d</code>	The dimensionality of the data.
<code>avgRadius</code>	The average radius of the micro-clusters.
<code>modelSeed</code>	Random seed for the model.
<code>instanceSeed</code>	Random seed for the instances.

**Value**

<code>description</code>	The name of the class of the DSD object.
<code>options</code>	The CLI params defined when creating the DSD object.
<code>javaObj</code>	The underlying Java object associated with DSD_MOA objects.

**References**

Albert Bifet, Geoff Holmes, Bernhard Pfahringer, Philipp Kranen, Hardy Kremer, Timm Jansen, Thomas Seidl. Journal of Machine Learning Research (JMLR) Workshop and Conference Proceedings. Volume 11: Workshop on Applications of Pattern Analysis (2010).

**See Also**

[DSD\\_Gaussian\\_Static](#), [DSD\\_ReadStream](#),  
[DSD\\_DataFrame](#), [write\\_stream](#)

---

DSD_ReadStream	<i>DataStreamData: ReadStream (from an open connection or a file name)</i>
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---

**Description**

A DSD object that reads a data stream from an R connection.

**Usage**

```
DSD_ReadStream(x, sep=",", loop=FALSE)
```

**Arguments**

<code>x</code>	An open connection, or a file path/URL to be opened as a connection.
<code>sep</code>	The character string that separates dimensions in data points in the stream.
<code>loop</code>	If enabled, the object will loop through the stream when the end has been reached. If disabled, the object will warn the user upon reaching the end.

**Details**

DSD\_ReadStream uses `read.table()` to read in data from an R connection. The connection is responsible for maintaining where the stream is currently being read from. In general, the connections will consist of files stored on disk but have many other possibilities (see [connection](#)).

The `get_points()` method can pass additional params to `read.table()` to alter how the reading is done. By default, the `comment.char` is set to an empty string for performance reasons.

**Value**

description	The name of the class of the DSD object.
con	The connection where the data stream is being read from.
sep	The character string that separates dimensions in data points in the stream.
loop	A flag that determines whether or not the stream will loop.

**See Also**

[DSD\\_MOA](#), [DSD\\_ReadStream](#), [DSD\\_Gaussian\\_Static](#),  
[DSD\\_DataFrame](#), [write\\_stream](#) [reset\\_stream](#)

**Examples**

```
# creating data and writing it to disk
dsd <- DSD_Gaussian_Static(k=3, d=5)
write_stream(dsd, "dsd_data.txt", n=100, sep=",")

# reading the same data back in
dsd2 <- DSD_ReadStream("dsd_data.txt", sep=",")
```

---

get_centers	<i>get_centers</i>
-------------	--------------------

---

**Description**

Gets the centers of micro-clusters (if available) from a DSC object

**Usage**

```
get_centers(x, ...)
```

**Arguments**

x	The DSC object the centers are being requested from.
...	Additional parameters to pass to <code>get_centers()</code> .

**Details**

Each DSC object has a unique way for returning data points, but they all are called through the generic function, `get_centers()`. This is done by using the S3 class system.

All of the `DSC_MOA` classes use the same function that first checks for the existence of micro-clusters (as deemed in the MOA framework), extracts those micro-clusters, then manually extracts the center from each of those micro-clusters and packages them into a matrix.

**Value**

A matrix of data that contains the centers of the micro-clusters (these can be either the centroids or medoids depending on the algorithm). There are some DSC classes that do not generate micro-clusters—an error will be given in these cases.

The number of columns in the matrix is dependent on the data stream that was used as input data to the DSC object, and the number of rows will differ upon the algorithm being used.

**See Also**

[DSClusterer](#), [DSC\\_CobWeb](#), [DSC\\_ClusTree](#), [DSC\\_StreamKM](#), [DSC\\_DenStream](#), [DSC\\_Clustream](#), [DSDData](#)

**Examples**

```
# setting up the objects
dsd <- DSD_Gaussian_Static()
dsc <- DSC_DenStream(initPoints=100)
cluster(dsc, dsd, 500)

# getting the centers
d <- get_centers(dsc)
```

---

get\_points

*get\_points*


---

**Description**

Gets points from a DSD object.

**Usage**

```
get_points(x, n=1, ...)
```

**Arguments**

x	The DSD object.
n	The number of data points being requested.
...	Additional parameters to pass to <code>get_points()</code>

**Details**

Each DSD object has a unique way for returning data points, but they all are called through the generic function, `get_points()`. This is done by using the S3 class system. See the man page for the specific DSD class on the semantics for each implementation of `get_points()`.

**Value**

Returns a matrix of `x$d` columns and `n` rows.

**See Also**

[DSD\\_Gaussian\\_Static](#), [DSD\\_ReadStream](#),  
[DSD\\_DataFrame](#), [DSD\\_MOA](#), [write\\_stream](#)

**Examples**

```
dsd <- DSD_Gaussian_Static()
d <- get_points(dsd, 100)
```

---

reset_stream	<i>reset_stream</i>
--------------	---------------------

---

### Description

Resets the counter in a DSD object to the beginning.

### Usage

```
reset_stream(x)
```

### Arguments

**x** Either a `DSD_ReadStream` or `DSD_DataFrame` object.

### Details

Resets the counter of the stream object that is passed in. For `DSD_ReadStream` objects, this is done by calling `seek()` on the underlying connection. For `DSD_DataFrame`, the counter stored in the environment variable is moved back to 1.

### See Also

[DSD\\_ReadStream](#), [DSD\\_DataFrame](#), [write\\_stream](#)

### Examples

```
# initializing the objects
dsd <- DSD_Gaussian_Static(k=3, d=2)
replayer <- DSD_DataFrame(get_points(dsd, 100), k=3)
dsc1 <- DSC_Clustream()
dsc2 <- DSC_CobWeb()

# clustering the same data in 2 DSC objects
cluster(dsc1, replayer, 100)
reset_stream(replayer) # resetting the dsd to its original state
cluster(dsc2, replayer, 100)
```

---

write_stream	<i>write_stream</i>
--------------	---------------------

---

### Description

Writes points to a connection from a DSD object.

### Usage

```
write_stream(dsd, con, n=100, sep=",", col.names=FALSE,
row.names=FALSE, ...)
```



**Arguments**

<code>dsd</code>	The DSD object that will generate the data points for output.
<code>con</code>	The R connection to be written to.
<code>n</code>	The number of data points to be written to the connection.
<code>sep</code>	The character that will separate attributes in a data point.
<code>col.names</code>	A flag that determines if column names will be output.
<code>row.names</code>	A flag that determines if row names will be output.
<code>...</code>	Additional parameters that are passed to <code>write.table()</code> .

**Value**

There is no value returned from this operation.

**See Also**

`write.table`, `DSD_Gaussian_Static`, `DSD_ReadStream`,  
`DSD_DataFrame`, `DSD_MOA`,

**Examples**

```
# creating data and writing it to disk
dsd <- DSD_Gaussian_Static(k=3, d=5)
write_stream(dsd, "dsd_data.txt", n=100, sep=",")
```

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