

# Package ‘jDirichletMixtureModels’

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**Title** Dirichlet Mixture Models for Clustering using Julia backend

**Version** 0.0.0.9000

**Description** This package provides utilities for clustering using Dirichlet Process mixture models using Julia for backend computation.

It supports a number of existing conjugate distribution pairs, as well as user-specified distributions. Currently, the package only supports clustering using conjugate distributions using the methods in Markov Chain Sampling Methods for Dirichlet Process Mixture Models by Radford Neal. Clustering using non-conjugate distributions is under active development.

Julia is a high-level, high-performance dynamic programming language for numerical computing. This package is based on the Julia package DirichletMixtureModels (see <https://github.com/krylea/DirichletMixtureModels.jl>), which is being co-developed alongside this package.

**URL** <https://github.com/nsdumont/jDirichletMixtureModels>

**Depends** R (>= 3.4.0)

**Imports** JuliaCall

**Suggests** data.table, ggplot2, scatterplot3d, testthat

**License** MIT + file LICENSE

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 6.0.1.9000

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aggregationData	<i>Aggregation Data</i>
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### Description

N=788, k=7, D=2

### Usage

```
data(aggregationData, package = "jDirichletMixtureModels")
```

### Format

An object of class matrix with 788 rows and 2 columns.

**Source**

Clustering basic benchmark

**References**

A. Gionis, H. Mannila, and P. Tsaparas, Clustering aggregation. ACM Transactions on Knowledge Discovery from Data (TKDD), 2007. 1(1): p. 1-30.

**Examples**

```
## Not run:
Xdata <- data(aggregationData, package = "jDirichletMixtureModels")
states <- dmm.cluster(model, Xdata)
dmm.plot(states[[1]]$labeledData)

## End(Not run)
```

---

dmm.addfile	<i>Add a Julia file to be accessible.</i>
-------------	---

---

**Description**

This function must be called before running the dmm.cluster function using a JModel (see dmm.JModel).

**Usage**

```
dmm.addfile(filename)
```

**Arguments**

filename	The name of the Julia file.
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dmm.BaseModel	<i>Create a model using built-in conjugate models</i>
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**Description**

Create an model object to be used in the dmm.cluster function, using the packages built-in conjugate models. Function dmm.model is an alternative method.

**Usage**

```
model <- dmm.BaseModel(typoname, params)
```

**Arguments**

typename	A string. The name of the predefined conjugate prior you wish to use. Options listed under details. "MultivariateNormalModel" is the default.
params	A list of the hyperparameter values for the likelihood functions. Many model have default params values and thus can be made without passing any params. See documentation for what parameters a given model may take.
data	In lieu of explicit hyperparameter values, some models can infer good hyperparameter values from given data. This option is supported for MultivariateNormalModel and UnivariateNormalModel, as well as UnivariateNormalKnownSigma.

**Details**

Built-in models available are: "MultivariateNormalModel" (default), "UnivariateNormalModel", "UnivariateNormalKnownSigma", "UnivariateExponentialModel".

**Value**

A model object of type BaseModel which can be passed to dmm.cluster.

---

dmm.benchmark	<i>To get MCMC computation times</i>
---------------	--------------------------------------

---

**Description**

This function is the same as dmm.cluster except instead of returning the states it returns the time it took to do preprocessing computations, the MCMC computation, and the postprocessing computations. Currently available for testing purposes.

**Usage**

```
dmm.benchmark(model, Xdata, alpha = 1, m_prior = 3, m_post = 3,
  iters = 5000, burnin = 200, shuffled = TRUE)
```

**Arguments**

model	An object returned by dmm.model().
Xdata	A 1D array of length N (univariate case) or 2D array of size N-by-d (multivariate case), where d is the dimensionality of the data and N is the number of observations.
alpha	A float. The concentration parameter. Default is 1.0.
m_prior	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
m_post	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
iters	An integer. Number of iterations. Default is 5000.
burnin	An integer. Amount of burn-in. Default is 200.
shuffled	A logical. Whether or not to shuffle the data. Default is true.

## Details

Performs `iters` iterations of Algorithm 2 (in conjugate case) or Algorithm 8 (in non-conjugate case) from Neal(2000) to generate possible clusters for the data in `Xdata`, using the model in `model`, with concentration parameter `alpha`. In the 1D case, `Xdata` is assumed to be a 1D array of floats. In the 2D case, `Xdata` is assumed to be a `dxN` array of floats, where the data is `d`-dimensional and `N` is the number of datapoints. Returns a dataframe of the time it took to do preprocessing computations, the MCMC computation, and the postprocessing computations.

## Value

A dataframe of the time in seconds it took to do preprocessing computations, the MCMC computation, and the postprocessing computations.

---

```
dmm.benchmark.BaseModel
```

*To get MCMC computation times*

---

## Description

This function is the same as `dmm.cluster` except instead of returning the states it returns the time it took to do preprocessing computations, the MCMC computation, and the postprocessing computations. Currently available for testing purposes.

## Usage

```
## S3 method for class 'BaseModel'
dmm.benchmark(model, Xdata, alpha = 1, iters = 5000,
  burnin = 200, shuffled = TRUE)
```

## Arguments

<code>model</code>	An object returned by <code>dmm.model()</code> .
<code>Xdata</code>	A 1D array of length <code>N</code> (univariate case) or 2D array of size <code>N-by-d</code> (multivariate case), where <code>d</code> is the dimensionality of the data and <code>N</code> is the number of observations.
<code>alpha</code>	A float. The concentration parameter. Default is 1.0.
<code>iters</code>	An integer. Number of iterations. Default is 5000.
<code>burnin</code>	An integer. Amount of burn-in. Default is 200.
<code>shuffled</code>	A logical. Whether or not to shuffle the data. Default is true.
<code>m_prior</code>	An integer. Optionally paramter only used in non-conjugate case. Default is 3.
<code>m_post</code>	An integer. Optionally paramter only used in non-conjugate case. Default is 3.

## Details

Performs `iters` iterations of Algorithm 2 (in conjugate case) or Algorithm 8 (in non-conjugate case) from Neal(2000) to generate possible clusters for the data in `Xdata`, using the model in `model`, with concentration parameter `alpha`. In the 1D case, `Xdata` is assumed to be a 1D array of floats. In the 2D case, `Xdata` is assumed to be a `dxN` array of floats, where the data is `d`-dimensional and `N` is the number of datapoints. Returns a dataframe of the time it took to do preprocessing computations, the MCMC computation, and the postprocessing computations.

**Value**

A dataframe of the time in seconds it took to do preprocessing computations, the MCMC computation, and the postprocessing computations.

---

```
dmm.benchmark.JConjugateModel
```

*To get MCMC computation times*

---

**Description**

This function is the same as `dmm.cluster` except instead of returning the states it returns the time it took to do preprocessing computations, the MCMC computation, and the postprocessing computations. Currently available for testing purposes.

**Usage**

```
## S3 method for class 'JConjugateModel'
dmm.benchmark(model, Xdata, alpha = 1,
  m_prior = 3, m_post = 3, iters = 5000, burnin = 200,
  shuffled = TRUE)
```

**Arguments**

<code>model</code>	An object returned by <code>dmm.model()</code> .
<code>Xdata</code>	A 1D array of length <code>N</code> (univariate case) or 2D array of size <code>N</code> -by- <code>d</code> (multivariate case), where <code>d</code> is the dimensionality of the data and <code>N</code> is the number of observations.
<code>alpha</code>	A float. The concentration parameter. Default is 1.0.
<code>m_prior</code>	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
<code>m_post</code>	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
<code>iters</code>	An integer. Number of iterations. Default is 5000.
<code>burnin</code>	An integer. Amount of burn-in. Default is 200.
<code>shuffled</code>	A logical. Whether or not to shuffle the data. Default is true.

**Details**

Performs `iters` iterations of Algorithm 2 (in conjugate case) or Algorithm 8 (in non-conjugate case) from Neal(2000) to generate possible clusters for the data in `Xdata`, using the model in `model`, with concentration parameter `alpha`. In the 1D case, `Xdata` is assumed to be a 1D array of floats. In the 2D case, `Xdata` is assumed to be a `dxN` array of floats, where the data is `d`-dimensional and `N` is the number of datapoints. Returns a dataframe of the time it took to do preprocessing computations, the MCMC computation, and the postprocessing computations.

**Value**

A dataframe of the time in seconds it took to do preprocessing computations, the MCMC computation, and the postprocessing computations.

---

dmm.benchmark.JNonConjugateModel

*To get MCMC computation times*


---

## Description

This function is the same as `dmm.cluster` except instead of returning the states it returns the time it took to do preprocessing computations, the MCMC computation, and the postprocessing computations. Currently available for testing purposes.

## Usage

```
## S3 method for class 'JNonConjugateModel'
dmm.benchmark(model, Xdata, alpha = 1,
  m_prior = 3, m_post = 4, iters = 5000, burnin = 200,
  shuffled = TRUE)
```

## Arguments

<code>model</code>	An object returned by <code>dmm.model()</code> .
<code>Xdata</code>	A 1D array of length <code>N</code> (univariate case) or 2D array of size <code>N</code> -by- <code>d</code> (multivariate case), where <code>d</code> is the dimensionality of the data and <code>N</code> is the number of observations.
<code>alpha</code>	A float. The concentration parameter. Default is 1.0.
<code>m_prior</code>	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
<code>m_post</code>	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
<code>iters</code>	An integer. Number of iterations. Default is 5000.
<code>burnin</code>	An integer. Amount of burn-in. Default is 200.
<code>shuffled</code>	A logical. Whether or not to shuffle the data. Default is true.

## Details

Performs `iters` iterations of Algorithm 2 (in conjugate case) or Algorithm 8 (in non-conjugate case) from Neal(2000) to generate possible clusters for the data in `Xdata`, using the model in `model`, with concentration parameter `alpha`. In the 1D case, `Xdata` is assumed to be a 1D array of floats. In the 2D case, `Xdata` is assumed to be a `dxN` array of floats, where the data is `d`-dimensional and `N` is the number of datapoints. Returns a dataframe of the time it took to do preprocessing computations, the MCMC computation, and the postprocessing computations.

## Value

A dataframe of the time in seconds it took to do preprocessing computations, the MCMC computation, and the postprocessing computations.

---

dmm.cluster	<i>Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.</i>
-------------	--

---

## Description

Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.

## Usage

```
dmm.cluster(model, Xdata, alpha=1.0, m_prior=3, m_post=3, iters=5000, burnin=200, shuffled=TRUE)
```

## Arguments

model	An object returned by <code>dmm.model()</code> .
Xdata	A 1D array of length N (univariate case) or 2D array of size N-by-d (multivariate case), where d is the dimensionality of the data and N is the number of observations.
alpha	A float. The concentration parameter. Default is 1.0.
m_prior	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
m_post	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
iters	An integer. Number of iterations. Default is 5000.
burnin	An integer. Amount of burn-in. Default is 200.
shuffled	A logical. Whether or not to shuffle the data. Default is true.

## Details

Performs `iters` iterations of Algorithm 2 (in conjugate case) or Algorithm 8 (in non-conjugate case) from Neal(2000) to generate possible clusters for the data in `Xdata`, using the model in `model`, with concentration parameter `alpha`. In the 1D case, `Xdata` is assumed to be a 1D array of floats. In the 2D case, `Xdata` is assumed to be a `dxN` array of floats, where the data is `d`-dimensional and `N` is the number of datapoints. Returns a list of states. The elements of the list are all states post-burnin iteration, with the default being a burnin of 200. By default, this array is shuffled so that it may be used to approximate I.I.D draws from the posterior.

A single state from the returned list of states has fields `data` and `clusters`. `data` is a dataframe consisting of the `Xdata` and their cluster labels. `clusters` is a `data.table` (if the user has the `data.table` package loaded) or a list.

If `clusters` is a `data.table`, each row refers to a cluster. Columns are the cluster label, the population, and the rest of the columns are parameters.

If `clusters` is a list, each element of the list refers to a cluster, `clusters[[i]]` is a list containing of the above information for cluster `i` as elements. Each single item in `clusters` is a list with fields `cluster`, `population`, and `params`. E.g. `clusters[[1]]$population` is the population of cluster 1. The `params` field (`clusterInfo[[i]]$params`) is itself a list of each of the parameters

To see a formatted summary of all the clusters in a given state use the `dmm.summarize(clusters)` function.

To see a plot of the labeled data in a given state use the `dmm.plot(data)` function.



**Value**

A list of states (i.e. `state = states[[i]]`). A state is itself a list. A state has two fields: `labeledData` and `clusterInfo`.

`labeledData` is a `data.frame` of the `Xdata` data points and their cluster labels. `clusterInfo` is either a list or a `data.table` (if the `data.table` package is loaded by the user). It contains (1) cluster labels, (2) the number of data points (i.e. population) of each cluster, and (3) all of the parameters for each cluster.

**Examples**

```
dmm.setup()
model <- dmm.BaseModel(data = Xdata)
states <- dmm.cluster(model, Xdata)
```

---

`dmm.cluster.BaseModel`    *Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.*

---

**Description**

Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.

**Usage**

```
dmm.cluster.BaseModel(model, Xdata, alpha=1.0, iters=5000, burnin=200, shuffled=TRUE)
```

**Arguments**

<code>model</code>	An object returned by <code>dmm.model()</code> .
<code>Xdata</code>	A 1D array of length <code>N</code> (univariate case) or 2D array of size <code>N-by-d</code> (multivariate case), where <code>d</code> is the dimensionality of the data and <code>N</code> is the number of observations.
<code>alpha</code>	A float. The concentration parameter. Default is 1.0.
<code>iters</code>	An integer. Number of iterations. Default is 5000.
<code>burnin</code>	An integer. Amount of burn-in. Default is 200.
<code>shuffled</code>	A logical. Whether or not to shuffled the data. Default is true.
<code>m_prior</code>	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
<code>m_post</code>	An integer. Optionally parameter only used in non-conjugate case. Default is 3.

**Details**

Performs `iters` iterations of Algorithm 2 (in conjugate case) or Algorithm 8 (in non-conjugate case) from Neal(2000) to generate possible clusters for the data in `Xdata`, using the model in `model`, with concentration parameter `alpha`. In the 1D case, `Xdata` is assumed to be a 1D array of floats. In the 2D case, `Xdata` is assumed to be a `dxN` array of floats, where the data is `d`-dimensional and `N` is the number of datapoints. Returns a list of states. The elements of the list are all states post-burnin iteration, with the default being a burnin of 200. By default, this array is shuffled so that it may be used to approximate I.I.D draws from the posterior.

A single state from the returned list of states has fields `data` and `clusters`. `data` is a dataframe consisting of the `Xdata` and their cluster labels. `clusters` is a `data.table` (if the user has the `data.table` package loaded) or a list.

If `clusters` is a `data.table`, each row refers to a cluster. Columns are the cluster label, the population, and the rest of the columns are parameters.

If `clusters` is a list, each element of the list refers to a cluster, `clusters[[i]]` is a list containing of the above information for cluster `i` as elements. Each single item in `clusters` is a list with fields `cluster`, `population`, and `params`. E.g. `clusters[[1]]$population` is the population of cluster 1. The `params` field (`clusters[[i]]$params`) is itself a list of each of the parameters

To see a formatted summary of all the clusters in a given state use the `dmm.summarize(clusters)` function.

To see a plot of the labeled data in a given state use the `dmm.plot(data)` function.

### Value

A list of states (i.e. `state = states[[i]]`). A state is itself a list. A state has two fields: `data` and `clusters`.

`data` is a `data.frame` of the `Xdata` data points and their cluster labels. `clusters` is either a list or a `data.table` (if the `data.table` package is loaded by the user). It contains (1) cluster labels, (2) the number of data points (i.e. population) of each cluster, and (3) all of the parameters for each cluster.

### Examples

```
dmm.setup()
model <- dmm.BaseModel(data = Xdata)
states <- dmm.cluster(model, Xdata)
```

---

`dmm.cluster.JConjugateModel`

*Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.*

---

### Description

Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.

### Usage

```
## S3 method for class 'JConjugateModel'
dmm.cluster(model, Xdata, alpha = 1, m_prior = 3,
  m_post = 3, iters = 5000, burnin = 200, shuffled = TRUE)
```

### Arguments

<code>model</code>	An object returned by <code>dmm.model()</code> .
<code>Xdata</code>	A 1D array of length <code>N</code> (univariate case) or 2D array of size <code>N-by-d</code> (multivariate case), where <code>d</code> is the dimensionality of the data and <code>N</code> is the number of observations.
<code>alpha</code>	A float. The concentration parameter. Default is 1.0.

m_prior	An integer. Optionally paramter only used in non-conjugate case. Default is 3.
m_post	An integer. Optionally paramter only used in non-conjugate case. Default is 3.
iters	An integer. Number of iterations. Default is 5000.
burnin	An integer. Amount of burn-in. Default is 200.
shuffled	A logical. Whether or not to shuffled the data. Default is true.

## Details

Performs `iters` iterations of Algorithm 2 (in conjugate case) or Algorithm 8 (in non-conjugate case) from Neal(2000) to generate possible clusters for the data in `Xdata`, using the model in `model`, with concentration parameter `alpha`. In the 1D case, `Xdata` is assumed to be a 1D array of floats. In the 2D case, `Xdata` is assumed to be a `dxN` array of floats, where the data is `d`-dimensional and `N` is the number of datapoints. Returns a list of states. The elements of the list are all states post-burnin iteration, with the default being a burnin of 200. By default, this array is shuffled so that it may be used to approximate I.I.D draws from the posterior.

A single state from the returned list of states has fields `data` and `clusters`. `data` is a dataframe consisting of the `Xdata` and their cluster labels. `clusters` is a `data.table` (if the user has the `data.table` package loaded) or a list.

If `clusters` is a `data.table`, each row refers to a cluster. Columns are the cluster label, the population, and the rest of the columns are parameters.

If `clusters` is a list, each element of the list refers to a cluster, `clusters[[i]]` is a list containing of the above information for cluster `i` as elements. Each single item in `clusters` is a list with fields `cluster`, `population`, and `params`. E.g. `clusters[[1]]$population` is the population of cluster 1. The `params` field (`clusters[[i]]$params`) is itself a list of each of the parameters

To see a formatted summary of all the clusters in a given state use the `dmm.summarize(clusters)` function.

To see a plot of the labeled data in a given state use the `dmm.plot(data)` function.

## Value

A list of states (i.e. `state = states[[i]]`). A state is itself a list. A state has two fields: `data` and `clusters`.

`data` is a `data.frame` of the `Xdata` data points and their cluster labels. `clusters` is either a list or a `data.table` (if the `data.table` package is loaded by the user). It contains (1) cluster labels, (2) the number of data points (i.e. population) of each cluster, and (3) all of the parameters for each cluster.

## Examples

```
dmm.setup()
dmm.addfile(filename)
model <- dmm.JConjugateModel(pdf_name, sample_name, marg_name, params)
states <- dmm.cluster(model, Xdata)
```

---

dmm.cluster.JNonConjugateModel

*Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.*

---

## Description

Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.

## Usage

```
## S3 method for class 'JNonConjugateModel'
dmm.cluster(model, Xdata, alpha = 1,
  m_prior = 3, m_post = 4, iters = 5000, burnin = 200,
  shuffled = TRUE)
```

## Arguments

model	An object returned by <code>dmm.model()</code> .
Xdata	A 1D array of length N (univariate case) or 2D array of size N-by-d (multivariate case), where d is the dimensionality of the data and N is the number of observations.
alpha	A float. The concentration parameter. Default is 1.0.
m_prior	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
m_post	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
iters	An integer. Number of iterations. Default is 5000.
burnin	An integer. Amount of burn-in. Default is 200.
shuffled	A logical. Whether or not to shuffle the data. Default is true.

## Details

Performs `iters` iterations of Algorithm 2 (in conjugate case) or Algorithm 8 (in non-conjugate case) from Neal(2000) to generate possible clusters for the data in `Xdata`, using the model in `model`, with concentration parameter `alpha`. In the 1D case, `Xdata` is assumed to be a 1D array of floats. In the 2D case, `Xdata` is assumed to be a `dxN` array of floats, where the data is `d`-dimensional and `N` is the number of datapoints. Returns a list of states. The elements of the list are all states post-burnin iteration, with the default being a burnin of 200. By default, this array is shuffled so that it may be used to approximate IID draws from the posterior.

A single state from the returned list of states has fields `data` and `clusters`. `data` is a dataframe consisting of the `Xdata` and their cluster labels. `clusters` is a `data.table` (if the user has the `data.table` package loaded) or a list.

If `clusters` is a `data.table`, each row refers to a cluster. Columns are the cluster label, the population, and the rest of the columns are parameters.

If `clusters` is a list, each element of the list refers to a cluster, `clusters[[i]]` is a list containing of the above information for cluster `i` as elements. Each single item in `clusters` is a list with fields `cluster`, `population`, and `params`. E.g. `clusters[[1]]$population` is the population of cluster 1. The `params` field (`clusters[[i]]$params`) is itself a list of each of the parameters

To see a formatted summary of all the clusters in a given state use the `dmm.summarize(clusters)` function.

To see a plot of the labeled data in a given state use the `dmm.plot(data)` function.

## Value

A list of states (i.e. `state = states[[i]]`). A state is itself a list. A state has two fields: `data` and `clusters`.

`data` is a `data.frame` of the `Xdata` data points and their cluster labels. `clusters` is either a list or a `data.table` (if the `data.table` package is loaded by the user). It contains (1) cluster labels, (2) the number of data points (i.e. population) of each cluster, and (3) all of the parameters for each cluster.

## Examples

```
dmm.setup()
dmm.addfile(filename)
model <- dmm.JNonConjugateModel(pdf_name, sample_name, params)
states <- dmm.cluster(model, Xdata)
```

---

<code>dmm.cluster.RModel</code>	<i>Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.</i>
---------------------------------	--

---

## Description

Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.

## Usage

```
## S3 method for class 'RModel'
dmm.cluster(model, Xdata, alpha = 1, m_prior = 3,
  m_post = 3, iters = 5000, burnin = 200, shuffled = TRUE)
```

## Arguments

<code>model</code>	An object returned by <code>dmm.model()</code> .
<code>Xdata</code>	A 1D array of length <code>N</code> (univariate case) or 2D array of size <code>N-by-d</code> (multivariate case), where <code>d</code> is the dimensionality of the data and <code>N</code> is the number of observations. Use a Dirichlet Mixture Model on data to get cluster labels and cluster parameter values.
<code>alpha</code>	A float. The concentration parameter. Default is 1.0.
<code>m_prior</code>	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
<code>m_post</code>	An integer. Optionally parameter only used in non-conjugate case. Default is 3.
<code>iters</code>	An integer. Number of iterations. Default is 5000.
<code>burnin</code>	An integer. Amount of burn-in. Default is 200.
<code>shuffled</code>	A logical. Whether or not to shuffle the data. Default is true.
<code>model</code>	An object returned by <code>dmm.model()</code> .
<code>Xdata</code>	A 1D array of length <code>N</code> (univariate case) or 2D array of size <code>N-by-d</code> (multivariate case), where <code>d</code> is the dimensionality of the data and <code>N</code> is the number of observations.

## Details

Performs `iters` iterations of Algorithm 2 (in conjugate case) or Algorithm 8 (in non-conjugate case) from Neal(2000) to generate possible clusters for the data in `Xdata`, using the model in `model`, with concentration parameter `alpha`. In the 1D case, `Xdata` is assumed to be a 1D array of floats. In the 2D case, `Xdata` is assumed to be a `dxN` array of floats, where the data is `d`-dimensional and `N` is the number of datapoints. Returns a list of states. The elements of the list are all states post-burnin iteration, with the default being a burnin of 200. By default, this array is shuffled so that it may be used to approximate I.I.D draws from the posterior.

A single state from the returned list of states has fields `data` and `clusters`. `data` is a dataframe consisting of the `Xdata` and their cluster labels. `clusters` is a `data.table` (if the user has the `data.table` package loaded) or a list.

If `clusters` is a `data.table`, each row refers to a cluster. Columns are the cluster label, the population, and the rest of the columns are parameters.

If `clusters` is a list, each element of the list refers to a cluster, `clusters[[i]]` is a list containing of the above information for cluster `i` as elements. Each single item in `clusters` is a list with fields `cluster`, `population`, and `params`. E.g. `clusters[[1]]$population` is the population of cluster 1. The `params` field (`clusters[[i]]$params`) is itself a list of each of the parameters

To see a formatted summary of all the clusters in a given state use the `dmm.summarize(clusters)` function.

To see a plot of the labeled data in a given state use the `dmm.plot(data)` function.

## Value

A list of states (i.e. `state = states[[i]]`). A state is itself a list. A state has two fields: `data` and `clusters`.

`data` is a `data.frame` of the `Xdata` data points and their cluster labels. `clusters` is either a list or a `data.table` (if the `data.table` package is loaded by the user). It contains (1) cluster labels, (2) the number of data points (i.e. population) of each cluster, and (3) all of the parameters for each cluster.

---

dmm.JConjugateModel	Create a conjugate model using Julia functions
---------------------	--

---

## Description

Create an model object to be used in the `dmm.cluster` function, using user given Julia functions. Must call `dmm.addfile` to import files, in which the Julia functions are stored, before using `dmm.cluster` on a `JModel`. Functions `dmm.JModel` and `dmm.model` are alternatives.

## Usage

```
dmm.addfile(filename)
model <- dmm.JConjugateModel(pdf_name, sample_name, marg_name, params)
```

## Arguments

pdf_name	A string. The name of the Julia function in <code>filename</code> that returns the probability density function likelihood. The function should be of the form <code>pdf_name(y::Float64, theta::Tuple)</code> or <code>pdf_name(y::Array{Float64,1}, theta::Tuple, params::Tuple)</code> .
----------	---

sample_name	A string. The name of the Julia function in filename that returns the sample posterior function. The function should be of the form <code>sample_name(y::Float64, params::Tuple)</code> , <code>sample_name(y::Array{Float64,1}, params::Tuple)</code> or <code>sample_name(y::Array{Float64,2}, params::Tuple)</code> .
marg_name	A string. The name of the Julia function in filename that returns the marginal likelihood. The function should be of the form <code>marg_name(y::Float64, params::Tuple)</code> .
params	A list of all hyperparameters needed for the above three functions.
isconjugate	A logical. TRUE (default) if the user specified model is conjugate, FALSE if not.

### Value

A model object of type JModel which can be passed to `dmm.cluster`.

### Examples

```
dmm.addfile(filename)
# The following all make conjugate models using Julia functions
model <- dmm.model(pdf_name, sample_name, marg_name, params, isconjugate=TRUE)
model <- dmm.Jmodel(pdf_name, sample_name, marg_name, params, isconjugate=TRUE)
model <- dmm.JConjugateModel(pdf_name, sample_name, marg_name, params)
```

---

dmm.JModel	<i>Create a model using Julia functions</i>
------------	---

---

### Description

Create an model object to be used in the `dmm.cluster` function, using user given Julia functions. Must call `dmm.addfile` to import files, in which the Julia functions are stored, before using `dmm.cluster` on a JModel. Functions `dmm.JConjugateModel` and `dmm.JNonConjugateModel` are alternatives.

### Usage

```
\code{dmm.addfile(filename)}
\code{model <- dmm.model(pdf_name, sample_name, marg_name, params, isconjugate=TRUE)}
\code{model <- dmm.Jmodel(pdf_name, sample_name, marg_name, params, isconjugate=TRUE)}
\code{model <- dmm.JConjugateModel(pdf_name, sample_name, marg_name, params)}
\code{model <- dmm.JNonConjugateModel(pdf_name, sample_name, params)}
```

### Arguments

pdf_name	A string. The name of the Julia function in filename that returns the probability density function likelihood. The function should be of the form <code>pdf_name(y::Float64, theta::Tuple)</code> or <code>pdf_name(y::Array{Float64,1}, theta::Tuple, params::Tuple)</code> .
sample_name	A string. The name of the Julia function in filename that returns the sample posterior function for conjugate case or the sample prior for nonconjugate case. The function should be of the form <code>sample_name(y::Float64, params::Tuple)</code> , <code>sample_name(y::Array{Float64,1}, params::Tuple)</code> or <code>sample_name(y::Array{Float64,2}, params::Tuple)</code> .
marg_name	A string. For conjugate case only. The name of the Julia function in filename that returns the marginal likelihood. The function should be of the form <code>marg_name(y::Float64, params::Tuple)</code> .
params	A list of all hyperparameters needed for the above three functions.
isconjugate	A logical. TRUE (default) if the user specified model is conjugate, FALSE if not.

**Details**

marg\_name is only required for conjugate models.

**Value**

A model object of type JModel which can be passed to dmm.cluster.

**Examples**

```
dmm.addfile(filename)
# The following all make models using Julia functions
model <- dmm.model(pdf_name, sample_name, marg_name, params, isconjugate=TRUE)
model <- dmm.Jmodel(pdf_name, sample_name, marg_name, params, isconjugate=TRUE)
model <- dmm.JConjugateModel(pdf_name, sample_name, marg_name, params)
model <- dmm.JNonConjugateModel(pdf_name, sample_name, params)
```

---

```
dmm.JNonConjugateModel
```

*Create a nonconjugate model using Julia functions*

---

**Description**

Create an model object to be used in the dmm.cluster function, using user given Julia functions. Must call dmm.addfile to import files, in which the Julia functions are stored, before using dmm.cluster on a JModel. Functions dmm.JModel and dmm.model are alternatives..

**Usage**

```
dmm.addfile(filename)
model <- dmm.model(pdf_name, sample_name, marg_name, params, isconjugate=TRUE)
```

**Arguments**

pdf_name	A string. The name of the Julia function in filename that returns the probability density function likelihood. The function should be of the form pdf_name(y::Float64, theta::Tuple) or pdf_name(y::Array{Float64,1}, theta::Tuple, params::Tuple).
sample_name	A string. The name of the Julia function in filename that returns the sample prior. The function should be of the form sample_name(y::Float64, params::Tuple), sample_name(y::Array{Float64,1}, params::Tuple) or sample_name(y::Array{Float64,2}, params::Tuple).
params	A list of all hyperparameters needed for the above three functions.

**Value**

A model object of type JModel which can be passed to dmm.cluster.



**Examples**

```
dmm.addfile(filename)
# The following all make nonconjugate models using Julia functions
model <- dmm.model(pdf_name, sample_name, params, isconjugate=FALSE)
model <- dmm.Jmodel(pdf_name, sample_name, params, isconjugate=FALSE)
model <- dmm.JNonConjugateModel(pdf_name, sample_name, params)
```

---

dmm.model

---

*Create a model*


---

**Description**

Make a model object to be passed to `dmm.cluster`. If not passed any arguments a conjugate multivariate normal likelihood model with default parameters will be used.

**Usage**

```
dmm.model(...)
```

**Details**

Depending on what arguments are passed to `dmm.model()` either a `BaseModel` (see `dmm.BaseModel`), a `RModel` (see `dmm.RModel`), or `JModel` (see `dmm.JModel`) will be made.

\*NOTE: The use of `RModels` is not available in the current version of `jDirichletMixtureModels`\*

**Value**

A model object which can be passed to `dmm.cluster`.

**Examples**

```
## Not run:
# Example of using a multivariate normal conjugate prior (the default)
model <- dmm.model()

# Example of using a built-in conjugate prior
model <- dmm.model(tyename, params)

# Using a user specified R functions
model <- dmm.model(pdf_fct, sample_fct, marg_fct, params, isconjugate=TRUE)

# Using a user specified Julia functions located in a .jl file called filename
dmm.addfile(filename)
model <- dmm.model(pdf_name, sample_name, marg_name, params, isconjugate=TRUE)

## End(Not run)
```

---

dmm.model.character	<i>Create a model object</i>
---------------------	------------------------------

---

### Description

If a string is passed as the first input to `dmm.model` it will create and return either a `BaseModel` (if passed one string, plus an optional `params`) (see `dmm.BaseModel`), or a `JModel` (see `dmm.JModel`)

### Usage

```
## S3 method for class 'character'
dmm.model(...)
```

---

dmm.model.function	<i>Create a model object</i>
--------------------	------------------------------

---

### Description

If a function is passed as the first input to `dmm.model` it will create and return an `Rmodel` (see `dmm.RModel`).

### Usage

```
## S3 method for class 'function'
dmm.model(...)
```

---

dmm.model.NULL	<i>Create a model object</i>
----------------	------------------------------

---

### Description

If nothing is passed to `dmm.model` it will create and return a `BaseModel` using a multivariate normal conjugate model with default parameters (see `dmm.BaseModel`).

### Usage

```
## S3 method for class 'NULL'
dmm.model(...)
```

dmm.plot

*Plot labeledData for a state returned by dmm***Description**

Plot labeledData for a state returned by dmm

**Usage**

```
dmm.plot(labeledData)
```

Given the data from a single state returned by `dmm.cluster(...)`, plot it. Can do 2D, 1D, or 3D plots. `\code{ggplot2}` recommended for 2D plots. `\code{scatterplot3d}` required for 3D plots.

**Arguments**

`labeledData`      The data from a single state. Given `states <- dmm.cluster(...)`, this input is `states[[i]]$data`.

**Examples**

```
states <- dmm.cluster(model,Xdata,...)
dmm.plot(states[[1]]$data)
```

dmm.RModel

*Create a model using R functions***Description**

*\*NOTE: This function is not available in the current version of jDirichletMixtureModels.\** Create an model object to be used in the `dmm.cluster` function, using user given R functions.

**Usage**

```
dmm.RModel(pdf_func, sample_func = NULL, marg_func = NULL, params,
            isconjugate = TRUE)
```

**Arguments**

`pdf_func`      A function that takes as input (`data`, `likelihoodparams`, `params`). It returns the value of the probability density function likelihood at (`data`, `likelihoodparams`) given `params`.

`params`      A list of all hyperparameters needed for the above three functions.

`isconjugate`    A logical. TRUE (default) if the user specified model is conjugate, FALSE if not.

`sample_fct`    Optional for nonconjugate models. A function takes as input (`data`, `params`). It returns the value of the sample posterior function at data given `params`.

`marg_fct`      Only needed for conjugate models. A function takes as input (`data`, `params`). It returns the value of the marginal likelihood function at data given `params`.

**Details**

```
@usage model <- dmm.Rmodel(pdf_fct, sample_fct, marg_fct, params, isconjugate=TRUE)
```

**Value**

A model object of type RModel which can be passed to `dmm.cluster`.

```
##@export
```

---

```
dmm.setup
```

*Do initial setup for jDirichletMixtureModels package.*

---

**Description**

`dmm.setup` does the initial setup for `jDirichletMixtureModels` package.

**Usage**

```
dmm.setup()
```

**Arguments**

```
...
```

arguments passed to `JuliaCall::julia_setup`. Most notable: the first input should be the path to the folder in which the Julia bin file is located on the user's machine. It will default to using the macOS path for the most recent Julia version.

**Examples**

```
## Not run:
dmm.setup()

## End(Not run)
```

---

```
dmm.state
```

*Formats a single DMM OutputState Julia object into a list of labeled data and list of cluster info*

---

**Description**

Formats a single DMM OutputState Julia object into a list of labeled data and list of cluster info

**Usage**

```
dmm.state(juliastate, paramnames)
```

---

dmm.stateAsTable	<i>Formats a single DMM OutputState Julia object into a list of labeled data and data.table of cluster info</i>
------------------	---

---

### Description

Formats a single DMM OutputState Julia object into a list of labeled data and data.table of cluster info

### Usage

```
dmm.stateAsTable(juliastate, paramnames)
```

---

dmm.states	<i>Constructing list of all states from dmm.cluster run (excluding burnin).</i>
------------	---

---

### Description

Constructing list of all states from dmm.cluster run (excluding burnin).

### Usage

```
dmm.states(juliastates, paramnames = NULL)
```

### Arguments

juliastates	The object returned by Julia code
paramnames	Optionally. A list of the parameter names. Returned by Julia code for most built-in models.

### Details

Each item in the list (i.e. `state = states[[i]]`) is a state. A state is also a list. A state has two fields: `data` and `clusters`.

`data` is a `data.frame` of the data points and their cluster labels. `clusters` is either a list or a `data.table` (if the `data.table` package is loaded by the user). It contains (1) cluster labels, (2) the number of data points (i.e. population) of each cluster, and (3) all of the parameters for each cluster.

If `clusters` is a `data.table`, each row refers to a cluster. Columns are the cluster label, the population, and the rest of the columns are parameters.

If `clusters` is a list, each element of the list refers to a cluster, `clusters[[i]]` is a list containing of the above information for cluster `i` as elements. E.g. `clusters[[1]]$population` is the population of cluster 1. The `params` field (`clusters[[i]]$params`) is itself a list of each of the parameters

---

dmm.summarize	<i>Summerize given a single state's cluster info</i>
---------------	--

---

**Description**

Summerize given a single state's cluster info

**Usage**

```
dmm.summarize(clusterInfo)
```

**Arguments**

clusterInfo      A list or data.table. Given states <- dmm.cluster(...), this input is states[[i]]\$clusters

**Examples**

```
states <- dmm.cluster(model, data)
dmm.summarize(states[[1]]$clusters)
```

---

dmm.summarize.data.table	<i>Summerize given a single state's cluster info</i>
--------------------------	--

---

**Description**

Summerize given a single state's cluster info

**Usage**

```
dmm.summarize(clusterInfo)
```

**Arguments**

clusterInfo      A data.table. Given states <- dmm.cluster(...), this input is states[[i]]\$clusters

**Examples**

```
states <- dmm.cluster(model, data)
dmm.summarize(states[[1]]$clusters)
```

---

dmm.summarize.list	<i>Summerize given a single state's cluster info</i>
--------------------	--

---

**Description**

Summerize given a single state's cluster info

**Usage**

```
dmm.summarize(clusterInfo)
```

**Arguments**

clusterInfo      A list. Given states <- dmm.cluster(...), this input is states[[i]]\$clusters

**Examples**

```
states <- dmm.cluster(model, data)
dmm.summarize(states[[1]]$clusters)
```

---

mouse	<i>Mouse Data</i>
-------	-------------------

---

**Description**

Clusters that look like Mickey mouse's head.

**Usage**

```
data(mouse, package = "jDirichletMixtureModels")
```

**Format**

An object of class matrix with 500 rows and 2 columns.

**Source**

[mouse.csv](#)

**Examples**

```
## Not run:
Xdata <- data(mouse, package = "jDirichletMixtureModels")
states <- dmm.cluster(model, Xdata)
dmm.plot(states[[1]]$labeledData)

## End(Not run)
```

---

syntheticGaussian1	<i>Synthetic Gaussian Cluster Data 1</i>
--------------------	--

---

**Description**

Synthetic 2-d data with N=5000 vectors and k=15 Gaussian clusters with different degree of cluster overlapping

**Usage**

```
data(syntheticGaussian1, package = "jDirichletMixtureModels")
```

**Format**

An object of class matrix with 5000 rows and 2 columns.

**Source**

Clustering basic benchmark

**References**

P. Fränti and O. Virtajoki, "Iterative shrinking method for clustering problems", Pattern Recognition, 39 (5), 761-765, May 2006.

**Examples**

```
## Not run:
Xdata <- data(syntheticGaussian1, package = "jDirichletMixtureModels")
states <- dmm.cluster(model, Xdata)
dmm.plot(states[[1]]$labeledData)

## End(Not run)
```

---

syntheticGaussian2	<i>Synthetic Gaussian Cluster Data 2</i>
--------------------	--

---

**Description**

Synthetic 2-d data with N=5000 vectors and k=15 Gaussian clusters with different degree of cluster overlapping

**Usage**

```
data(syntheticGaussian2, package = "jDirichletMixtureModels")
```

**Format**

An object of class matrix with 5000 rows and 2 columns.



**Source**

Clustering basic benchmark

**References**

P. Fränti and O. Virtajoki, "Iterative shrinking method for clustering problems", Pattern Recognition, 39 (5), 761-765, May 2006.

**Examples**

```
## Not run:  
Xdata <- data(syntheticGaussian2, package = "jDirichletMixtureModels")  
states <- dmm.cluster(model, Xdata)  
dmm.plot(states[[1]]$labeledData)  
  
## End(Not run)
```

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