

# Package ‘FuzzyDBScan’

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**Title** Run and Predict a Fuzzy DBScan

**Version** 0.0.3

**Description** An interface for training Fuzzy DBScan with both Fuzzy Core and Fuzzy Border. Therefore, the package provides a method to initialize and run the algorithm and a function to predict new data w.t.h. of 'R6'.  
The package is build upon the paper ``Fuzzy Extensions of the DBScan algorithm" from Ienco and Bordogna (2018) <[doi:10.1007/s00500-016-2435-0](https://doi.org/10.1007/s00500-016-2435-0)>.  
A predict function assigns new data according to the same criteria as the algorithm itself. However, the prediction function freezes the algorithm to preserve the trained cluster structure and treats each new prediction object individually.

**License** LGPL-3

**Depends** R (>= 4.0.0)

**Imports** ggplot2, R6, data.table, dbscan, checkmate

**Suggests** testthat (>= 3.0.0), rmarkdown, factoextra, spelling

**Encoding** UTF-8

**RoxygenNote** 7.2.2

**Language** en-US

**NeedsCompilation** no

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Fuzzy\_DBScan

*Fuzzy DBScan*

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## Description

This object implements fuzzy DBScan with both, fuzzy cores and fuzzy borders. Additionally, it provides a predict function.

## Details

A method to initialize and run the algorithm and a function to predict new data. The package is build upon the paper "Fuzzy Extensions of the DBScan algorithm" from Ienco and Bordogna. The predict function assigns new data based on the same criteria as the algorithm itself. However, the prediction function freezes the algorithm to preserve the trained cluster structure and treats each new prediction object individually. Note, that border points are included to the cluster.

## Public fields

`data` [data.frame](#) | [matrix](#)

The data to be clustered by the algorithm. Allowed are only [numeric](#) columns.

`eps` [numeric](#)

The size (radius) of the epsilon neighborhood. If the radius contains 2 numbers, the fuzzy cores are calculated between the minimum and the maximum radius. If epsilon is a single number, the algorithm loses the fuzzy core property. If the length of `pts` is also 1L, the algorithm equals to non-fuzzy DBScan.

`pts` [numeric](#)

number of maximum and minimum points required in the `eps` neighborhood for core points (excluding the point itself). If the length of the argument is 1, the algorithm loses its fuzzy border property. If the length of `eps` is also 1L, the algorithm equals to non-fuzzy DBScan.

`clusters` [factor](#)

Contains the assigned clusters per observation in the same order as in `data`.

`dense` [numeric](#)

Contains the assigned density estimates per observation in the same order as in `data`.

`point_def` [character](#)

Contains the assigned definition estimates per observation in the same order as in `data`. Possible are "Core Point", "Border Point" and "Noise".

`results` [data.table](#)

A table where each column indicates for the probability of the new data to belong to a respective cluster.

## Methods

### Public methods:

- [FuzzyDBScan\\$new\(\)](#)
- [FuzzyDBScan\\$predict\(\)](#)

- `FuzzyDBScan$plot()`
- `FuzzyDBScan$clone()`

**Method** `new()`: Create a FuzzyDBScan object. Apply the fuzzy DBScan algorithm given the data `dta`, the range of the radius `eps` and the range of the Points `pts`.

*Usage:*

```
FuzzyDBScan$new(dta, eps, pts)
```

*Arguments:*

`dta` [data.frame](#) | [matrix](#)

The data to be clustered by the algorithm. Allowed are only [numeric](#) columns.

`eps` [numeric](#)

The size (radius) of the epsilon neighborhood. If the radius contains 2 numbers, the fuzzy cores are calculated between the minimum and the maximum radius. If epsilon is a single number, the algorithm loses the fuzzy core property. If the length of `pts` is also 1L, the algorithm equals to non-fuzzy DBScan.

`pts` [numeric](#)

number of maximum and minimum points required in the `eps` neighborhood for core points (excluding the point itself). If the length of the argument is 1, the algorithm loses its fuzzy border property. If the length of `eps` is also 1L, the algorithm equals to non-fuzzy DBScan.

**Method** `predict()`: Predict new data with the initialized algorithm.

*Usage:*

```
FuzzyDBScan$predict(new_data, cmatrix = TRUE)
```

*Arguments:*

`new_data` [data.frame](#) | [matrix](#)

The data to be predicted by the algorithm. Allowed are only [numeric](#) columns which should match to `self$dta`.

`cmatrix` [logical](#)

Indicating whether the assigned cluster should be returned in form of a matrix where each column indicates for the probability of the new data to belong to a respective cluster. The object will have the same shape as the `results` field. If set to `FALSE` the shape of the returned assigned clusters is a two-column [data.table](#) with one column indicating the assigned cluster and the second column indicating the respective probability of the new data.

**Method** `plot()`: Plot clusters and soft labels on two features.

*Usage:*

```
FuzzyDBScan$plot(x, y)
```

*Arguments:*

`x` [character](#)

Feature to plot on the x-axis.

`y` [character](#)

Feature to plot on the y-axis.

**Method** `clone()`: The objects of this class are cloneable with this method.

*Usage:*

```
FuzzyDBScan$clone(deep = FALSE)
```

*Arguments:*

`deep` Whether to make a deep clone.

## References

Ienco, Dino, and Gloria Bordogna. Fuzzy extensions of the DBScan clustering algorithm. *Soft Computing* 22.5 (2018): 1719-1730.

## Examples

```
# load factoextra for data and ggplot for plotting
library(factoextra)
dta = multishapes[, 1:2]
eps = c(0, 0.2)
pts = c(3, 15)
# train DBScan based on data, ep and pts
cl = FuzzyDBScan$new(dta, eps, pts)
# Plot DBScan for x and y
library(ggplot2)
cl$plot("x", "y")
# produce test data
x <- seq(min(dta$x), max(dta$x), length.out = 50)
y <- seq(min(dta$y), max(dta$y), length.out = 50)
p_dta = expand.grid(x = x, y = y)
# predict on test data and plot results
p = cl$predict(p_dta, FALSE)
ggplot(p, aes(x = p_dta[, 1], y = p_dta[, 2], colour = as.factor(cluster))) +
  geom_point(alpha = p$dense)
```

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