

Package: daltoolboxdp (via r-universe)

September 5, 2024

Title Data Pre-Processing Extensions

Version 1.0.767

Description An important aspect of data analytics is related to data management support for artificial intelligence. It is related to preparing data correctly. This package provides extensions to support data preparation in terms of both data sampling and data engineering. Overall, the package provides researchers with a comprehensive set of functionalities for data science based on experiment lines, promoting ease of use, extensibility, and integration with various tools and libraries. Information on Experiment Line is based on Ogasawara et al. (2009) <[doi:10.1007/978-3-642-02279-1_20](https://doi.org/10.1007/978-3-642-02279-1_20)>.

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URL <https://github.com/cefet-rj-dal/daltoolboxdp>

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RoxygenNote 7.3.1

Imports daltoolbox, leaps, FSelector, doBy, glmnet, smotefamily

Repository <https://cefet-rj-dal.r-universe.dev>

RemoteUrl <https://github.com/cefet-rj-dal/daltoolboxdp>

RemoteRef HEAD

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bal_oversampling	<i>Oversampling</i>
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Description

Oversampling balances the class distribution of a dataset by increasing the representation of the minority class in the dataset. It wraps the smotefamily library.

Usage

```
bal_oversampling(attribute)
```

Arguments

attribute The class attribute to target balancing using oversampling.

Value

A bal_oversampling object.

Examples

```
data(iris)
mod_iris <- iris[c(1:50,51:71,101:111),]

bal <- bal_oversampling('Species')
bal <- daltoolbox::fit(bal, mod_iris)
adjust_iris <- daltoolbox::transform(bal, mod_iris)
table(adjust_iris$Species)
```

bal_subsampling	<i>Subsampling</i>
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Description

Subsampling balances the class distribution of a dataset by reducing the representation of the majority class in the dataset.

Usage

```
bal_subsampling(attribute)
```

Arguments

attribute The class attribute to target balancing using subsampling

Value

A `bal_subsampling` object.

Examples

```
data(iris)
mod_iris <- iris[c(1:50,51:71,101:111),]

bal <- bal_subsampling('Species')
bal <- daltoolbox::fit(bal, mod_iris)
adjust_iris <- daltoolbox::transform(bal, mod_iris)
table(adjust_iris$Species)
```

Description

Feature selection is a process of selecting a subset of relevant features from a larger set of features in a dataset for use in model training. The `FeatureSelection` class in R provides a framework for performing feature selection.

Usage

```
fs(attribute)
```

Arguments

`attribute` The target variable.

Value

An instance of the `FeatureSelection` class.

Examples

```
#See ?fs_fss for an example of feature selection
```

`fs_fss`*Forward Stepwise Selection*

Description

Forward stepwise selection is a technique for feature selection in which attributes are added to a model one at a time based on their ability to improve the model's performance. It stops adding once the candidate addition does not significantly improve model adjustment. It wraps the leaps library.

Usage

```
fs_fss(attribute)
```

Arguments

`attribute` The target variable.

Value

A `fs_fss` object.

Examples

```
data(iris)
myfeature <- daltoolbox::fit(fs_fss("Species"), iris)
data <- daltoolbox::transform(myfeature, iris)
head(data)
```

`fs_ig`*Information Gain*

Description

Information Gain is a feature selection technique based on information theory. It measures the information obtained for the target variable by knowing the presence or absence of a feature. It wraps the FSelector library.

Usage

```
fs_ig(attribute)
```

Arguments

`attribute` The target variable.

Value

A fs_ig object.

Examples

```
data(iris)
myfeature <- daltoolbox::fit(fs_ig("Species"), iris)
data <- daltoolbox::transform(myfeature, iris)
head(data)
```

fs_lasso

Feature Selection using Lasso

Description

Feature selection using Lasso regression is a technique for selecting a subset of relevant features. It wraps the glmnet library.

Usage

```
fs_lasso(attribute)
```

Arguments

attribute The target variable.

Value

A fs_lasso object.

Examples

```
data(iris)
myfeature <- daltoolbox::fit(fs_lasso("Species"), iris)
data <- daltoolbox::transform(myfeature, iris)
head(data)
```

`fs_relief`*Relief*

Description

Feature selection using Relief is a technique for selecting a subset of relevant features. It calculates the relevance of a feature by considering the difference in feature values between nearest neighbors of the same and different classes. It wraps the FSelector library.

Usage

```
fs_relief(attribute)
```

Arguments

`attribute` The target variable.

Value

A `fs_relief` object.

Examples

```
data(iris)
myfeature <- daltoolbox::fit(fs_relief("Species"), iris)
data <- daltoolbox::transform(myfeature, iris)
head(data)
```

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