Package ‘applicable’

July 12, 2020

Title A Compilation of Applicability Domain Methods

Version 0.0.1.2

Description A modeling package compiling applicability domain methods in R. It combines different methods to measure the amount of extrapolation new samples can have from the training set. See Netzeva et al (2005) <doi:10.1177/026119290503300209> for an overview of applicability domains.

URL https://github.com/tidymodels/applicable,
   https://applicable.tidymodels.org

BugReports https://github.com/tidymodels/applicable/issues

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**ames_new**

**Recent Ames Iowa Houses**

**Description**

More data related to the set described by De Cock (2011) where data where data were recorded for 2,930 properties in Ames IA.

**Details**

This data sets includes three more properties added since the original reference. There are less fields in this data set; only those that could be transcribed from the assessor’s office were included.

**Value**

ames_new  a tibble

**Source**


https://www.cityofames.org/government/departments-divisions-a-h/city-assessor


Description

`apd_hat_values()` fits a model.

Usage

```r
apd_hat_values(x, ...)
```

## Default S3 method:
apd_hat_values(x, ...)

## S3 method for class 'data.frame'
apd_hat_values(x, ...)

## S3 method for class 'matrix'
apd_hat_values(x, ...)

## S3 method for class 'formula'
apd_hat_values(formula, data, ...)

## S3 method for class 'recipe'
apd_hat_values(x, data, ...)

Arguments

- **x** Depending on the context:
  - A **data frame** of predictors.
  - A **matrix** of predictors.
  - A **recipe** specifying a set of preprocessing steps created from `recipes::recipe()`.

- **...** Not currently used, but required for extensibility.

- **formula** A formula specifying the predictor terms on the right-hand side. No outcome should be specified.

- **data** When a **recipe** or **formula** is used, data is specified as:
  - A **data frame** containing the predictors.

Value

A `apd_hat_values` object.
Examples

predictors <- mtcars[, -1]

# Data frame interface
mod <- apd_hat_values(predictors)

# Formula interface
mod2 <- apd_hat_values(mpg ~ ., mtcars)

# Recipes interface
library(recipes)
rec <- recipe(mpg ~ ., mtcars)
rec <- step_log(rec, disp)
mod3 <- apd_hat_values(rec, mtcars)

---

apd_pca  Fit a apd_pca

Description

apd_pca() fits a model.

Usage

apd_pca(x, ...)

## Default S3 method:
apd_pca(x, ...)

## S3 method for class 'data.frame'
apd_pca(x, threshold = 0.95, ...)

## S3 method for class 'matrix'
apd_pca(x, threshold = 0.95, ...)

## S3 method for class 'formula'
apd_pca(formula, data, threshold = 0.95, ...)

## S3 method for class 'recipe'
apd_pca(x, data, threshold = 0.95, ...)

Arguments

x  Depending on the context:

* A **data frame** of predictors.
* A **matrix** of predictors.
**apd_similarity**

- A **recipe** specifying a set of preprocessing steps created from `recipes::recipe()`.

... Not currently used, but required for extensibility.

**threshold**

A number indicating the percentage of variance desired from the principal components. It must be a number greater than 0 and less or equal than 1.

**formula**

A formula specifying the predictor terms on the right-hand side. No outcome should be specified.

**data**

When a **recipe** or **formula** is used, data is specified as:

- A **data frame** containing the predictors.

### Details

The function computes the principal components that account for up to either 95% or the provided **threshold** of variability. It also computes the percentiles of the absolute value of the principal components. Additionally, it calculates the mean of each principal component.

### Value

A `apd_pca` object.

### Examples

```r
predictors <- mtcars[, -1]

# Data frame interface
mod <- apd_pca(predictors)

# Formula interface
mod2 <- apd_pca(mpg ~ ., mtcars)

# Recipes interface
library(recipes)
rec <- recipe(mpg ~ ., mtcars)
rec <- step_log(rec, disp)
mod3 <- apd_pca(rec, mtcars)
```

---

**apd_similarity**

*Applicability domain methods using binary similarity analysis*

### Description

`apd_similarity()` is used to analyze samples in terms of similarity scores for binary data. All features in the data should be binary (i.e. zero or one).
Usage

`apd_similarity(x, ...)`

## Default S3 method:
apd_similarity(x, quantile = NA_real_, ...)

## S3 method for class 'data.frame'
apd_similarity(x, quantile = NA_real_, ...)

## S3 method for class 'matrix'
apd_similarity(x, quantile = NA_real_, ...)

## S3 method for class 'formula'
apd_similarity(formula, data, quantile = NA_real_, ...)

## S3 method for class 'recipe'
apd_similarity(x, data, quantile = NA_real_, ...)

Arguments

- **x**: Depending on the context:
  - A **data frame** of binary predictors.
  - A **matrix** of binary predictors.
  - A **recipe** specifying a set of preprocessing steps created from `recipes::recipe()`.

- **...**: Options to pass to `proxyC::simil()`, such as method. If no options are specified, `method = "jaccard"` is used.

- **quantile**: A real number between 0 and 1 or NA for how the similarity values for each sample versus the training set should be summarized. A value of NA specifies that the mean similarity is computed. Otherwise, the appropriate quantile is computed.

- **formula**: A formula specifying the predictor terms on the right-hand side. No outcome should be specified.

- **data**: When a **recipe** or **formula** is used, data is specified as:
  - A **data frame** containing the binary predictors. Any predictors with no 1’s will be removed (with a warning).

Details

The function computes measures of similarity for different samples points. For example, suppose samples A and B both contain $p$ binary variables. First, a 2x2 table is constructed between A and B across their elements. The table will contain $p$ entries across the four cells (see the example below). From this, different measures of likeness are computed.

For a training set of $n$ samples, a new sample is compared to each, resulting in $n$ similarity scores. These can be summarized into a single value; the median similarity is used by default by the scoring function.
For this method, the computational methods are fairly taxing for large data sets. The training set must be stored (albeit in a sparse matrix format) so object sizes may become large.

By default, the computations are run in parallel using all possible cores. To change this, call the setThreadOptions function in the RcppParallel package.

**Value**

A `apd_similarity` object.

**References**


**Examples**

```r

data(qsar_binary)

jacc_sim <- apd_similarity(binary_tr)
jacc_sim

# plot the empirical cumulative distribution function (ECDF) for the training set:
library(ggplot2)
autoplot(jacc_sim)

# Example calculations for two samples:
A <- as.matrix(binary_tr[1,])
B <- as.matrix(binary_tr[2,])
xtab <- table(A, B)
xtab

# Jaccard statistic
xtab[2, 2] / (xtab[1, 2] + xtab[2, 1] + xtab[2, 2])

# Hamman statistic
((xtab[1, 1] + xtab[2, 2]) - (xtab[1, 2] + xtab[2, 1])) / sum(xtab)

# Faith statistic
(xtab[1, 1] + xtab[2, 2]/2) / sum(xtab)

# Summarize across all training set similarities
mean_sim <- score(jacc_sim, new_data = binary_unk)
mean_sim

```

**autplot.apd_pca**  
*Plot the distribution function for pcas*
Description
Plot the distribution function for pcas

Usage
```r
## S3 method for class 'apd_pca'
autoplot(object, ...)
```

Arguments
- `object` An object produced by `apd_pca`.
- `...` An optional set of `dplyr` selectors, such as `dplyr::matches()` or `dplyr::starts_with()` for selecting which variables should be shown in the plot.

Value
A `ggplot` object that shows the distribution function for each principal component.

Examples
```r
library(ggplot2)
library(dplyr)
library(modeldata)
data(biomass)

biomass_ad <- apd_pca(biomass[, 3:8])

autoplot(biomass_ad)
# Using selectors in `...`
autoplot(biomass_ad, distance) + scale_x_log10()
autoplot(biomass_ad, matches("PC[1-2]"))
```

---

Description
Plot the cumulative distribution function for similarity metrics

Usage
```r
## S3 method for class 'apd_similarity'
autoplot(object, ...)
```
Arguments

object

... An object produced by apd_similarity.

Value

A ggplot object that shows the cumulative probability versus the unique similarity values in the training set. Not that for large samples, this is an approximation based on a random sample of 5,000 training set points.

Examples

set.seed(535)
tr_x <- matrix(sample(0:1, size = 20 * 50, prob = rep(.5, 2), replace = TRUE), ncol = 20)
model <- apd_similarity(tr_x)

---

binary  

Description

Binary QSAR Data

Details

These data are from two different sources on quantitative structure-activity relationship (QSAR) modeling and contain 67 predictors that are either 0 or 1. The training set contains 4,330 samples and there are five unknown samples (both from the Mutagen data in the QSARdata package).

Value

binary_tr,binary_ukn

data frame frames with 67 columns

Examples

data(qsar_binary)
str(binary_tr)
OkCupid Binary Predictors

Description

OkCupid Binary Predictors

Details

Data originally from Kim (2015) includes a training and test set consistent with Kuhn and Johnson (2020). Predictors include ethnicity indicators and a set of keywords derived from text essay data.

Value

okc_binary_train, okc_binary_test
data frame frames with 61 columns

Source


Examples

data(okc_binary)
str(okc_binary_train)

print.apd_hat_values

Print number of predictors and principal components used.

Description

Print number of predictors and principal components used.

Usage

```r
## S3 method for class 'apd_hat_values'
print(x, ...)
```

Arguments

- `x` A `apd_hat_values` object.
- `...` Not currently used, but required for extensibility.
Value

None

Examples

```r
model <- apd_hat_values(~ Sepal.Length + Sepal.Width, iris)
print(model)
```

---

**Description**

Print number of predictors and principal components used.

**Usage**

```r
## S3 method for class 'apd_pca'
print(x, ...)
```

**Arguments**

- `x`  
  A `apd_pca` object.
- `...`  
  Not currently used, but required for extensibility.

**Value**

None

**Examples**

```r
model <- apd_pca(~ Sepal.Length + Sepal.Width, iris)
print(model)
```
print.apd_similarity  
*Print number of predictors and principal components used.*

**Description**

Print number of predictors and principal components used.

**Usage**

```r
## S3 method for class 'apd_similarity'
print(x, ...)
```

**Arguments**

- `x`: A `apd_similarity` object.
- `...`: Not currently used, but required for extensibility.

**Value**

None

**Examples**

```r
set.seed(535)
tr_x <- matrix(sample(0:1, size = 20 * 50, prob = rep(.5, 2),
        replace = TRUE), ncol = 20)
model <- apd_similarity(tr_x)
print(model)
```

---

**score**  
*A scoring function*

**Description**

A scoring function

**Usage**

```r
score(object, ...)
```

**Examples**

```r
score(object, ...)
```
score.apd_hat_values

Arguments

object Depending on the context:

• A data frame of predictors.
• A matrix of predictors.
• A recipe specifying a set of preprocessing steps created from recipes::recipe().

... Not currently used, but required for extensibility.

Value

A tibble of predictions.

Description

Score new samples using hat values

Usage

## S3 method for class 'apd_hat_values'
score(object, new_data, type = "numeric", ...)

Arguments

object A apd_hat_values object.
new_data A data frame or matrix of new predictors.
type A single character. The type of predictions to generate. Valid options are:

• "numeric" for a numeric value that summarizes the hat values for each sample across the training set.

... Not used, but required for extensibility.

Value

A tibble of predictions. The number of rows in the tibble is guaranteed to be the same as the number of rows in new_data. For type = "numeric", the tibble contains two columns hat_values and hat_values_pctls. The column hat_values_pctls is in percent units so that a value of 11.5 indicates that, in the training set, 11.5 percent of the training set samples had smaller values than the sample being scored.
Examples

```r
train_data <- mtcars[1:20,]
test_data <- mtcars[21:32,]

hat_values_model <- apd_hat_values(train_data)

hat_values_scoring <- score(hat_values_model, new_data = test_data)
hat_values_scoring
```

---

**score.apd_pca**  
*Predict from a apd_pca*

**Description**

Predict from a apd_pca

**Usage**

```r
# S3 method for class 'apd_pca'
score(object, new_data, type = "numeric", ...)
```

**Arguments**

- `object` A apd_pca object.
- `new_data` A data frame or matrix of new samples.
- `type` A single character. The type of predictions to generate. Valid options are:
  - "numeric" for numeric predictions.
- `...` Not used, but required for extensibility.

**Details**

The function computes the principal components of the new data and their percentiles as compared to the training data. The number of principal components computed depends on the threshold given at fit time. It also computes the multivariate distance between each principal component and its mean.

**Value**

A tibble of predictions. The number of rows in the tibble is guaranteed to be the same as the number of rows in `new_data`. 
Examples

```r
train <- mtcars[1:20,]
test <- mtcars[21:32, -1]

# Fit
mod <- apd_pca(mpg ~ cyl + log(drat), train)

# Predict, with preprocessing
score(mod, test)
```

Description

Score new samples using similarity methods

Usage

```r
## S3 method for class 'apd_similarity'
score(object, new_data, type = "numeric", add_percentile = TRUE, ...)
```

Arguments

- `object` A `apd_similarity` object.
- `new_data` A data frame or matrix of new predictors.
- `type` A single character. The type of predictions to generate. Valid options are:
  - "numeric" for a numeric value that summarizes the similarity values for each sample across the training set.
- `add_percentile` A single logical; should the percentile of the similarity score relative to the training set values by computed?
- `...` Not used, but required for extensibility.

Value

A tibble of predictions. The number of rows in the tibble is guaranteed to be the same as the number of rows in `new_data`. For `type = "numeric"`, the tibble contains a column called "similarity". If `add_percentile = TRUE`, an additional column called `similarity_pctl` will be added. These values are in percent units so that a value of 11.5 indicates that, in the training set, 11.5 percent of the training set samples had smaller values than the sample being scored.
Examples

data(qsar_binary)

jacc_sim <- apd_similarity(binary_tr)

mean_sim <- score(jacc_sim, new_data = binary_unk)
mean_sim
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