Package ‘PivotalR’

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Type Package

Title A Fast, Easy-to-Use Tool for Manipulating Tables in Databases and a Wrapper of MADlib

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Depends R (>= 2.14.0), methods, Matrix

Suggests DBI, RPostgreSQL, shiny, testthat, tools, rpart, randomForest, topicmodels

Description Provides an R interface for the Pivotal Data stack running on 'PostgreSQL', 'Greenplum' or 'Apache HAWQ (incubating)' databases with parallel and distributed computation ability for big data processing. 'PivotalR' provides an R interface to various database operations on tables or views. These operations are almost the same as the corresponding native R operations. Thus users of R do not need to learn 'SQL' when they operate on objects in the database. It also provides a wrapper for 'Apache MADlib (incubating)', which is an open-source library for parallel and scalable in-database analytics.

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PivotalR is a package that enables users of R, the most popular open source statistical programming language and environment to interact with the Pivotal (Greenplum) Database as well as Pivotal HD/HAWQ for Big Data analytics. It does so by providing an interface to the operations on tables/views in the database. These operations are almost the same as those of data.frame. Thus the users of R do not need to learn SQL when they operate on the objects in the database. The latest code, along with a training video and a quick-start guide, are available at https://github.com/pivotalsoftware/PivotalR.

This package enables R users to easily develop, refine and deploy R scripts that leverage the parallelism and scalability of the database as well as in-database analytics libraries to operate on big data sets that would otherwise not fit in R memory - all this without having to learn SQL because the package provides an interface that they are familiar with.

The package also provides a wrapper for MADlib. MADlib is an open-source library for scalable in-database analytics. It provides data-parallel implementations of mathematical, statistical and machine-learning algorithms for structured and unstructured data. The number of machine learning algorithms that MADlib covers is quickly increasing.

As an R front-end to the PostgreSQL-like databases, this package minimizes the amount of data transferred between the database and R. All the big data is stored in the database. The user enters their familiar R syntax, and the package translates it into SQL queries and sends the SQL query into database for parallel execution. The computation result, which is small (if it is as big as the original data, what is the point of big data analytics?), is returned to R to the user.

On the other hand, this package also gives the usual SQL users the access of utilizing the powerful
analytics and graphics functionalities of R. Although the database itself has difficulty in plotting, the result can be analyzed and presented beautifully with R.

This current version of PivotalR provides the core R infrastructure and data frame functions as well as over 50 analytical functions in R that leverage in-database execution. These include

* Data Connectivity - db.connect, db.disconnect, db.Rquery
* Data Exploration - db.data.frame, subsets
* R language features - dim, names, min, max, nrow, ncol, summary etc
* Reorganization Functions - merge, by (group-by), samples
* Transformations - as.factor, null replacement
* Algorithms - linear regression and logistic regression wrappers for MADlib

**Note**

This package is different from PL/R, which is another way of using R with PostgreSQL-like databases. PL/R enables the users to run R scripts from SQL. In the parallel Greenplum database, one can use PL/R to implement parallel algorithms.

However, PL/R still requires non-trivial knowledge of SQL to use it effectively. It is mostly limited to explicitly parallel jobs. And for the end user, it is still a SQL interface.

This package does not require any knowledge of SQL, and it works for both explicitly and implicitly parallel jobs by employing the open-source MADlib library. It is much more scalable. And for the end user, it is a pure R interface with the conventional R syntax.

**Author(s)**

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Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**References**


**See Also**

madlib.lm Linear regression
madlib.glm Linear, logistic and multinomial logistic regressions
madlib.summary summary of a table in the database.
Examples

```r
## Not run:
## get the help for the package
help("PivotalR-package")

## get help for a function
help(madlib.lm)

## create multiple connections to different databases
db.connect(port = 5433) # connection 1, use default values for the parameters
db.connect(dbname = "test", user = "qianh", password = "", host = "remote.machine.com", madlib = "madlib07", port = 5432) # connection 2

db.list() # list the info for all the connections

## list all tables/views that has "ornst" in the name
db.objects("ornst")

## list all tables/views
db.objects(conn.id = 1)

## create a table and the R object pointing to the table
## using the example data that comes with this package
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone")

## OR if the table already exists, you can create the wrapper directly
## x <- db.data.frame("abalone")

dim(x) # dimension of the data table

names(x) # column names of the data table

madlib.summary(x) # look at a summary for each column

lk(x, 20) # look at a sample of the data

## look at a sample sorted by id column
lookat(sort(x, decreasing = FALSE, x$id), 20)

lookat(sort(x, FALSE, NULL), 20) # look at a sample ordered randomly

## linear regression Examples

## fit one different model to each group of data with the same sex
fit1 <- madlib.lm(rings ~ . - id | sex, data = x)

fit1 # view the result

lookat(mean((x$rings - predict(fit1, x))^2)) # mean square error

## plot the predicted values v.s. the true values
```
PivotalR-package

ap <- x$rings # true values
ap$pred <- predict(fit1, x) # add a column which is the predicted values

## If the data set is very big, you do not want to load all the
## data points into R and plot. We can just plot a random sample.
random.sample <- lk(sort(ap, FALSE, "random"), 1000) # sort randomly

plot(random.sample) # plot a random sample

## fit a single model to all data treating sex as a categorical variable --------
y <- x # make a copy, y is now a db.data.frame object
y$sx <- as.factor(y$sx) # y becomes a db.Rquery object now
fit2 <- madlib.nlm(rings ~ . - id, data = y)

fit2 # view the result

lookat(mean((y$rings - predict(fit2, y))^2)) # mean square error

## logistic regression Examples --------

## fit one different model to each group of data with the same sex
fit3 <- madlib glm(rings < 10 ~ . - id | sex, data = x, family = "binomial")

fit3 # view the result

## the percentage of correct prediction
lookat(mean((y$rings < 10) == predict(fit3, y)))

## fit a single model to all data treating sex as a categorical variable --------
y <- x # make a copy, y is now a db.data.frame object
y$sx <- as.factor(y$sx) # y becomes a db.Rquery object now
fit4 <- madlib glm(rings < 10 ~ . - id, data = y, family = "binomial")

fit4 # view the result

## the percentage of correct prediction
lookat(mean((y$rings < 10) == predict(fit4, y)))

## Group by Examples --------

## mean value of each column except the "id" column
lk(by(x[, -1], x$sx, mean))

## standard deviation of each column except the "id" column
lookat(by(x[, -1], x$sx, sd))

## Merge Examples --------

## create two objects with different rows and columns
key(x) <- "id"
y <- x[1:300, 1:6]
z <- x[201:400, c(1,2,4,5)]
## abalone

**Abalone data set**

### Description

An example data.frame which is used by examples in this user manual

```r
## get 100 rows
m <- merge(y, z, by = c("id", "sex"))
lookat(m, 20)

## operator Examples -------
y <- x$length + x$height + 2.3
z <- x$length * x$height / 3
lk(y < z, 20)

## -----------------------------------------------
## Deal with NULL values

delete("null_data")
x <- as.db.data.frame(null.data, "null_data")

## OR if the table already exists, you can create the wrapper directly
## x <- db.data.frame("null_data")

dim(x)
names(x)

## ERROR, because of NULL values
fit <- madlib.lm(sf_mrtg_pct_assets - ., data = x)

## remove NULL values
y <- x # make a copy
for (i in 1:10) y <- y[!is.na(y[i]),]
dim(y)
fit <- madlib.lm(sf_mrtg_pct_assets - ., data = y)
fit

## Or we can replace all NULL values
x[is.na(x)] <- 45

## End(Not run)
```
Usage

data(abalone)

Format

Given is the attribute name, attribute type, the measurement unit and a brief description. The number of rings is the value to predict: either as a continuous value or as a classification problem.

Name / Data Type / Measurement Unit / Description

Id / integer / – / index of each observation
Sex / nominal / – / M, F, and I (infant)
Length / continuous / mm / Longest shell measurement
Diameter / continuous / mm / perpendicular to length
Height / continuous / mm / with meat in shell
Whole weight / continuous / grams / whole abalone
Shucked weight / continuous / grams / weight of meat
Viscera weight / continuous / grams / gut weight (after bleeding)
Shell weight / continuous / grams / after being dried
Rings / integer / – / +1.5 gives the age in years

Details

Predicting the age of abalone from physical measurements. The age of abalone is determined by cutting the shell through the cone, staining it, and counting the number of rings through a microscope – a boring and time-consuming task. Other measurements, which are easier to obtain, are used to predict the age. Further information, such as weather patterns and location (hence food availability) may be required to solve the problem.

From the original data examples with missing values were removed (the majority having the predicted value missing), and the ranges of the continuous values have been scaled for use with an ANN (by dividing by 200).

Note

Lazy data loading is enabled in this package. So the user does not need to explicitly run data(abalone) to load the data. It will be loaded whenever it is used.

Source

Aggregate functions

Examples

## Not run:

```r
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
## The user does not need to run data(abalone) to load the data
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", key = "id",
                      distributed.by = "id", conn.id = cid,
                      verbose = FALSE)

## preview the actual data
lk(x)

## preview the actual data ordered by id
lk(sort(x, FALSE, x$id))

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

Description

An aggregate function is a function where the values of multiple rows are grouped together as input to calculate a single value of more significant meaning or measurement. The aggregate functions included are mean, sum, count, max, min, standard deviation, and variance. Also included is a function to compute the mean value of each column and a function to compute the sum of each column.

Usage

```r
## S4 method for signature 'db.obj'
mean(x, ...)

## S4 method for signature 'db.obj'
sum(x, ..., na.rm = FALSE)

## S4 method for signature 'db.obj'
count(x)
```
## S4 method for signature 'db.object'
max(x, ..., na.rm = FALSE)

## S4 method for signature 'db.object'
min(x, ..., na.rm = FALSE)

## S4 method for signature 'db.object'
sd(x)

## S4 method for signature 'db.object'
var(x)

## S4 method for signature 'db.object'
colMeans(x, na.rm = FALSE, dims = 1, ...)

## S4 method for signature 'db.object'
colSums(x, na.rm = FALSE, dims = 1, ...)

colAgg(x)

\texttt{db.array(x, \ldots)}

### Arguments

\textbf{x} 
A \texttt{db.object} object. The signature of the method. For \texttt{db.array}, \texttt{x} can also be a normal R object like double value.

\textbf{...} 
Further arguments passed to or from other methods. This is currently not implemented.

\textbf{na.rm} 
Logical. Should missing values (including 'NaN') be removed? This is currently not implemented.

\textbf{dims} 
Integer: Which dimensions are regarded as 'rows' or 'columns' to sum over. This is currently not implemented and the default behavior is to sum over columns.

### Details

For aggregate functions: \texttt{mean}, \texttt{sum}, \texttt{count}, \texttt{max}, \texttt{min}, \texttt{sd}, and \texttt{var}, the signature \texttt{x} must be a reference to a single column in a table.

For aggregate functions: \texttt{colMeans}, \texttt{colSums}, and \texttt{colAgg} the signature \texttt{x} can be a \texttt{db.object} referencing to a single column or a single table, or can be a \texttt{db.Rquery} referencing to multiple columns in a table.

### Value

For \texttt{mean}, a \texttt{db.Rquery} which is a SQL query to extract the average of a column of a table. Actually, it can work on multiple columns, so it is the same as \texttt{colMeans}.

For \texttt{sum}, a \texttt{db.Rquery} which is a SQL query to extract the sum of a column of a table. Actually, it can work on multiple columns, so it is the same as \texttt{colSums}.
Aggregate functions

For count, a `db.Rquery` which is a SQL query to extract the count of a column of a table.
For max, a `db.Rquery` which is a SQL query to extract the max of a column of a table.
For min, a `db.Rquery` which is a SQL query to extract the min of a column of a table.
For sd, a `db.Rquery` which is a SQL query to extract the standard deviation of a column of a table.
For var, a `db.Rquery` which is a SQL query to extract the variance of a column of a table.
For colMeans, a `db.Rquery` which is a SQL query to extract the mean of multiple columns of a table.
For colSums, a `db.Rquery` which is a SQL query to extract the sum of multiple columns of a table.
For colAgg, a `db.Rquery` which is a SQL query to retrieve the column values as an array aggregate.
For db.array, a `db.Rquery` which is a SQL query which combine all columns into an array.

Author(s)
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Maintainer: Frank McQuillan, <fmcquillan@pivotal.io>

See Also
by, `db.obj-method` is usually used together with aggregate functions.

Examples
```r
## Not run:
## get the help for a method
## help("mean,db.obj-method")

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)

## get the mean of a column
mean(x$diameter)

## get the sum of a column
sum(x$height)

## get the number of entries in a column
count(x$id)
```
## AIC

### AIC methods for Madlib regression objects

#### Description

Functions to extract the AIC and log-likelihood for regression models fit in Madlib.

#### Usage

```r
## S3 method for class 'lm.madlib'
extractAIC(fit, scale=0, k=2, ...)
## S3 method for class 'lm.madlib.grps'
extractAIC(fit, scale=0, k=2, ...)
## S3 method for class 'lm.madlib'
logLik(object, ...)
## S3 method for class 'lm.madlib.grps'
logLik(object, ...)
```
# Arguments

- **fit, object**
  The regression model object, of class `lm.madlib` or `logregr.madlib`, fit using
  `madlib.lm` or `madlib.glm` respectively.
- **scale**
  The scale parameter for the model. Currently unused.
- **k**
  Numeric, specifying the equivalent degrees of freedom part in the AIC formula.
- **...**
  Other arguments, not used.

# Details

See the documentation for [AIC](https://cran.r-project.org/) and [extractAIC](https://cran.r-project.org/).

# Value

For ungrouped regressions, `logLik` returns an object of class `logLik`, and `extractAIC` returns a
length-2 numeric vector giving the edf and AIC.
For grouped regressions, logLik and extractAIC return a list giving the output of these methods for each of the component models. Similarly, AIC for a grouped regression returns a vector of the AICs for each of the component models.

Author(s)

Author: Hong Ooi, Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

AIC, extractAIC, logLik.

Examples

## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)

fit <- madlib.glm(rings < 10 ~ . - id | sex, data = x, family =
                   "binomial")

AIC(fit)

AIC(fit[[1]])

db.disconnect(cid, verbose = FALSE)

## End(Not run)

Arith-methods

Arithmetic Operators for db.obj objects

Description

These binary operators perform arithmetic on db.obj objects.
Usage

```r
## S4 method for signature 'db.obj,db.obj'
e1 + e2
## S4 method for signature 'db.obj,db.obj'
e1 - e2
## S4 method for signature 'db.obj,ANY'
e1 - e2
## S4 method for signature 'db.obj,db.obj'
e1 * e2
## S4 method for signature 'db.obj,db.obj'
e1 / e2
## S4 method for signature 'db.obj,db.obj'
e1 %% e2
## S4 method for signature 'db.obj,db.obj'
e1 %% e2
## S4 method for signature 'db.obj,db.obj'
e1 ^ e2
## S4 method for signature 'numeric,db.obj'
e1 + e2
## S4 method for signature 'character,db.obj'
e1 + e2
## S4 method for signature 'numeric,db.obj'
e1 - e2
## S4 method for signature 'character,db.obj'
e1 - e2
## S4 method for signature 'numeric,db.obj'
e1 * e2
## S4 method for signature 'numeric,db.obj'
e1 / e2
## S4 method for signature 'numeric,db.obj'
e1 %% e2
## S4 method for signature 'numeric,db.obj'
e1 %% e2
## S4 method for signature 'numeric,db.obj'
e1 ^ e2
## S4 method for signature 'numeric,numeric'
e1 + e2
## S4 method for signature 'db.obj,character'
e1 + e2
## S4 method for signature 'db.obj,numeric'
e1 - e2
## S4 method for signature 'db.obj,character'
e1 - e2
## S4 method for signature 'db.obj,numeric'
e1 + e2
## S4 method for signature 'db.obj,numeric'
e1 / e2
## S4 method for signature 'db.obj,numeric'
e1 / e2
```
Arith-methods

```r
E1 %% E2
## S4 method for signature 'db.object,numeric'
E1 %% E2
## S4 method for signature 'db.object,numeric'
E1 ^ E2
```

**Arguments**

- `e1, e2`: numeric or `db.object` object.

**Value**

- `db.Rquery` object, which contains the SQL query that computes the arithmetic operations.

**Note**

A meaningful expression is generated only when the `.col.data_type` is "numeric", otherwise a "NULL" value is generated.

"-" and "+" support computing the arithmetic computations between dates, timestamps, times etc.

**Author(s)**

- Author: Predictive Analytics Team at Pivotal Inc.
- Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

**See Also**

- `db.Rquery` contains a SQL query that does the operations.

**Examples**

```r
## Not run:
## get the help for a method
## help("+,db.object,db.object-method")

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)
x$rings <- (x$rings + 2.3) * 3 # change the values
x$area <- x$length * x$height # add a new column
```
Array length

```r
lk(x$area, 10) # view the actual values computed in database
fit <- madlib.lm(rings ~ area, data = x)
## -----------------------------------------------
db.disconnect(cid, verbose = FALSE)
## End(Not run)
```

---

**array.len**

*Get the length of the array in an array column*

**Description**

The column of a table in database can be an array. This function measures the length of the array.

**Usage**

```r
array.len(x)
```

**Arguments**

- `x` A *db.obj* object.

**Value**

An integer, which is the length of the array.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

**See Also**

- *db.array* combines columns of a table/view into an array.
- *as.list* expands the *db.obj* columns into a list of separated *db.Rquery* objects.
- *cbind2* and *cbind* combine multiple *db.obj* objects into one *db.obj* object.
Examples

## Not run:

```
## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)
y <- db.array(x[-2]) # put columns into an array
names(y) # "agg_opr"
array.len(y$agg_opr) # 9

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

arraydb.to.arrayr  

Convert strings extracted from database into arrays

Description

An array object in database is converted to a string when passed into R, for example `'{1.2, 3.4, 5.7}'`, and this function can convert the string to an array in R, for example `c(1.2, 3.4, 5.7)`. This function can also convert a vector of such strings into a two-dimensional array.

Usage

```
arraydb.to.arrayr(str, type = "double", ...)
```

Arguments

- **str**: A vector of strings, or a single string, that has multiple elements in it and delimited by ",".
- **type**: The type of the return value of this function. Default is "double". It can be "character", "double", "logical", "integer", "numeric" etc. All types other than "character", "logical" and "integer" will be treated as "numeric".
- **...**: Further arguments passed to or from other methods. Currently, no more parameters can be passed and this is kept for backwards compatibility.

Details

When R reads in data from a table in the database, the result is a data.frame object. However, if the original data table has a column which is the array type, the array is automatically converted into a string and data.frame object has a corresponding column of strings, each of which starts with "\" and ends with "\"", and all the original array elements are casted into strings delimited by ",".
For example, the array in database `array['ab', 'c d', '"axx, t"]' becomes a string in R '
(ab, c d, "axx, t")'.
This function deals with such strings and turn them into familiar arrays that users can directly use.

**Value**

A two dimensional array, whose element’s type is decided by the function argument `type`.

**Note**

1. The returned value is a two dimensional array, even if `str` is a single string.
2. Although this function is for the strings extracted from database, it can actually deal with strings like "a, b, c", which do not start or end with curly brackets.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**See Also**

`lk` or `link{lookat}` extracts the data of a table

**Examples**

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## Example 1 --------
str <- '1.2, 3.4, 5.6'
arraydb.to.arrayr(str, "double") # (1.2, 3.4, 5.6)

str <- 'a, b, "c, d"'
arraydb.to.arrayr(str, "character") # c("a", "b", "c, d")

## Example 2 --------

## table_in_database has a column of arrays
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
x$col.array <- db.array(x[,3:10])
dat <- lk(x, nrows = 50, array = FALSE) # extract the actual data
arraydb.to.arrayr(dat$col.array, "double") # an array of 50 rows

## ---------------------------------------------------------------
db.disconnect(cid, verbose = FALSE)
```
as.db.data.frame

Convert other objects into a db.data.frame object

Description

Methods for function as.db.data.frame in package PivotalR. When x is a file name or data.frame, the method puts the data into a table in the database. When x is a db.Rquery object, it is converted into a table. When x is a db.data.frame object, a copy of the table/view that x points to is created.

Usage

## S4 method for signature 'character'
as.db.data.frame(x, table.name = NULL,
                              verbose = TRUE, conn.id = 1, add.row.names = FALSE, key = character(0),
                                  distributed.by = NULL, append = FALSE, is.temp = FALSE, ...)

## S4 method for signature 'data.frame'
as.db.data.frame(x, table.name = NULL, verbose =
                                TRUE, conn.id = 1, add.row.names = FALSE, key = character(0),
                                    distributed.by = NULL, append = FALSE, is.temp = FALSE, ...)

## S4 method for signature 'db.Rquery'
as.db.data.frame(x, table.name = NULL, verbose =
                                    TRUE, is.view = FALSE, is.temp = FALSE, pivot = TRUE, distributed.by =
                                        NULL, nrow = NULL, field.types = NULL, na.as.level = FALSE,
                                            factor.full = rep(FALSE, length(names(x))))

## S4 method for signature 'db.data.frame'
as.db.data.frame(x, table.name = NULL, verbose =
                                TRUE, is.view = FALSE, is.temp = FALSE, distributed.by = NULL, nrow =
                                    NULL, field.types = NULL)

as.db.Rview(x)

Arguments

x

The signature of this method.

When it is of type character, it should be a file name.

When it is of type data.frame, it is the data.frame that already exists in the current R session.

When it is of type db.Rquery, it represents a series of operations on a existing db.data.frame object. See db.Rquery for more.

For as.db.Rview, x must be a db.Rquery object.
as.db.data.frame

table.name  A string, the name of the table to be created. The returned db.data.frame object is pointing to this table. When table.name is NULL, a random name is used, which also avoids the name conflicts.

verbose   A logical, default is TRUE, whether to print some prompt messages.

conn.id    An integer, default is 1. The ID of the connection. See db.list for more information.

add.row.names A logical, default is FALSE, whether to add a column named "row.names" is added to the newly created table as the first column, which is just the row number of the original data.frame or file.

key       A string, default is character(0). The primary key column name. When it is not character(0), a primary key is created for this column.

distributed.by A string, default is NULL. It is a column name or multiple column names separated by comma. When creating tables in a Greenplum database [1], the user can choose to specify whether he want to distributed the table onto multiple segments according the values of some columns. When this parameter is NULL, the data is distributed randomly, and when this parameter is an empty string code"", Greenplum database automatically chooses a column and distribute the data according to that column.

append A logical, default is FALSE. Whether to append the content of a file or data.frame to an existing table in the database.

nrow   An integer, default is NULL. How many rows of data extracted from a db.Rquery object is used to create the new table. NULL means using all the rows.

is.temp A logical, default is FALSE, whether the created table/view should be a temporary table/view.

... Extra parameters used to create the table inside the database. We support the following parameters: header = FALSE, nrows = 50, sep = ",", eol = "\n", skip = 0.

header is a logical indicating whether the first data line (but see skip) has a header or not. If missing, it value is determined following read.table convention, namely, it is set to TRUE if and only if the first row has one fewer field that the number of columns.

nrows When creating table from file or data.frame, the function will try to infer the data type of each column using the first nrows rows of the data.

sep specifies the field separator, and its default is ",".

eol specifies the end-of-line delimiter, and its default is "\n".

skip specifies number of lines to skip before reading the data, and it defaults to 0.

field.types A list of key=value pairs, where the value is a string of data type.
Force the new table to use the data type for the column key.

is.view A logical, default is FALSE, whether to create a view instead of a table.

pivot A logical, default is TRUE, whether to create dummy columns for a column that has been denoted as "factor". See as.factor for more details.

na.as.level A logical value, default is FALSE. Whether to treat NA value as a level in a categorical variable or just ignore it.
field.types A list of key=value pairs, where the value is a string of data type. Force the new table to use the data type for the column key.

factor.full A vector of logical values with the length of the column number. All FALSE by default. When the function creates dummy variables for a factor (categorical) variable, whether to create n dummies or n-1 dummies, where n is the number of levels of the factor. For some regression problem, we need to create dummy variables for all the distinct values of the categorical variable.

Value
A db.data.frame object. It points to a table whose name is given by table.name in connection conn.id.

Note
All the as.db.data.frame accept the option field.types.

Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References

See Also
db.data.frame creates an object pointing to a table/view in the database.
lk looks at data from the table
db.Rquery this type of object represents operations on an existing db.data.frame object.

Examples
## Not run:
## get the help for a method
## help("as.db.data.frame")
## help("as.db.data.frame,db.Rquery-method")

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

## preview of a table
lk(x, nrows = 10) # extract 10 rows of data
## as.environment

Evaluate expressions within the context of a database table or view

### Description

These functions allow a `db.table` or `db.view` object to be treated as an environment, in a manner analogous to data frames.

### Usage

```r
## S3 method for class 'db.obj'
as.environment(x, ...)
## S3 method for class 'db.obj'
with(data, expr, ...)
```

### Arguments

- `x`, `data`  
  A `db.obj` object to treat as an environment.

- `expr`  
  For `with`, an R expression to evaluate in the context of a database table or view.

- `...`  
  Other arguments; unused.

### Value

For `as.environment`, the created environment. Note that no data is transferred to the client; all objects in the environment are queries pointing back to the host.

For `with`, a `db.Rquery` stored query object representing the expression. Use `lk`, `lookat` or `as.data.frame` to execute the query on the host and retrieve its contents.
as.factor-methods

Author(s)

Author: Hong Ooi, Pivotal Inc. <hooi@pivotal.io>
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

as.environment

Description

Convert one column of a db.obj object into a categorical variable. When madlib.lm or madlib.glm are applied onto a db.obj with categorical columns, dummy columns will be created and fitted. The reference level for regressions can be selected using relevel.

Usage

## S4 method for signature 'db.obj'
as.factor(x)

## S4 method for signature 'db.obj'
relevel(x, ref, ...)

Arguments

x A db.obj object. It must have only one column.
ref A single value, which is the reference level that is used in the regressions.
... Other arguments passed into the result. Not implemented yet.

Value

A db.Query object. It has only one column which is categorical. By default, a reference level is automatically selected in regressions, which is usually the minimum of all levels, but one can easily change the reference level using relevel.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

madlib.lm and madlib.glm can fit categorical variables

When as.db.data.frame creates a table/view, it can create dummy variables for a categorical variable.
Examples

## Not run:

```r
## get help for a method
## help("as.factor,db.obj-method")
```

```r
## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a temporary table from the example data.frame "abalone"
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

## set sex to be a categorical variable
x$sex <- as.factor(x$sex)

fit1 <- madlib.lm(rings ~ . - id, data = x) # linear regression

fit2 <- madlib.glm(rings < 10 ~ . - id, data = x, family = "binomial") # logistic regression

## another temporary table
z <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

## specify factor during fitting
fit3 <- madlib.lm(rings ~ as.factor(sex) + length + diameter, data = z)

## as.factor is automatically used onto text column
## so as.factor is not necessary
fit4 <- madlib.glm(rings < 10 ~ sex + length + diameter, data = z, family = "binomial")

## using relevel to change the reference level
x$sex <- relevel(x$sex, ref = "M")

madlib.lm(rings ~ . - id, data = x) # use "M" as the reference level

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

### Apply a Function to a db.data.frame Split by column(s)

**Description**

`by` is equivalent to "group by" in SQL language. It groups the data according the value(s) of one or multiple columns, and then apply an aggregate function onto each group of the data.
Usage

```r
## S4 method for signature 'db.obj'
by(data, INDICES, FUN, ..., simplify = TRUE)
```

Arguments

- `data`: A `db.obj` object. It represents a table/view in the database if it is an `db.data.frame` object, or a series of operations applied on an existing `db.data.frame` object if it is a `db.Rquery` object.
- `INDICES`: A list of `db.Rquery` objects. Each of the list element selects one or multiple columns of data. When the value is `NULL`, no grouping of data is done, and the aggregate function `FUN` will be applied onto all the data.
- `FUN`: A function, which will be applied onto each group of the data. The result of `FUN` can be of `db.obj` type or any other data types that R supports.
- `...`: Extra arguments passed to `FUN`, currently not implemented.
- `simplify`: Not implemented yet.

Value

The type of the returned value depends on the return type of `FUN`.

If the return type of `FUN` is a `db.obj` object, then this function returns a `db.Rquery` object, which is actually the SQL query that does the "GROUP BY". It computes the group-by values. The result can be viewed using `lk` or `lookat`.

If the return type of `FUN` is not a `db.obj` object, then this function returns a list, which contains a number of sub-lists. Each sub-list contains two items: (1) `index`, an array of strings, a set of distinct values of the `INDICES` converted to string; and (2) `result`, the result produced by `FUN` applying onto the group of data that has the set of distinct values. The total number of sub-lists is equal to the total number of groups of data partitioned by `INDICES`.

Author(s)

- **Author**: Predictive Analytics Team at Pivotal Inc.
- **Maintainer**: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

See Also

- **Aggregate functions** lists all the supported aggregate functions.
- `lk` or `lookat` can display the actual result of this function.

Examples

```r
## Not run:
## help("by,db.obj-method") # display this doc

## set up the database connection
## Assume that .port is port number and .dbname is the database name
```
cbind2-methods

**Description**

`cbind2` or `cbind` combine two or multiple `db.obj` objects to form a new `db.obj`. And `as.list` does the opposite and expand a `db.obj` object into a list of `db.obj` objects with each one of them representing one column of the original `db.obj` object. `as.list` is usually used together with Reduce and Map.

**Usage**

```r
## S4 method for signature 'db.obj,db.obj'
cbind2(x, y)

## S4 method for signature 'db.obj'
as.list(x, array = FALSE, ...)
```

**Arguments**

- `x, y` The signature of the method. Both arguments are `db.obj` objects.
- `array` logical, default is FALSE. When it is TRUE, the array columns are also expanded and all the elements are put into the resulting list. Otherwise, an array column is treated as a single item in the result.
- `...` In `cbind` they can be anything that can form new columns together with `x`. In `as.list`, it is not implemented yet.
Value

cbind2 or cbind: A \texttt{db.Rquery} object, which contains all columns of \(x\) and \(y\).
\texttt{as.list}: A list of \texttt{db.Rquery} objects, which are the columns of \(x\)

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. \texttt{<fmcquillan@pivotal.io>}

See Also

db.array combines columns of a table/view into an array.
array.len measures the length of the array in an array column.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
delete('abalone', conn.id = cid)
x <- as.db.data.frame(abalone, 'abalone', conn.id = cid, verbose = FALSE)

fit <- madlib.lm(rings ~ . - id - sex, data = x)

## create a db.Rquery object that has two columns
z <- cbind(x$rings, predict(fit, x))

## plot prediction v.s. real value
plot(lookat(z, 100))

## expand the db.obj
unlist(Map(function(x)
  if (col.types(x) == "text")
    paste(lk(unique(x)), collapse="-", sep="")
  else
    lk(mean(x)),
    as.list(x))

## sum of all columns (excluding the 2nd column)
Reduce(function(left, right) left + right, as.list(x[-2]))

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```
clean.madlib.temp  

Delete all the result tables created during calculations of MADlib

Description

Some MADlib wrapper functions create result tables that cannot be dropped in the background, because other functions need to use these tables. For example, `madlib.arima` creates 3 result tables, which are needed by `predict.arima.css.madlib`. One can manually delete these 3 tables when they are not useful anymore using `delete.arima.css.madlib-method`. One can also choose to all such tables created by many such functions together using this function.

Usage

```
clean.madlib.temp(conn.id = 1)
```

Arguments

`conn.id`  
An integer, the connection ID of the database. See `db.connect` for more details.

Details

All such result tables created by MADlib wrapper functions start with "__madlib_temp_" followed by three random integers. This function deletes all such tables.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

See Also

- `madlib.arima` creates three tables with names starting with "__madlib_temp_" when it fits ARIMA model to time series
- `delete.arima.css.madlib-method` deletes the result of `madlib.arima` together with the model, residual and statistics tables.

Examples

```
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## use double values as the time stamp
## Any values that can be ordered will work
example_time_series <- data.frame(  
```
id = seq(0, 1000, length.out=length(ts)),
val = arima.sim(list(order=c(2,0,1),
ar=c(0.7, -0.3), ma=0.2), n=1000000) + 3.2)
x <- as.db.data.frame(example_time_series, field.types =
  list(id="double precision", val = "double precision"),
  conn.id = cid, verbose = FALSE)

dim(x)

names(x)

## use formula
s <- madlib.arima(val ~ id, x, order = c(2,0,1))

s

## delete all result tables
clean.madlib.temp(conn.id = 1)

## s still exists but the 3 tables (model, residuals, etc.) are deleted
s

db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

**coef**

*Extract model coefficients for Madlib regression objects*

**Description**

Functions to extract the coefficients for regression models fit in Madlib.

**Usage**

```
## S3 method for class 'lm.madlib' 
coef(object, ...) 

## S3 method for class 'lm.madlib.grps' 
coef(object, ...) 

## S3 method for class 'logregr.madlib' 
coef(object, ...) 

## S3 method for class 'logregr.madlib.grps' 
coef(object, ...) 
```

**Arguments**

- `object`  
  The regression model object, fit using `madlib.lm` or `madlib.glm`.

- `...`  
  Other arguments, not used.
Details

Extract the fitted coefficients for a linear or logistic regression model, or a grouped list of such models.

Value

For ungrouped regressions, a named numeric vector giving the fitted coefficients.
For grouped regressions, a list giving the coefficients for each of the component models.

Author(s)

Author: Hong Ooi, Pivotal Inc. <hooi@pivotal.io>
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

coef.

Examples

## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)

fit <- madlib.glm(rings < 10 ~ . - id | sex, data = x, family = "binomial")

coef(fit)
coef(fit[[1]])

db.disconnect(cid, verbose = FALSE)

## End(Not run)
Usage

```r
## S4 method for signature 'db.obj,db.obj'
e1 > e2
## S4 method for signature 'db.obj,db.obj'
e1 < e2
## S4 method for signature 'db.obj,db.obj'
e1 >= e2
## S4 method for signature 'db.obj,db.obj'
e1 <= e2
## S4 method for signature 'db.obj,db.obj'
e1 == e2
## S4 method for signature 'db.obj,db.obj'
e1 != e2
## S4 method for signature 'character,db.obj'
e1 > e2
## S4 method for signature 'character,db.obj'
e1 < e2
## S4 method for signature 'character,db.obj'
e1 >= e2
## S4 method for signature 'character,db.obj'
e1 <= e2
## S4 method for signature 'character,db.obj'
e1 == e2
## S4 method for signature 'character,db.obj'
e1 != e2
## S4 method for signature 'db.obj,character'
e1 > e2
## S4 method for signature 'db.obj,character'
e1 < e2
## S4 method for signature 'db.obj,character'
e1 >= e2
## S4 method for signature 'db.obj,character'
e1 <= e2
## S4 method for signature 'db.obj,character'
e1 == e2
## S4 method for signature 'db.obj,character'
e1 != e2
## S4 method for signature 'numeric,db.obj'
e1 > e2
## S4 method for signature 'numeric,db.obj'
e1 < e2
## S4 method for signature 'numeric,db.obj'
e1 >= e2
## S4 method for signature 'numeric,db.obj'
e1 <= e2
## S4 method for signature 'numeric,db.obj'
e1 == e2
## S4 method for signature 'numeric,db.obj'
```

e1 != e2
    ## S4 method for signature 'db.obj,numeric'
e1 > e2
    ## S4 method for signature 'db.obj,numeric'
e1 < e2
    ## S4 method for signature 'db.obj,numeric'
e1 >= e2
    ## S4 method for signature 'db.obj,numeric'
e1 <= e2
    ## S4 method for signature 'db.obj,numeric'
e1 == e2
    ## S4 method for signature 'db.obj,numeric'
e1 != e2
    ## S4 method for signature 'db.obj,logical'
e1 == e2
    ## S4 method for signature 'logical,db.obj'
e1 == e2
    ## S4 method for signature 'db.obj,logical'
e1 != e2
    ## S4 method for signature 'logical,db.obj'
e1 != e2
    ## S4 method for signature 'character,db.obj'
grepl(pattern, x, ignore.case = FALSE, perl = FALSE, fixed = FALSE, useBytes = FALSE)

Arguments

e1, e2 numeric, character or `db.obj` object.

pattern character string containing a regular expression (or character string for 'fixed = TRUE') to be matched in the given character vector.

x A `db.obj` object.

ignore.case if 'FALSE', the pattern matching is _case sensitive_ and if 'TRUE', case is ignored during matching.

perl logical. Should perl-compatible regexps be used? Not implemented yet.

fixed logical. If 'TRUE', 'pattern' is a string to be matched as is. Overrides all conflicting arguments.

useBytes logical. Not implemented yet.

Value

db.Query object, which contains the SQL query that computes the comparison operations.

Note

A meaningful expression is generated only when the .col.data_type is "character" or "numeric", otherwise a "NULL" value is generated.
**conn.eql**

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**See Also**

`db.Rquery` contains a SQL query that does the operations.

**Examples**

```r
## Not run:
## get the help for a method
## help("> db.obj.db.obj-method")

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)

lk(x[x$length > 10,])

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

**conn.eql**  
*Check whether two connections are the same*

**Description**

Two connections are regarded as equal if and only if they have the same database name, host, DBMS, and port number.

**Usage**

`conn.eql(conn.id1, conn.id2)`

**Arguments**

- `conn.id1`: An integer, a connection ID number.
- `conn.id2`: An integer, another connection ID number.
Value

A logical. TRUE if and only if the two connections have the same database name, host, DBMS, and port number.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

connection info has all functions that can extract information about the database connection.
db.connect creates connections to the databases.
db.disconnect disconnects an existing connection.
db.list lists all the current connections with their information.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid1 <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
cid2 <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

db.list() # list the above two connections

conn.eql(cid1, cid2) # returns TRUE

db.disconnect(cid1, verbose = FALSE)
db.disconnect(cid2, verbose = FALSE)

## End(Not run)
```

---

**conn.id**

Find out the connection ID of a db.obj object

Description

Each db.obj object contains the ID of the connection that its data resides on. This function returns the connection ID number. The user can also change the connection ID that a db.obj is associated with.

Usage

```r
conn.id(x)
conn.id(x) <- value
```
**conn.id**

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**Arguments**

- **x**  
  A `db.obj` object.

- **value**  
  An integer, the connection ID number. The user is allowed to change the connection ID that is associated with `x`.

**Value**

An integer, the connection ID associated with `x`

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**See Also**

- `db.connect` creates connections to the databases.
- `db.disconnect` disconnects an existing connection.
- `db.list` lists all the current connections with their information.
- `connection info` has all functions that can extract information about the database connection.
- `conn.eql` tests whether two connections are actually the same one.

**Examples**

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid1 <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
cid2 <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

db.list() # list the two connections
conn.eql(cid1, cid2) # returns TRUE

## use the example data to create a table in connection 1
delete("abalone", conn.id = cid2)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid1, verbose = FALSE)
db.disconnect(cid1) # disconnect connection 1

## lookat(x) # gives an error since connection 1 is disconnected
conn.id(x) <- cid2 # 1 and 2 are the same

lk(x) # gives what you want

db.disconnect(cid2, verbose = FALSE)
```
Print the content of a db.obj object

Description

A db.data.frame object’s content is the table/view name that it points to. A db.Rquery object’s content is the SQL query that represents the operations applied on an existing db.data.frame. This function is mainly for debugging. Normal user who is not familiar with SQL does not need to use it.

Usage

content(x)

Arguments

x A db.obj object, whose content will be returned.

Value

A string, the content of db.obj object x. A db.data.frame object’s content is the table/view name that it points to. A db.Rquery object’s content is the SQL query that represents the operations applied on an existing db.data.frame.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

See Also

db.obj, db.data.frame, db.table, db.view, db.Rquery explain the definitions of the class hierarchy of this package.

Examples

## Not run:

## set up the database connection

# Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

delete(“abalone”, conn.id = cid)

y <- as.db.data.frame(abalone, “abalone”, conn.id = cid) # create a table

x <- db.data.frame(“abalone”, conn.id = cid, key = “id”)
crossprod

Compute the matrix product of $X^T$ and $Y$.

Description

The function computes the cross product of two matrices. The matrix is stored in the table either as multiple columns of data or a column of arrays.

Usage

```r
## S4 method for signature 'db.obj,ANY'
crossprod(x, y = x)
```

Arguments

- **x**: A `db.obj` object. It either has multiple columns or a column of arrays, and thus forms a matrix.
- **y**: A `db.obj` object, default is the same as `x`. This represents the second matrix in the cross product.

Value

A `db.Rcrossprod` object, which is subclass of `db.Rquery`. It is actually a vectorized version of the resulting product matrix represented in an array. If you want to take a look at the actual values inside this matrix, `lk` or `lookat` can be used to extract the correct matrix format as long as the matrix can be loaded into the memory. Usually the resulting product matrix is not too large because the number $n$ of columns is usually not too large and the dimension of the resulting matrix is $n \times n$. 

```r
## actually, x and y are pointing the same table
eql(x, y) # returns TRUE

content(x)
content(x$id)
content(x$id < 10)
content(x[,1:5])
content(x == y) # this is different from eql(x, y)
content(sort(x, INDICES = x$id))
content(x[x$id<10,])
content(x[1:10,])
content(colSums(x))
content(by(x, NULL, sum))
content(by(x, x$sex, sum))

db.disconnect(cid, verbose = FALSE)
```

## End(Not run)
Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

db.array forms an array using columns

Examples

```r
## Not run:
## get the help for a method
## help("crossprod.db.obj-method")

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)
lookat(crossprod(x[, -c(1, 2)]))
x$darr <- db.array(1, x$length, x$diameter)
lookat(crossprod(x$darr))

## Create a function that does Principal Component Analysis in parallel.
## As long as the number of features of the data table is fewer than
## ~ 5000, the matrix t(x)
## the eigenvalues and eigenvectors. However, the step t(x)
## be done in-database in parallel, because x can be very big.
pca <- function (x, center = TRUE, scale = FALSE)
{
  y <- scale(x, center = center, scale = scale) # centering and scaling
  z <- as.db.data.frame(y, verbose = FALSE) # create an intermediate table to save computation
  m <- lookat(crossprod(z)) # one scan of the table to compute Z^T * Z
  d <- delete(z) # delete the intermediate table
  res <- eigen(m) # only this computation is in R
  n <- attr(y, "row.number") # save the computation to count rows

  ## return the result
  list(val = sqrt(res$values/(n-1)), # eigenvalues
       vec = res$vectors, # columns of this matrix are eigenvectors
            center = attr(y, "scaled:center"),
            scale = attr(y, "scaled:scale"))
}
```
### Description

Create a connection to a PostgreSQL or Greenplum (Pivotal) database. One can create multiple connections to multiple databases. The connections are indexed by an integer starting from 1.

### Usage

```r
db.connect(host = "localhost", user = Sys.getenv("USER"), dbname = user, password = "", port = 5432, madlib = "madlib", conn.pkg = "RPostgreSQL", default.schemas = NULL, verbose = TRUE, quick = FALSE)
```

### Arguments

- **host**  
  A string, default is "localhost". The name or IP of the host where the database is located.

- **user**  
  A string, default is the user's username. The username used to connect to the database.

- **dbname**  
  A string, default is the same as the username. The name of the database that you want to connect to.

- **password**  
  A string, default is "". The password string used to connect to the database.

- **port**  
  An integer, default is 5432. The port number used to connect to the database.

- **madlib**  
  A string, default is "madlib". The name of the schema where MADlib is installed.

- **conn.pkg**  
  A string, default is "RPostgreSQL". The name of the R package used to connect to the database. Currently, only RPostgreSQL is supported, but the support for other packages such as RODBC can be easily added.

- **default.schemas**  
  A string, default is NULL. The search path or default schemas of the database that you want to use. The string must be a set of schema names separated by commas. One can also use `db.default.schemas` or `db.search.path` to display or set the search path in the database.
`db.connect`

verbose A logical value, default is TRUE, whether to print some information while connecting to the database.

quick A logical value, default is FALSE. Whether to skip some of the argument checks to speed up the creation of the connection. Useful when using this function inside a function, where you have already validate all the arguments. It is not recommended to set this value to TRUE when you are using this function directly.

Value

An integer, the ID number for the newly created connection.

Note

Right only MADlib 0.6 or later is supported. If you have an older version of MADlib, you will not be able to use all the functions whose names start with "madlib.". However you can still use all the other functions.

Also, right now only PostgreSQL and Greenplum databases are supported.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

`db.disconnect` disconnects a connection.
`db.list` lists all active connections.
`connection info` the functions that extract information about the connection.
`conn.eql` tests whether two connections are the same.
`db.search.path` and `db.default.schemas` displays or sets the search path (i.e. default schemas) in the connected database.

Examples

```r
## Not run:
## connect to a database

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

db.disconnect(cid, verbose = FALSE)
## End(Not run)
```
**db.data.frame**

Create a *db.data.frame* object pointing to a table/view in the database

**Description**

This function creates an object of *db.data.frame*, which points to an existing table/view in the database. The operations that can be applied onto this class of objects are very similar to those of *data.frame*. No real data is loaded into R. The data transferred between the database and R is minimized, which is necessary when we deal with large data sets.

**Usage**

```r
db.data.frame(x, conn.id = 1, key = character(0), verbose = TRUE, is.temp = FALSE)
```

**Arguments**

- `x` A string. It is the name of an existing table/view in the database.
- `conn.id` An integer, default is 1. The ID number of the database connection where the table resides.
- `key` A string, default is `character(0)`. The name of the primary key column. A primary key is a column in a table which must contain a unique value which can be used to identify each and every row of a table uniquely.
- `verbose` A logical, default is `TRUE`. Whether to print a short message when the object in the database is created.
- `is.temp` A logical, default is `FALSE`. Whether the existing table/view in the database is temporary.

**Value**

A *db.data.frame* object. More precisely, a *db.table* object if it points to an existing table in the database, and a *db.view* object if it points to an existing view in the database.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

**See Also**

- `db.objects` lists all tables and views in a database together with their schema.
- `db.existsObject` tests whether a table/view exists in the database.
- `as.db.data.frame` creates a *db.data.frame* from a *data.frame*, a data file or a *db.Rquery*. 
Examples

## Not run:

```r
## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname)

## create a table using as.db.data.frame
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid)

## create an object pointing to the table
y <- db.data.frame("abalone", conn.id = cid)

## x and y point to the same table
eql(x, y) # returns TRUE

## create an object pointing to a table in a schema
db.q("create schema myschema", conn.id = cid)
z <- as.db.data.frame(abalone, "myschema.abalone", conn.id = cid)
db.q("drop schema myschema cascade", conn.id = cid)

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

### db.data.frame-class

**Class** "db.data.frame"

### Description

An object of this class points to a real table/view in the database. No data is transferred into R. Only a minimal amount of information is kept in the object.

### Objects from the Class

Objects can be created by calls of `db.data.frame` or `as.db.data.frame`. The object represents a real table/view in the database. Usually it is NOT recommended to directly manipulate the internal slots of these objects.

### Slots

- `.name`: Object of class "character". It is the table name if this `db.data.frame` was created using just a table name. It can also be a two-element array if this `db.data.frame` was created. This slot is obsolete.
- `.content`: Object of class "character". The table name. The function `content` can get this value.
.conn.id: Object of class "numeric", an integer. The ID number of the database connection where the table resides. The functions `conn.id` and `conn.id<-` can get and set this value.
.col.name: Object of class "character". The 1D array of column names of the table/view that this `db.data.frame` points to. The S4 method `names,db.obj-method` gets this value.
.col.data_type: Object of class "character". The 1D array of column data types of the table/view that this `db.data.frame` points to. This is not supposed to be used by the normal user.
.col.udt_name: Object of class "character". The 1D array of column udt names of the table/view that this `db.data.frame` points to. This is not to be used by normal users.
.table.type: Object of class "character". The information about the type of the table/view that this `db.data.frame` points to, for example, "BASE TABLE", "VIEW" or "LOCAL TEMPORARY".
.is.factor: Object of class "logical". An array of logical values which indicate whether each column of the table/view is a factor. This is not to be used by the normal users.
.factor.suffix: Object of class "character". An array of strings for every column. When creating dummy columns for a factor column, we add a random string in the names of the dummy columns to avoid naming conflicts. So a factor column's .factor.suffix is a random string, otherwise it is just an empty string. This is not to be used by the normal users. It is used only the MADlib wrapper functions that support categorical variables.
.factor.ref: The value of the factor reference level for the regressions. If it is NA, then the regressions automatically select a reference level.
.appear.name: Object of class "character". This is also related the factor columns. `print.lm.madlib` and `print.logregr.madlib` use this value for printing the names of the dummy columns. This is not to be used by the normal users.
.dummy: Object of class "character". An array of strings, The dummy column names which are used only for factor support.
.dummy.expr: Object of class "character". The SQL expressions used to create dummy column names which are used only for factor support.
.dist.by: A string, the distribution policy when using Greenplum database or HAWQ. It can be character(0), which means the data table is distributed randomly. Or it can be a string of column names separated by comma, which are the columns that are used in the "distributed by" when the table was created.

Extends

Class `db.obj`, directly.

Methods

Aggregate functions, `by,db.obj-method`, `dim,db.obj-method`, `dim,db.table-method`, `dim,db.view-method`, `dim,db.Rquery-method`, `names,db.obj-method`, `conn.id,conn.id<-`, `eql,key,key<-`, `merge,db.obj,db.obj-method`, `print,db.data.frame-method`, `show,db.data.frame-method`, `sort,db.obj-method`, `subset,db.obj-method`, `Arith methods,Compare methods,Logical methods,Extraction methods,Replacement methods, madlib.lm,madlib.glm,madlib.summary`
**db.disconnect**

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**See Also**

*db.data.frame* creates a *db.data.frame* object.

*as.db.data.frame* converts *db.Rquery* object, *data.frame*, or a data file into a *db.data.frame* object and at the same time creates a new table in the database.

*db.obj* is the superclass.

*db.table* and *db.view* are the sub-classes.

*db.Rquery* is another sub-class of *db.obj*.

*lk* or *lookat* display a part of the table

**Examples**

```r
## Not run:
showClass("db.data.frame")

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

delete("abalone", conn.id = cid)
as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)
x <- db.data.frame("abalone", conn.id = cid, verbose = FALSE) # x points to table "abalone"

lk(x)
db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

**Description**

Although all the database connections will be automatically closed when this package is unloaded, one can choose to disconnect a database connection himself.

**Usage**

`db.disconnect(conn.id = 1, verbose = TRUE, force = FALSE)`
Arguments

- **conn.id**: An integer, the ID of the connection that you want to disconnect.
- **verbose**: A logical, default is `TRUE`. Whether to print a message during disconnection.
- **force**: A logical, default is `FALSE`. Whether to remove the connection forcefully. This is useful when you lose the connection and cannot disconnect the connection normally.

Value

A logical, `TRUE` if the connection is successfully disconnected.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

See Also

- `db.connect` creates the database connection.
- `db.list` lists all active connections.
- `connection info` the functions that extract information about the connection.
- `conn.eql` tests whether two connections are the same.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

db.list()

## disconnect the connection
db.disconnect(cid, verbose = FALSE)

db.list()

## End(Not run)
```
db.existsObject

Test whether an object exists in the database

Description
Test whether a table or view exists in the database

Usage

```r
db.existsObject(name, conn.id = 1, is.temp = FALSE)
```

Arguments

- `name`: A string, the name of table or view
- `conn.id`: An integer, default is 1. The ID of the database connection.
- `is.temp`: A logical, default is FALSE. Whether this table/view is a temporary object.

Value
This function returns different types of results depending the input.

- If `name` has the format of `myschema.mytable`, the return value is a logical. It is TRUE if the table/view exists in the database.
- If `name` has the format of `mytable` and `is.temp = FALSE`, the return value is also a logical, which is TRUE if the table/view exists in the database.
- If `name` has the format of `mytable` and `is.temp = TRUE`, the return value is a list. The list has two elements. The first is a logical, which is TRUE if the table/view exists in the database. The second is a character array with 2 elements, whose first is the temporary schema name and the second is the table/view name.

Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also
objects to See Also as `help`, `??`

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
```
db.list()

db.existsObject("madlibtestdata.lin_ornstein", cid)

db.disconnect(cid, verbose = FALSE)

## End(Not run)

db.list  List all the currently active connections with their information

Description

List all the currently active connections with their information including the connection ID, host, user, database, DBMS (database management system), MADlib schema name in the database, and the R package name used to connect to the database.

Usage

db.list()

Value

No value is returned.

Note

Currently, only connection to PostgreSQL and Greenplum databases are supported. Support for other types of DBMS’s will be added in the future.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

db.connect connects to database.
db.disconnect disconnects a connection.
connection info the functions that extract information about the connection.
conn.eql tests whether two connections are the same.
### Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid1 <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
cid2 <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

db.list() # list the two connections
db.disconnect(cid1, verbose = FALSE)
db.disconnect(cid2, verbose = FALSE)

## End(Not run)
```

---

**db.obj-class**

*Abstract Class "db.obj"

---

**Description**

The super class of **db.data.frame** and **db.Rquery**

**Objects from the Class**

A virtual Class: No objects may be created from it.

**Methods**

See **db.data.frame** for all the available methods and functions.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. &lt;fmcquillan@pivotal.io&gt;

**See Also**

**db.data.frame** creates a **db.data.frame** object.
**as.db.data.frame** converts **db.Rquery** object, **data.frame**, or a data file into a **db.data.frame** object and at the same time creates a new table in the database.
**db.data.frame** and **db.Rquery** are the sub-classes.
**lk** or **lookat** displays a part of the table
**db.objects**

List all the existing tables/views in a database with their schema names

**Description**

This function lists all the existing tables and views in a database, together with their schema names.

**Usage**

db.objects(search = NULL, conn.id = 1)

**Arguments**

- **search**
  - A string, default is NULL. List all database objects whose names have the string in them. You can put regular expression here.
- **conn.id**
  - An integer, default is 1. The ID of the database connection.

**Value**

A character array. Each element has the format of 'schema_name.table_name'.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**See Also**

- `db.connect` creates a connection to a database.
- `db.existsObject` tests whether an object exists in the database

**Examples**

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table using as.db.data.frame
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid)

db.objects(conn.id = cid) # list all tables/views

## list all tables/views start with "madlibtestdata.lin"
## where "madlibtestdata" is the schema name
```
db.objects("~madlibtestdata.lin", cid)

db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

**db.q**  
*Execute a SQL query*

---

**Description**

This function sends SQL queries into the connected database to execute, and then extracts the result if there is any.

**Usage**

```r
db(..., nrows = 100, conn.id = 1, sep = " ", verbose = TRUE)
```

```r
.db(..., nrows = 100, conn.id = 1, sep = " ", verbose = TRUE)
```

```r
db.q(..., nrows = 100, conn.id = 1, sep = " ", verbose = TRUE)
```

**Arguments**

- **...** One or multiple SQL query strings. Multiple strings will be concatenated into one SQL query string.
- **nrows** An integer, default is 100. How many rows should be extracted? If it is NULL, "all" or non-positive value, all rows in the result will be loaded into R. For big dataset, you may not want to do this.
- **conn.id** An integer, default is 1. The ID of the connection. See `db.list` for how to list the existing database connections.
- **sep** A string, default is a space character " ". If multiple strings are used in ..., this string is used to separate them in the concatenation.
- **verbose** A logical, default is TRUE. Whether to output the SQL query that you are executing.

**Value**

A `data.frame` that contains the result if the result is not empty. Otherwise, it returns a logical value, which indicates whether the SQL query has been sent to the database successfully.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>
See Also
db.connect, db.objects, db.list.

Examples
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

db("show search_path", conn.id = cid)
.db("drop table if exists tr;",
   "create temp table tr (idx integer,
   val double precision);",
   "insert into tr values (1, 2.3), (2, 3.4)", conn.id = cid)
db.q("select * from tr", conn.id = cid)

db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

db.Rcrossprod-class  Class "db.Rcrossprod"

Description
This is the result generated by crossprod, and a sub-class of db.Rquery

Slots
As a sub-class of db.Rquery, this class contains all the slots that belong to db.Rquery. It also has one additional slot as is described in the following.

.is.crossprod: A vector of logical values, which has the same length as the number of columns. Whether each column is the result of crossprod.
.is.symmetric: A vector of logical values, which has the same length as the number of columns. Whether the column contains matrices that are symmetric.
.dim: Dimension of the matrix represented by this object.

Extends
Class "db.Rquery", directly.

Methods
All methods for db.data.frame can be applied onto this class.
db.Rquery-class

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

db.Rquery is the superclass.
lk or lookat display the matrix

Examples

### Not run:
showClass("db.Rcrossprod")

### set up the database connection
### Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

### x points to table "abalone"
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lookat(crossprod(x[, -c(1, 2)]))

x$arr <- db.array(1, x$length, x$diameter)
lookat(crossprod(x$arr))
db.disconnect(cid, verbose = FALSE)

### End(Not run)

---

db.Rquery-class  Class "db.Rquery" and its sub-class db.Rview-class

Description

An object of this class represents a series of operations applied onto an existing db.data.frame object. These operations are actually a SQL query, which one can choose to materialize in the database using as.db.data.frame. lk can fetch a part of the result of executing the SQL query. Thus one does not need to create a table for every step of the operations, and the data transferred between R and the database is minimized.
Objects from the Class

Objects can be created by almost all functions/methods that can be applied onto \texttt{db.data.frame} except \texttt{content, lk} and \texttt{delete}.

\texttt{db.Rview-class} is a sub-class of \texttt{db.Rquery-class}, and it behaves just like "view" in the databases except that it exists only in R. Usually there is no difference to use \texttt{db.Rview} or \texttt{db.Rquery}. \texttt{as.db.Rview} casts a \texttt{db.Rquery} object into a \texttt{db.Rview} object.

Usually it is NOT recommended to directly manipulate the internal slots of these objects.

Slots

- \texttt{.content}: Object of class "character". The SQL query that represents the operations. The function \texttt{content} can get this value.
- \texttt{.expr}: Object of class "character". An array of expression strings for columns of the table that the SQL query can be materialized into. It is not to used by the normal users.
- \texttt{.source}: Object of class "character". A string, the table/view name which this SQL query is originated. It is not to used by the normal users.
- \texttt{.parent}: Object of class "character". A string. In the SQL query it is the part after "from". It is not to used by the normal users.
- \texttt{.conn.id}: Object of class "numeric", an integer. The ID number of the database connection where \texttt{.source} resides. The functions \texttt{conn.id} and \texttt{conn.id<-} can get and set this value.
- \texttt{.col.name}: Object of class "character". An array of strings. The names of columns of the table that the SQL query can be materialized into. The S4 method \texttt{names, db.obj-method} gets this value.
- \texttt{.key}: Object of class "character". The name of the primary key column name in \texttt{.source}. Currently only one primary key column is supported. This value can be set during the creation of the object when using the function \texttt{db.data.frame}. The functions \texttt{key} and \texttt{key<-} can be used to get and set this value.
- \texttt{.col.data_type}: Object of class "character". The 1D array of column data types of the table that the SQL query can be materialized into. This is not supposed to be used by the normal user.
- \texttt{.col.udt_name}: Object of class "character". The 1D array of column udt names of the table that the SQL query can be materialized into. This is not to be used by normal users.
- \texttt{.where}: Object of class "character". The condition string used in "where" inside the SQL query.
- \texttt{.is.factor}: Object of class "logical". An array of logical values which indicate whether each column of the table that the SQL query can be materialized into is a factor. This is not to be used by the normal users.
- \texttt{.factor.suffix}: Object of class "character". An array of strings for every column. When creating dummy columns for a factor column, we add a random string in the names of the dummy columns to avoid naming conflicts. So a factor column's \texttt{.factor.suffix} is a random string, otherwise it is just an empty string. This is not to be used by the normal users. It is used only the MADlib wrapper functions that support categorical variables.
- \texttt{.factor.ref}: The value of the factor reference level for the regressions. If it is NA, then the regressions automatically select a reference level.
sort: Object of class "list". The list contains the information used for "order by" in the SQL
query.
  by: A string. The column names that are used in "order by".
  order: A string, "" or "desc"
  str: A string, the full "order by ..." string.

.is.agg: logical value, whether this object represents an aggregate operation.

dist.by: A string, the distribution policy for the original data table, which is used to construct
this db.Rquery object, when using Greenplum database or HAWQ. It can be character(8),
which means the original data table is distributed randomly. Or it can be a string of column
names separated by comma, which are the columns that are used in the "distributed by" when
the original table was created.

Extends

Class "db.obj", directly.

Methods

All methods for db.data.frame can be applied onto this class.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

db.data.frame creates a db.data.frame object.

as.db.data.frame converts db.Rquery object, data.frame, or a data file into a db.data.frame
object and at the same time creates a new table in the database.

as.db.Rview converts a db.Rquery object to a db.Rview object.

db.obj is the superclass.

Class db.data.frame is another sub-class of db.obj.

lk display a part of the table

Examples

## Not run:
showClass("db.Rquery")

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname)
delete("abalone", conn.id = cid)
db.search.path

x <- as.db.data.frame(abalone, "abalone", conn.id = cid)

## create several db.Rquery objects
y <- x[,1:2]
z <- x[x$rings > 10,]

dim(z) # get an error

lk(y)

lk(z)

## materialize a db.Rquery object
z <- as.db.data.frame(z, "abalone_rings_larger_10")
delete("abalone_rings_larger_10", conn.id = cid)

dim(z) # no error

db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

**db.search.path**

*Display or set the search path (i.e. default schemas) for a connected session to a database. The user can easily switch to a schema that he has the privilege to write.*

**Description**

Allow the user to check and set the search path for the session that he connects to the database. The search path is a set of schema names separated by commas. These are the default schemas that the programme will search and save tables if a schema name is not given together with the table name in the format of "schema_name.table_name".

**Usage**

```
db.search.path(conn.id = 1, set = NULL)
db.default.schemas(conn.id = 1, set = NULL)
```

**Arguments**

- **conn.id**
  - An integer, default is 1. The ID of the database connection.
- **set**
  - A string, default is NULL. The default schema names separated by commas.

**Value**

When set is NULL, this function prints the current connected session’s search path.
Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

db.connect connects to database, and the parameter default.schemas can be used to set the search path when connecting.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE, default.schemas = "public, madlib")

db.search.path()

db.search.path(set = "public, madlibtestdata")

db.disconnect(cid, verbose = FALSE)
## End(Not run)
```

---

**db.table-class**

*Class* "db.table"

Description

A sub-class of **db.data.frame** which points to tables in the database

Objects from the Class

Objects can be created by calls of **db.data.frame** or **as.db.data.frame**

Slots

As a sub-class, this class has all the slots of **db.data.frame**. Here we list the extra slots.

-.key: Object of class "character". The name of the primary key column name. Currently only one primary key column is supported. This value can be set during the creation of the object when using the function **db.data.frame**. The functions **key** and **key<-** can be used to get and set this value.

-.dim: Object of class "numeric". A two-integer array, the dimension information of the table that this object points to. The first integer is the total row number of the table, and the second is the number of columns of the table. **dim, db.table-method** gets this value.
**db.view-class**

**Extends**
Class "db.data.frame", directly. Class "db.obj", by class "db.data.frame", distance 2.

**Methods**
See *db.data.frame* for all the methods that can take this class of object as an object.

**Author(s)**
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**See Also**
*db.data.frame* creates a *db.data.frame* object.
*as.db.data.frame* converts *db.Rquery* object, *data.frame*, or a data file into a *db.data.frame* object and at the same time creates a new table in the database.
*db.data.frame* is the superclass.
*db.view* is the other subclass of *db.data.frame*
*db.Rquery* is another sub-class of *db.obj*.
*lk* or *lookat* display a part of the table

**Description**
A sub-class of *db.data.frame* which points to tables in the database

**Objects from the Class**
Objects can be created by calls of *db.data.frame* or *as.db.data.frame*

**Slots**
As a sub-class, this class has all the slots of *db.data.frame*. Here we list the extra slots.

.key: Object of class "character". The name of the primary key column name when the view is materialized. The view in the database does not have a primary key. Currently only one primary key column is supported. This value can be set during the creation of the object when using the function *db.data.frame*. The functions *key* and *key<-* can be used to get and set this value.

**Extends**
Class "db.data.frame", directly. Class "db.obj", by class "db.data.frame", distance 2.
Methods

See `db.data.frame` for all the methods that can take this class of object as an object.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

db.data.frame creates a db.data.frame object.
as.db.data.frame converts db.Rquery object, data.frame, or a data file into a db.data.frame object and at the same time creates a new table in the database.
db.data.frame is the superclass.
db.table is the other subclass of db.data.frame
db.Rquery is another subclass of db.obj.
ls or lookat display a part of the table

delete

```
Safely delete a db.obj object or a table/view in the database
```

Description

This function deletes a db.data.frame object together with the table/view that it points to. It deletes a db.Rquery object. It can also directly delete a table or view in the database. When applied onto some composite data objects, it deletes the data table wrapped by them.

Usage

```r
## S4 method for signature 'db.data.frame'
delete(x, cascade = FALSE)
## S4 method for signature 'db.Rquery'
delete(x)
## S4 method for signature 'character'
delete(x, conn.id = 1, is.temp = FALSE, cascade = FALSE)
## S4 method for signature 'arima.css.madlib'
delete(x)
## S4 method for signature 'summary.madlib'
delete(x)
```
Arguments

x  The signature of the method.
   A db.data.frame object, which points to a table or view in the database;
   Or a db.Query object, which represents some operations on an existing db.data.frame
   object;
   Or a string, the table/view name to delete in the database;
   Or an object which is the result of madlib.arima. In the this case, the result
   model tables wrapped by model, residuals and statistics will be deleted.
   Or an object which is the result of madlib.summary (a summary.madlib object).
   In this case the result table created in the database and wrapped by the attribute
   "summary" will be deleted.
   Or an object which is the result of madlib.lm (a lm.madlib or lm.madlib.grps
   object). In this case, the result model table wrapped by model will be deleted.
   Or an object which is the result of madlib.glm with family = "binomial"
   (a logregr.madlib or logregr.madlib.grps object). In this case, the result
   model table wrapped by model will be deleted.
   Or an object which is the result of generic.bagging. In this case, all result
   model tables will be deleted.
   Or an object which is the result of madlib.elnet. In this case all result model
   tables will be deleted.
delete

Or an object which is the result of madlib.rpart. All result tables will be deleted.

conn.id An integer, default is 1. The connection ID to the database.
is.temp A logical, default is FALSE. Whether the table/view is temporary.
cascade A logical, default is FALSE. Whether to delete objects together with all the objects depending on it.

Details

When a db.data.frame object is deleted, the table/view that is associated with it is also deleted.

Value

When x is db.data.frame or table/view name, this function returns a logical value. which is TRUE if the deletion is successful.
No value is returned if x is db.Rquery

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

db.data.frame creates an object pointing to a table/view in the database.
db.objects lists all tables and views in a database together with their schema.
db.existsObject tests whether a table/view exists in the database.
as.db.data.frame creates a db.data.frame from a data.frame, a data file or a db.Rquery.
madlib.lm, madlib.glm, madlib.summary, madlib.arima are MADlib wrapper functions whose results can be safely deleted by this function.

Examples

## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
delete("abalone", cid, is.temp = TRUE)
delete("abalone", cid, is.temp = FALSE)
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid)
lk(x, 10)
dim-methods

Dimension of a table

Description

Display the dimension of the table that a `db.table` object points to.

Usage

```r
## S4 method for signature 'db.table'
dim(x)
```

```r
## S4 method for signature 'db.view'
dim(x)
```

```r
## S4 method for signature 'db.Rquery'
dim(x)
```

Arguments

- `x`: A `db.obj`. Only for `db.table` object, this function gives the dimension of table that `x` points to. For `db.view` and `db.Rquery` objects, an error message is raised.

Value

A two-integer array, where the first integer is the number of rows and the second integer is the number of columns.
Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also
db.obj, db.data.frame, db.table, db.view, db.Rquery are the class hierarchy structure of this package.

Examples
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
delete(“abalone”, conn.id = cid)
x <- as.db.data.frame(abalone, ”abalone”, conn.id = cid, verbose = FALSE)
## preview of a table
lk(x, nrows = 10) # extract 10 rows of data

## get names of all columns
dim(x)

dim(x[,1:3])

db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

eql-methods  

Test if two objects point to the same table

Description
This function checks if two db.obj objects are the equivalent. For objects of class db.data.frame, they need to have the same associated table. For objects of other types, they need to have identical expressions and the same associated table.

Usage
## S4 method for signature 'db.obj,db.obj'
eql(e1,e2)
eql-methods

Arguments

e1, e2
The signature of the method. Both arguments are db.obj objects to be checked for equality.

Details

Objects of type db.data.frame are considered equal if they have the same content representation, and their associated tables have the same name, connected database, and type. Objects of other types derived from db.obj are considered equal if they have the same values for content representation, @.source, @.parent, @.expression, @.where, @.conn.id, @.col.data_type, @.is.factor and @.col.name. Two objects of different types are always considered not equal.

Value

A logical. Returns TRUE if the objects are equal.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

See Also

lk or lookat Displays the actual data in a db.obj object.

Examples

## Not run:

```r

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

delete("abalone", conn.id = cid)
as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)

x <- db.data.frame('abalone', conn.id = cid, key = 'id') # use default connection 1

y <- db.data.frame('abalone', conn.id = cid)

## Check for equality
eql(x,y) # This returns true

## create a db.Rquery object
z <- x[,] # x is a db.data.frame object, but z is not
eql(x,z) # This returns false

db.disconnect(cid, verbose = FALSE)
```

Extract database connection info

Utilities for extracting related information about a database connection

Description

For a given database connection, these functions return the user name, host, database name, info about database management system, connection, the version of MADlib installed on this database, the schema name of MADlib installation, and the R package that is used to connect to this database.

Usage

user(conn.id = 1)
host(conn.id = 1)
dbname(conn.id = 1)
dbms(conn.id = 1)
conn(conn.id = 1)
port(conn.id = 1)
madlib(conn.id = 1)
madlib.version(conn.id = 1)
schema.madlib(conn.id = 1)
conn.pkg(conn.id = 1)

Arguments

conn.id Default value is 1. The database connection ID number. It is an integer.

Value

For user, a string, which is the user name.
For host, a string, which is the host address.
For dbname, a string, which is the database name.
For dbms, a string, which is DBMS version information.
For conn, an object of DBI connection, which can be directly used with packages such as RPostgreSQL.

For port, an integer, which is the port number of the connection.

For madlib, a string, which is the MADlib version information.

For madlib.version, a string, exactly the same as madlib.

For schema.madlib, a string, which is the schema name of MADlib installation.

For conn.pkg, a string, which is the name of the R package that has been used to connect to this database.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

See Also

db.connect creates connections to the databases.
db.disconnect disconnects an existing connection.
db.list lists all the current connections with their information.
conn.eql tests whether two connections are actually the same one.

Examples

## Not run:

```r
## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid1 <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
cid2 <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

user(cid1)
host(cid2)
dbname(cid1) # use default connection 1
dbms(cid1)
madlib(cid1)
madlib.version(cid1)
schema.madlib(cid1)
conn.pkg(cid1)

## conn is mostly for other packages
con <- conn(cid1) # get the connection object
dbListTables(con) # directly use functions in package RPostgreSQL

## This package provides a better function to list all tables/views
db.objects(cid1) # list all tables/views with their schema in connection 1

db.disconnect(cid1, verbose = FALSE)
```
Extract-Replace-methods

Extract or replace a part of `db.obj` objects

Description

Operators acting on `db.obj` objects to extract or replace parts.

Usage

```r
## S4 method for signature 'db.obj'
$x$name
## S4 method for signature 'db.obj'
$x[[i, j, ...]]
## S4 method for signature 'db.obj,ANY,ANY,ANY'
$x[i, j, ...], drop = TRUE]
## S4 replacement method for signature 'db.obj,character'
$x$name <- value
## S4 replacement method for signature 'db.obj,integer'
$x$name <- value
## S4 replacement method for signature 'db.obj,numeric'
$x$name <- value
## S4 replacement method for signature 'db.obj,logical'
$x$name <- value
## S4 replacement method for signature 'db.obj,db.Rquery'
$x$name <- value
## S4 replacement method for signature 'db.obj,ANY,ANY,character'
$x[[i, j]] <- value
## S4 replacement method for signature 'db.obj,ANY,ANY,integer'
$x[[i, j]] <- value
## S4 replacement method for signature 'db.obj,ANY,ANY,numeric'
$x[[i, j]] <- value
## S4 replacement method for signature 'db.obj,ANY,ANY,logical'
$x[[i, j]] <- value
## S4 replacement method for signature 'db.obj,ANY,ANY,db.Rquery'
$x[[i, j]] <- value
## S3 replacement method for class 'db.obj'
$x[i, j] <- value
```

db.disconnect(cid2, verbose = FALSE)

## End(Not run)
Arguments

- **x**: A `db.obj` (either `db.table`, `db.view`, or `db.Rquery`) from which to extract element(s).
- **i, j, ...**: Indices specifying elements to extract or replace. Indices are ‘numeric’ or ‘character’ vectors or `db.Rquery` object or empty (missing) or ‘NULL’.
- **name**: A string. The column name.
- **value**: Any valid value, including `db.Rquery`, character, numeric, integer, and logical object. The value that is used to replace the part of the `db.obj`.
- **drop**: Not implemented yet.

Value

A `db.Rquery` object is returned. For the extraction methods, this is a SQL query to extract the requested subset. For the replacement methods, this is a SQL query representing the modified version of `x`.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

`subset, db.obj-method` Operator to extract elements

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

x$area <- x[,"length"] * x[,"height"] # add a new column

y <- x[,-c(1,2)] # use all columns except the first two

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```
**Mathematical functions that take `db.obj` objects as the argument**

**Description**

Functions that apply onto `db.obj` objects

**Usage**

```r
## S4 method for signature 'db.obj'
ex(x)
## S4 method for signature 'db.obj'
abs(x)
## S4 method for signature 'db.obj'
log(x, ...)
## S4 method for signature 'db.obj'
log10(x)
## S4 method for signature 'db.obj'
sign(x)
## S4 method for signature 'db.obj'
sqrt(x)
## S4 method for signature 'db.obj'
factorial(x)
## S4 method for signature 'db.obj'
sin(x)
## S4 method for signature 'db.obj'
cos(x)
## S4 method for signature 'db.obj'
tan(x)
## S4 method for signature 'db.obj'
asin(x)
## S4 method for signature 'db.obj'
acos(x)
## S4 method for signature 'db.obj'
atan(x)
## S4 method for signature 'db.obj,db.obj'
atan2(y, x)
## S4 method for signature 'db.obj,numeric'
atan2(y, x)
## S4 method for signature 'numeric,db.obj'
atan2(y, x)
```

**Arguments**

`x` `y` `db.obj` object. The function applies to each column of the `db.obj` object. If a column is an array, then the function applies onto each element of the array. If
the data type of the column makes no sense to be used in the function, then a null value is returned.

... Extra parameters. Not implemented.

Value

*db.Rquery* object, which contains the SQL query that computes the operations.

Note

A meaningful expression is generated only when the `.col.data_type` is "numeric", otherwise a "NULL" value is generated.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. &lt;fmcquillan@pivotal.io&gt;

See Also

*db.Rquery* contains a SQL query that does the operations.

Examples

```r
## Not run:
## get the help for a method
## help("*,db.obj,db.obj-method")

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname)

## create a table from the example data.frame "abalone"
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid)
x$rings <- exp(x$rings) # change the values
x$area <- log((x$length + 1) * (x$height + 1)) # add a new column
lk(x$area, 10) # view the actual values computed in database
fit <- madlib.lm(rings ~ area, data = x)
db.disconnect(cid, verbose = FALSE)
## End(Not run)
```
This function runs bootstrap aggregating for a given training function.

Description

A generic function to do bootstrap aggregating for a given machine learning model. The user might need to write a wrapper for the training function so that they could satisfy the format requirements described in the following.

Usage

generic.bagging(train, data, nbags = 10, fraction = 1)

Arguments

- `train`: A training function. It must have only one argument `data`. Given the data, it produces the model.
- `data`: A `db.obj` object, which wraps the data in the database.
- `nbags`: An integer, default is 10. The number of bagging sampling.
- `fraction`: A double, default is 1. The fraction of data in each bagging sample.

Value

A `bagging.model` object, which is actually a list of fitted models.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

References


See Also

- `predict.bagging.model` makes predictions using the result of this function.
- `generic.cv` for cross-validation
- `sample,db.obj-method` samples data from a table
Examples

## Not run:

```r
## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

delete("abalone", conn.id = cid)
as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)
y <- db.data.frame("abalone", conn.id = cid)

fit <- generic.bagging(function(data) {
  madlib.lm(rings ~ . - id - sex, data = data)
}, data = y, nbags = 25, fraction = 0.7)
pred <- predict(fit, newdata = y) # make prediction
lookat(mean((y$rings - pred)^2)) # mean squared error
db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

generic.cv

Generic cross-validation for supervised learning algorithms

Description

This function runs cross-validation for a given supervised learning model, which is specified by the training function, prediction function, and metric function. The user might need to write wrappers for the functions so that they satisfy the format requirements described in the following. This function works on both in-memory and in-database data.

Usage

```r
generic.cv(train, predict, metric, data, params = NULL, k = 10,
approx.cut = TRUE, verbose = TRUE, find.min = TRUE)
```

Arguments

- **train**
  A training function. Its first argument must be a `db.obj` object which is the wrapper for the data in database. Given the data, it produces the model. It can also have other parameters that specifies the model, and these parameters must appear in the list `params`.

- **predict**
  A prediction function. It must have only two arguments, which are the fitted model (the first argument) and the new data input for prediction (the second argument).
metric A metric function. It must have only two arguments. The first argument is the prediction and the second is the data that contains the actual value. This function should measure the difference between the predicted and actual values and produce a single numeric value.

data A db.obj object, which wraps the data in the database, used for cross-validation. Or a data.frame, which contains data in memory.

params A list, default is NULL. The values of each parameters used by the training function. An array of values for each parameter is an element in the list. The value arrays for different parameters do not have to be the same length. The arrays of shorter lengths are circularly expanded to the length of the longest element.

k An integer, default is 10. The cross-validation fold number.

approx.cut A boolean, default is TRUE. Whether to cut the data into k pieces in an approximate way, which is faster than the accurate way. For big data sets, cutting the data into k pieces in an approximate way does not affect the result. See details for more.

verbose A logical value, default is TRUE. Whether to print

find.min A logical value, default is TRUE. Whether the best set of parameters produces the mode with the minimum metric value. Then a model will be trained on the whole data set using the best set of parameters. If it is FALSE, the parameter set with the maximum metric value will be used. This is ignored if params is NULL.

Details

In order to cut the data table into k equal pieces, a column of unique id for every row needs to be attached to the data so that one can cut the data using different ranges of the row id. For example, for a 1000 rows data table, when id is 1-100, 101-200, ..., one can cut the data into 10 pieces. The id should be randomly assigned to the rows for cross-validation to use. Note that the original data is not touched in this process, instead all the data is copied to a new temporary table with the id column created in the new table. Because a unique id is to be randomly assigned to each row, this process cannot be easily parallelized.

When approx.cut is TRUE, which is the default, a column of uniform random integer instead of consecutive integers is created in the temporary table. We apply the same method to cut the data using the different ranges of this column, for example, 1-100, 101-200, etc. Apparently, the k pieces of data do not have an exact equal size, and the sizes of them are only approximately equal. However, for big data sets, the differences are relatively small and should not affect the result. This process does not generate unique ID’s for the rows, but can be easily parallelized, so this method is much faster for big data sets.

Value

If params is NULL, this function returns a list, which contains two elements: err and err.std, which are the errors and its standard deviation.

If params is not NULL, this function returns a cv.generic object, which is a list that contains the following items:

metric A list, which contains
- avg The average metric value for each set of parameters.
The standard deviation for the metric values of each set of parameters.

vals A matrix that contains all the metric value measured, whose rows correspond to different set of parameters, and columns correspond to different folds of cross-validation.

params A data.frame, which contains all the parameter sets.

best The fit that has the optimum metric value.

best.params A list, the set of parameters that produces the optimum metric value.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

generic.bagging does the bootstrap aggregate computation.

Examples

## Not run:

```r
## set up the database connection
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## -----------------------------------------------
dat <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

err <- generic.cv(function(data) {
  madlib.lm(rings ~ . - id - sex, data = data)
}, predict,
  function(predicted, data) {
    lookat(mean((data$rings - predicted)^2))
  }, data = dat, verbose = FALSE)

## -----------------------------------------------
x <- matrix(rnorm(100*20),100,20)
y <- rnorm(100, 0.1, 2)

dat <- data.frame(x, y)
delete("eldata", conn.id = cid)
z <- as.db.data.frame(dat, "eldata", conn.id = cid, verbose = FALSE)

g <- generic.cv(
  train = function (data, alpha, lambda) {
    madlib.elnet(y ~ ., data = data, family = "gaussian",
```
alpha = alpha, lambda = lambda,
control = list(random.stepsize=TRUE))
},
predict = predict,
metric = function (predicted, data) {
  lk(mean((data$y - predicted)^2))
},
data = z,
params = list(alpha=1, lambda=seq(0, 0.2, 0.1)),
k = 5, find.min = TRUE, verbose = FALSE)

plot(g$params$lambda, g$metric$avg, type = 'b')
g$best

## --------------------------------------------------------------

db.disconnect(cid, verbose = FALSE)
## End(Not run)

getTree.rf.madlib  MADlib wrapper function for Random Forest

Description

This function is a wrapper of MADlib’s random forest model get_tree function. The model built using \texttt{madlib.randomForest} is passed as input to this function.

Usage

\texttt{getTree.rf.madlib(object, k=1, \ldots)}

Arguments

- \texttt{object}  A random forest model object built using \texttt{madlib.randomForest}.
- \texttt{k}  Id of the tree to be retrieved. Can range between 1 and maximum number of trees in the forest. default is 1.
- \ldots  Arguments to be passed to or from other methods.

Value

A data frame object similar to R’s \texttt{getTree} result.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>
groups

References


See Also

madlib.randomForest function to train a random forest model.
print.rf.madlib function to print summary of a model fitted through madlib.randomForest
predict.rf.madlib is a wrapper for MADlib’s predict function for random forests.
madlib.lm, madlib.glm, madlib.summary, madlib.arima, madlib.elnet, madlib.rpart are all MADlib wrapper functions.

Examples

## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

lk(x, 10)

## decision tree using abalone data, using default values of minsplit, ## maxdepth etc.
key(x) <- "id"
fit <- madlib.randomForest(rings < 10 ~ length + diameter + height + whole + shell,
data=x)

fit
getTree.rf.madlib(fit, k=2)

db.disconnect(cid)

## End(Not run)

---

groups

Summary information for Logistic Regression output

Description

The function prints the value of each element in the Logistic Regression output object.
Usage

```r
## S3 method for class 'lm.madlib'
groups(x)

## S3 method for class 'lm.madlib.grps'
groups(x)

## S3 method for class 'logregr.madlib'
groups(x)

## S3 method for class 'logregr.madlib.grps'
groups(x)
```

Arguments

- `x` The result of `madlib.lm` or `madlib glm`.

Value

A list that contains the value of each grouping column. The elements of the list are the same as the grouping columns. If `x` is a `lm.madlib` object with one group’s information in it, the elements of the resulting list contain one value for each grouping column. If `x` is `lm.madlib.grps`, which contains multiple groups' information, then each element of the resulting list is a vector with the length equal to the number of different groups. `logregr.madlib` and `logregr.madlib.grps` have the similar interpretation of the results.

If no grouping column is used, this function returns `NULL`.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

- `madlib glm` wrapper for MADlib linear and logistic regressions.
- `madlib.lm` wrapper for MADlib linear regression.
- `predict.lm.madlib`, `predict.lm.madlib.grps`, `predict.logregr.madlib`, `predict.logregr.madlib.grps` make predictions for new data.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
```
## Description
This function launches a shiny server which provides a graphical interface for PivotalR. Press Ctrl+c to stop the shiny server.

## Usage

```r
pivotalr()
```

## Details
The graphical interface for PivotalR is very easy to use. Just follow the instructions on screen. The GUI is still at a very early stage and has only very limited functionality. We will add more functionalities into the GUI in the future versions.

## Author(s)

**Author:** Predictive Analytics Team at Pivotal Inc.

**Maintainer:** Frank McQuillan, Pivotal Inc. \(<\texttt{fmcquillan@pivotal.io}>\)

## References


ifelse  

*Conditional Element Selection*

**Description**

'ifelse' returns a value with the same shape as 'test' which is filled with elements selected from either 'yes' or 'no' depending on whether the element of 'test' is 'TRUE' or 'FALSE'.

**Usage**

```r
## S4 method for signature 'db.obj'
ifelse(test, yes, no)
```

**Arguments**

- `test` A `db.obj` object, which has only one column. The column can be casted into boolean values.
- `yes` A normal value or a `db.obj` object. It is the returned value when `test` is TRUE.
- `no` The returned value when `test` is FALSE.

**Value**

A `db.obj` which has the same length in-database as `test`.

**Author(s)**

- Author: Hong Ooi, Pivotal Inc.
- Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

**See Also**

- `db.obj`

**Examples**

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

## create a new db.obj with one-column,
## and values "small" or "big"
z <- ifelse(x$rings < 10, "small", "big")
```
is.db.data.frame

Check if an object is of type db.data.frame

Description

This function checks if the input is of type db.data.frame.

Usage

is.db.data.frame(x)

Arguments

x

The input can be of any type.

Details

is.db.data.frame() returns TRUE if x is of type db.data.frame. Otherwise, it returns FALSE.

Value

A logical. Returns TRUE if the input is of type db.data.frame

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

as.db.data.frame Convert an object into another object of type db.data.frame.

Examples

## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
tmp <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
x <- db.data.frame(content(tmp), conn.id = cid, key = 'id')
is.factor-methods

## is.factor-methods

Detect whether a 
\texttt{db.obj} object is a categorical object

### Description

This function detects whether a \texttt{db.obj} object is a categorical object.

### Usage

```r
## S4 method for signature 'db.obj'
is.factor(x)
```

### Arguments

- **x**
  - A \texttt{db.obj} object.

### Value

A logical value. When all columns of \texttt{db.obj} are categorical variables, this function returns \texttt{TRUE}.

### Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcqillan@pivotal.io>

### See Also

- \texttt{as.factor,db.obj-method} converts a column \texttt{db.obj} of into categorical variables.

### Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
```
is.na-method

## set sex to be a categorical variable
x$sex <- as.factor(x$sex)

is.factor(x$sex)

is.factor(x)

db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

**is.na-method**  
*Query if the entries in a table are NULL*

---

### Description

This function is equivalent to an SQL query that checks if the entries in a table are NULL.

### Usage

```r
## S4 method for signature 'db.obj'
is.na(x)
```

### Arguments

- `x`  
The signature of the method. A db.obj object.

### Details

`is.na()` creates a db.Rquery object where the NULL entries in a db.obj object are TRUE, and other the entries are FALSE.

### Value

The return value is a db.Rquery object.

### Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

### See Also

`lk` or `lookat` Displays the contents of a db.obj object.
### Get or set the primary key for a table

This function gets or sets the primary key for a `db.obj` table.

#### Usage

```r
key(x)
key(x) <- value
```

#### Arguments

- `x` is a `db.obj` object.
- `value` must be a string.

#### Details

`key()` will return the primary key of a table. If the primary key is not set, `key()` will return the character `0`. If `key()` is being used to set the primary key, then `value` must be a string, and it must match one of the column names in the table.
If this function is used to change the primary key to a new column name, this function does NOT check if all the values in that column are unique.

Value

The return value is the primary key of the table.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

lk or lookat Displays the contents of a db.obj object.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
tmp <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
x <- db.data.frame(content(tmp), key = 'id', conn.id = cid, verbose = FALSE)

## getting the primary key
key(x) # Display the primary key for x

## Changing the primary key
key(x) <- 'length'

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

Description

These binary operators perform logical operations on db.obj objects
Usage

```r
## S4 method for signature 'db.obj'
!x

## S4 method for signature 'db.obj,db.obj'
e1 & e2

## S4 method for signature 'db.obj,db.obj'
e1 | e2

## S4 method for signature 'db.obj,logical'
e1 & e2

## S4 method for signature 'db.obj,logical'
e1 | e2

## S4 method for signature 'logical,db.obj'
e1 & e2

## S4 method for signature 'logical,db.obj'
e1 | e2
```

Arguments

- `e1, e2` logical or `db.obj` object.
- `x` `db.obj` object.

Value

- `db.Rquery` object, which contains the SQL query that computes the logical operations.

Note

A meaningful expression is generated only when the `.col.data_type` is "boolean", otherwise a "NULL" value is generated.

Author(s)

- Author: Predictive Analytics Team at Pivotal Inc.
- Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

- `db.Rquery` contains a SQL query that does the operations.
**madlib.arima**

**Examples**

```r
## Not run:
## get the help for a method
## help("|,db.obj,db.obj-method")

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

lk(x$s-length > 10 & x$s-height < 2,]

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

**madlib.arima**  
*Wrapper for MADlib’s ARIMA model fitting function*

**Description**

Apply ARIM model fitting onto a table that contains time series data. The table must have two columns: one for the time series values, and the other for the time stamps. The time stamp can be anything that can be ordered. This is because the rows of a table does not have inherent order and thus needs to be ordered by the extra time stamp column.

**Usage**

```r
## S4 method for signature 'db.Rquery,db.Rquery'
madlib.arima(x, ts, by = NULL,
order=c(1,1,1), seasonal = list(order = c(0,0,0), period = NA),
include.mean = TRUE, method = "CSS", optim.method = "LM",
optim.control = list(), ...)

## S4 method for signature 'formula,db.obj'
madlib.arima(x, ts, order=c(1,1,1),
seasonal = list(order = c(0,0,0), period = NA), include.mean = TRUE,
method = "CSS", optim.method = "LM", optim.control = list(), ...)
```

**Arguments**

- `x`  
  A formula with the format of time series value ~ time stamp | grouping col_1 + ... + grouping col_n. Or a `db.Rquery` object, which is the time series value. Grouping is not implemented yet. Both time stamp and time series can be valid expressions.
We must specify the time stamp because the table in database has no order of rows, and we have to order they according the given time stamps.

**ts**
If x is a formula object, this must be a `db.obj` object, which contains both the time series and time stamp columns. If x is a `db.Rquery` object, this must be another `db.Rquery` object, which is the time stamp and can be a valid expression.

**by**
A list of `db.Rquery`, the default is NULL. The grouping columns. Right now, this functionality is not implemented yet.

**order**
A vector of 3 integers, default is c(1,1,1). The ARIMA orders p, d, q for AR, I and MA.

**seasonal**
A list of order and period, default is list(order = c(0,0,0), period = NA). The seasonal orders and period. Currently not implemented.

**include.mean**
A logical value, default is TRUE. Whether to estimate the mean value of the time series. If the integration order d (the second element of order) is not zero, `include.mean` is set to FALSE in the calculation.

**method**
A string, the fitting method. The default is "CSS", which uses conditional-sum-of-squares to fit the time series. Right now, only "CSS" is supported.

**optim.method**
A string, the optimization method. The default is "LM", the Levenberg-Marquardt algorithm. Right now, only "LM" is supported.

**optim.control**
A list, default is list(). The control parameters of the optimizer. For `optim.method="LM"`, it can have the following optional parameters:
- max_iter: Maximum number of iterations to run learning algorithm (Default = 100)
- tau: Computes the initial step size for gradient algorithm (Default = 0.001)
- e1: Algorithm-specific threshold for convergence (Default = 1e-15)
- e2: Algorithm-specific threshold for convergence (Default = 1e-15)
- e3: Algorithm-specific threshold for convergence (Default = 1e-15)
- hessian_delta: Delta parameter to compute a numerical approximation of the Hessian matrix (Default = 1e-6)

**Details**
Given a time series of data X, the Autoregressive Integrated Moving Average (ARIMA) model is a tool for understanding and, perhaps, predicting future values in the series. The model consists of three parts, an autoregressive (AR) part, a moving average (MA) part, and an integrated (I) part where an initial differencing step can be applied to remove any non-stationarity in the signal. The model is generally referred to as an ARIMA(p, d, q) model where parameters p, d, and q are non-negative integers that refer to the order of the autoregressive, integrated, and moving average parts of the model respectively.

MADlib’s ARIMA function implements a parallel version of the LM algorithm to maximize the conditional log-likelihood, which is suitable for big data.
\textbf{Value}\\
Returns an \texttt{arima.css.madlib} object, which is a list that contains the following items:
\begin{itemize}
  \item \texttt{coef} \hspace{1em} A vector of double values. The fitting coefficients of AR, MA and mean value (if \texttt{include.mean} is TRUE).
  \item \texttt{s.e.} \hspace{1em} A vector of double values. The standard errors of the fitting coefficients.
  \item \texttt{series} \hspace{1em} A string, the data source table or SQL query.
  \item \texttt{time.stamp} \hspace{1em} A string, the name of the time stamp column.
  \item \texttt{time.series} \hspace{1em} A string, the name of the time series column.
  \item \texttt{sigmaN2} \hspace{1em} the MLE of the innovations variance.
  \item \texttt{loglik} \hspace{1em} the maximized conditional log-likelihood (of the differenced data).
  \item \texttt{iter.num} \hspace{1em} An integer, how many iterations of the LM algorithm is used to fit the time series with ARIMA model.
  \item \texttt{exec.time} \hspace{1em} The time spent on the MADlib ARIMA fitting.
  \item \texttt{residuals} \hspace{1em} A \texttt{db.data.frame} object that points to the table that contains all the fitted innovations.
  \item \texttt{model} \hspace{1em} A \texttt{db.data.frame} object that points to the table that contains the coefficients and standard error. This table is needed by \texttt{predict.arima.css.madlib}.
  \item \texttt{statistics} \hspace{1em} A \texttt{db.data.frame} object that points to the table that contains information including log-likelihood, sigma^2 etc. This table is needed by \texttt{predict.arima.css.madlib}.
  \item \texttt{call} \hspace{1em} A language object. The matched function call.
\end{itemize}

\textbf{Author(s)}

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. \texttt{<fmcquillan@pivotal.io>}

\textbf{References}


\textbf{See Also}

\texttt{madlib.lm, madlib.glm, madlib.summary} are MADlib wrapper functions.
\texttt{delete} deletes the result of this function together with the model, residual and statistics tables.
\texttt{print.arima.css.madlib, show.arima.css.madlib} and \texttt{summary.arima.css.madlib} prints the result in a pretty format.
\texttt{predict.arima.css.madlib} makes forecast of the time series based upon the result of this function.
# madlib.arima

## Examples

```r
## Not run:
library(PivotalR)

corpus <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## set up the database connection
## Assume that .port is port number and .dbname is the database name
x <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

corpus <- data.frame(id =
  c(1, 2, 3, 4, 5, 6, 7, 8, 9, 10),
  val = c(10, 20, 30, 40, 50, 60, 70, 80, 90, 100))

define the table for the data
# create a new table called corpus
create_table(corpus, name = "corpus")

define the variable for the data
# create a new variable called x
create_variable(x, name = "x")

define the model
# create a new model called arima
create_model(arima, name = "arima")

define the query
# create a new query called query
create_query(query, name = "query")

## use formula

```r
s <- madlib.arima(val ~ id, x, order = c(2,0,1))
```

## delete s and the 3 tables: model, residuals and statistics

delete(s)

s # s does not exist any more

## do not use formula

```r
s <- madlib.arima(x$val, x$id, order = c(2,0,1))
```

## 10 forecasts

```r
pred <- predict(s, n.ahead = 10)
```

lookat(sort(pred, F, pred$step_ahead), "all")

## Use expressions

```r
s <- madlib.arima(val+2 ~ I(id + 1), x, order = c(2,0,1))
```
madlib.elnet

MADlib's elastic net regularization for generalized linear models

Description

This function wraps MAlib's elastic net regularization for generalized linear models. Currently linear and logistic regressions are supported.

Usage

madlib.elnet(formula, data, family = c("gaussian", "linear", "binomial", "logistic"), na.action = NULL, na.as.level = FALSE, alpha = 1, lambda = 0.1, standardize = TRUE, method = c("fista", "igd", "sgd", "cd"), control = list(), glmnet = FALSE, ...)

Arguments

formula A formula (or one that can be coerced to that class), specifies the dependent and independent variables.
data A db.obj object. Currently, this parameter is mandatory. If it is an object of class db.Query or db.view, a temporary table will be created, and further computation will be done on the temporary table. After the computation, the temporary will be dropped from the corresponding database.
family A string which indicates which form of regression to apply. Default value is "gaussian". The accepted values are: "gaussian" or "linear": Linear regression; "binomial" or "logistic": Logistic regression. The support for other families will be added in the future.
na.action A string which indicates what should happen when the data contain NAs. Possible values include na.omit, "na.exclude", "na.fail" and NULL. Right now, na.omit has been implemented. When the value is NULL, nothing is done on the R side and NA values are filtered on the MADlib side. User defined na.action function is allowed.
na.as.level A logical value, default: FALSE. Whether to treat NA value as a level in a categorical variable or just ignore it.
alpha A numeric value in [0,1], elastic net mixing parameter. The penalty is defined as (1-alpha)/2||beta||_2^2+alpha||beta||_1. 'alpha=1' is the lasso penalty, and 'alpha=0' the ridge penalty.
lambda A positive numeric value, the regularization parameter.
standardize A logical, default: TRUE. Whether to normalize the data. Setting this to TRUE usually yields better results and faster convergence.
method

A string, default: "fista". Name of optimizer, "fista", "sgd" or "cd". "fista" means the fast iterative shrinkage-thresholding algorithm [1], and "sgd" implements the stochastic gradient descent algorithm [2]. "cd" implements the coordinate descent algorithm [5].

control

A list, which contains the control parameters for the optimizers.

(1) If method is "fista", the allowed control parameters are:
- maxNstepsize A numeric value, default is 4.0. Initial backtracking step size. At each iteration, the algorithm first tries \( \text{stepsize} = \text{max.stepsize} \), and if it does not work out, it then tries a smaller step size, \( \text{stepsize} = \text{stepsize} / \text{eta} \), where \( \text{eta} \) must be larger than 1. At first glance, this seems to perform repeated iterations for even one step, but using a larger step size actually greatly increases the computation speed and minimizes the total number of iterations. A careful choice of \( \text{max.stepsize} \) can decrease the computation time by more than 10 times.
- eta A numeric value, default is 2. If stepsize does not work \( \text{stepsize} / \text{eta} \) is tried. Must be greater than 1.
- use.active.set A logical value, default is FALSE. If use_active_set is TRUE, an active-set method is used to speed up the computation. Considerable speedup is obtained by organizing the iterations around the active set of features – those with nonzero coefficients. After a complete cycle through all the variables, we iterate on only the active set until convergence. If another complete cycle does not change the active set, we are done, otherwise the process is repeated.
- activeset.tolerance A numeric value, default is the value of the tolerance argument (see below). The value of tolerance used during active set calculation.
- random.stepsize A logical value, default is FALSE. Whether to add some randomness to the step size. Sometimes, this can speed up the calculation.

(2) If method is "sgd", the allowed control parameters are:
- stepsize The default is 0.01.
- step.decay A numeric value, the actual setpsize used for current step is \( (\text{previous stepsize}) / \exp(\text{setp.decay}) \). The default value is 0, which means that a constant stepsize is used in SGD.
- threshold A numeric value, default is 1e-10. When a coefficient is really small, set this coefficient to be 0. Due to the stochastic nature of SGD, we can only obtain very small values for the fitting coefficients. Therefore, threshold is needed at the end of the computation to screen out tiny values and hard-set them to zeros. This is accomplished as follows: (1) multiply each coefficient with the standard deviation of the corresponding feature; (2) compute the average of absolute values of re-scaled coefficients; (3) divide each rescaled coefficient with the average, and if the resulting absolute value is smaller than threshold, set the original coefficient to zero.
- parallel A logical value, the default is TRUE. Whether to run the computation on multiple segments. SGD is a sequential algorithm in nature. When running in a distributed manner, each segment of the data runs its own SGD model and then the models are averaged to get a model for each iteration. This averaging might slow down the convergence speed, although we also acquire the ability to process large datasets on multiple machines. This algorithm, therefore, provides the parallel option to allow you to choose whether to do parallel computation.

(3) The common control parameters for both "fista" and "sgd" optimizers:
- **max.iter** An integer, default is 100. The maximum number of iterations that are allowed.

- **tolerance** A numeric value, default is 1e-4. The criteria to end iterations. Basically 1e-4 will produce results with 4 significant digits. Both the "fista" and "sgd" optimizers compute the average difference between the coefficients of two consecutive iterations, and when the difference is smaller than tolerance or the iteration number is larger than max.iter, the computation stops.

- **warmup** A logical value, default is FALSE. If warmup is TRUE, a series of lambda values, which is strictly descent and ends at the lambda value that the user wants to calculate, is used. The larger lambda gives very sparse solution, and the sparse solution again is used as the initial guess for the next lambda’s solution, which speeds up the computation for the next lambda. For larger data sets, this can sometimes accelerate the whole computation and may be faster than computation on only one lambda value.

- **warmup.lambda** A vector of numeric values, default is NULL. The lambda value series to use when warmup is True. The default is NULL, which means that lambda values will be automatically generated. The warmup lambda values start from a large value and end at the lambda value.

- **warmup.lambda.no** An integer Default: 15. How many lambdas are used in warm-up. If warmup.lambda is not NULL, this value is overridden by the number of provided lambda values.

- **warmup.tolerance** A numeric value, the value of tolerance used during warmup. The default is the same as the tolerance value.

(4) The control parameters for "cd" optimizer include max.iter, tolerance, use.active.set and verbose for family = "gaussian"
max.iter, tolerance, use.active.set, verbose, warmup, warmup.lambda.no for family = "binomial"

All parameters have been explained above. The only one left is verbose.

- **verbose** A logical value, whether to output the warning message for "cd" optimizer. See the note section for details.

A logical value, default is FALSE. The R package glmnet states that "Note also that for gaussian, glmnet standardizes y to have unit variance before computing its lambda sequence (and then unstandardizes the resulting coefficients); if you wish to reproduce/compare results with other software, best to supply a standardized y." So if the user wants to compare the result of this function with that of glmnet, he can set this value to be TRUE, which tells this function to do the same data transformation as glmnet in the "gaussian" case so that one can easily compare the results. When family = "binomial", this parameter is ignored.

More arguments, currently not implemented.

**Details**

the objective function for "gaussian" is

1/2 RSS/nobs + lambda*penalty,

and for the other models it is

-loglik/nobs + lambda*penalty.
Value

An object of `elnet.madlib` class, which is actually a list that contains the following items:

- `coef`: A vector, the fitting coefficients.
- `intercept`: A numeric value, the intercept.
- `y.scl`: A numeric value, which is used to scale the dependent values. In the "gaussian" case, it is 1 if `glmnet` is FALSE, and it is the standard deviation of the dependent variable if `glmnet` is TRUE.
- `loglik`: A numeric value, the log-likelihood of the fitting result.
- `standardize`: The `standardize` value in the arguments.
- `iter`: An integer, the iteration number used.
- `ind.str`: A string. The independent variables in an array format string.
- `terms`: A `terms` object, describing the terms in the model formula.
- `model`: A `db.data.frame` object, which wraps the result table of this function. When `method = "cd"`, there is no result table, because all the results are in R side.
- `call`: A language object. The function call that generates this result.
- `alpha`: The `alpha` in the arguments.
- `lambda`: The `lambda` in the arguments.
- `method`: The `method` string in the arguments.
- `family`: The `family` string in the arguments.
- `appear`: An array of strings, the same length as the number of independent variables. The strings are used to print a clean result, especially when we are dealing with the factor variables, where the dummy variable names can be very long due to the inserting of a random string to avoid naming conflicts, see `as.factor, db.obj-method` for details. The list also contains `dummy` and `dummy.expr`, which are also used for processing the categorical variables, but do not contain any important information.
- `max.iter, tolerance`: The `max.iter` and `tolerance` in the control.

Note

The coordinate descent (method = "cd") algorithm is currently only available in PivotalR. In the future, we will also implement it in MADlib. The idea is to do some part of the computation in memory. Due to the memory usage limitation of the database, this method cannot handle the fitting where the number of features is too large (a couple of thousands).

Note

It is strongly recommended that you run this function on a subset of the data with a limited `max.iter` before applying it to the full data set with a large `max.iter` if the data set is big. In the pre-run, you can adjust the parameters to get the best performance and then apply the best set of parameters to the whole data set.
Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <mcquillan@pivotal.io>

References


See Also

generic.cv does k-fold cross-validation. See the examples there about how to use elastic net together with cross-validation.

Examples

```R
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- matrix(rnorm(100*20),100,20)
y <- rnorm(100, 0.1, 2)

dat <- data.frame(x, y)
delete("eldata")
z <- as.db.data.frame(dat, "eldata", conn.id = cid, verbose = FALSE)

fit <- madlib.elnet(y ~ ., data = z, alpha = 0.2, lambda = 0.05, control = list(random.stepsize=TRUE))

fit

lk(mean((z$y - predict(fit, z))^2)) # mean square error

fit <- madlib.elnet(y ~ ., data = z, alpha = 0.2, lambda = 0.05, method = "cd")

fit
```
**Description**

The wrapper function for MADlib's generalized linear regression [7] including the support for multiple families and link functions. Heteroskedasticity test is implemented for linear regression. One or multiple columns of data can be used to separate the data set into multiple groups according to the values of the grouping columns. The requested regression method is applied onto each group, which has fixed values of the grouping columns. Multinomial logistic regression is not implemented yet. Categorical variables are supported. The computation is parallelized by MADlib if the connected database is Greenplum/HAWQ database. The regression computation can also be done on a column which contains an array as its value in the data table.

**Usage**

```r
madlib.glm(formula, data, family = gaussian, na.action = NULL, control = list(), ...)
```

**Arguments**

- **formula**: An object of class `formula` (or one that can be coerced to that class): a symbolic description of the model to be fitted. The details of model specification are given under ‘Details’.
- **data**: An object of `db.obj` class. Currently, this parameter is mandatory. If it is an object of class `db.Rquery` or `db.view`, a temporary table will be created, and further computation will be done on the temporary table. After the computation, the temporary will be dropped from the corresponding database.
- **family**: A string which indicates which form of regression to apply. Default value is “gaussian”. The accepted values are: `gaussian(identity)` (default for gaussian family), `gaussian(log)`, `gaussian(inverse)`, `binomial(logit)` (default for binomial family), `binomial(probit)`, `poisson(log)` (default for poisson family), `poisson(identity)`, `poisson(sqrt)`, `Gamma(inverse)` (default for Gamma family), `Gamma(identity)`, `Gamma(log)`, `inverse.gaussian(1/mu^2)` (default for inverse.gaussian family), `inverse.gaussian(log)`, `inverse.gaussian(identity)`, `inverse.gaussian(identity)`. 
- **na.action**: A string which indicates what should happen when the data contain NAs. Possible values include `na.omit`, "na.exclude", "na.fail" and NULL. Right now, `na.omit` has been implemented. When the value is NULL, nothing is done on the R side and NA values are filtered on the MADlib side. User defined `na.action` function is allowed.
control

A list, extra parameters to be passed to linear or logistic regressions.

na.as.level: A logical value, default is FALSE. Whether to treat NA value as a level in a categorical variable or just ignore it.

For the linear regressions, the extra parameter is hetero. A logical, default is FALSE. If it is TRUE, then Breusch-Pagan test is performed on the fitting model and the corresponding test statistic and p-value are computed.

For logistic regression, one can pass the following extra parameters:

method: A string, default is "irls" (iteratively reweighted least squares [3]), other choices are "cg" (conjugate gradient descent algorithm [4]) and "igd" (stochastic gradient descent algorithm [5]). These algorithm names for logistic regression, namely family=binomial(logit) and use.glm=FALSE in the control list.

max.iter: An integer, default is 10000. The maximum number of iterations that the algorithms will run.

tolerance: A numeric value, default is 1e-5. The stopping threshold for the iteration algorithms.

use.glm: Whether to call MADlib’s GLM function even when the family is gaussian(identity) or binomial(logit). For these two cases, the default behavior is to call MADlib’s linear regression or logistic regression respectively, which might give better performance under certain circumstances. However, if use.glm is TRUE, then the generalized linear function will be used.

Further arguments passed to or from other methods. Currently, no more parameters can be passed to the linear regression and logistic regression.

Details

See madlib.lm for more details.

Value

For the return value of linear regression see madlib.lm for details.

For the logistic regression, the returned value is similar to that of the linear regression. If there is no grouping (i.e. no | in the formula), the result is a logregr.madlib object. Otherwise, it is a logregr.madlib.grps object, which is just a list of logregr.madlib objects.

If MADlib’s generalized linear regression function is used (use.glm=TRUE for family=binomial(logit)), the return value is a glm.madlib object without grouping or a glm.madlib.grps object with grouping.

A logregr.madlib or glm.madlib object is a list which contains the following items:

grouping column(s)

When there are grouping columns in the formula, the resulting list has multiple items, each of which has the same name as one of the grouping columns. All of these items are vectors, and they have the same length, which is equal to the number of distinct combinations of all the grouping column values. Each row of these items together is one distinct combination of the grouping values. When there is no grouping column in the formula, none of such items will appear in the resulting list.
**coef**  A numeric matrix, the fitting coefficients. Each row contains the coefficients for the linear regression of each group of data. So the number of rows is equal to the number of distinct combinations of all the grouping column values.

**log_likelihood**  A numeric array, the log-likelihood for each fitting to the groups. Thus the length of the array is equal to grps.

**std_err**  A numeric matrix, the standard error for each coefficients. The row number is equal to grps.

**z_stats,t_stats**  A numeric matrix, the z-statistics or t-statistics for each coefficient. Each row is for a fitting to a group of the data.

**p_values**  A numeric matrix, the p-values of z_stats. Each row is for a fitting to a group of the data.

**odds_ratios**  Only for logregr.madlib object. A numeric array, the odds ratios [6] for the fittings for all groups.

**condition_no**  Only for logregr.madlib object. A numeric array, the condition number for all combinations of the grouping column values.

**num_iterations**  An integer array, the iteration number used by each fitting group.

**grp.cols**  An array of strings. The column names of the grouping columns.

**has.intercept**  A logical, whether the intercept is included in the fitting.

**ind.vars**  An array of strings, all the different terms used as independent variables in the fitting.

**ind.str**  A string. The independent variables in an array format string.

**call**  A language object. The function call that generates this result.

**col.name**  An array of strings. The column names used in the fitting.

**appear**  An array of strings, the same length as the number of independent variables. The strings are used to print a clean result, especially when we are dealing with the factor variables, where the dummy variable names can be very long due to the inserting of a random string to avoid naming conflicts, see as.factor, db.obj-method for details. The list also contains dummy and dummy.expr, which are also used for processing the categorical variables, but do not contain any important information.

**model**  A db.data.frame object, which wraps the result table of this function.

**terms**  A terms object, describing the terms in the model formula.

**nobs**  The number of observations used to fit the model.

**data**  A db.obj object, which wraps all the data used in the database. If there are fittings for multiple groups, then this is only the wrapper for the data in one group.

**origin.data**  The original db.obj object. When there is no grouping, it is equal to data above, otherwise it is the "sum" of data from all groups.

Note that if there is grouping done, and there are multiple logregr.madlib objects in the final result, each one of them contains the same copy model.
**Note**

See `madlib.lm`'s note for more about the formula format.

For logistic regression, the dependent variable MUST be a logical variable with values being TRUE or FALSE.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**References**


**See Also**

`madlib.lm`, `madlib.summary`, `madlib.arima` are MADlib wrapper functions.

`as.factor` creates categorical variables for fitting.

`delete` safely deletes the result of this function.

**Examples**

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

source_data <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

lk(source_data, 10)

## linear regression conditioned on nation value
```
## Examples for using GLM model

```r
## i.e. grouping
fit <- madlib(glm(rings ~ . - id | sex, data = source_data, heteroskedasticity = T)
fit

## logistic regression

## logistic regression
## The dependent variable must be a logical variable
## Here it is y < 10.
fit <- madlib(glm(rings < 10 ~ . - id ~ 1, data = source_data, family = binomial)
fit <- madlib(glm(rings < 10 ~ sex + length + diameter, 
data = source_data, family = "logistic")

## 3rd example
## The table has two columns: x is an array, y is double precision
dat <- source_data
dat$arr <- db.array(source_data[-c(1,2)])
array.data <- as.db.data.frame(dat)

## Fit to y using every element of x
## This does not work in R's lm, but works in madlib.lm
fit <- madlib(glm(rings < 10 ~ arr, data = array.data, family = binomial)
fit <- madlib(glm(rings < 10 ~ arr[1:2], data = array.data, family = binomial)
fit <- madlib(glm(rings < 10 ~ arr[1:7] + sex | id
fit <- madlib(glm(rings < 10 ~ arr[8] + sex | id

## 4th example
## Step-wise feature selection
start <- madlib(glm(rings < 10 ~ . - id - sex, data = source_data, family = "binomial")
## step(start)

# Examples for using GLM model

fit <- madlib(glm(rings < 10 ~ . - id - sex, data = source_data, family = binomial(probit), 
control = list(max.iter = 10))
fit <- madlib(glm(rings ~ . - id | sex, data = source_data, family = poisson(log), 
control = list(max.iter = 10))
fit <- madlib(glm(rings ~ . - id, data = source_data, family = Gamma(inverse), 
control = list(max.iter = 10))

db.disconnect(cid, verbose = FALSE)

## End(Not run)
Wrapper for MADlib’s Kmeans clustering function

Description

The wrapper function for MADlib’s kmeans clustering [1]. Clustering refers to the problem of partitioning a set of objects according to some problem-dependent measure of similarity. Each centroid represents a cluster that consists of all points to which this centroid is closest. The computation is parallelized by MADlib if the connected database is Greenplum/HAWQ database.

Usage

madlib.kmeans(
  x, centers, iter.max = 10, nstart = 1, algorithm = "Lloyd", key,
  fn.dist = "squared_dist_norm2", agg.centroid = "avg", min.frac = 0.001,
  kmeanspp = FALSE, seeding.sample.ratio=1.0, ...)

Arguments

x An object of db.obj class. Currently, this parameter is mandatory. If it is an object of class db.Rquery or db.view, a temporary table will be created, and further computation will be done on the temporary table. After the computation, the temporary will be dropped from the corresponding database. Data points and predefined centroids (if used) are expected to be stored row-wise, and each point should be of numeric type.

centers A number, a matrix or db.data.frame object. If it is a number, this sets the number of target centroids and the random (or kmeans++) seeding method is used. Otherwise, this parameter is used for initial centers. If it is a matrix, its rows will denote the initial centroid coordinates. Else, this parameter will point to a table in the connected database that contains the initial centroids.

iter.max The maximum number of iterations allowed.

nstart If centers is a number, this parameter specifies how many random sets should be chosen.

algorithm The algorithm to compute the kmeans. Currently disabled (default: “Lloyd”) and kept for the future implementations.

key Name of the column (from the table that is pointed by x) that contains the ids for each point.

fn.dist The distance function used by MADlib to compute the objective function.

agg.centroid The aggregate function used by MADlib to compute the objective function.

min.frac The minimum fraction of centroids reassigned to continue iterating.

kmeanspp Whether to call MADlib’s kmeans++ centroid seeding method.
seeding.sample.ratio
  The proportion of subsample of original dataset to use for kmeans++ centroid seeding method.

... Further arguments passed to or from other methods. Currently, no more parameters can be passed to madlib.kmeans.

Details
See madlib.kmeans for more details.

Value
For the return value of kmeans clustering see madlib.kmeans for details.
MADlib kmeans clustering output is similar to that of the kmeans output of the kmeans function of R package stats. madlib.kmeans also returns an object of class "kmeans" which has a print and a fitted method. It is a list with at least the following components:

cluster A vector of integers (from 1:k) indicating the cluster to which each point is allocated.
centers A matrix of cluster centres.
withinss Vector of within-cluster sum of squares, one component per cluster.
tot.withinss Total within-cluster sum of squares, i.e. sum(withinss).
size The number of points in each cluster.
iter The number of (outer) iterations.

Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References

See Also
madlib.lm, madlib.summary, madlib.arima are MADlib wrapper functions.

Examples
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
madlib.lda

Wrapper for MADlib's Latent Dirichlet Allocation

Description

This function is a wrapper for MADlib's Latent Dirichlet Allocation. The computation is parallelized by MADlib if the connected database is distributed. Please refer to MADlib documentation for details of the algorithm implementation [1].

Usage

madlib.lda(data, topic_num, alpha, beta, iter_num = 20, nstart = 1, best = TRUE,...)

Arguments

data An object of db.obj class. This is the database table containing the documents on which the algorithm will train. The text of each document should be tokenized into 'words'.
topic_num Number of topics.
alpha Dirichlet parameter for the per-doc topic multinomial.
beta      Dirichlet parameter for the per-topic word multinomial.
iter_num  Number of iterations.
nstart    Number of repeated random starts.
best      If TRUE only the model with the minimum perplexity is returned.
...       Other optional parameters. Not implemented.

Value
An lda.madlib object or a list of them, which is a list that contains the following items:

assignments The per-document topic assignments.
document_sums The per-document topic counts.
model_table  The db.table object for accessing the model table in the database.
output_table The db.table object for accessing the output table in the database.
tf_table     The db.table object for accessing the term frequency table in the database.
topic_sums   The per-topic sum of assignments.
topics       The per-word association with topics.

Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References

See Also
predict.lda.madlib is used for prediction-labelling test documents using a learned lda.madlib model.
perplexity.lda.madlib is used for computing the perplexity of a learned lda.madlib model.

Examples

## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

dat <- db.data.frame("_madlib_pivotalr_lda_data_", conn.id = cid, verbose = FALSE)

output.db <- madlib.lda(dat, 2, 0.1, 0.1, 50)
perplexity.db <- perplexity.lda.madlib(output.db)
print(perplexity.db)

## Run LDA multiple times and get the best one
output.db <- madlib.lda(dat, 2, 0.1, 0.1, 50, nstart=2)
perplexity.db <- perplexity.lda.madlib(output.db)
print(perplexity.db)

## Run LDA multiple times and keep all models
output.db <- madlib.lda(dat, 2, 0.1, 0.1, 50, nstart=2, best=FALSE)
perplexity.db <- perplexity.lda.madlib(output.db[[1]])
print(perplexity.db)
perplexity.db <- perplexity.lda.madlib(output.db[[2]])
print(perplexity.db)

db.disconnect(cid)

## End(Not run)

---

madlib.lm  

**Linear regression with grouping support, heteroskedasticity**

**Description**

The wrapper function for MADlib linear regression. Heteroskedasticity can be detected using the Breusch-Pagan test. One or multiple columns of data can be used to separated the data set into multiple groups according to the values of the grouping columns. Linear regression is applied onto each group, which has fixed values of the grouping columns. Categorial variables are supported, see details below. The computation is parallelized by MADlib if the connected database is Greenplum database. The regression computation can also be done on a column that is an array in the data table.

**Usage**

madlib.lm(formula, data, na.action = NULL, hetero = FALSE, na.as.level = FALSE, ...)

**Arguments**

- **formula**: an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted. The details of model specification are given under 'Details'.
- **data**: An object of db.obj class. Currently, this parameter is mandatory. If it is an object of class db.query or db.view, a temporary table will be created, and further computation will be done on the temporary table. After the computation, the temporary will be dropped from the corresponding database.
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madlib.lm

**na.action**
A string which indicates what should happen when the data contain NaNs. Possible values include `na.omit`, "na.exclude", "na.fail" and NULL. Right now, `na.omit` has been implemented. When the value is NULL, nothing is done on the R side and NaN values are filtered on the MADlib side. User defined `na.action` function is allowed.

**hetero**
A logical value with default value FALSE. If it is TRUE, then Breusch-Pagan test is performed on the fitting model and the corresponding test statistic and p-value are computed. See [1] for more details.

**na.as.level**
A logical value, default is FALSE. Whether to treat NA value as a level in a categorical variable or just ignore it.

... More parameters can be passed into this function. Currently, it is just a place holder and any parameter here is not used.

**Details**

For details about how to write a formula, see `formula` for details. "|" can be used at the end of the formula to denote that the fitting is done conditioned on the values of one or more variables. For example, `y ~ x + sin(z) | v + w` will do the fitting each distinct combination of the values of `v` and `w`.

Both the linear regression (this function) and the logistic regression (`madlib.glm`) support categorical variables. Use `as.factor,db.obj-method` to denote that a variable is categorical, and the corresponding dummy variables are created and fitted. See `as.factor,db.obj-method` for more.

**Value**

If there is no grouping (i.e. no `|` in the formula), the result is a `lm.madlib` object. Otherwise, it is a `lm.madlib.grps` object, which is just a list of `lm.madlib` objects.

A `lm.madlib` object is a list which contains the following items:

**grouping column(s)**
When there are grouping columns in the formula, the resulting list has multiple items, each of which has the same name as one of the grouping columns. All of these items are vectors, and they have the same length, which is equal to the number of distinct combinations of all the grouping column values. Each row of these items together is one distinct combination of the grouping values. When there is no grouping column in the formula, none of such items will appear in the resulting list.

**coef**
A numeric matrix, the fitting coefficients. Each row contains the coefficients for the linear regression of each group of data. So the number of rows is equal to the number of distinct combinations of all the grouping column values. The number of columns is equal to the number features (including intercept if it presents in the formula).

**r2**
A numeric array. R2 values for all combinations of the grouping column values.

**std_err**
A numeric matrix, the standard error for each coefficients.

**t_stats**
A numeric matrix, the t-statistics for each coefficient, which is the absolute value of the ratio of `std_err` and `coef`. 

p_values  A numeric matrix, the p-values of t_stats. Each row is for a fitting to a group of the data.

condition_no  A numeric array, the condition number for all combinations of the grouping column values.

bp_stats  A numeric array when hetero = TRUE, the Breusch-Pagan test statistics for each combination of the grouping column values.

bp_p_value  A numeric array when hetero = TRUE, the Breusch-Pagan test p value for each combination of the grouping column values.

grps  An integer, the number of groups that the data is divided into according to the grouping columns in the formula.

grp.cols  An array of strings. The column names of the grouping columns.

has.intercept  A logical, whether the intercept is included in the fitting.

ind.vars  An array of strings, all the different terms used as independent variables in the fitting.

ind.str  A string. The independent variables in an array format string.

call  A language object. The function call that generates this result.

col.name  An array of strings. The column names used in the fitting.

appear  An array of strings, the same length as the number of independent variables. The strings are used to print a clean result, especially when we are dealing with the factor variables, where the dummy variable names can be very long due to the inserting of a random string to avoid naming conflicts, see as.factor, db.obj-method for details. The list also contains dummy and dummy.expr, which are also used for processing the categorical variables, but do not contain any important information.

model  A db.data.frame object, which wraps the result table of this function.

terms  A terms object, describing the terms in the model formula.

nobs  The number of observations used to fit the model.

data  A db.obj object, which wraps all the data used in the database. If there are fittings for multiple groups, then this is only the wrapper for the data in one group.

origin.data  The original db.obj object. When there is no grouping, it is equal to data above, otherwise it is the "sum" of data from all groups.

Note that if there is grouping done, and there are multiple lm.madlib objects in the final result, each one of them contains the same copy model.

Note

| is not part of standard R formula object, but many R packages use | to add their own functionalities into formula object. However, | has different meanings and usages in different packages. The user must be careful that usage of | in PivotalR-package may not be the same as the others.
Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References

[2] Documentation of linear regression in MADlib v0.6, http://doc.madlib.net/v0.6/group__grp__linreg.html.

See Also

madlib.glm, madlib.summary, madlib.arima are MADlib wrapper functions.

as.factor creates categorical variables for fitting.
delete safely deletes the result of this function.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

## linear regression conditioned on nation value
## i.e. grouping
fit <- madlib.lm(rings ~ . - id | sex, data = x, heteroskedasticity = T)
fit

## use I(.) for expressions
fit <- madlib.lm(rings ~ length + diameter + shell + I(diameter^2),
data = x, heteroskedasticity = T)
fit # display the result

## Another example
fit <- madlib.lm(rings ~ . - id | sex + (id < 2000), data = x)

## 3rd example
## The table has two columns: x is an array, y is double precision
dat <- x
dat$arr <- db.array(x[-c(1,2)])
array.data <- as.db.data.frame(dat)

## Fit to y using every element of x
## This does not work in R’s lm, but works in madlib.lm
fit <- madlib.lm(rings ~ arr, data = array.data)
```
```r
madlib.randomForest

fit <- madlib.lm(rings ~ arr - arr[1], data = array.data)
fit <- madlib.lm(rings ~ . - arr[1:2], data = array.data)
fit <- madlib.lm(as.integer(rings < 10) ~ . - arr[1:2], data = array.data)

## 4th example
## Step-wise feature selection
start <- madlib.lm(rings ~ . - id - sex, data = x)
## step(start)

db.disconnect(cid)

## End(Not run)
```

---

### Description

This function is a wrapper of MADlib's random forest model training function. The resulting forest is stored in a table in the database, and one can also view the result from R using `print.rf.madlib`.

### Usage

```r
madlib.randomForest(formula, data, id = NULL, ntree = 100, mtry = NULL, importance = FALSE, nPerm = 1, na.action = NULL, control, na.as.level = FALSE, verbose = FALSE, ...)
```

### Arguments

- **formula**: A formula object, intercept term will automatically be removed. Factors will not be expanded to their dummy variables. Grouping syntax is also supported, see `madlib.lm` and `madlib.glm` for more details.
- **data**: A `db.obj` object, which wraps the data in the database.
- **id**: A string, the index for each row. If `key` has been specified for `data`, the key will be used as the ID unless this argument is also specified. We have to have this specified so that `predict.rf.madlib`'s result can be compared with the original data.
- **ntree**: An integer, maximum number of trees to grow in the random forest model, default is 100.
- **mtry**: An integer, number of features randomly selected for each split.
- **importance**: A boolean, whether or not to calculate variable importance, default is `FALSE`.
- **nPerm**: An integer, number of times to permute each feature value while calculating variable importance, default is 1.
na.action  A function, which filters the NULL values from the data. Not implemented yet.
control  A list, which includes parameters for the fit. Supported parameters include:
'minsplit' - minimum number of observations that must be present in a node for a split to be attempted. default is minsplit=20
'minbucket' - Minimum number of observations in any terminal node, default is min_split/3
'maxdepth' - Maximum depth of any node, default is maxdepth=10
'nbins' - Number of bins to find possible node split threshold values for continuous variables, default is 100 (Must be greater than 1)
'max_surrogates' - Number of surrogate splits at each node in the trees constructed.
na.as.level  A boolean, indicating if NULL value for a categorical variable is treated as a distinct level, default is na.as.level=false
verbose  A boolean, indicating whether or not to print more info, default is verbose=false
...  Arguments to be passed to or from other methods.

Value
An S3 object of type rf.madlib in the case of non-grouping, and of type rf.madlib.grp in the case of grouping.

Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References

See Also
print.rf.madlib function to print summary of a model fitted through madlib.randomForest
predict.rf.madlib is a wrapper for MADlib’s predict function for random forests.
madlib.lm, madlib.glm, madlib.summary, madlib.arima, madlib.elnet, madlib.rpart are all MADlib wrapper functions.

Examples
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

## decision tree using abalone data, using default values of msplit, 
## maxdepth etc.
key(x) <- "id"
fit <- madlib.randomForest(rings < 10 ~ length + diameter + height + whole + shell, 
data=x)
fit

## Another example, using grouping
fit <- madlib.randomForest(rings < 10 ~ length + diameter + height + whole + shell | sex, 
data=x)
fit
db.disconnect(cid)

## End(Not run)

### madlib.rpart

**MADlib wrapper function for Decision Tree**

**Description**

This function is a wrapper of MADlib’s decision tree model training function. The resulting tree is stored in a table in the database, and one can also view the result from R using `plot.dt.madlib`, `text.dt.madlib` and `print.dt.madlib`.

**Usage**

```r
madlib.rpart(formula, data, weights = NULL, id = NULL, na.action = NULL, parms, 
control, na.as.level = FALSE, verbose = FALSE, ...)
```

**Arguments**

- **formula**: A formula object, intercept term will automatically be removed. Factors will not be expanded to their dummy variables. Grouping syntax is also supported, see `madlib.lm` and `madlib glm` for more details.
- **data**: A `db.obj` object, which wraps the data in the database.
- **weights**: A string, the column name for the weights.
- **id**: A string, the index for each row. If `key` has been specified for `data`, teh key will be used as the ID unless this argument is also specified. We have to have this specified so that `predict.dt.madlib`’s result can be compared with the original data.
- **na.action**: A function, which filters the NULL values from the data. Not implemented yet.
- **parms**: A list, which includes parameters for the splitting function. Supported parameters include: ‘split’ specifying which split function to use. Options are ‘gini’, ‘misclassification’ and ‘entropy’ for classification, and ‘mse’ for regression. Default is ‘gini’ for classification and ‘mse’ for regression.
control A list, which includes parameters for the fit. Supported parameters include:
'minsplit' - minimum number of observations that must be present in a node for
a split to be attempted. default is minsplit=20
'minbucket' - Minimum number of observations in any terminal node, default is
min_split/3
'maxdepth' - Maximum depth of any node, default is maxdepth=10
'nbins' - Number of bins to find possible node split threshold values for contin-
uous variables, default is 100 (Must be greater than 1)
'cp' - Cost complexity parameter, default is cp=0.01
'n_folds' - Number of cross-validation folds
'max_surrogates' - The number of surrogates number
na.as.level A boolean, indicating if NULL value for a categorical variable is treated as a
distinct level, default is na.as.level=false
verbose A boolean, indicating whether or not to print more info, default is verbose=false
... Arguments to be passed to or from other methods.

Value
An S3 object of type dt.madlib in the case of non-grouping, and of type dt.madlib.grp in the case of
grouping.

Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References

See Also
plot.dt.madlib, text.dt.madlib, print.dt.madlib are visualization functions for a model fit-
ted through madlib.rpart
predict.dt.madlib is a wrapper for MADlib’s predict function for decision trees.
madlib.lm, madlib.glm, madlib.summary, madlib.arima, madlib.elnet are all MADlib wrap-
er functions.

Examples
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
madlib.summary

```r
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

## decision tree using abalone data, using default values of minsplit, 
## maxdepth etc.
key(x) <- "id"
fit <- madlib.rpart(rings < 10 ~ length + diameter + height + whole + shell, 
data=x, parms = list(split='gini'), control = list(cp=0.005))
fit

## Another example, using grouping
fit <- madlib.rpart(rings < 10 ~ length + diameter + height + whole + shell | sex, 
data=x, parms = list(split='gini'), control = list(cp=0.005))
fit
db.disconnect(cid)

## End(Not run)
```

---

**madlib.summary**

**Data summary function**

**Description**

'summary' is a generic function used to produce summary statistics of any data table. The function invokes particular methods' from the MADlib library to provide an overview of the data. The computation is parallelized by MADlib if the connected database is Greenplum database.

**Usage**

```r
madlib.summary(x, target.cols = NULL, grouping.cols = NULL, 
get.distinct = TRUE, get.quartiles = TRUE, 
ntile = NULL, n.mfv = 10, estimate = TRUE, 
interactive = FALSE)

## S4 method for signature 'db.obj'
summary(object, target.cols = NULL, grouping.cols = NULL, 
get.distinct = TRUE, get.quartiles = TRUE, 
ntile = NULL, n.mfv = 10, estimate = TRUE, 
interactive = FALSE)
```

**Arguments**

- `x`: An object of `db.obj` class. Currently, this parameter is mandatory. If it is an object of class `db.query` or `db.view`, a temporary table will be created, and further computation will be done on the temporary table. After the computation, the temporary will be dropped from the corresponding database.

- `target.cols`: Vector of string. Default value is NULL. Column names in the table for which the summary is desired. When NULL all summary of all columns are returned.
grouping.cols List of string. Default value is NULL. Column names in the table by which to group the data. When NULL no grouping of data is performed.

get.distinct Logical. Default value is TRUE. Are distinct values required in the summary?

get.quartiles Logical. Default value is TRUE. Are quartile values required in the summary?

ntile Vector of floats. Default value is NULL. Vector of quantiles required as part of the summary.

n.mfv Integer. Default value is 10. How many ‘most-frequent-values’ (MFVs) to compute?

estimate Logical. Default value is TRUE. Should an estimated computation be used to compute values for distincts and MFVs (as opposed to an exact but slow method)?

interactive Logical. Default is FALSE. If x is of type db.view, then extracting data from it would actually compute the view, which might take a longer time, especially for large data sets. When interactive is TRUE, this function will ask the user whether to continue to extract data from the view.

Value

A data.frame object. Each column in the table (or target.cols) is a row in the result data frame. Each column of the data frame is described below:

group_by character. Group-by column names (NA if none provided)

group_by_value character. Values of the group-by columns (NA if no grouping)

target_column character. Targeted column values for which summary is requested

column_number integer. Physical column number for the target column in the database

data_type character. Data type of target column. Standard database descriptors will be displayed

row_count numeric. Number of rows for the target column

distinct_values numeric. Number of distinct values in the target column

missing_values numeric. Number of missing values in the target column

blank_values numeric. Number of blank values (blanks are defined as values with only whitespace)

fraction_missing numeric. Percentage of total rows that are missing. Will be expressed as a decimal (e.g. 0.3)

fraction_blank numeric. Percentage of total rows that are blank. Will be expressed as a decimal (e.g. 0.3)

mean numeric. Mean value of target column (if target is numeric, else NA)

variance numeric. Variance of target columns (if target is numeric, else NA for strings)

min numeric. Min value of target column (for strings this is the length of the shortest string)
The `data.frame` has an extra attribute named "summary", which is a `db.data.frame` object and wraps the result table created by MADlib inside the database. One can access this object using `attr(res, "summary"),` where `res` is the result of this function.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

**See Also**

`madlib.lm, madlib.glm, madlib.arima` are MADlib wrapper functions.
`delete` safely deletes the result of this function.

**Examples**

```r
### Not run:
### get the help for a method
### help("madlib.summary")

### set up the database connection
### Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
delete("abalone", conn.id = cid)
as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)
x <- db.data.frame("abalone", conn.id = cid, verbose = FALSE)

lk(x, 10)

# madlib.summary
summary_result <- madlib.summary(x)
print(summary_result)

# madlib.summary
summary_result <- madlib.summary(x, target.cols=c('rings', 'length', 'diameter'),
                                  grouping.cols=c('sex'),
                                  grouping.cols=c('class'),
                                  grouping.cols=c('species'))
```
Description

This function wraps MADlib's SVM for classification, regression and novelty detection.

Usage

madlib.svm (formula, data,
    na.action = NULL, na.as.level = FALSE,
    type = c("classification", "regression", "one-class"),
    kernel = c("gaussian", "linear", "polynomial"),
    degree = 3, gamma = NULL, coef0 = 1.0, class.weight = NULL,
    tolerance = 1e-10, epsilon = NULL, cross = 0, lambda = 0.01,
    control = list(), verbose = FALSE, ...)

Arguments

formula an object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted. The details of model specification are given under 'Details'.

data An object of db.obj class. Currently, this parameter is mandatory. If it is an object of class db.Rquery or db.view, a temporary table will be created, and further computation will be done on the temporary table. After the computation, the temporary will be dropped from the corresponding database.

na.action A string which indicates what should happen when the data contain NAs. Possible values include na.omit, "na.exclude", "na.fail" and NULL. Right now, na.omit has been implemented. When the value is NULL, nothing is done on the R side and NA values are filtered on the MADlib side. User defined na.action function is allowed.

na.as.level A logical value, default is FALSE. Whether to treat NA value as a level in a categorical variable or just ignore it.
**madlib.svm**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Default Value</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>type</strong></td>
<td>A string, default: &quot;classification&quot;. Indicate type of analysis to perform: &quot;classification&quot;, &quot;regression&quot; or &quot;one-class&quot;.</td>
<td></td>
</tr>
<tr>
<td><strong>kernel</strong></td>
<td>A string, default: &quot;gaussian&quot;. Type of kernel. Currently three kernel types are supported: 'linear', 'gaussian', and 'polynomial'.</td>
<td></td>
</tr>
<tr>
<td><strong>degree</strong></td>
<td>Default: 3. The parameter needed for polynomial kernel</td>
<td></td>
</tr>
<tr>
<td><strong>gamma</strong></td>
<td>Default: 1/num_features. The parameter needed for gaussian kernel</td>
<td></td>
</tr>
<tr>
<td><strong>coef0</strong></td>
<td>Default: 1.0. The independent term in polynomial kernel</td>
<td></td>
</tr>
<tr>
<td><strong>class.weight</strong></td>
<td>Default: 1.0. Set the weight for the positive and negative classes. If not given, all classes are set to have weight one. If class_weight = balanced, values of y are automatically adjusted as inversely proportional to class frequencies in the input data i.e. the weights are set as n_samples / (n_classes * bincount(y)). Alternatively, class_weight can be a mapping, giving the weight for each class. Eg. For dependent variable values 'a' and 'b', the class_weight can be a: 2, b: 3. This would lead to each 'a' tuple's y value multiplied by 2 and each 'b' y value will be multiplied by 3. For regression, the class weights are always one.</td>
<td></td>
</tr>
<tr>
<td><strong>tolerance</strong></td>
<td>Default: 1e-10. The criterion to end iterations. The training stops whenever the difference between the training models of two consecutive iterations is smaller than tolerance or the iteration number is larger than max_iter.</td>
<td></td>
</tr>
<tr>
<td><strong>epsilon</strong></td>
<td>Default: [0.01]. Determines the epsilon for epsilon-SVR. Ignored during classification. When training the model, differences of less than epsilon between estimated labels and actual labels are ignored. A larger epsilon will yield a model with fewer support vectors, but will not generalize as well to future data. Generally, it has been suggested that epsilon should increase with noisier data, and decrease with the number of samples. See [5].</td>
<td></td>
</tr>
<tr>
<td><strong>cross</strong></td>
<td>Default: 0. Number of folds (k). Must be at least 2 to activate cross validation. If a value of k &gt; 2 is specified, each fold is then used as a validation set once, while the other k - 1 folds form the training set.</td>
<td></td>
</tr>
<tr>
<td><strong>lambda</strong></td>
<td>Default: [0.01]. Regularization parameter. Must be non-negative.</td>
<td></td>
</tr>
<tr>
<td><strong>control</strong></td>
<td>A list, which contains the more control parameters for the optimizer.</td>
<td></td>
</tr>
<tr>
<td></td>
<td>- <strong>init.stepsize</strong>: Default: [0.01]. Also known as the initial learning rate. A small value is usually desirable to ensure convergence, while a large value provides more room for progress during training. Since the best value depends on the condition number of the data, in practice one often searches in an exponential grid using built-in cross validation; e.g., &quot;init_stepsize = [1, 0.1, 0.001]&quot;. To reduce training time, it is common to run cross validation on a subsampled dataset, since this usually provides a good estimate of the condition number of the whole dataset. Then the resulting init_stepsize can be run on the whole dataset.</td>
<td></td>
</tr>
<tr>
<td></td>
<td>- <strong>decay.factor</strong>: Default: [0.9]. Control the learning rate schedule: 0 means constant rate; &lt;1 means inverse scaling, i.e. stepsizes = init_stepsize / iteration; &gt;0 means exponential decay, i.e. stepsizes = init_stepsize * decay_factor*iteration.</td>
<td></td>
</tr>
<tr>
<td></td>
<td>- <strong>max.iter</strong>: Default: [100]. The maximum number of iterations allowed.</td>
<td></td>
</tr>
<tr>
<td></td>
<td>- <strong>norm</strong>: Default: 'L2'. Name of the regularization, either 'L2' or 'L1'.</td>
<td></td>
</tr>
</tbody>
</table>
|               | - **eps.table**: Default: NULL. Name of the input table that contains values of epsilon for different groups. Ignored when grouping_col is NULL. Define this
input table if you want different epsilon values for different groups. The table consists of a column named epsilon which specifies the epsilon values, and one or more columns for grouping_col. Extra groups are ignored, and groups not present in this table will use the epsilon value specified in parameter epsilon.

- validation.result: Default: NULL. Name of the table to store the cross validation results including the values of parameters and their averaged error values. For now, metric like 0-1 loss is used for classification and mean square error is used for regression. The table is only created if the name is not NULL.

verbose
A logical value, default: FALSE. Verbose output of the results of training.

...  
More parameters can be passed into this function. Currently, it is just a placeholder and any parameter here is not used.

Details
For details about how to write a formula, see formula for details. "|" can be used at the end of the formula to denote that the fitting is done conditioned on the values of one or more variables. For example, y ~ x + sin(z) | v + w will do the fitting each distinct combination of the values of v and w.

Value
If there is no grouping (i.e. no | in the formula), the result is a svm.madlib object. Otherwise, it is a svm.madlib.grps object, which is just a list of svm.madlib objects.

A svm.madlib object is a list which contains the following items:

coeff  
A vector, the fitting coefficients.

groups  
An integer, the number of groups that the data is divided into according to the grouping columns in the formula.

groups.cols  
An array of strings. The column names of the grouping columns.

has.intercept  
A logical, whether the intercept is included in the fitting.

ind.vars  
An array of strings, all the different terms used as independent variables in the fitting.

ind.str  
A string. The independent variables in an array format string.

call  
A language object. The function call that generates this result.

col.name  
An array of strings. The column names used in the fitting.

appear  
An array of strings, the same length as the number of independent variables. The strings are used to print a clean result, especially when we are dealing with the factor variables, where the dummy variable names can be very long due to the inserting of a random string to avoid naming conflicts, see as.factor, db.obj-method for details. The list also contains dummy and dummy.expr, which are also used for processing the categorical variables, but do not contain any important information.

model  
A db.data.frame object, which wraps the model table of this function.

model.summary  
A db.data.frame object, which wraps the summary table of this function.
model.random  A `db.data.frame` object, which wraps the kernel table of this function. Only created when non-linear kernel is used.

terms  A `terms` object, describing the terms in the model formula.
nobs  The number of observations used to fit the model.
data  A `db.obj` object, which wraps all the data used in the database. If there are fittings for multiple groups, then this is only the wrapper for the data in one group.
origin.data  The original `db.obj` object. When there is no grouping, it is equal to `data` above, otherwise it is the "sum" of `data` from all groups.

Note that if there is grouping done, and there are multiple `svm.madlib` objects in the final result, each one of them contains the same copy `model`.

Note

`|` is not part of standard R formula object, but many R packages use `|` to add their own functionalities into formula object. However, `|` has different meanings and usages in different packages. The user must be careful that usage of `|` in `PivotalR-package` may not be the same as the others.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

`madlib.lm`, `madlib.summary`, `madlib.arima` are MADlib wrapper functions.
`as.factor` creates categorical variables for fitting.
`delete` safely deletes the result of this function.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

data <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(data, 10)

## svm regression
## i.e. grouping on multiple columns
fit <- madlib.svm(length ~ height + shell | sex + (rings > ?), data = data, type = "regression")
fit

## use I(.) for expressions
fit <- madlib.svm(rings > 7 ~ height + shell + diameter + I(diameter^2),
```

margins

Compute the marginal effects of regression models

Description

margins calculates the marginal effects of the variables given the result of regressions (madlib.lm, madlib.glm etc). Vars lists all the variables used in the regression model. Terms lists the specified terms in the original model. Vars and Terms are only used in margins's dydx option.

Usage

## S3 method for class 'lm.madlib'
margins(model, dydx = ~Vars(model), newdata = model$data, at.mean = FALSE, factor.continuous = FALSE, na.action = NULL, ...)

## S3 method for class 'lm.madlib.grps'
margins(model, dydx = ~Vars(model), newdata = lapply(model, function(x) x$data), at.mean = FALSE, factor.continuous = FALSE, na.action = NULL, ...)

## S3 method for class 'logregr.madlib'
margins(model, dydx = ~Vars(model), newdata = model$data, at.mean = FALSE, factor.continuous = FALSE, na.action = NULL, ...)

## S3 method for class 'logregr.madlib.grps'
margins(model, dydx = ~Vars(model), newdata = lapply(model, function(x) x$data), at.mean = FALSE, factor.continuous = FALSE, na.action = NULL, ...)

## S3 method for class 'margins'
print(x, digits = max(3L,getOption("digits") - 3L), ...)

data = data, type = "classification")
fit # display the result

## Adding new column for training
dat <- data
dat$arr <- db.array(data[,c(1,2)])
array.data <- as.db.data.frame(dat)
fit <- madlib.svm(rings > 7 ~ arr, data = array.data)

db.disconnect(cid)

## End(Not run)
Vars(model)
Terms(term = NULL)

Arguments

model  The result of `madlib.lm`, `madlib.glm`, which represents a regression model for the training data.
dydx  A formula, and the default is ~ Vars(model), which tells the function to compute the marginal effects for all the variables that appear in the model. ~ . will compute the marginal effects of all variables in newdata. Use the normal formula to specify which variables’ marginal effects are to be computed.
newdata  A db.obj object, which represents the data in the database. The default is the data used to train the regression model, but the user can freely use other data sets.
at.mean  A logical, the default is FALSE. Whether to compute the marginal effects at the mean values of the variables.
factor.continuous  A logical, the default is FALSE. Whether to compute the marginal effects of factors by treating them as continuous variables. See "details" for more explanation.
na.action  A string which indicates what should happen when the data contain NAs. Possible values include `na.omit`, "na.exclude", "na.fail" and NULL. Right now, `na.omit,db.obj-method` has been implemented. When the value is NULL, nothing is done on the R side and NA values are filtered out and omitted on the MADlib side. User defined na.action function is allowed, and see `na.omit,db.obj-method` for the preferred function interface.
...  Other arguments, not implemented.
x  The result of margins function, which is of the class "margins".
digits  A non-null value for ‘digits’ specifies the minimum number of significant digits to be printed in values. The default, ‘NULL’, uses ‘getOption("digits")’. (For the interpretation for complex numbers see `signif`.) Non-integer values will be rounded down, and only values greater than or equal to 1 and no greater than 22 are accepted.
term  A vector of integers, the default is NULL. When term=1, compute the marginal effects of the 1-th term. Even if this term contains multiple variables, we treat it as a variable independent of all others. When term=NULL, the marginal effects of all terms are calculated. In the final result, marginal effect results for ".term.1", ".term.2" etc will be shown. By comparing with `names(model$coef)`, one can easily figure out which term corresponds to which expression. (Intercept) term’s marginal effect cannot be computed using this (One can create an extra column that equals 1 and use it as a variable without using intercept by add -1 into the fitting formula).
Details
For a continuous variable, its marginal effects is just the first derivative of the response function with respect to the variable. For a categorical variable, it is usually more meaningful to compute the finite difference of the response function for the variable being 1 and 0. The finite difference marginal effect measures how much more the response function would be compared with the reference category. The reference category for a categorical variable can be changed by `relevel`.

Value
margins function returns a `margins` object, which is a `data.frame`. It contains the following item:

- **Estimate** The marginal effect values for all variable that have been specified in `dydx`.
- **Std. Error** The standard errors for the marginal effects.
- **t value, z value** The t statistics (for linear regression) or z statistics (for logistic regression).
- **Pr(>|t|), Pr(>|z|)** The corresponding p values.

`vars` returns a vector of strings, which are the variable names that have been used in the regression model.

Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References

See Also
`relevel` changes the reference category.
`madlib.lm, madlib(glm)` compute linear and logistic regressions.

Examples
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merge-method

Computing a join on two tables

Description

This method is equivalent to a database join on two tables, and the merge can be by common column or row names. It supports the equivalent of inner, left-outer, right-outer, and full-outer join operations. This method is similar to `merge.data.frame`.

Usage

```r
## S4 method for signature 'db.obj,db.obj'
merge(x, y, by = intersect(names(x), names(y)),
      by.x = by, by.y = by, all = FALSE, all.x = all, all.y = all,
      key = x@key, suffixes = c("_x","_y"), ...)
```

Arguments

- `x,y` The signature of the method. Both argument are `db.obj` objects, and their associated tables will be merged.
- `by,by.x,by.y` specifications of the columns used for merging. See 'Details'.
- `all` logical; `all = L` is shorthand for `all.x = L` and `all.y = L`, where `L` is either `TRUE` or `FALSE`.

```r
margins(fit)
margins(fit, at.mean = TRUE)
margins(fit, factor.continuous = TRUE)
margins(fit, dydx = ~ Vars(model) + Terms())

fit <- madlib.glm(rings < 10 ~ length + diameter*sex, data = dat, family = "logistic")
margins(fit, ~ length + sex)
margins(fit, ~ length + sex.M, at.mean = TRUE)
margins(fit, ~ length + sex.I, factor.continuous = TRUE)
margins(fit, ~ Vars(model) + Terms())

## create a data table that has two columns
## one of them is an array column
dat1 <- cbind(db.array(dat[,,-c(1,2,10)]), dat[,10])
names(dat1) <- c("x", "y")
delete("abalone_array", conn.id = cid)
dat1 <- as.db.data.frame(dat1, "abalone_array")

fit <- madlib.glm(y < 10 ~ x[-1], data = dat1, family = "logistic")
margins(fit, ~ x[2:5])

db.disconnect(cid, verbose = FALSE)
```

## End(Not run)
merge-method

all.x  logical; if TRUE, then extra rows will be added to the output, one for each row in x that has no matching row in y. These rows will have NAs in those columns that are usually filled with values from y. The default is FALSE, so that only rows with data from both x and y are included in the output.

all.y  logical; analogous to all.x.

key  specifies the primary key of the newly created table.

suffixes  a character vector of length 2 specifying the suffixes to be used for making unique the names of columns in the result which not used for merging (appearing in by etc).

...  arguments to be passed to or from methods.

Details

See merge.data.frame. Note that merge.data.frame supports an incomparables argument, which is not yet supported here.

Value

A db.Rquery object, which expresses the join operation.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

merge.data.frame a merge operation for two data frames.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create sample databases
authors <- data.frame(
  surname = I(c("Tukey", "Venables", "Tierney", "Ripley", "McNeil")),
  nationality = c("US", "Australia", "US", "UK", "Australia"),
  deceased = c("yes", rep("no", 4)))

books <- data.frame(
  name = I(c("Tukey", "Venables", "Tierney",
             "Ripley", "Ripley", "McNeil", "R Core")),
  title = c("Exploratory Data Analysis",
            "Modern Applied Statistics ...",
            "LISP-STAT"),
```
na.action

"Spatial Statistics", "Stochastic Simulation",
"Interactive Data Analysis",
"An Introduction to R"),
other.author = c(NA, "Ripley", NA, NA, NA,
"Venables & Smith")

delete("books", conn.id = cid)
delete("authors", conn.id = cid)
as.db.data.frame(books, 'books', conn.id = cid, verbose = FALSE)
as.db.data.frame(authors, 'authors', conn.id = cid, verbose = FALSE)

## Cast them as db.data.frame objects
a <- db.data.frame('authors', conn.id = cid, verbose = FALSE)
b <- db.data.frame('books', conn.id = cid, verbose = FALSE)

## Merge them together
ml <- merge(a, b, by.x = "surname", by.y = "name", all = TRUE)

db.disconnect(cid, verbose = FALSE)

## End(Not run)

na.action

Functions for filtering NA values in data

Description

'na.omit' returns the object with incomplete cases removed.

Usage

## S4 method for signature 'db.obj'
na.omit(object, vars = NULL, ...)

Arguments

object A db.obj object, which wraps a data table in the connected database (db.data.frame), or some operations on a data table (db.Rquery).

vars An array of strings, default is NULL. The names of the columns that the user wants to filter NA values. If it is NULL, all rows that contains NULL in any column will be filtered out.

... Further arguments, not implemented yet.

Value

A db.Rquery object, which wraps the operation that filters the NA values from the columns vars in object.
names-methods

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

madlib.lm, madlib.glm for linear and logistic regressions.

Examples

## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

delete("abalone", conn.id = cid)
dat <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)

madlib.lm(rings ~ . - sex - id, data = dat, na.action = na.omit)
db.disconnect(cid, verbose = FALSE)
## End(Not run)

name-methods  The Names of an object

Description

This function gives the names of a db.obj, which are the column names of a db.table or db.view. The names are returned as a list.

Usage

## S4 method for signature 'db.obj'
names(x)
## S4 replacement method for signature 'db.obj'
names(x) <- value

Arguments

x  A db.obj. The input data frame for which the column names are required.
value  An array of strings. New names to replace the names of x.
null.data

Value

Returns a string with the list of the column names of data frame. The names are ordered.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also
db.obj, db.data.frame, db.table, db.view, db.Rquery are the class hierarchy structure of this package.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)
## preview of a table
lk(x, nrows = 10) # extract 10 rows of data

## get names of all columns
names(x)

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

null.data A Data Set with lots of NA values

Description

An example data.frame which is used by examples in this user manual

Usage
data(null.data)

Format

This data has 104 columns and 2000 rows.
Details

This data set has lots of NA values in it. By using `as.db.data.frame`, one can put the data set into the connected database. All the NA values will be converted into NULL values.

The MADlib wrapper functions like `madlib.lm` and `link{madlib.glm}` will throw an error if there are NULL values in the data. So one needs to clean up the data before using the regression functions supplied by MADlib.

Note

Lazy data loading is enabled in this package. So the user does not need to explicitly run `data(null.data)` to load the data. It will be loaded whenever it is used.

Examples

```r
## Not run

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data frame "abalone"
delete("null_data", conn.id = cid)
x <- as.db.data.frame(null.data, "null_data", conn.id = cid, verbose = FALSE)

## ERROR, because of NULL values
fit <- madlib.lm(sf.mrtg_pct_assets ~ ris_asset + lnrcd + lncrb +
                 lnconoth + lnconrp + intmsrfv + lnrnr1a + lnrnr2a +
                 lnrnr3a, data = x)

## select columns
y <- x[,c("sf.mrtg_pct_assets","ris_asset", "lnrcd","lncrb",
           "lnconoth","lnconrp","intmsrfv","lnnr1a","lnnr2a",
           "lnnr3a")]

dim(y)

## remove NULL values
for (i in 1:10) y <- y[!is.na(y[i]),]

dim(y)

fit <- madlib.lm(sf.mrtg_pct_assets ~ ., data = y)

fit
db.disconnect(cid, verbose = FALSE)

## End(Not run)
```
**Description**

Perplexity is a measurement of how well a probability distribution or probability model predicts a sample. This function computes the perplexity of the prediction by \texttt{link[predict.madlib.lda]}

**Usage**

```r
## S3 method for class 'lda.madlib'
perplexity(object, predict_output_table, ...)
```

**Arguments**

- `object`: The result of \texttt{madlib.lda}.
- `predict_output_table`: The result of \texttt{predict} on the \texttt{madlib.lda} object.
- `...`: Arguments passed to or from other methods, not implemented yet.

**Value**

A numeric value that indicates the perplexity of the LDA prediction.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. \texttt{<fmcquillan@pivotal.io>}

**See Also**

- \texttt{madlib.lda} builds a topic model using a set of documents.

**Examples**

```r
## Not run:
## Please see the examples in madlib.lda doc.

## End(Not run)
Description

This is a visualization function which plots the result of `madlib.rpart`. This function internally calls R’s `plot.rpart` function.

Usage

```r
## S3 method for class 'dt.madlib'
plot(x, uniform = FALSE, branch = 1, compress = FALSE,
     nspace, margin = 0, minbranch = 0.3, ...)
```

Arguments

- `x` The fitted tree from the result of `madlib.rpart`
- `uniform` A boolean, if TRUE, uses uniform vertical spacing of the nodes.
- `branch` A double value, between 0 and 1, to control the shape of the branches from parent to child.
- `compress` A boolean, if FALSE, the leaf nodes will be at the horizontal plot coordinate of 1:nleaves. Use TRUE for a more compact arrangement.
- `nspace` A double value, indicating the amount of extra space between a node with children and a leaf. default is `branch`
- `margin` A double value, indicating the amount of extra space to leave around the borders of the tree.
- `minbranch` A double value, specifying the minimum length for a branch.
- `...` Arguments to be passed to or from other methods.

Value

The coordinates of the nodes are returned as a list, with components x and y.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. 
	<fmcquillan@pivotal.io>

References

See Also

`madlib.rpart` is the wrapper for MADlib’s `tree_train` function for decision trees. `text.dt.madlib`, `print.dt.madlib` are other visualization functions.

`madlib.lm`, `madlib glm`, `madlib.rpart`, `madlib.summary`, `madlib arima`, `madlib elnet` are all MADlib wrapper functions.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

## decision tree using abalone data, using default values of minsplit, 
## maxdepth etc.
key(x)<-"id"
fit <- madlib.rpart(rings < 10 ~ length + diameter + height + whole + shell, 
  data=x, parms = list(split='gini'), control = list(cp=0.005))
fit

plot(fit, uniform =TRUE)
text(fit)
db.disconnect(cid)

## End(Not run)
```

---

### predict

*Generate the `db.Rquery` object that can calculate the predictions*

#### Description

Generate the `db.Rquery` object that can calculate the predictions for linear/logistic regressions. The actual result can be viewed using `lk`.

#### Usage

```r
## S3 method for class 'lm.madlib'
predict(object, newdata, ...)

## S3 method for class 'lm.madlib.grps'
predict(object, newdata, ...)
```
# S3 method for class 'logregr.madlib'
predict(object, newdata, type = c("response", "prob"), ...)

# S3 method for class 'logregr.madlib.grps'
predict(object, newdata, type = c("response", "prob"), ...)

# S3 method for class 'glm.madlib'
predict(object, newdata, type = c("response", "prob"), ...)

# S3 method for class 'glm.madlib.grps'
predict(object, newdata, type = c("response", "prob"), ...)

## Arguments

- **object**: The result of `madlib.lm` and `madlib.glm`.
- **newdata**: A `db.obj` object, which contains the information about the real data in the database.
- **type**: A string, default is "response". It produces the predicted results for the `newdata`. The alternative value is "prob", which is only used for binomial{logit} to compute the probabilities. A string, default is "response", which produces the TRUE or FALSE prediction. If it is "prob", this function computes the probabilities for TRUE cases.
- **...**: Extra parameters. Not implemented yet.

## Value

A `db.Rquery` object, which contains the SQL query to compute the predictions.

## Author(s)

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

## See Also

- `madlib.lm` linear regression
- `madlib.glm` logistic regression
- `lk` view the actual result
- `groups.lm.madlib`, `groups.lm.madlib.grps`, `groups.logregr.madlib`, `groups.logregr.madlib.grps` extract grouping column information from the fitted model(s).
Examples

## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create db.table object pointing to a data table
delete("abalone", conn.id = cid)
x <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)

## Example 1 --------
fit <- madlib.lm(rings ~ . - sex - id, data = x)

fit

pred <- predict(fit, x) # prediction

content(pred)

ans <- x$rings # the actual value

lk((ans - pred)^2, 10) # squared error

lk(mean((ans - pred)^2)) # mean squared error

## Example 2 --------

y <- x
y$sex <- as.factor(y$sex)
fit <- madlib.lm(rings ~ . - id, data = y)

lk(mean((y$rings - predict(fit, y))^2))

## Example 3 --------

fit <- madlib.lm(rings ~ . - id | sex, data = x)

fit

pred <- predict(fit, x)

content(pred)

ans <- x$rings

lk(mean((ans - pred)^2))
## predict.arima

Forecast from MADlib’s ARIMA fits

### Description

Forecast from models fit by `madlib.arima`

### Usage

```r
## S3 method for class 'arima.css.madlib'
predict(object, n.ahead = 1, ...)
```

### Arguments

- **object**
  - The result of `madlib.arima`.

- **n.ahead**
  - The number of steps ahead for which prediction is required.

- **...**
  - Arguments passed to or from other methods, not implemented yet.
**predict.bagging.model**

**Value**

A `db.table` object, which points to a table that contains the forecasted values. The table has two columns: steps_ahead and forecast_value. One can use the function `lk` to look at the values.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. `<mcquillan@pivotal.io>`

**See Also**

`madlib.arima` fits ARIMA model to a time series.

**Examples**

```r
## Not run:
## Please see the examples in madlib.arima doc.
## End(Not run)
```

---

**predict.bagging.model**  *Make predictions using the result of generic.bagging*

**Description**

Make predictions using bootstrap aggregating models

**Usage**

```r
## S3 method for class 'bagging.model'
predict(object, newdata, combine = "mean", ...)
```

**Arguments**

- `object`  
  A `bagging.model`, which is the result of `generic.bagging`.
- `newdata`  
  A `db.obj` object, which wraps the data in the database.
- `combine`  
  A string, default is "mean". The other choice is "vote". How to summarize the predictions of the multiple models in the fitting result of `generic.bagging`. "mean" will produce the average of the predictions, while "vote" will select the prediction with the most votes.
- `...`  
  Extra parameters. Not implemented yet.

**Value**

A `db.Query` object, which contains the SQL query to compute the prediction. One can use the function `lk` to look at the values.
Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

generic.bagging generates the models of bootstrap aggregating.
predict.lm.madlib and predict.logregr.madlib produce predictions for linear and logistic models.

Examples

## Not run:

```r
# set up the database connection
# Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
y <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

fit <- generic.bagging(function(data) {
  madlib.lm(rings ~ . - id - sex, data = data)
}, data = y, nbags = 25, fraction = 0.7)

pred <- predict(fit, newdata = y) # make prediction

lookat(mean((y$rings - pred)^2)) # mean squared error

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

predict.dt.madlib  

**Compute the predictions of the model produced by madlib.rpart**

Description

This is actually a wrapper for MADlib’s predict function of decision tree. It accepts the result of madlib.rpart, which is a representation of decision tree, and compute the predictions for new data sets.

Usage

```r
## S3 method for class 'dt.madlib'
predict(object, newdata, type = c("response", "prob"),
  ...)  ```
**predict.dt.madlib**

**Arguments**

- **object**
  A `dt.madlib` object, which is the result of `madlib.rpart`.

- **newdata**
  A `db.obj` object, which contains the data used for prediction. If it is not given, then the data set used to train the model will be used.

- **type**
  A string, default is "response". For regressions, this will generate the fitting values. For classification, this will generate the predicted class values. There is an extra option "prob" for classification tree, which computes the probabilities of each class.

  ... Other arguments. Not implemented yet.

**Value**

A `db.obj` object, which wraps a table that contains the predicted values and also a valid ID column. For type='response', the predicted column has the fitted value (regression tree) or the predicted classes (classification tree). For type='prob', there are one column for each class, which contains the probabilities for that class.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

**References**


**See Also**

`madlib.lm, madlib.glm, madlib.rpart, madlib.summary, madlib.arima, madlib.elnet` are all MADlib wrapper functions.

`predict.lm.madlib, predict.logregr.madlib, predict.elnet.madlib, predict.arima.css.madlib` are all predict functions related to MADlib wrapper functions.

**Examples**

```
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

key(x) <- "id"
fit <- madlib.rpart(rings < 10 ~ length + diameter + height + whole + shell,
   data=x, parms = list(split='gini'), control = list(cp=0.005))

predict(fit, x, 'r')
```
predict.elnet.madlib

Description

Prediction from models fit by `madlib.elnet`

Usage

```r
## S3 method for class 'elnet.madlib'
predict(object, newdata, type = c("response", "prob"), ...)
```

Arguments

- `object`: The result of `madlib.elnet`
- `newdata`: A `db.obj` object, which wraps the data in the database.
- `type`: A string, default is "response". The other option is "prob". The prediction for "gaussian"/"linear" family of `madlib.elnet` result is always for the dependent variable. The prediction for "binomial"/"logistic" family is TRUE/FALSE values for "response", and probabilities of TRUE for "prob".
- `...`: Extra parameters. Not implemented yet.

Value

A `db.Query` object, which contains the SQL query to compute the prediction. One can use the function `lk` to look at the values.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

- `madlib.elnet` Wrapper for MADlib elastic net regularization.
- `predict.lm.madlib` and `predict.logregr.madlib` produce predictions for linear and logistic models.

Examples

```r
## see the examples in madlib.elnet
```
**predict.lda**

*Prediction function for MADlib's LDA models*

**Description**

Labelling test documents using a learned LDA model built by `linl{}madlib.lda`

**Usage**

```r
## S3 method for class 'lda.madlib'
predict(object, data, docid, words, ...)
```

**Arguments**

- `object` The result of `madlib.lda`.
- `data` An object of `db.obj` class. This is the database table containing the documents on which the algorithm will predict. The text of each document should be tokenized into 'words'.
- `docid` Text name of the column containing the id of the documents.
- `words` Column name of the input data table containing the vector of words/tokens in the documents.
- `...` Arguments passed to or from other methods, not implemented yet.

**Value**

A `db.table` object, which points to a table that contains the predicted values. The table has the following columns: `docid` `wordcount` `words` `counts` `topic_count` `topic_assignment`

One can use the function `lk` to look at the values.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

**See Also**

`madlib.lda` builds a topic model using a set of documents.

**Examples**

```r
## Not run:
## Please see the examples in madlib.lda doc.

## End(Not run)```
predict.rf.madlib  

**Description**

This is actually a wrapper for MADlib’s predict function of random forests. It accepts the result of `madlib.randomForest`, which is a representation of a random forest model, and computes the predictions for new data sets.

**Usage**

```r
## S3 method for class 'rf.madlib'
predict(object, newdata, type = c("response", "prob"),
    ...)
```

**Arguments**

- `object`  
  A `rf.madlib` object, which is the result of `madlib.randomForest`.
- `newdata`  
  A `db.obj` object, which contains the data used for prediction. If it is not given, then the data set used to train the model will be used.
- `type`  
  A string, default is "response". For regressions, this will generate the fitting values. For classification, this will generate the predicted class values. There is an extra option "prob" for classification models, which computes the probabilities of each class.
- `...`  
  Other arguments. Not implemented yet.

**Value**

A `db.obj` object, which wraps a table that contains the predicted values and also a valid ID column. For `type='response'`, the predicted column has the fitted value (regression model) or the predicted classes (classification model). For `type='prob'`, there are one column for each class, which contains the probabilities for that class.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

**References**

See Also

madlib.lm, madlib.glm, madlib.randomForest, madlib.rpart, madlib.summary, madlib.arima, madlib.elnet are all MADlib wrapper functions.

predict.lm, predict.logregr, predict.elnet, madlib.predict.arima, css, madlib.predict.dt, madlib.predict.rf are all predict functions related to MADlib wrapper functions.

Examples

## Not run:

```r
## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.data.frame(abalone, conn.id = cid, verbose = FALSE)

dt(key(x) <- "id"
fit <- madlib.randomForest(rings < 10 ~ length + diameter + height + whole + shell,
data=x)

predict(fit, x, 'r')

db.disconnect(cid)

## End(Not run)
```

---

**Description**

These functions read the actual data from a database table or operation, returning a `data.frame` or other object as appropriate. `lookat` and `lk` are actually the same.

**Usage**

```r
lookat(x, nrows = 100, array = TRUE, conn.id = 1, drop = TRUE)

lk(x, nrows = 100, array = TRUE, conn.id = 1, drop = TRUE)
```

```r
## S3 method for class 'db.table'
as.data.frame(x, row.names = NULL, optional = FALSE,
             nrows = NULL, stringsAsFactors = default.stringsAsFactors(), array = TRUE, ...)

## S3 method for class 'db.view'
```
as.data.frame(x, row.names=NULL, optional=FALSE, nrows
    = NULL, stringsAsFactors = default.stringsAsFactors(), array = TRUE,
    ...)  
## S3 method for class 'db.Rquery'
as.data.frame(x, row.names = NULL, optional =
    FALSE, nrows = NULL, stringsAsFactors = default.stringsAsFactors(),
    array = TRUE, ...)  

Arguments

  x         A db.data.frame (includes db.table and db.view) object, which points to a
             table or view in the database; or a db.Rquery object, which represents some
             operations on a db.data.frame object. If x is a string, which means a table
             name, this function directly reads data from the table without having to wrap it
             with a db.data.frame class object.
  conn.id   An integer, the ID of the connection where the table resides.
  nrows     An integer, how many rows of data to retrieve. If it is NULL or “all” or a non-
             positive value, then all data in the table will be send into R. Be careful, you do
             not want to do this if the data table is very large.
  array     Logical, default is TRUE. This decides how to parse columns that have array as
             their elements. When TRUE, each element in the array is extracted and put into a
             new column. Otherwise, the array is read in as a string.
  stringsAsFactors Logical, whether character variables should be converted to factors.
  drop      Whether to coerce single-column tables to a vector of the appropriate class.
  row.names,optional,...  
             For compatibility with the as.data.frame generic; not used.

Details

  When x is a db.data.frame object, this function reads the data in a table or view in the connected
  database.
  When x is a db.Rquery object, this function reads the result of some operations on a db.data.frame
  object.
  When x is a db.Rcrossprod object, this function output a matrix to R. If the matrix is symmetric,
  it is returned as dspMatrix. Otherwise, it is returned as dgeMatrix. If there are multiple matrices
  in x, a list is returned and each element of the list is a matrix.
  The as.data.frame method calls lookat with nrows = NULL to perform the conversion to a data
  frame. In this case drop is set to FALSE, ie the result will always be a data frame.

Value

  For db.data.frame and db.Rquery objects, a data frame. Each column in the table becomes a
  column of the returned data.frame. A column of arrays is converted into a column of strings,
  see arraydb.to.arrayr for more details. Single-column tables created with lookat or 1k will be
  coerced to a vector if drop == TRUE.
For `db.Rcrossprod` objects, a matrix or list of matrix objects as appropriate (see above).

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Hong Ooi, Pivotal Inc. `<hooi@pivotal.io>` wrote the `as.data.frame` methods.
Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

**See Also**

`arraydb.to.arrayr` convert strings extracted from database into arrays.

**Examples**

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

## preview of a table
lk(x, nrows = 10) # extract 10 rows of data

## do some operations and preview the result
y <- (x[, 1:2] + 1.2) * 2
lk(y, 20)

## table abalone has a column named "id"
lk(sort(x, INDICES = x$id), 20) # the preview is ordered by "id" value

## use as.data.frame
as.data.frame(x, 10)

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

**Description**

This function displays the results of logistic regression in a pretty format.
Usage

```r
## S3 method for class 'logregr.madlib'
print(x, digits = max(3L, getOption("digits") - 3L), ...)

## S3 method for class 'logregr.madlib.grp5s'
print(x, digits = max(3L, getOption("digits") - 3L), ...)

## S3 method for class 'logregr.madlib'
show(object)

## S3 method for class 'logregr.madlib.grp5s'
show(object)

## S3 method for class 'glm.madlib'
print(x, digits = max(3L, getOption("digits") - 3L), ...)

## S3 method for class 'glm.madlib.grp5s'
print(x, digits = max(3L, getOption("digits") - 3L), ...)

## S3 method for class 'glm.madlib'
show(object)

## S3 method for class 'glm.madlib.grp5s'
show(object)
```

Arguments

- `x`, object
  - The logistic regression result object to be printed.
- `digits`
  - A non-null value for ‘digits’ specifies the minimum number of significant digits to be printed in values. The default, ‘NULL’, uses ‘getOption("digits")’. (For the interpretation for complex numbers see `signif`.) Non-integer values will be rounded down, and only values greater than or equal to 1 and no greater than 22 are accepted.
- `...`
  - Further arguments passed to or from other methods. This is currently not implemented.

Value

No value is returned

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

madlib.glm Wrapper for MADlib linear and logistic regression

Examples

```r
## Not run:
## see the examples in madlib.glm
## End(Not run)
```

print-methods

Display the connection information associated with a db object

Description

This function displays the SQL table, database, host, and connection information associated with a db.table or db.view object.

Usage

```r
## S4 method for signature 'db.data.frame'
print(x)

## S4 method for signature 'db.Rquery'
print(x)

## S4 method for signature 'db.data.frame'
show(object)

## S4 method for signature 'db.Rquery'
show(object)
```

Arguments

- `x` The signature of the method. A db.data.frame (includes db.table and db.view) object, which points to a table or view in the database; or a db.Rquery object, which represents some operations on a db.data.frame object.
- `object` The signature of the method. A db.data.frame (includes db.table and db.view) object, which points to a table or view in the database; or a db.Rquery object, which represents some operations on a db.data.frame object.
**Details**

When the signature `x` is either a `db.data.frame` object or a `db.Rquery` object, this function displays the name of connected SQL database, the SQL database host, and the connection ID.

When the signature `x` is a `db.data.frame` object, the function also displays the associated table. When the signature `x` is a `db.Rquery` object, this function displays the temporary status of the input, and the table that it is derived from.

**Value**

This function returns nothing.

**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**See Also**

`lk` or `lookat` Displays the contents of an associated table.

**Examples**

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

## create a table from the example data.frame "abalone"
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

## printing db.data.frame object
x # Display the associated table, and database information for x
db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

**Description**

This function displays the results of `madlib.arima` in a pretty format.
Usage

```r
## S3 method for class 'arima.css.madlib'
print(x, digits = max(3L,getOption("digits")
- 3L), ...)
```

```r
## S3 method for class 'arima.css.madlib'
show(object)
```

Arguments

- **x, object**  
  The ARIMA fitting result object of `madlib.arima`

- **digits**  
  A non-null value for 'digits' specifies the minimum number of significant digits to be printed in values. The default, 'NULL', uses 'getOption("digits")'. (For the interpretation for complex numbers see `signif`). Non-integer values will be rounded down, and only values greater than or equal to 1 and no greater than 22 are accepted.

- ...  
  Further arguments passed to or from other methods. This is currently not implemented.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

See Also

- `madlib.arima` Wrapper for MADlib ARIMA model fitting

Examples

```r
## Not run:
## Please see the examples in madlib.arima doc

## End(Not run)
```

Description

This function prints the result of `madlib.rpart` to the screen. It internally calls R’s `print.rpart` function.
Usage

```r
## S3 method for class 'dt.madlib'
print(x, digits = max(3L, getOption("digits") - 3L),
      ...
)
```

Arguments

- `x` The fitted tree from the result of `madlib.rpart`
- `digits` The number of digits to print for numerical values.
- `...` Arguments to be passed to or from other methods.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References


See Also

- `madlib.rpart` is the wrapper for MADlib’s `tree_train` function for decision trees. `plot.dt.madlib`, `text.dt.madlib` are other visualization functions.
- `madlib.lm, madlib.glm, madlib.rpart, madlib.summary, madlib.arima, madlib.elnet` are all MADlib wrapper functions.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
1k(x, 10)

## decision tree using abalone data, using default values of msplit, 
## maxdepth etc.
key(x)<-"id"
fit <- madlib.rpart(rings < 10 ~ length + diameter + height + whole + shell, 
                    data=x, parms = list(split='gini'), control = list(cp=0.005))

print(fit)
```
print.elnet.madlib

Display the results from madlib.elnet function in a pretty format

Description

This function prints the results from `madlib.elnet` in a human-readable format.

Usage

## S3 method for class 'elnet.madlib'
print(x, digits = max(3L,getOption("digits") - 3L), ...)

## S3 method for class 'elnet.madlib'
show(object)

Arguments

- `x`, `object`  The `elnet.madlib` object to be printed.
- `digits`  A non-null value for 'digits' specifies the minimum number of significant digits to be printed in values. The default, 'NULL', uses 'getOption("digits")'. (For the interpretation for complex numbers see `signif`.) Non-integer values will be rounded down, and only values greater than or equal to 1 and no greater than 22 are accepted.
- `...`  Further arguments passed to or from other methods. This is currently not implemented.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

- `madlib.elnet` Wrapper for MADlib elastic net regularization.

Examples

## see the examples in madlib.elnet
print.lm.madlib  
Display results of linear regression

Description
This function displays the results of linear regression in a pretty format.

Usage

```r
# S3 method for class 'lm.madlib'
print(x, digits = max(3L,getOption("digits") - 3L),
     ...)  

# S3 method for class 'lm.madlib.grps'
print(x, digits = max(3L,getOption("digits") - 3L),  
     ...)  

# S3 method for class 'lm.madlib'
show(object)  

# S3 method for class 'lm.madlib.grps'
show(object)
```

Arguments

- `x`, `object` The linear regression result object to be printed.
- `digits` A non-null value for ‘digits’ specifies the minimum number of significant digits to be printed in values. The default, ‘NULL’, uses ‘getOption("digits")’. (For the interpretation for complex numbers see `signif`.) Non-integer values will be rounded down, and only values greater than or equal to 1 and no greater than 22 are accepted.
- `...` Further arguments passed to or from other methods. This is currently not implemented.

Value
No value is returned

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

madlib.lm Wrapper for MADlib linear regression
print.none.obj  Function used in GUI to print absolutely nothing

Description

This function prints nothing and is used only in GUI.

Usage

```r
## S3 method for class 'none.obj'
print(x, ...)
```

Arguments

- `x`: A `none.obj` object. The content of this object does not matter. It is used to return a value which makes the GUI print nothing on the screen.
- `...`: Not used.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>

See Also

PivotR launches the GUI for PivotalR.

print.rf.madlib  Print the result of madlib.randomForest

Description

This function prints the result of `madlib.randomForest` to the screen. It internally calls R’s print function for random forests.

Usage

```r
## S3 method for class 'rf.madlib'
print(x, digits = max(3L,getOption("digits") - 3L),
      ...)
```
Arguments

- **x**: The fitted forest from the result of `madlib.randomForest`
- **digits**: The number of digits to print for numerical values.
- **...**: Arguments to be passed to or from other methods.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References


See Also

- `madlib.randomForest` is the wrapper for MADlib’s forest_train function for random forests.
- `madlib.lm`, `madlib.glm`, `madlib.rpart`, `madlib.summary`, `madlib.arima`, `madlib.elnet`, `madlib.rpart` are all MADlib wrapper functions.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

## random forest using abalone data, using default values of minsplit, maxdepth etc.
key(x)<-"id"
fit <- madlib.randomForest(rings < 10 ~ length + diameter + height + whole + shell,
data=x)

print(fit)
db.disconnect(cid)

## End(Not run)
```
print.summary.madlib  Display the results from summary function in a pretty format

Description

This function prints the results from madlib.summary in a human-readable format.

Usage

```r
## S3 method for class 'summary.madlib'
print(x, digits = max(3L,getOption("digits") - 3L),...)
```

```r
## S3 method for class 'summary.madlib'
show(object)
```

Arguments

- `x`, `object`  The summary result object to be printed.
- `digits`  A non-null value for ‘digits’ specifies the minimum number of significant digits to be printed in values. The default, ‘NULL’, uses ‘getOption("digits")’. (For the interpretation for complex numbers see `signif`.) Non-integer values will be rounded down, and only values greater than or equal to 1 and no greater than 22 are accepted.
- `...`  Further arguments passed to or from other methods. This is currently not implemented.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.

Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

madlib.summary Wrapper for MADlib linear and logistic regression

Examples

```r
## see the examples in madlib.summary
```
Description

Functions to extract the residuals for regression models fit in Madlib.

Usage

```r
## S3 method for class 'lm.madlib'
residuals(object, ...)

## S3 method for class 'lm.madlib.grps'
residuals(object, ...)

## S3 method for class 'logregr.madlib'
residuals(object, ...)

## S3 method for class 'logregr.madlib.grps'
residuals(object, ...)

## S3 method for class 'glm.madlib'
residuals(object, ...)

## S3 method for class 'glm.madlib.grps'
residuals(object, ...)
```

Arguments

- `object` The regression model object, of class `lm.madlib`, `lm.madlib.grps` or `logregr.madlib`, `logregr.madlib.grps` obtained using `madlib.lm` or `madlib(glm)` respectively.
- `...` Other arguments, not used.

Details

See the documentation for `residuals`

Value

For ungrouped regressions, `residuals` returns an object of class `db.Rquery`

For grouped regressions, `residuals` returns a list of `db.Rquery` objects giving the output of these methods for each of the component models. Similarly, AIC for a grouped regression returns a vector of the AICs for each of the component models.
Author(s)

Author: Predictive Analytics Team, Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

See Also

AIC, extractAIC, logLik.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

fit <- madlib.glm(rings < 10 ~ . - id | sex, data = x, family = "binomial")
residuals(fit)
db.disconnect(cid)

## End(Not run)
```

---

**Row_actions**

*Compute the sum or mean of all columns in one row of a table*

Description

This function returns a db.Rquery object, which produces the sum or mean value of all columns of one row when executed in database.

Usage

```r
## S4 method for signature 'db.obj'
rowSums(x, na.rm = FALSE, dims = 1, ...)
## S4 method for signature 'db.obj'
rowMeans(x, na.rm = FALSE, dims = 1, ...)
```
Arguments

x A db.obj object, which has only one column. The column can be casted into boolean values.

na.rm logical. Should missing values (including 'NaN') be omitted from the calculations? Not implemented yet.

dims integer: Which dimensions are regarded as 'rows' or 'columns' to sum over. For 'row*', the sum or mean is over dimensions 'dims+1, ...'; for 'col*' it is over dimensions '1:dims'. Not implemented yet.

... Other arguments. Not implemented yet.

Value

A db.Rquery object which, when executed, computes the mean or sum of all columns on every row of a table.

Author(s)

Author: Hong Ooi, Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

sum, db.obj-method, colSums, db.obj-method compute the sum of each column.

mean, db.obj-method, colMeans, db.obj-method compute the mean values column-wise.

Examples

## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

lk(x, 10)

row.sum <- rowSums(x[,2]) # the second column is text
row.avg <- rowMeans(x[,2])

## lokk at 10 results
lk(row.sum, 10)
lk(row.avg, 10)

db.disconnect(cid, verbose = FALSE)

## End(Not run)
Methods for sampling rows of data from a table/view randomly

Description

This method samples rows of data from a table/view randomly. The sampled result is stored in a temporary table.

Usage

```r
## S4 method for signature 'db.obj'
sample(x, size, replace = FALSE, prob = NULL, ...)
```

Arguments

- `x`: A `db.obj` object, which is the wrapper to the data table.
- `size`: An integer. The size of the random sample. When `replace` is `FALSE`, `size` must be smaller than the data table/view's total row number.
- `replace`: A logical value, default is `FALSE`. When it is `TRUE`, the data is sampled with replacement, which means a row might be sampled for multiple times. When it is `FALSE`, each row can only be sampled at most once.
- `prob`: A vector of double values, default is `NULL`. The probabilities of each row to sample. Not implemented yet.
- `...`: Extra parameters. Not implemented.

Details

When `replace` is `FALSE`, the data is just sorted randomly (see `sort.db.obj-method`) and selected, which is similar to `sort(x, 'random')`. When `replace` is `TRUE`, we have to scan the table multiple times to select repeated items.

Value

A `db.data.frame` object, which is a wrapper to a temporary table. The table contains the sampled data.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmquillan@pivotal.io>

See Also

- `generic.bagging` uses `sample`
Examples

## Not run:

```r
# set up the database connection
# Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

y <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(y, 10)

dim(y)

a <- sample(y, 20)

dim(a)

lookat(a)

b <- sample(y, 40, replace = TRUE)

dim(b)

lookat(b)

delete(b)

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

**scale**

**Scaling and centering of tables**

**Description**

scale centers and/or scales the columns of a numeric table.

**Usage**

```r
# S4 method for signature 'db.obj'
scale(x, center = TRUE, scale = TRUE)
```

**Arguments**

- `x` A `db.obj` object. It represents a table/view in the database if it is an `db.data.frame` object, or a series of operations applied on an existing `db.data.frame` object if it is a `db.Rquery` object.
center either a logical value or a numeric vector of length equal to the number of columns of 'x'.

scale either a logical value or a numeric vector of length equal to the number of columns of 'x'.

Details

The value of 'center' determines how column centering is performed. If 'center' is a numeric vector with length equal to the number of columns of 'x', then each column of 'x' has the corresponding value from 'center' subtracted from it. If 'center' is 'TRUE' then centering is done by subtracting the column means (omitting 'NA's) of 'x' from their corresponding columns, and if 'center' is 'FALSE', no centering is done.

The value of 'scale' determines how column scaling is performed (after centering). If 'scale' is a numeric vector with length equal to the number of columns of 'x', then each column of 'x' is divided by the corresponding value from 'scale'. If 'scale' is 'TRUE' then scaling is done by dividing the (centered) columns of 'x' by their standard deviations if 'center' is 'TRUE', and the root mean square otherwise. If 'scale' is 'FALSE', no scaling is done.

The root-mean-square for a (possibly centered) column is defined as sqrt(sum(x^2)/(n-1)), where x is a vector of the non-missing values and n is the number of non-missing values. In the case 'center = TRUE', this is the same as the standard deviation, but in general it is not. (To scale by the standard deviations without centering, use 'scale(x, center = FALSE, scale = lookat(sd(x)))'.)

Value

A db.Rquery object. It computes the centering and/or scaling of codex for each column including array elements. The result can be viewed using 1k or lookat.

The numeric centering and scalings used (if any) are returned as attributes "scaled:center" and "scaled:scale". The number of rows in the table is also returned as the attribute "row.number".

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

db.array creates an array column for a db.Rquery object.

Examples

## Not run:
## help("scale,db.obj-method") # display this doc

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname)
x <- as.db.data.frame(abalone, conn.id = cid)
lk(x, 10)

s <- scale(x[-c(1,2)]) # scale all numeric columns
centers <- attr(s, "scaled:center")
scales <- attr(s, "scaled:scale")

## create the scaled table
delete("scaled_abalone")
y <- as.db.data.frame(s, "scaled_abalone")

lk(y, 10)
db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

### sort

**Sort a table or view by a set of columns**

#### Description

This function is used to sort a table of view in the database.

#### Usage

```r
## S4 method for signature 'db.obj'
sort(x, decreasing = FALSE, INDICES, ...)
```

#### Arguments

- **x**: The signature of the method. A db.obj (includes db.table and db.view) object, which points to a table or view in the database.
- **decreasing**: A logical, with default value as FALSE. Should the sort be increasing or decreasing?
- **INDICES**: A list of db.Rquery objects. Each of the list element selects one or multiple columns of x. NULL to order by random().
- **...**: Further arguments passed to or from other methods. This is currently not implemented.

#### Value

A db.Rquery object. It is the query object used to sort the db.obj in the database.

#### Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>
See Also

`by` has similar syntax to this function.

`lk` or `lookat` to view portion of the data table

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

y <- sort(x, decreasing = FALSE, list(x$id, x$sex))
# get the SQL query to be run
content(y)
# get the sorted output
lk(y, 20)

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

subset-methods

Extract a subset of a table or view

Description

This function extracts a subset of a `db.obj` which could either be a `db.table` or `db.view` object.

Usage

```r
## S4 method for signature 'db.obj'
subset(x, subset, select)
```

Arguments

- `x` A `db.obj` (either `db.table` or `db.view`) object from which to extract element(s).
- `subset, select` Indices specifying elements to extract or replace. Indices are ‘numeric’ or ‘character’ vectors or empty (missing) or ‘NULL’. Numeric values are coerced to integer as by ‘as.integer’ (and hence truncated towards zero).

Value

A `db.Rquery` object is returned which is a SQL query to extract the requested subset.
**Author(s)**

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

**See Also**

[-methods Operator to extract elements](#)

**Examples**

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)
lk(x[1:3])
lk(subset(x, 1:3))

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

**Description**

The function prints the value of each element in the Logistic Regression output object.

**Usage**

```r
## S3 method for class 'logregr.madlib'
summary(object, ...)

## S3 method for class 'logregr.madlib.grps'
summary(object, ...)

## S3 method for class 'glm.madlib'
summary(object, ...)

## S3 method for class 'glm.madlib.grps'
summary(object, ...)
```
Arguments

object          Logistic regression object
...            Further arguments passed to or from other methods. This is currently not implemented.

Value

The function returns the logregr.madlib or logregr.madlib.grps object passed to the function.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

madlib.glm wrapper for MADlib linear and logistic regressions.
madlib.lm wrapper for MADlib linear regression

Examples

## see the examples in madlib.glm

summary.arima.madlib  Summary information for MADlib’s ARIMA model

Description

The function prints the result of madlib.arima in a pretty format

Usage

## S3 method for class ‘arima.css.madlib’
summary(object, ...)

Arguments

object          The ARIMA fitting result object of madlib.arima
...            Further arguments passed to or from other methods. This is currently not implemented.

Value

The function returns the object passed to the function
Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also
madlib.arima Wrapper for MADlib ARIMA model fitting
print.arima.css.madlib print the ARIMA result

Examples
## Not run:
## Please see the examples in madlib.arima doc
## End(Not run)

summary.elnet.madlib  Summary information for Elastic net regularization output

Description
The function prints the value of each element in the output object of madlib.elnet.

Usage
## S3 method for class 'elnet.madlib'
summary(object, ...)

Arguments
object A elnet.madlib object produced by madlib.elnet.
...
Further arguments passed to or from other methods. This is currently not implemented.

Value
The function returns the elnet.madlib object in the argument.

Author(s)
Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also
madlib.elnet Wrapper for MADlib elastic net regularization.
Summary information for Linear Regression output

Description

The function prints the value of each element in the Linear Regression output object.

Usage

```r
## S3 method for class 'lm.madlib'
summary(object, ...)
```

```r
## S3 method for class 'lm.madlib.grps'
summary(object, ...)
```

Arguments

- `object` Linear regression object
- `...` Further arguments passed to or from other methods. This is currently not implemented.

Value

The function returns the `lm.madlib` or `lm.madlib.grps` object passed to the function.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. `<fmquillan@pivotal.io>`

See Also

- `madlib.lm` Wrapper for MADlib linear regression

Examples

```r
## see the examples in madlib.lm
```
text.dt.madlib  Add labels onto the figure generated by plot.dt.madlib

Description

This is a function which adds labels to the plot generated by plot.dt.madlib.

Usage

```r
## S3 method for class 'dt.madlib'
text(x, splits = TRUE, label, FUN = text, all = FALSE,
     pretty = NULL, digits =getOption("digits") - 3L, use.n = FALSE, fancy
     = FALSE, fwidth = 0.8, fheight = 0.8, bg = par("bg"), minlength = 1L, ...)
```

Arguments

- `x`: The fitted tree from the result of `madlib.rpart`
- `splits`: A boolean, if TRUE, labels the splits with the criterion for the split.
- `label`: This is currently ignored.
- `FUN`: The name of a labeling function, e.g. `text`
- `all`: A boolean, if TRUE, labels all the nodes, otherwise just the terminal nodes.
- `pretty`: An alternative to the `minlength` argument.
- `digits`: Number of significant digits to include in numeric labels.
- `use.n`: A boolean, if TRUE, adds to label (\#events level1/\#events level2/etc. for classification and n for regression)
- `fancy`: A boolean, if TRUE, represents internal nodes by ellipses and leaves by rectangles.
- `fwidth`: Controls the width of the ellipses and rectangles if fancy=TRUE.
- `fheight`: Controls the height of the ellipses and rectangles if fancy=TRUE.
- `bg`: The color used to paint the background if fancy=TRUE.
- `minlength`: The length to use for factor labels.
- `...`: Other graphical parameters to be supplied as input to this function (see `par`).

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

References

See Also

madlib.rpart is the wrapper for MADlib’s tree_train function for decision trees. plot.dt.madlib, print.dt.madlib are visualization functions for a model fitted through madlib.rpart

predict.dt.madlib is a wrapper for MADlib’s predict function for decision trees.

madlib.lm, madlib.glm, madlib.summary, madlib.arima, madlib.elnet are all MADlib wrapper functions.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)

key(x) <- "id"

fit <- madlib.rpart(rings < 10 ~ length + diameter + height + whole + shell,
               data=x, parms = list(split='gini'), control = list(cp=0.005))

plot(fit, uniform=TRUE)
text(fit, use.n=TRUE, all=TRUE)

db.disconnect(cid)

## End(Not run)
```

Type Cast functions

**Cast columns of db.obj objects to other types**

**Description**

Coerce db.obj object columns into other types. col.types displays the types of each column. as.Date converts to date (no time of day); as.time converts to time of day (no date); as.timestamp converts to both date and time; as.interval converts to time interval. db.date.style can display or set the date style for a particular connection.

Usage

```r
## S4 method for signature 'db.obj'
as.integer(x, ...)

## S4 method for signature 'db.obj'
as.character(x, array = TRUE, ...)
```
## Type Cast functions

```r
## S4 method for signature 'db.obj'
as.double(x, ...)

## S4 method for signature 'db.obj'
as.logical(x, ...)

## S4 method for signature 'db.obj'
as.numeric(x, ...)

## S4 method for signature 'db.obj'
as.Date(x, ...)

db.date.style(conn.id = 1, set = NULL)

as.time(x, ...)

as.timestamp(x, ...)

as.interval(x, ...)

col.types(x)
```

**Arguments**

- `x`: A `db.obj` object. All columns of the object will be converted into the target type. If the column contains arrays, the array will be converted into an array of the target type.
- `array`: A logical, default is `TRUE`. If `array` is `TRUE`, then an array column is converted into an array of strings. If `array` is `FALSE`, then an array column is converted into a string column, i.e. `array[1,2,3]` is casted into `'\{1,2,3\}'`.
- `...`: further arguments passed to or from other methods. This is currently not implemented.
- `conn.id`: An integer, default is 1. The connection ID for the database connection.
- `set`: A string, default is `NULL`. It can be "us" or "european". If it is `NULL`, `db.date.style` displays the current date style in the connected database. Otherwise, it sets the date style for the connected database.

**Value**

- A `db.Query` object, which is a SQL query which combine all columns into an array.
- `col.types` returns a vector of characters, which are the column types of `x`.

**Author(s)**

- Author: Predictive Analytics Team at Pivotal Inc.
- Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>
unique-methods

See Also

by, db.obj-method is usually used together with aggregate functions.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

z <- as.integer(x > 1)
lookat(z, 10)

z <- as.integer(x[,2] == "M")
lookat(z, 10)

col.types(x)
col.types(z)

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

## unique-methods

The Unique of an object

**Description**

This function gives the unique values of a db.obj, which are the column unique of a db.table or db.view.

**Usage**

```r
## S4 method for signature 'db.obj'
unique(x, incomparables = FALSE, ...)
```

**Arguments**

- **x**: A db.obj object, which the column unique are to be computed. The object has to have only one column otherwise an error will be raised.
- **incomparables**: Not implemented.
- **...**: Not implemented.
Value

An `db.Rquery`, whose column is the unique value of the column.

Note

This function applies only onto `db.obj` with one column. If you want to put the unique values from multiple columns together, you have to use `db.array`.

Author(s)

Author: Predictive Analytics Team at Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. <fmcquillan@pivotal.io>

See Also

`db.obj`, `db.data.frame`, `db.table`, `db.view`, `db.Rquery` are the class hierarchy structure of this package.

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

## get unique of all columns
unique(x$sex)

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

**vcov**

vcov methods for Madlib regression objects

Description

Functions to extract the variance-cocariance matrix for regression models fit in Madlib.
Usage

```r
## S3 method for class 'lm.madlib.grps'
vcov(object, na.action = NULL, ...)
## S3 method for class 'lm.madlib'
vcov(object, na.action = NULL, ...)
## S3 method for class 'logregr.madlib.grps'
vcov(object, na.action = NULL, ...)
## S3 method for class 'logregr.madlib'
vcov(object, na.action = NULL, ...)
```

Arguments

- `object` The regression model object, of class `lm.madlib`, `lm.madlib.grps`, `logregr.madlib`, `logregr.madlib.grps`.
- `na.action` A function, default is `NULL`. Possible choice is `na.omit.db.obj-method`.
- `...` Other arguments, not used.

Value

For `lm.madlib` and `logregr.madlib` objects, this function returns the variance-covariance matrix of the main parameters.

For `lm.madlib.grps` and `logregr.madlib.grps` objects, which are a list of models for multiple groups of data, returns a list, each of which is the variance-covariance matrix for the model of each group of data.

Author(s)

Author: Hong Ooi, Pivotal Inc.
Maintainer: Frank McQuillan, Pivotal Inc. `<fmcquillan@pivotal.io>`

See Also

- `madlib_lm`, `madlib_glm` for MADlib regression wrappers

Examples

```r
## Not run:

## set up the database connection
## Assume that .port is port number and .dbname is the database name
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)

x <- as.db.data.frame(abalone, conn.id = cid, verbose = FALSE)
lk(x, 10)

fit <- madlib.glm(rings < 10 ~ . - id | sex, data = x, family = "binomial")
```
vcov(fit)

vcov(fit[[1]])

db.disconnect(cid, verbose = FALSE)

## End(Not run)
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