Package ‘PivotalR’

March 13, 2021

Type Package
Title A Fast, Easy-to-Use Tool for Manipulating Tables in Databases
and a Wrapper of MADlib
Version 0.1.18.5
Date 2021-03-11
Maintainer Orhan Kislal <okislal@vmware.com>
Author Predictive Analytics Team at VMware.
<user@madlib.apache.org>
Depends R (>= 3.5.0), methods, Matrix, semver
Suggests DBI, RPostgreSQL, shiny, testthat, tools, rpart,
randomForest, topicmodels
Description Provides an R interface for the 'VMware Data Stack' running on
'PostgreSQL' or 'Greenplum' 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', which is an
open-source library for parallel and scalable in-database analytics.
License GPL (>= 2)
LazyLoad yes
LazyData yes
NeedsCompilation yes
Repository CRAN
Date/Publication 2021-03-12 23:00:06 UTC

R topics documented:

PivotalR-package . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . 4
abalone . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . 8
Aggregate functions . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . . 10
<table>
<thead>
<tr>
<th>R topics documented:</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>AIC</td>
<td>13</td>
</tr>
<tr>
<td>Arith-methods</td>
<td>15</td>
</tr>
<tr>
<td>array.len</td>
<td>17</td>
</tr>
<tr>
<td>arraydb.to.arrayr</td>
<td>19</td>
</tr>
<tr>
<td>as.db.data.frame</td>
<td>20</td>
</tr>
<tr>
<td>as.environment</td>
<td>24</td>
</tr>
<tr>
<td>as.factor-methods</td>
<td>25</td>
</tr>
<tr>
<td>by</td>
<td>26</td>
</tr>
<tr>
<td>cbind2-methods</td>
<td>28</td>
</tr>
<tr>
<td>clean.madlib.temp</td>
<td>29</td>
</tr>
<tr>
<td>coef</td>
<td>31</td>
</tr>
<tr>
<td>Compare-methods</td>
<td>32</td>
</tr>
<tr>
<td>conn.eql</td>
<td>35</td>
</tr>
<tr>
<td>conn.id</td>
<td>36</td>
</tr>
<tr>
<td>content</td>
<td>38</td>
</tr>
<tr>
<td>crossprod</td>
<td>39</td>
</tr>
<tr>
<td>db.connect</td>
<td>41</td>
</tr>
<tr>
<td>db.data.frame</td>
<td>43</td>
</tr>
<tr>
<td>db.data.frame-class</td>
<td>44</td>
</tr>
<tr>
<td>db.disconnect</td>
<td>46</td>
</tr>
<tr>
<td>db.existsObject</td>
<td>48</td>
</tr>
<tr>
<td>db.list</td>
<td>49</td>
</tr>
<tr>
<td>db.obj-class</td>
<td>50</td>
</tr>
<tr>
<td>db.objects</td>
<td>51</td>
</tr>
<tr>
<td>db.q</td>
<td>52</td>
</tr>
<tr>
<td>db.Rcrossprod-class</td>
<td>53</td>
</tr>
<tr>
<td>db.Rquery-class</td>
<td>54</td>
</tr>
<tr>
<td>db.search.path</td>
<td>57</td>
</tr>
<tr>
<td>db.table-class</td>
<td>58</td>
</tr>
<tr>
<td>db.view-class</td>
<td>59</td>
</tr>
<tr>
<td>delete</td>
<td>60</td>
</tr>
<tr>
<td>dim-methods</td>
<td>63</td>
</tr>
<tr>
<td>eql-methods</td>
<td>64</td>
</tr>
<tr>
<td>Extract database connection info</td>
<td>66</td>
</tr>
<tr>
<td>Extract-Replace-methods</td>
<td>68</td>
</tr>
<tr>
<td>Func-methods</td>
<td>70</td>
</tr>
<tr>
<td>generic.bagging</td>
<td>72</td>
</tr>
<tr>
<td>generic.cv</td>
<td>73</td>
</tr>
<tr>
<td>getTree.rf.madlib</td>
<td>76</td>
</tr>
<tr>
<td>groups</td>
<td>77</td>
</tr>
<tr>
<td>GUI</td>
<td>79</td>
</tr>
<tr>
<td>ifelse</td>
<td>80</td>
</tr>
<tr>
<td>is.db.data.frame</td>
<td>81</td>
</tr>
<tr>
<td>is.factor-methods</td>
<td>82</td>
</tr>
<tr>
<td>is.na-method</td>
<td>83</td>
</tr>
<tr>
<td>key</td>
<td>84</td>
</tr>
<tr>
<td>Logical-methods</td>
<td>85</td>
</tr>
<tr>
<td>madlib.arima</td>
<td>87</td>
</tr>
</tbody>
</table>
R topics documented:

madlib.elnet ................................................. 90
madlib.glm .................................................. 96
madlib.kmeans .............................................. 100
madlib.lda ................................................... 103
madlib.lm ..................................................... 105
madlib.randomForest ......................................... 109
madlib.rpart ................................................ 111
madlib.summary ............................................. 113
madlib.svm .................................................. 116
margins ....................................................... 120
merge-method ............................................... 123
na.action .................................................... 125
names-methods ............................................... 126
null.data .................................................... 127
perplexity.lda ............................................... 128
plot.dt.madlib ............................................. 129
predict ...................................................... 131
predict.arima ............................................... 134
predict.bagging.model ..................................... 135
predict.dt.madlib .......................................... 136
predict.elnet.madlib ....................................... 137
predict.lda .................................................. 138
predict.rf.madlib .......................................... 139
preview ...................................................... 141
print ......................................................... 143
print-methods ............................................... 145
print.arima.madlib ......................................... 146
print.dt.madlib ............................................ 147
print.elnet.madlib ......................................... 148
print.lm.madlib ............................................. 149
print.none.obj ............................................. 150
print.rf.madlib ............................................ 151
print.summary.madlib ..................................... 152
residuals ..................................................... 153
Row_actions ................................................ 154
sample-methods ............................................ 156
scale ......................................................... 157
sort ......................................................... 159
subset-methods ............................................. 160
summary ..................................................... 161
summary.arima.madlib ..................................... 162
summary.elnet.madlib ..................................... 163
summary.lm.madlib ......................................... 164
text.dt.madlib ............................................. 165
Type Cast functions ........................................ 166
unique-methods ............................................. 168
vcov ......................................................... 169

Index ......................................................... 172
**PivotalR-package**

An R font-end to PostgreSQL and Greenplum database, and wrapper for in-database parallel and distributed machine learning open-source library MADlib

**Description**

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/greenplum-db/PivotalR](https://github.com/greenplum-db/PivotalR).

**Details**

<table>
<thead>
<tr>
<th>Package:</th>
<th>PivotalR</th>
</tr>
</thead>
<tbody>
<tr>
<td>Type:</td>
<td>Package</td>
</tr>
<tr>
<td>Version:</td>
<td>0.1.18</td>
</tr>
<tr>
<td>Date:</td>
<td>2016-09-15</td>
</tr>
<tr>
<td>License:</td>
<td>GPL (&gt;= 2)</td>
</tr>
<tr>
<td>Depends:</td>
<td>methods, DBI, RPostgreSQL</td>
</tr>
</tbody>
</table>

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)

Author: Predictive Analytics Team at Pivotal Inc. <user@madlib.apache.org>, with contributions from Data Scientist Team at Pivotal Inc.
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 = "qianh1", 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
```
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$sex <- as.factor(y$sex) # y becomes a db.Rquery object now
fit2 <- madlib.lm(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((x$rings < 10) == predict(fit3, x)))

## 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$sex <- as.factor(y$sex) # 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$sex, mean))

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

## 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)

---

**Abalone**

**Abalone data set**

**Description**

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

**Usage**

```r
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

<table>
<thead>
<tr>
<th>Name</th>
<th>Data Type</th>
<th>Measurement Unit</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id</td>
<td>integer</td>
<td>–</td>
<td>index of each observation</td>
</tr>
<tr>
<td>Sex</td>
<td>nominal</td>
<td>–</td>
<td>M, F, and I (infant)</td>
</tr>
<tr>
<td>Length</td>
<td>continuous</td>
<td>mm</td>
<td>Longest shell measurement</td>
</tr>
<tr>
<td>Diameter</td>
<td>continuous</td>
<td>mm</td>
<td>perpendicular to length</td>
</tr>
<tr>
<td>Height</td>
<td>continuous</td>
<td>mm</td>
<td>with meat in shell</td>
</tr>
<tr>
<td>Whole</td>
<td>continuous</td>
<td>grams</td>
<td>whole abalone</td>
</tr>
<tr>
<td>Shucked</td>
<td>continuous</td>
<td>grams</td>
<td>weight of meat</td>
</tr>
<tr>
<td>Viscera</td>
<td>continuous</td>
<td>grams</td>
<td>gut weight (after bleeding)</td>
</tr>
<tr>
<td>Shell</td>
<td>continuous</td>
<td>grams</td>
<td>after being dried</td>
</tr>
<tr>
<td>Rings</td>
<td>integer</td>
<td>–</td>
<td>+1.5 gives the age in years</td>
</tr>
</tbody>
</table>

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


Examples

```r
## Not run:

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

functions to perform a calculation on multiple values and return a single value

### 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.obj'
max(x, ..., na.rm = FALSE)

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

## S4 method for signature 'db.obj'
.sd(x)
```
Aggregate functions

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

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

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

colAgg(x)

db.array(x, ...)

Arguments

- **x** A db.obj object. The signature of the method.
  - For db.array, x can also be a normal R object like double value.
- **...** further arguments passed to or from other methods This is currently not implemented.
- **na.rm** logical. Should missing values (including 'NaN') be removed? This is currently not implemented.
- **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: mean, sum, count, max, min, sd, and var, the signature x must be a reference to a single column in a table.

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

Value

- For mean, a 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 colMeans.
- For sum, a 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 colSums.
- 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.
 Aggregate functions

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)

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

See Also

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

Examples

## 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)

## get the max value of a column
max(x$diameter)

## get the min value of a column
min(x$diameter)

## get the standard deviation of the values in column
sd(x$diameter)

## get the variance of the values in column
### 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, ...)

AIC(object, ..., k=2)

## S3 method for class 'logregr.madlib'
extractAIC(fit, scale=0, k=2, ...)

## S3 method for class 'logregr.madlib.grps'
extractAIC(fit, scale=0, k=2, ...)

## S3 method for class 'logregr.madlib'
logLik(object, ...)
```
## S3 method for class 'logregr.madlib.grps'
logLik(object, ...)
AIC(object, ..., k=2)

## S3 method for class 'glm.madlib'
extractAIC(fit, scale=0, k=2, ...)

## S3 method for class 'glm.madlib.grps'
extractAIC(fit, scale=0, k=2, ...)

## S3 method for class 'glm.madlib'
logLik(object, ...)

## S3 method for class 'glm.madlib.grps'
logLik(object, ...)

## S3 method for class 'glm.madlib.grps'
AIC(object, ..., k=2)

### 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 and extractAIC.

### 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

```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
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)
```

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 '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,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
## S4 method for signature 'db.obj,numeric'
e1 %/% e2
## S4 method for signature 'db.obj,numeric'
e1 ^ e2

**Arguments**

e1, e2 numeric or db.obj 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.
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

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:

```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)
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', 'cd', '"axx,t"'] becomes a string in R '{ab,cd,\"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>
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'

```r
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'

```r
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'

```r
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 an existing `db.data.frame` object. See `db.Rquery` for more.
  - For `as.db.Rview`, `x` must be a `db.Rquery` object.

- **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, its 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.

nrowsWhen 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>
as.db.data.frame

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

## do some operations and preview the result
y <- (x[, -2] + 1.2) * 2
lk(y, 20, FALSE)

## table abalone has a column named "id"
lk(sort(x, INDICES = x$id), 20) # the preview is ordered by "id" value

## create a copied table
## x[,] converts x from db.data.frame object to db.Rquery object
z <- as.db.data.frame(x[,])

## Force the data type, use random table name
z1 <- as.db.data.frame(x$rings, field.types = list(rings="integer"))

db.disconnect(cid, verbose = FALSE)

## End(Not run)
**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.

**Author(s)**

Author: Hong Ooi, Pivotal Inc. `<hooi@pivotal.io>`

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

**See Also**

`as.environment`
as.factor-methods

Convert one column of a db.obj object into a categorical variable

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.Rquery 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:
## get help for a method
## help("as.factor,db.obj-method")
## set up the database connection
## Assume that .port is port number and .dbname is the database name

```r
cid <- db.connect(port = .port, dbname = .dbname, verbose = FALSE)
```

## create a temporary table from the example data.frame "abalone"

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

## set sex to be a categorical variable

```r
x$sex <- as.factor(x$sex)
```

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

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

## another temporary table

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

## specify factor during fitting

```r
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

```r
fit4 <- madlib.glm(rings < 10 ~ sex + length + diameter, data = z, family = "binomial")
```

## using relevel to change the reference level

```r
x$sex <- relevel(x$sex, ref = "M")
madlib.lm(rings ~ . - id, data = x) # use "M" as the reference level
```

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

## End(Not run)

---

### by

#### 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 a 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 1k 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.

1k or lookat can display the actual result of this function.

Examples

## 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
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)
## mean values for each column
lk(by(x, x$sex, mean))

## No need to compute the mean of id and sex
lk(by(x[,][-c(1,2)], x$sex, mean))
lk(by(x[,][-c(1,2)], x[,]2, mean)) # the same
lk(by(x[,][-c(1,2)], x[,"sex"], mean)) # the same

## The return type of FUN is not db.obj
dat <- x

## Fit linear model to each group of data
by(dat, dat$sex, function(x) madlib.lm(rings ~ . - id - sex, data = x))
db.disconnect(cid, verbose = FALSE)

## End(Not run)

### 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 db.Rquery object, which contains all columns of x and y.
as.list: A list of db.Rquery objects, which are the columns of x
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.
array.len measures the length of the array in an array column.

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)

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)
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

```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)

## 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"),
```
coef

## 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

coeff.

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)
c 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
```


### 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.Rquery** 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

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

Arguments

```r
cconn.id1 An integer, a connection ID number.
cconn.id2 An integer, another connection ID number.
```
conn.id

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
collection 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
## 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(x) 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
conn.id(x)
conn.id(x) <- value
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)
```
content

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. <fmcquillan@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")
## 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)

---

**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

```
## 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

`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 $n \times n$. 
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

## 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$arr <- db.array(1, x$length, x$diameter)
lookat(crossprod(x$arr))

## -----------------------------------------------------
## 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"))
}
## create a data table with a random name

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

## exclude id and sex columns

p <- pca(dat[, -c(1,2)])

p$val # eigenvalues

db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

**db.connect**

Create a connection to a database

**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**

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.
### 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

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:

## 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 \code{db.data.frame} or \code{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

\code{.name}: Object of class "character". It is the table name if this \code{db.data.frame} was created using just a table name. It can also be a two-element array if this \code{db.data.frame} was created. This slot is obsolete.

\code{.content}: Object of class "character". The table name. The function \code{content} can get this value.
**db.data.frame-class**

- **.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 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.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`
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

## 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)

---

**db.disconnect**

**Disconnect a connection to a database**

**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

## 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)

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. <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` 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.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

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

```r
## 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)
```

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 `db.data.frame` except `content`, `lk` and `delete`.

`db.Rview-class` is a sub-class of `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 `db.Rview` or `db.Rquery`. `as.db.Rview` casts a `db.Rquery` object into a `db.Rview` object.

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

**Slots**

`.content`: Object of class "character". The SQL query that represents the operations. The function `content` can get this value.

`.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.

`.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.

`.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.

`.conn.id`: Object of class "numeric", an integer. The ID number of the database connection where `.source` resides. The functions `conn.id` and `conn.id<-` can get and set this value.

`.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 `names,db.obj-method` gets this value.

`.key`: Object of class "character". The name of the primary key column name in `.source`. 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.

`.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.

`.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.

`.where`: Object of class "character". The condition string used in "where" inside the SQL query.

`.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.

`.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.
.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(0), 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)
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.xs

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 sub-class of `db.obj`.
`lk` or `lookat` display a part of the table

delete

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)
```
## S4 method for signature 'lm.madlib'
delete(x)

## S4 method for signature 'lm.madlib.grps'
delete(x)

## S4 method for signature 'logregr.madlib'
delete(x)

## S4 method for signature 'logregr.madlib.grps'
delete(x)

## S4 method for signature 'bagging.model'
delete(x)

## S4 method for signature 'elnet.madlib'
delete(x)

## S4 method for signature 'dt.madlib'
delete(x)

## S4 method for signature 'dt.madlib.grps'
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.Rquery` 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 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.
Or an object which is the result of \texttt{madlib.rpart}. All result tables will be deleted.

\begin{description}
\item[conn.id] An integer, default is 1. The connection ID to the database.
\item[is.temp] A logical, default is \texttt{FALSE}. Whether the table/view is temporary.
\item[cascade] A logical, default is \texttt{FALSE}. Whether to delete objects together with all the objects depending on it.
\end{description}

Details

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

Value

When \(x\) is \texttt{db.data.frame} or table/view name, this function returns a logical value, which is \texttt{TRUE} if the deletion is successful.

No value is returned if \(x\) is \texttt{db.Rquery}

Author(s)

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

See Also

\texttt{db.data.frame} creates an object pointing to a table/view in the database.
\texttt{db.objects} lists all tables and views in a database together with their schema.
\texttt{db.existsObject} tests whether a table/view exists in the database.
\texttt{as.db.data.frame} creates a \texttt{db.data.frame} from a \texttt{data.frame}, a data file or a \texttt{db.Rquery}.
\texttt{madlib.lm}, \texttt{madlib.glm}, \texttt{madlib.summary}, \texttt{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)
```
y <- as.db.data.frame(abalone, "abalone", conn.id = cid, is.temp = TRUE)

lk(y, 10)

db.existsObject("abalone", cid, is.temp = TRUE)

db.existsObject("abalone", cid, is.temp = FALSE)

delete("abalone", cid)

p <- db.objects()
p[p == "abalone"]

## Example: delete multiple tables
## all table in public schema start with "ornste"
to.delete <- db.objects("public.ornste", conn.id = cid)
for (table.name in to.delete) delete(table.name, conn.id = cid)

db.disconnect(cid, verbose = FALSE)

## End(Not run)

dim-methods                            Dimension of a table

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

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

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

## 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)
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. <fmcquillan@pivotal.io>

See Also

lk or lookat Displays the actual data in 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)
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. <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.

conn.eql tests whether two connections are actually the same one.

Examples

## 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)

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)
db.disconnect(cid2, verbose = FALSE)

## End(Not run)

---

**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'
## S4 method for signature 'db.obj'
x$name
## S4 method for signature 'db.obj,ANY,ANY,ANY'
x[i, j, ..., drop = TRUE]
## S4 replacement method for signature 'db.obj,character'
x$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
```
Extract-Replace-methods

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)
Description

Functions that apply onto `db.obj` objects

Usage

```r
## S4 method for signature 'db.obj'
exp(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

\texttt{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. <fmcquillan@pivotal.io>

See Also

\texttt{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 boostrap 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. <fmcquillan@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

```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)

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 boostrap aggregate computation.

Examples

## Not run:

## 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",
    },
    predict,
    function(predicted, data) {
      lookat(mean((data$rings - predicted)^2))
    }, data = dat, verbose = FALSE)

## ----------------------------------------------------------------------
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 madlib.randomForest is passed as input to this function.

Usage
getTree.rf.madlib(object, k=1, ...)

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

Value
A data frame object similar to R’s 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)

<table>
<thead>
<tr>
<th>groups</th>
<th>Summary information for Logistic Regression output</th>
</tr>
</thead>
</table>

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 a `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)
```
## create a table from the example data.frame
delete("abalone", conn.id = cid)
source_data <- as.db.data.frame(abalone, "abalone", conn.id = cid, verbose = FALSE)
lk(source_data, 10)

## logistic regression
fit <- madlib.glm(rings < 10 ~ . - id | sex, data = source_data, family = "binomial")
groups(fit) # all grouping column values
groups(fit[[1]]) # the first model's grouping column value
db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

### GUI

**Graphical interface for PivotalR based upon shiny**

### 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()
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. <fmcquillan@pivotal.io>

### References


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

## 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. <fmcquillan@pivotal.io>

See Also
db.obj

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
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')
## getting the primary key

```r
is.db.data.frame(x) # check if x is of type db.data.frame

db.disconnect(cid, verbose = FALSE)

## End(Not run)
```

---

**is.factor-methods**  
*Detect whether a db.obj object is a categorical object*

---

### Description

This function detects whether a db.obj object is a categorical object.

### Usage

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

### Arguments

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

### Value

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

### Author(s)

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

### See Also

- `as.factor`, `db.obj-method` converts a column 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)
```
## 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. <fmcqillan@pivotal.io>

### See Also

- `lk` or `lookat` Displays the contents of a `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)

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

#Query which entries of x are NULL
is.na(x)

y <- x
y[is.na(y)] <- 3

z <- x
z[is.na(x$height),"height"] <- 23

db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

key

*Get or set the primary key for a table*

Description

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

Usage

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 = '/quotesingle.Varid'/quotesingle.Var, 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)
```

Logical-methods

Logical operations for db.obj objects

Description

These binary operators perform logical operations on db.obj objects

Usage

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

## 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.

**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$rings[x$length > 10 & x$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.Query`, 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)

... Other optional parameters. Not implemented.

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.

Value

Returns an `arima.css.madlib` object, which is a list that contains the following items:

coef A vector of double values. The fitting coefficients of AR, MA and mean value (if include.mean is TRUE).

s.e. A vector of double values. The standard errors of the fitting coefficients.

series A string, the data source table or SQL query.
### madlib.arima

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>time.stamp</td>
<td>A string, the name of the time stamp column.</td>
</tr>
<tr>
<td>time.series</td>
<td>A string, the name of the time series column.</td>
</tr>
<tr>
<td>sigma2</td>
<td>the MLE of the innovations variance.</td>
</tr>
<tr>
<td>loglik</td>
<td>the maximized conditional log-likelihood (of the differenced data).</td>
</tr>
<tr>
<td>iter.num</td>
<td>An integer, how many iterations of the LM algorithm is used to fit the time series with ARIMA model.</td>
</tr>
<tr>
<td>exec.time</td>
<td>The time spent on the MADlib ARIMA fitting.</td>
</tr>
<tr>
<td>residuals</td>
<td>A <code>db.data.frame</code> object that points to the table that contains all the fitted innovations.</td>
</tr>
<tr>
<td>model</td>
<td>A <code>db.data.frame</code> object that points to the table that contains the coefficients and standard error. This table is needed by <code>predict.arima.css.madlib</code>.</td>
</tr>
<tr>
<td>statistics</td>
<td>A <code>db.data.frame</code> object that points to the table that contains information including log-likelihood, sigma^2 etc. This table is needed by <code>predict.arima.css.madlib</code>.</td>
</tr>
<tr>
<td>call</td>
<td>A language object. The matched function call.</td>
</tr>
</tbody>
</table>

### Author(s)

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

### References


### See Also

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

### Examples

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

## 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)

dim(x)

names(x)

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

s

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

s # s does not exist any more

## do not use formula
s <- madlib.arima(x$val, x$id, order = c(2,0,1))

s

lookat(sort(s$residuals, F, s$residuals$tstamp), 10)

lookat(s$model)

lookat(s$statistics)

## 10 forecasts
pred <- predict(s, n.ahead = 10)

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

## Use expressions
s <- madlib.arima(val+2 ~ I(id + 1), x, order = c(2,0,1))

db.disconnect(cid, verbose = FALSE)

## End(Not run)
**Description**

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

**Usage**

```r
call = 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.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" 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 is `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", "igd"/"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: - `max.stepsize` A numeric value, default is 4.0. Initial backtracking step size. At each iteration, the algorithm first tries stepsize = max.stepsize, and if it does
not work out, it then tries a smaller step size, \text{stepsize} = \text{stepsize/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.

- \text{eta} A numeric value, default is 2. If \text{stepsize} does not work \text{stepsize / eta} is tried. Must be greater than 1.

- \text{use\_active\_set} A logical value, default is \text{FALSE}. If \text{use\_active\_set} is \text{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.

- \text{activeset\_tolerance} A numeric value, default is the value of the tolerance argument (see below). The value of tolerance used during active set calculation.

- \text{random\_stepsize} A logical value, default is \text{FALSE}. Whether to add some randomness to the step size. Sometimes, this can speed up the calculation.

(2) If \text{method} is "sgd", the allowed control parameters are:

- \text{stepsize} The default is 0.01.

- \text{step\_decay} A numeric value, the actual setpsize used for current step is (previous stepsize) / exp\text{setp.decay}. The default value is 0, which means that a constant stepsize is used in SGD.

- \text{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.

- \text{parallel} A logical value, the default is \text{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:

- \text{max\_iter} An integer, default is 100. The maximum number of iterations that are allowed.

- \text{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 \text{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.lambdas`: 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.lambdas` 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.

---

### glmnet

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

\[ \frac{1}{2} \text{RSS}/nobs + \lambda \text{penalty}, \]

and for the other models it is

\[ -\text{loglik}/nobs + \lambda \text{penalty}. \]

### 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.
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. <fmcquillan@pivotal.io>
madlib.elnet

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
db.disconnect(cid, verbose = FALSE)

## End(Not run)
```
Generalized Linear Regression by MADlib in databases

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

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(inverse).
- **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 itertion 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.
**madlib.glm**

**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
## 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 - arr[1:2], data = array.data, family = binomial)
fit <- madlib.glm(rings < 10 ~ arr[1:7] + sex | id

fit <- madlib.glm(rings < 10 ~ arr - arr[8] + sex | id

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

## step(start)

# Example 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)

madlib.kmeans

Wrapper for MADlib's Kmeans clustering function
**madlib.kmeans**

**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**

```r
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 parameters 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

```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)

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

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

seed.matrix <- matrix(
c(14.23,1.71,2.43,15.6,127,2.8,3.06,0.28,2.29,5.64,1.04,3.92,1065,
  13.2,1.78,2.14,11.2,2.65,2.76,0.26,1.28,4.38,1.05,3.49,1050),
  byrow=T, nrow=2)
```
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

```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)

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

```r
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 separate 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. Categorical 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

```r
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.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.
- **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 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, \( y \sim x + \sin(z) \mid v + w \) 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 \texttt{as.factor,db.obj-method} to denote that a variable is categorical, and the corresponding dummy variables are created and fitted. See \texttt{as.factor,db.obj-method} for more.

Value

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

A \texttt{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 \texttt{std_err} and \texttt{coef}.

- **p_values**
  - A numeric matrix, the p-values of \texttt{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 \texttt{hetero = TRUE}, the Breusch-Pagan test statistics for each combination of the grouping column values.

- **bp_p_value**
  - A numeric array when \texttt{hetero = TRUE}, the Breusch-Pagan test p value for each combination of the grouping column values.
madlib.lm

| 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

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)

## 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)

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)
```
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.
madlib.randomForest

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 minsplit, 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
madlib.rpart

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
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', 'misclssification' 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 continuous 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
madlib.rpart

### Arguments

- `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

[1] Documentation of decision tree in MADlib 1.6, [https://madlib.apache.org/docs/latest/](https://madlib.apache.org/docs/latest/)

### See Also

- `plot.dt.madlib`, `text.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)
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
```
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

```
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, object**: 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.

- **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 \texttt{db.view}, then extracting data from it would actually compute the view, which might take a longer time, especially for large data sets. When \texttt{interactive} is TRUE, this function will ask the user whether to continue to extract data from the view.

Value

A \texttt{data.frame} object. Each column in the table (or \texttt{target.cols}) is a row in the result data frame. Each column of the data frame is described below:

- \texttt{group_by} character. Group-by column names (\texttt{NA} if none provided)
- \texttt{group_by_value} character. Values of the group-by columns (\texttt{NA} if no grouping)
- \texttt{target_column} character. Targeted column values for which summary is requested
- \texttt{column_number} integer. Physical column number for the target column in the database
- \texttt{data_type} character. Data type of target column. Standard database descriptors will be displayed
- \texttt{row_count} numeric. Number of rows for the target column
- \texttt{distinct_values} numeric. Number of distinct values in the target column
- \texttt{missing_values} numeric. Number of missing values in the target column
- \texttt{blank_values} numeric. Number of blank values (blanks are defined as values with only whitespace)
- \texttt{fraction_missing} numeric. Percentage of total rows that are missing. Will be expressed as a decimal (e.g. 0.3)
- \texttt{fraction_blank} numeric. Percentage of total rows that are blank. Will be expressed as a decimal (e.g. 0.3)
- \texttt{mean} numeric. Mean value of target column (if target is numeric, else \texttt{NA})
- \texttt{variance} numeric. Variance of target columns (if target is numeric, else \texttt{NA} for strings)
- \texttt{min} numeric. Min value of target column (for strings this is the length of the shortest string)
- \texttt{max} numeric. Max value of target column (for strings this is the length of the longest string)
- \texttt{first_quartile} numeric. First quartile (25th percentile, valid only for numeric columns)
- \texttt{median} numeric. Median value of target column (valid only for numeric columns)
- \texttt{third_quartile} numeric. Third quartile (75th percentile, valid only for numeric columns)
- \texttt{quantile_array} numeric. Percentile values corresponding to \texttt{ntile_array}
- \texttt{most_frequent_values} character. Most frequent values
- \texttt{mfv_frequencies} character. Frequency of the most frequent values

The \texttt{data.frame} has an extra attribute names "\texttt{summary}" which is a \texttt{db.data.frame} object and wraps the result table created by MADlib inside the database. One can access this object using \texttt{attr(res,"summary")}, where \texttt{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.
delele safely deletes the result of this function.

Examples

## 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 = .id)
as.db.data.frame(abalone, "abalone", conn.id = .id, verbose = FALSE)
x <- db.data.frame("abalone", conn.id = .id, 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"),
get.distinct=FALSE,
get.quartiles=TRUE,
ntile=c(0.1, 0.6),
n.mfv=5,
estimate=TRUE,
interactive=FALSE)

print(summary_result)

db.disconnect(cid, verbose = FALSE)

## End(Not run)
Description

This function wraps MADlib’s SVM for classification, regression and novelty detection.

Usage

```r
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.
- `type`: A string, default: "classification". Indicate type of analysis to perform: "classification", "regression" or "one-class".
- `kernel`: A string, default: "gaussian". Type of kernel. Currently three kernel types are supported: 'linear', 'gaussian', and 'polynomial'.
- `degree`: Default: 3. The parameter needed for polynomial kernel
- `gamma`: Default: 1/num_features. The parameter needed for gaussian kernel
- `coef0`: Default: 1.0. The independent term in polynomial kernel
- `class.weight`: 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.

tolerance 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.

epsilon 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].

cross Default: 0. Number of folds (k). Must be at least 2 to activate cross validation. If a value of k > 2 is specified, each fold is then used as a validation set once, while the other k - 1 folds form the training set.

lambda Default: [0.01]. Regularization parameter. Must be non-negative.

control A list, which contains the more control parameters for the optimizer.
  - init.stepsize: 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., "init_stepsize = [1, 0.1, 0.001]". 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.
  - decay.factor: Default: [0.9]. Control the learning rate schedule: 0 means constant rate; <-1 means inverse scaling, i.e., stepsize = init_stepsize / iteration; > 0 means exponential decay, i.e., stepsize = init_stepsize * decay_factor^iteration.
  - max.iter: Default: [100]. The maximum number of iterations allowed.
  - norm: Default: 'L2'. Name of the regularization, either 'L2' or 'L1'.
  - 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, \( y \sim x + \sin(z) | v + w \) 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:

- coef: A vector, the fitting coefficients.
- 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 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

## 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 > 7), data = data, type = "regression")

fit

## use I(.) for expressions
fit <- madlib.svm(rings > 7 ~ height + shell + diameter + I(diameter^2),
                 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)
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),
...

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=i`, compute the marginal effects of the i-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

## 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)

## create a data table in database and the R wrapper
delete("abalone", conn.id = cid)
dat <- as.db.data.frame(abalone, "abalone", conn.id = cid)

fit <- madlib.lm(rings ~ length + diameter*sex, data = dat)
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  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

## 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.
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. &lt;fmcquillan@pivotal.io&gt;

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",
            "Spatial Statistics", "Stochastic Simulation",
            "Interactive Data Analysis",
            "An Introduction to R"),
  other.author = c(NA, "Ripley", NA, 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
m1 <- 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**

```r
## 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`.

**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:

```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)
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)
```

names-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

```r
## 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`.

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>
null.data

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
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 + lncrd + lnauto +
                 lnconoth + lnconrp + intmsrfv + lnrenr1a + lnrenr2a +
                 lnrenr3a, data = x)

## select columns
y <- x[,c("sf_mrtg_pct_assets","ris_asset", "lncrd","lnauto",
             "lnconoth","lnconrp","intmsrfv","lnrenr1a","lnrenr2a",
             "lnrenr3a")]
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)
```

---

**perplexity.lda**

Perplexity of LDA predictions

Description

Perplexity is a measurement of how well a probability distribution or probability model predicts a sample. This functions computes the perplexity of the prediction by `link(predict.madlib.lda)`
Usage

```r
## S3 method for class 'lda.madlib'
perplexity(object, predict_output_table, ...)
```

Arguments

- `object` The result of `madlib.lda`.
- `predict_output_table` The result of `predict` on the `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. <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)
```

---

plot.dt.madlib

Plot the result of `madlib.rpart`

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 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))

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

```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 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))

## predictions for one group of data where sex = I
idx <- which(groups(fit)["sex"] == "I")  # which sub-model
predI <- predict(fit[idx], x[x$sex == "I",])  # predict on part of data

## Example 3 --------

## plot the predicted values v.s. the true values
ap <- ans  # true values
ap$pred <- pred  # 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, NULL), 1000)  # sort randomly

plot(random.sample)

## glm prediction

fit <- madlib.glm(rings ~ . - id | sex, data = x, family = poisson(log),
predict.arima

\[
\text{control} = \text{list}(\text{max.iter} = 20)
\]

\[
p \leftarrow \text{predict}(f)
\]

\[
\text{lk}(p, 10)
\]

db.disconnect(cid, verbose = FALSE)

## End(Not run)

---

predict.arima  

*Forecast from MADlib's ARIMA fits*

### Description

Forecast from models fit by `linlk(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.

### 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. <fmcquillan@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 boostrap aggregating models

Usage

## 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.Rquery 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:

## 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)
predict.dt.madlib

**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"),
         ...)```

**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.
predict.elnet.madlib

Author(s)

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

db.disconnect(cid)

## End(Not run)
predict.lda

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.Rquery` 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
```

 predicts.lda

**Prediction function for MADlib's LDA models**

Description

Labelling test documents using a learned LDA model built by `linlk(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**  
Compute the predictions of the model produced by `madlib.randomForest`

**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

## 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 val-
ues. 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 col-
umn. 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. <fmcquillan@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, madlib.predict.logregr, madlib.predict.elnet, madlib.predict.arima.css, madlib,
predict.dt, madlib.predict.rf 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.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)
```

---

**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 `lk` 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 form database into arrays.
### Display results of logistic regression

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.grps'
print(x, digits = max(3L, getOption("digits")
  - 3L), ...)  

## S3 method for class 'logregr.madlib'
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

## 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)
```

print.arima.madlib

Display results of ARIMA fitting of `madlib.arima`

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), ...)  

## 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. <fmcquillan@pivotal.io>

See Also

madlib.arima Wrapper for MADlib ARIMA model fitting

Examples

## Not run:
## Please see the examples in madlib.arima doc
## End(Not run)

print.dt.madlib

Print the result of madlib.rpart

Description

This function prints the result of madlib.rpart to the screen. It internally calls R’s print.rpart function.

Usage

## 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)
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))

print(fit)

db.disconnect(cid)

## End(Not run)
```

**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

```r
## 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

## 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

Examples

```r
## see the examples in madlib.lm
```

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
PivotalR 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

## 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.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.
 residuals

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
## see the examples in madlib.summary

residuals

Residuals methods for Madlib regression objects

Description
Functions to extract the residuals for regression models fit in Madlib.

Usage

## 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. <fmcquillan@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

## 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.
...

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)
sample-methods

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
## 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,FALSE,"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. <fmcquillan@pivotal.io>
See Also

generic.bagging uses sample

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)

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

## 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

```r
## 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

## 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

## 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 Logistics 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>
summary.lm.madlib

See Also

madlib.elnet Wrapper for MADlib elastic net regularization.

Examples

```r
## see the examples in madlib.elnet
```

---

summary.lm.madlib  

**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, ...)

## 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. `<fmcquillan@pivotal.io>`

**See Also**

madlib.lm Wrapper for MADlib linear regression

**Examples**

```r
## see the examples in madlib.lm
```
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

Type Cast functions

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, ...)
```
## 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.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.Rquery** 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>
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)
```

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)
Index

! =, character, db.obj-method (Compare-methods), 85
!=, db.obj, character-method (Compare-methods), 32
!=, db.obj, db.obj-method (Compare-methods), 32
!=, db.obj, logical-method (Compare-methods), 32
!=, db.obj, numeric-method (Compare-methods), 32
!=, logical, db.obj-method (Compare-methods), 32
!=, numeric, db.obj-method (Compare-methods), 32

* `tree`
  getTree.rf.madlib, 76
  madlib.randomForest, 109
  madlib.rpart, 111
  plot.dt.madlib, 129
  print.dt.madlib, 147
  print.rf.madlib, 151
  text.dt.madlib, 165

* GUI
  GUI, 79
  print.none.obj, 150

* IO
  print, 143
  print-methods, 145
  print.arima.madlib, 146
  print.elnet.madlib, 148
  print.lm.madlib, 149
  print.none.obj, 150
  print.summary.madlib, 152

* classes
  db.data.frame-class, 44
  db.obj-class, 50
  db.Rcrossprod-class, 53
  db.Rquery-class, 54
  db.table-class, 58
  db.view-class, 59

* connection
  db.connect, 41
  db.disconnect, 46

* data operation
  array.len, 17
  by, 26
  cbind2-methods, 28
  Extract-Replace-methods, 68
  key, 84
  merge-method, 123
  null.data, 127
  preview, 141
  subset-methods, 160
  Type Cast functions, 166

* database
  arraydb.to.arrayr, 19
  as.db.data.frame, 20
  clean.madlib.temp, 29
  conn.eql, 35
  conn.id, 36
  content, 38
  db.connect, 41
  db.data.frame, 43
  db.data.frame-class, 44
  db.disconnect, 46
  db.existsObject, 48
  db.list, 49
  db.obj-class, 50
  db.objects, 51
  db.q, 52
  db.Rcrossprod-class, 53
  db.Rquery-class, 54
  db.search.path, 57
  db.table-class, 58
  db.view-class, 59
  delete, 60
  dim-methods, 63
INDEX

eql-methods, 64
Extract database connection info, 66
is.db.data.frame, 81
is.na-method, 83
key, 84
merge-method, 123
names-methods, 126
null.data, 127
preview, 141
sample-methods, 156
sort, 159
subset-methods, 160
unique-methods, 168

* datasets
  abalone, 8

* factor
  as.factor-methods, 25

* madlib
  groups, 77
  madlib.arima, 87
  madlib.elnet, 90
  madlib.glm, 96
  madlib.kmeans, 100
  madlib.lda, 103
  madlib.lm, 105
  madlib.summary, 113
  madlib.svm, 116
  perplexity.lda, 128
  predict.arima, 134
  predict.bagging.model, 135
  predict.dt.madlib, 136
  predict.elnet.madlib, 137
  predict.lda, 138
  predict.rf.madlib, 139
  predict, 131
  sample-methods, 156
  scale, 157
  summary, 161
  summary.arima.madlib, 162
  summary.elnet.madlib, 163
  summary.lm.madlib, 164

* methods
  as.db.data.frame, 20
  as.factor-methods, 25
  by, 26
  Compare-methods, 32
crossprod, 39
  Func-methods, 70
generic.bagging, 72
generic.cv, 73
groups, 77
  is.factor-methods, 82
  is.na-method, 83
Logical-methods, 85
margins, 120
perplexity.lda, 128
predict, 131
predict.arima, 134
predict.bagging.model, 135
predict.dt.madlib, 136
predict.elnet.madlib, 137
predict.lda, 138
predict.rf.madlib, 139
sample-methods, 156
scale, 157
summary, 161
summary.arima.madlib, 162
summary.elnet.madlib, 163
summary.lm.madlib, 164

* package
  PivotalR-package, 4

* stats
  Aggregate functions, 10
generic.bagging, 72
generic.cv, 73
groups, 77
madlib.arima, 87
madlib.elnet, 90
madlib.glm, 96
madlib.lm, 105
madlib.summary, 113
madlib.svm, 116
margins, 120
perplexity.lda, 128
predict, 131
predict.arima, 134
predict.bagging.model, 135
predict.dt.madlib, 136
predict.elnet.madlib, 137
predict.lda, 138
predict.rf.madlib, 139
sample-methods, 156
scale, 157
summary, 161
summary.arima.madlib, 162
summary.elnet.madlib, 163
summary.lm.madlib, 164

* utilities
  print.elnet.madlib, 148
  print.none.obj, 150
  print.summary.madlib, 152

* utility
  array.len, 17
  arraydb.to.arrayr, 19
  as.db.data.frame, 20
  cbind2-methods, 28
  clean.madlib.temp, 29
  coef, 31
  conn.eql, 35
  conn.id, 36
  content, 38
  db.connect, 41
  db.data.frame, 43
  db.disconnect, 46
  db.existsObject, 48
  db.list, 49
  db.objects, 51
  db.q, 52
  db.search.path, 57
  delete, 60
  eql-methods, 64

Extract database connection info,
abalone, 8
abs (Func-methods), 70
acos, db.obj-method (Func-methods), 70
Aggregation functions, 10, 27, 45
AIC, 13, 14, 154
Arith methods, 45
Arith methods (Arith-methods), 15
array.len, 17, 29
arraydb.to.array, 19, 142
as.character (Type Cast functions), 166
as.character, db.obj-method (Type Cast functions), 166
as.data.frame, 24
as.data.frame.db.Rquery (preview), 141
as.data.frame.db.table (preview), 141
as.data.frame.db.view (preview), 141
as.Date, db.obj-method (Type Cast functions), 166
as.data.frame, 20, 25, 43, 44, 46, 50, 54, 56, 58–60, 62, 81, 127
as.data.frame, character-method (as.data.frame), 20
as.data.frame, data.frame-method (as.data.frame), 20
as.data.frame, db.data.frame-method (as.data.frame), 20
as.data.frame, db.data.frame-method (as.data.frame), 20
as.data.frame, db.data.frame-method (as.data.frame), 20
as.data.frame, db.Rquery-method (as.data.frame), 20
as.db.Rview, 55, 56
as.db.Rview (as.db.data.frame), 20
as.double (Type Cast functions), 166
as.double, db.obj-method (Type Cast functions), 166
as.environment, 24, 24
as.factor, 22, 99, 108, 119
as.factor, db.obj-method (as.factor-methods), 25
as.factor-methods, 25
as.integer (Type Cast functions), 166
as.integer, db.obj-method (Type Cast functions), 166
as.interval (Type Cast functions), 166
as.list, 18
as.list, db.obj-method (cbind2-methods), 28
as.logical (Type Cast functions), 166
as.logical, db.obj-method (Type Cast functions), 166
as.numeric (Type Cast functions), 166
as.numeric, db.obj-method (Type Cast functions), 166
as.time (Type Cast functions), 166
as.timestamp (Type Cast functions), 166
asin, db.obj-method (Func-methods), 70
atan, db.obj-method (Func-methods), 70
atan2, db.obj, db.obj-method (Func-methods), 70
atan2, db.obj, numeric-method (Func-methods), 70
atan2, numeric, db.obj-method (Func-methods), 70
by, 26, 160
by, db.obj-method (by), 26
cbind, 18
cbind (cbind2-methods), 28
cbind2, 18
cbind2 (cbind2-methods), 28
cbind2.db.obj, db.obj-method (cbind2-methods), 28
cbind2-methods, 28
clean.madlib.temp, 29
col.types (Type Cast functions), 166
colAgg (Aggregate functions), 10
colMeans, db.obj-method (Aggregate functions), 10
colSums, db.obj-method (Aggregate functions), 10
Compare methods, 45
Compare methods (Compare-methods), 32
Compare-methods, 32
cconn (Extract database connection info), 66
cconn, 35, 37, 42, 47, 49, 67
INDEX

conn.id, 36, 45, 55
conn.id<- (conn.id), 36
connection info, 36, 37, 42, 47, 49
connection info (Extract database connection info), 66
content, 38, 44, 55, 65
cos, db.obj-method (Func-methods), 70
count (Aggregate functions), 10
count, db.obj-method (Aggregate functions), 10
crossprod, 39, 53
crossprod, db.obj, ANY-method (crossprod), 39
data.frame, 141
db (db.q), 52
db.array, 18, 29, 40, 158, 169
db.array (Aggregate functions), 10
db.connect, 30, 36, 37, 41, 47, 49, 51, 53, 58, 67
db.data.frame, 20, 23, 38, 43, 46, 50, 53–56, 58–60, 62, 64, 65, 89, 94, 98, 107, 114, 118, 125, 127, 141, 142, 156, 158, 169
db.data.frame-class, 44
db.date.style (Type Cast functions), 166
db.default.schemas, 41, 42
db.default.schemas (db.search.path), 57
db.disconnect, 36, 37, 42, 46, 49, 67
db.existsObject, 43, 48, 51, 62
db.list, 21, 36, 37, 42, 47, 49, 52, 53, 67
db.obj, 15, 16, 18, 24, 25, 27, 28, 32, 34, 38, 39, 43, 46, 56, 59, 60, 64, 65, 68, 70, 72–74, 80, 82, 85–87, 91, 109, 111, 121, 125, 127, 135, 136, 138, 140, 156, 158, 166, 167, 169
db.obj-class, 50
db.objects, 43, 51, 53, 62
db.q, 52
db.Rcrossprod, 39, 142
db.Rcrossprod-class, 53
db.Rquery-class, 54
db.Rview-class (db.Rquery-class), 54
db.search.path, 41, 42, 57
db.table, 24, 38, 46, 60, 64, 127, 134, 139, 141, 169
db.table-class, 58
db.view, 24, 38, 46, 59, 64, 127, 141, 169
db.view-class, 59
dbms (Extract database connection info), 66
dbname (Extract database connection info), 66
delete, 55, 60, 89, 99, 108, 115, 119
delete, arima.css.madlib-method (delete), 60
delete, bagging.model-method (delete), 60
delete, character-method (delete), 60
delete, db.data.frame-method (delete), 60
delete, db.Rquery-method (delete), 60
delete, dt.madlib-method (delete), 60
delete, dt.madlib.grps-method (delete), 60
delete, elnet.madlib-method (delete), 60
delete, lm.madlib-method (delete), 60
delete, lm.madlib.grps-method (delete), 60
delete, logregr.madlib-method (delete), 60
delete, logregr.madlib.grps-method (delete), 60
delete, summary.madlib-method (delete), 60
delete, summary.madlib.grps-method (delete), 60
dim, db.Rquery-method (dim-methods), 63
dim, db.table-method (dim-methods), 63
dim, db.view-method (dim-methods), 63
dim-methods, 63
dspMatrix, 142
eql, 45
eql (eql-methods), 64
eql, db.obj, db.obj-method (eql-methods), 64
eql-methods, 64
exp (Func-methods), 70
exp, db.obj-method (Func-methods), 70
Extract database connection info, 66
Extract-Replace-methods, 68
extractAIC, 14, 154
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
extractAIC.glm.madlib (AIC), 13
extractAIC.lm.madlib (AIC), 13
extractAIC.logregr.madlib (AIC), 13
Extraction methods, 45
Extraction methods
(Extract-Replace-methods), 68
factorial (Func-methods), 70
factorial, db.obj-method (Func-methods), 70
formula, 96, 105, 106, 116, 118
Func-methods, 70
generic.bagging, 61, 72, 75, 135, 157
generic.cv, 72, 73, 95
getTree.rf.madlib, 76
grepl (Compare-methods), 32
grepl, character, db.obj-method
(Compare-methods), 32
groups, 77
groups.lm.madlib, 132
groups.lm.madlib.grps, 132
groups.logregr.madlib, 132
groups.logregr.madlib.grps, 132
GUI, 79
help, 48
host (Extract database connection info), 66
ifelse, 80
ifelse, db.obj-method (ifelse), 80
is.db.data.frame, 81
is.factor, db.obj-method
(is.factor-methods), 82
is.factor-methods, 82
is.na, db.obj-method (is.na-method), 83
is.na-method, 83
key, 45, 55, 58, 59, 84, 109, 111
key<- (key), 84
lk, 20, 23, 24, 27, 39, 46, 50, 54–56, 59, 60,
65, 83, 85, 131, 132, 134, 135, 138, 139, 146, 158, 160
lk (preview), 141
log (Func-methods), 70
log, db.obj-method (Func-methods), 70
log10 (Func-methods), 70
log10, db.obj-method (Func-methods), 70
Logical methods, 45
Logical methods (Logical-methods), 85
Logical-methods, 85
logLik, 14, 154
logLik.glm.madlib (AIC), 13
logLik.lm.madlib (AIC), 13
logLik.logregr.madlib (AIC), 13
lookat, 24, 27, 39, 46, 50, 54, 59, 60, 65, 83,
85, 146, 158, 160
lookat (preview), 141
madlib (Extract database connection info), 66
madlib.arima, 30, 61, 62, 77, 87, 99, 102,
108, 110, 112, 115, 119, 130, 134,
137, 140, 146–148, 151, 162, 163,
166
madlib.arima, db.Rquery, db.Rquery-method
(madlib.arima), 87
madlib.arima, formula, db.obj-method
(madlib.arima), 87
madlib.elnet, 61, 77, 90, 110, 112, 130, 137,
138, 140, 148, 149, 151, 163, 164,
166
madlib.glm, 5, 14, 25, 31, 45, 61, 62, 77, 78,
89, 96, 106, 108–112, 115, 120, 122,
125, 130, 132, 137, 140, 144, 148,
151, 153, 162, 166, 170
madlib.kmeans, 100, 102
madlib.lm, 103, 129, 139
madlib.lm, 5, 14, 25, 31, 45, 61, 62, 77, 78,
89, 97–99, 102, 105, 109–112, 115,
119, 120, 122, 125, 127, 130, 132,
137, 140, 148, 150, 151, 153, 162,
164, 166, 170
madlib.randomForest, 76, 77, 109, 139, 140,
151
madlib.rpart, 62, 77, 110, 111, 129, 130,
136, 137, 140, 147, 148, 151, 165,
166
madlib.summary, 5, 45, 61, 62, 77, 89, 99,
102, 108, 110, 112, 113, 119, 130,
137, 140, 148, 151–153, 166
madlib.svm, 116
margins, 120
Math Functions (Func-methods), 70
max, db.obj-method (Aggregate functions), 10
mean, db.obj-method (Aggregate functions), 10
merge, db.obj, db.obj-method
(merge-method), 123
sort, 159
sort, db.obj-method (sort), 159
sqrt (Func-methods), 70
sqrt, db.obj-method (Func-methods), 70
subset, db.obj-method (subset-methods),
    160
subset-methods, 160
sum, db.obj-method (Aggregate
    functions), 10
summary, 161
summary, db.obj-method (madlib.summary),
    113
summary.arima.css.madlib, 89
summary.arima.css.madlib
    (summary.arima.madlib), 162
summary.arima.madlib, 162
summary.elnet.madlib, 163
summary.lm.madlib, 164
tan, db.obj-method (Func-methods), 70
Terms (margins), 120
terms, 94, 98, 107, 118
text.dt.madlib, 111, 112, 130, 148, 165
Type Cast functions, 166
unique, db.obj-method (unique-methods),
    168
unique-methods, 168
user (Extract database connection
    info), 66
var (Aggregate functions), 10
var, db.obj-method (Aggregate
    functions), 10
Vars (margins), 120
vcov, 169
with (as.environment), 24