Package ‘COBRA’

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Type Package

Title Nonlinear Aggregation of Predictors

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Description This package performs prediction for regression-oriented problems, aggregating in a non-linear scheme any basic regression machines suggested by the context and provided by the user. If the user has no valuable knowledge on the data, four defaults machines wrappers are implemented so as to cover a minimal spectrum of prediction methods. If necessary, the computations may be parallelized. The method is described in Biau, Fischer, Guedj and Malley (2013), "COBRA: A Nonlinear Aggregation Strategy".

License GPL (>= 2)

URL http://www.lsta.upmc.fr/doct/guedj/index.html

Suggests snowfall, lars, ridge, tree, randomForest

NeedsCompilation yes

Repository CRAN

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Description

The function COBRA delivers prediction outcomes for a testing sample on the basis of a training sample and a bunch of basic regression machines. By default, those machines are wrappers to the R packages \texttt{lars}, \texttt{ridge}, \texttt{tree} and \texttt{randomForest}, covering a minimal spectrum in contemporary prediction methods for regression. However the most interesting way to use COBRA is to use any regression method suggested by the context (see argument \texttt{machines}). COBRA may natively parallelize the computations (use option \texttt{parallel}).

Details

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License: GPL (>= 2)

Author(s)

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References

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See Also

COBRA

Examples

\begin{verbatim}
n <- 500
d <- 30
ntrain <- 400
X <- replicate(d,2*runif(n = n)-1)
Y <- X[,1]^2 + X[,3]^3 + exp(X[,10]) + rnorm(n = n, sd = .1)
train.design <- as.matrix(X[1:ntrain,])
\end{verbatim}
train.responses <- y[1:ntrain]
test <- as.matrix(X[-(1:ntrain),])
test.responses <- y[-(1:ntrain)]

## using the default machines
if(require(lars) &
& require(tree) &
& require(ridge) &
& require(randomForest))
{
  res <- COBRA(train.design = train.design,
               train.responses = train.responses,
               test = test)

  print(cbind(res$predict, test.responses))
  plot(test.responses, res$predict, xlab="Responses", ylab="Predictions", pch=3, col=2)
  abline(0,1,lty=2)
}

## using own machines
machines.names <- c("Soothsayer", "Dummy")
machines <- matrix(nr = n, nc = 2, data = 0)
machines[,1] <- y+rnorm(n = n, sd=.1)  # soothsayer
machines[,2] <- mean(train.responses)  # dummy prediction, averaging train.responses

res2 <- COBRA(train.design = train.design,
               train.responses = train.responses,
               test = test,
               machines = machines,
               machines.names = machines.names)

print(cbind(res2$predict, test.responses))
plot(test.responses, res2$predict, xlab="Responses", ylab="Predictions", pch=3, col=2)
abline(0,1,lty=2)

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COBRA

Description

The function COBRA delivers prediction outcomes for a testing sample on the basis of a training sample and a bunch of basic regression machines. By default, those machines are wrappers to
the R packages *lars*, *ridge*, *tree* and *randomForest*, covering a somewhat wide spectrum in contemporary prediction methods for regression. However the most interesting way to use COBRA is to use any regression method suggested by the context (see argument `machines`). COBRA may natively parallelize the computations (use option `parallel`).

**Usage**

```r
cobra(train.design, train.responses, split, test, machines, machines.names, logGrid = FALSE, grid = 200, alpha.machines, parallel = FALSE, nb.cpus = 2, plots = FALSE, savePlots = FALSE, logs = FALSE, progress = TRUE, path = "")
```

**Arguments**

- `train.design`: Mandatory. The design matrix for the training sample.
- `train.responses`: Mandatory. The responses vector for the training sample.
- `split`: Optional. How should COBRA cut the training sample?
- `test`: Mandatory. The design matrix of the testing sample.
- `machines`: Optional. Regression basic machines provided by the user. This should be a matrix, whose number of rows is the length of the training sample (ntrain) plus the length of the testing sample (ntest), and with as many columns as machines. Element (i,j) of this matrix is assumed to be r_j(X_i), the (scalar) prediction of machine j for query point X_i, where i is from 1 to ntrain+ntest.
- `machines.names`: Optional. If `machines` is provided, a list including the names of the machines.
- `logGrid`: Optional. If TRUE, parameter epsilon is generated according to a logarithmic scale. This should be TRUE if the user has a clue about the small magnitude of predictions.
- `grid`: Optional. How many points should be used in the discretization scheme for calibrating the parameter epsilon.
- `alpha.machines`: Optional. Coerce COBRA to use exactly `alpha.machines`. Obviously this should be an integer between 1 and the total number of machines.
- `parallel`: Optional. If TRUE, computations will be dispatched over available cpus.
COBRA

nb.cpus

Optional. If parallel, how many cpus should be used. Obviously this should not exceed the number of available cpus!

plots

Optional. If TRUE, explanatory plots about calibrating epsilon and alpha (see publication) are generated according to the path variable.

savePlots

Optional. If TRUE, plots are saved as .pdf files according to path, otherwise they pop up in the R IDE.

logs

Optional. If TRUE, quadratic risks over the training sample for all machines and COBRA are written in the file "risks.txt" according to the path variable.

progress

Optional. If TRUE, a progress bar and final quadratic errors are printed.

path

Optional. If savePlots and either plots or logs are TRUE, where should the corresponding files be created?

Details

For most users, options grid and split should be set to their default values.

Value

Returns a list including only

predict

The vector of predicted values.

Note

Caution: If your data is ordered, you should shuffle the observations before calling COBRA since the algorithm assumes all data points are independent and identically distributed.

Author(s)

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References

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See Also

COBRA-package

Examples

```r
n <- 500
d <- 30
ntrain <- 400
X <- replicate(d,2*runif(n=n)-1)
Y <- X[,1]^2 + X[,3]^3 + exp(X[,10]) + rnorm(n = n, sd = .1)
train.design <- as.matrix(X[1:ntrain,])
```
train.responses <- Y[1:ntrain]
test <- as.matrix(X[-(1:ntrain),])
test.responses <- Y[-(1:ntrain)]

## using the default machines
if(require(lars) && require(tree) && require(ridge) && require(randomForest)) {
  res <- COBRA(train.design = train.design,
                train.responses = train.responses,
                test = test)
  print(cbind(res$predict, test.responses))
  plot(test.responses, res$predict, xlab="Responses", ylab="Predictions", pch=3, col=2)
  abline(0,1, lty=2)
}

## using own machines
machines.names <- c("Soothsayer", "Dummy")
machines <- matrix(nr = n, nc = 2, data = 0)
machines[,1] <- Y+rnorm(n = n, sd=.1)  ## soothsayer
machines[,2] <- mean(train.responses)  ## dummy prediction, averaging train.responses
res2 <- COBRA(train.design = train.design,
               train.responses = train.responses,
               test = test,
               machines = machines,
               machines.names = machines.names)
print(cbind(res2$predict, test.responses))
plot(test.responses, res2$predict, xlab="Responses", ylab="Predictions", pch=3, col=2)
abline(0,1, lty=2)
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