

Package ‘ICEbox’

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Description Implements Individual Conditional Expectation (ICE) plots, a tool for visualizing the model estimated by any supervised learning algorithm. ICE plots refine Friedman's partial dependence plot by graphing the functional relationship between the predicted response and a covariate of interest for individual observations. Specifically, ICE plots highlight the variation in the fitted values across the range of a covariate of interest, suggesting where and to what extent they may exist.

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URL <https://github.com/kapelner/ICEbox>

BugReports <https://github.com/kapelner/ICEbox/issues>

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additivityLineup	<i>Lineup plot for additivity</i>
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Description

This function creates a lineup plot to assess the additivity of a predictor's effect. It uses a nonparametric bootstrap approach to generate null plots.

Usage

```
additivityLineup(
  backfit_obj,
  fitMethod,
  realICE,
  figs = 10,
  colorvecfcn,
  usecolorvecfcn_inreal = FALSE,
  null_predictfcn,
  ...
)
```

Arguments

<code>backfit_obj</code>	An object of class <code>backfitter</code> .
<code>fitMethod</code>	A function that accepts <code>X</code> and <code>y</code> and returns a fitted model.
<code>realICE</code>	The <code>ice</code> object for the real data.
<code>figs</code>	The total number of plots in the lineup (including the real one). Default is 10.
<code>colorvecfcn</code>	Optional function to generate a color vector for the curves.
<code>usecolorvecfcn_inreal</code>	If <code>TRUE</code> , use <code>colorvecfcn</code> for the real plot.
<code>null_predictfcn</code>	Optional prediction function for the null models.
<code>...</code>	Additional arguments passed to <code>plot.ice</code> .

Value

An object of class `additivityLineup` (invisibly).

backfitter

*Backfitting for Additive Models***Description**

Fits a model of the form $\hat{f}(x) = \hat{g}_1(x_S) + \hat{g}_2(x_C)$ using backfitting.

Usage

```
backfitter(
  X,
  y,
  predictor,
  fitMethod,
  predictfcn,
  eps = 0.01,
  iter.max = 10,
  verbose = TRUE,
  ...
)
```

Arguments

<code>X</code>	The design matrix.
<code>y</code>	The response vector.
<code>predictor</code>	The name or index of the predictor of interest (x_S).
<code>fitMethod</code>	A function that accepts <code>X</code> and <code>y</code> and returns a fitted model.
<code>predictfcn</code>	A function that accepts <code>object</code> and <code>newdata</code> and returns predictions.

eps	Convergence threshold.
iter.max	Maximum number of iterations.
verbose	If TRUE, prints progress messages.
...	Additional arguments passed to <code>fitMethod</code> .

Value

An object of class `backfitter`.

clusterICE

Clustering of ICE and d-ICE curves by kmeans.

Description

Clustering of ICE and d-ICE curves by kmeans. All curves are centered to have mean 0 and then kmeans is applied to the curves with the specified number of clusters.

Usage

```
clusterICE(
  ice_obj,
  nClusters,
  plot = TRUE,
  plot_margin = 0.05,
  colorvec,
  plot_pdp = FALSE,
  x_quantile = FALSE,
  avg_lwd = 3,
  centered = FALSE,
  plot_legend = FALSE,
  main = NULL,
  num_cores = 1,
  ...
)
```

Arguments

ice_obj	Object of class <code>ice</code> or <code>dice</code> to cluster.
nClusters	Number of clusters to find.
plot	If TRUE, plots the clusters.
plot_margin	Extra margin to pass to <code>ylim</code> as a fraction of the range of cluster centers.
colorvec	Optional vector of colors to use for each cluster.
plot_pdp	If TRUE, the PDP (ice object) or d-PDP (dice object) is plotted with a dotted black line and highlighted in yellow.

x_quantile	If TRUE, the plot is drawn with the x-axis taken to be quantile(gridpts). If FALSE, the predictor's original scale is used.
avg_lwd	Average line width to use when plotting the cluster means. Line width is proportional to the cluster's size.
centered	If TRUE, all cluster means are shifted to be 0 at the minimum value of the predictor. If FALSE, the original cluster means are used.
plot_legend	If TRUE a legend mapping line colors to the proportion of the data in each cluster is added to the plot.
main	Optional title for the plot.
num_cores	Integer number of cores to use for parallel operations. Default is 1.
...	Additional arguments for plotting.

Value

A list with the following elements:

cl	The output of the kmeans call (a list of class kmeans).
plot	The ggplot object used for plotting (if plot = TRUE).

See Also

[ice](#), [dice](#)

Examples

```
## Not run:
require(ICEbox)
require(randomForest)
require(MASS) #has Boston Housing data, Pima

data(Boston) #Boston Housing data
X = Boston
y = X$medv
X$medv = NULL

## build a RF:
bh_rf = randomForest(X, y)

## Create an 'ice' object for the predictor "age":
bh.ice = ice(object = bh_rf, X = X, y = y, predictor = "age",
              frac_to_build = .1)

## cluster the curves into 2 groups.
clusterICE(bh.ice, nClusters = 2, plot_legend = TRUE)

## cluster the curves into 3 groups, start all at 0.
clusterICE(bh.ice, nClusters = 3, plot_legend = TRUE, center = TRUE)

## End(Not run)
```

colSds_cpp

Efficient Column Standard Deviations

Description

Efficient Column Standard Deviations

Usage

```
colSds_cpp(x, n_cores = 1L)
```

Arguments

x	Numeric Matrix
n_cores	Number of cores to use

derivative_cpp

Efficient Numerical Derivative for Matrix (Row-wise)

Description

Computes the first derivative using centered differences, mirroring sfsmisc::D1tr.

Usage

```
derivative_cpp(x, gridpts, n_cores = 1L)
```

Arguments

x	Numeric Matrix (smoothed values)
gridpts	Grid points corresponding to columns of x
n_cores	Number of cores to use

dice	<i>Creates an object of class dice.</i>
------	---

Description

Estimates the partial derivative function for each curve in an `ice` object. See Goldstein et al (2013) for further details.

Usage

```
dice(
  ice_obj,
  DerivEstimator = NULL,
  use_supsmu = FALSE,
  verbose = TRUE,
  num_cores = 1,
  sg_poly_order = 2,
  sg_window_size = NULL
)
```

Arguments

<code>ice_obj</code>	Object of class <code>ice</code> . This function generates partial derivative estimates for each row in <code>ice_obj\$ice_curves</code> .
<code>DerivEstimator</code>	Optional function with a single argument <code>y</code> . Returns the estimated partial derivative of a function sampled at the points (<code>ice_obj\$gridpts,y</code>). If <code>NULL</code> , the default uses a Savitzky-Golay filter to estimate the first derivative.
<code>use_supsmu</code>	If <code>TRUE</code> , uses the old <code>supsmu</code> based derivative estimation logic. This is much slower than the default Savitzky-Golay filter.
<code>verbose</code>	If <code>TRUE</code> , prints messages about the procedure's progress.
<code>num_cores</code>	Integer number of cores to use for parallel derivative estimation. Defaults to 1.
<code>sg_poly_order</code>	Polynomial order for Savitzky-Golay filter. Default is 2.
<code>sg_window_size</code>	Window size for Savitzky-Golay filter. Default is 30% of the grid.

Value

A list of class `dice` with the following elements. Most are passed directly through from `ice_object` and exist to enable various plotting facilities.

<code>d_ice_curves</code>	Matrix of dimension <code>nrow(Xice)</code> by <code>length(gridpts)</code> . Each row corresponds to an observation's d-ICE curve, estimated at the values of <code>predictor</code> in <code>gridpts</code> .
<code>xj</code>	The actual values of <code>predictor</code> observed in the data in the order of <code>Xice</code> .
<code>actual_deriv</code>	Vector of length <code>nrow(Xice)</code> containing the estimated partial derivatives at the value of the predictor actually found in <code>Xice</code> .

sd_deriv	Vector of length <code>length(gridpts)</code> with the cross-observation sd of partial derivative estimates. For instance <code>sd_deriv[1]</code> equals <code>sd(d_ice_curves[,1])</code> .
logodds	Passed from <code>ice_object</code> . If TRUE, <code>d_ice_curves</code> are estimated derivatives of the centered log-odds.
gridpts	Passed from <code>ice_object</code> .
predictor	Passed from <code>ice_object</code> .
xlab	Passed from <code>ice_object</code> .
nominal_axis	Passed from <code>ice_object</code> .
range_y	Passed from <code>ice_object</code> .
Xice	Passed from <code>ice_object</code> .
dpdp	The estimated partial derivative of the PDP.

References

Goldstein, A., Kapelner, A., Bleich, J., and Pitkin, E., Peeking Inside the Black Box: Visualizing Statistical Learning With Plots of Individual Conditional Expectation. (2014) Journal of Computational and Graphical Statistics, in press

See Also

[ice](#), [dice](#)

Examples

```
## Not run:
# same examples as for 'ice', but now create a derivative estimate as well.
require(ICEbox)
require(randomForest)
require(MASS) #has Boston Housing data, Pima

##### regression example
data(Boston) #Boston Housing data
X = Boston
y = X$medv
X$medv = NULL

## build a RF:
bhd_rf_mod = randomForest(X, y)

## Create an 'ice' object for the predictor "age":
bhd.ice = ice(object = bhd_rf_mod, X = X, y = y, predictor = "age", frac_to_build = .1)

# make a dice object:
bhd.dice = dice(bhd.ice)

#### classification example
data(Pima.te) #Pima Indians diabetes classification
y = Pima.te$type
```

```

X = Pima.te
X$type = NULL

## build a RF:
pima_rf = randomForest(x = X, y = y)

## Create an 'ice' object for the predictor "skin":
# For classification we plot the centered log-odds. If we pass a predict
# function that returns fitted probabilities, setting logodds = TRUE instructs
# the function to set each ice curve to the centered log-odds of the fitted
# probability.
pima.ice = ice(object = pima_rf, X = X, predictor = "skin", logodds = TRUE,
                predictfcn = function(object, newdata){
                  predict(object, newdata, type = "prob")[, 2]
                }
)
# make a dice object:
pima.dice = dice(pima.ice)

## End(Not run)

```

ice *Creates an object of class ice.*

Description

Creates an ice object with individual conditional expectation curves for the passed model object, X matrix, predictor, and response. See Goldstein et al (2013) for further details.

Usage

```

ice(
  object,
  X,
  y,
  predictor,
  predictfcn,
  verbose = TRUE,
  frac_to_build = 1,
  indices_to_build = NULL,
  num_grid_pts,
  logodds = FALSE,
  probit = FALSE,
  num_cores = 1,
  ...
)

```

Arguments

object	The fitted model to estimate ICE curves for.
X	The design matrix we wish to estimate ICE curves for. Rows are observations, columns are predictors. Typically this is taken to be object's training data, but this is not strictly necessary.
y	Optional vector of the response values object was trained on. It is used to compute y-axis ranges that are useful for plotting. If not passed, the range of predicted values is used and a warning is printed.
predictor	The column number or variable name in X of the predictor of interest, ($x_S = X[, j]$).
predictFcn	Optional function that accepts two arguments, object and newdata, and returns an N vector of object's predicted response for data newdata. If this argument is not passed, the procedure attempts to find a generic predict function corresponding to <code>class(object)</code> .
verbose	If TRUE, prints messages about the procedure's progress.
frac_to_build	Number between 0 and 1, with 1 as default. For large X matrices or fitted models that are slow to make predictions, specifying <code>frac_to_build</code> less than 1 will choose a subset of the observations to build curves for. The subset is chosen such that the remaining observations' values of predictor are evenly spaced throughout the quantiles of the full <code>X[, predictor]</code> vector.
indices_to_build	Vector of indices, $\subset \{1, \dots, nrow(X)\}$ specifying which observations to build ICE curves for. As this is an alternative to setting <code>frac_to_build</code> , both cannot be specified.
num_grid_pts	Optional number of values in the range of predictor at which to estimate each curve. If missing, the curves are estimated at each unique value of predictor in the X observations we estimate ICE curves for.
logodds	If TRUE, for classification creates PDPs by plotting the centered log-odds implied by the fitted probabilities. We assume that the generic or passed predict function returns probabilities, and so the flag tells us to transform these to centered logits after the predictions are generated. Note: <code>probit</code> cannot be TRUE.
probit	If TRUE, for classification creates PDPs by plotting the probit implied by the fitted probabilities. We assume that the generic or passed predict function returns probabilities, and so the flag tells us to transform these to probits after the predictions are generated. Note: <code>logodds</code> cannot be TRUE.
num_cores	Integer number of cores to use for parallel prediction. Defaults to 1.
...	Other arguments to be passed to object's generic predict function.

Value

A list of class `ice` with the following elements:

gridpts	Sorted values of predictor at which each curve is estimated. Duplicates are removed – by definition, elements of <code>gridpts</code> are unique.
---------	---

ice_curves	Matrix of dimension nrow(X) by length(gridpts). Each row corresponds to an observation's ICE curve, estimated at the values of predictor in gridpts.
xj	The actual values of predictor observed in the data in the order of Xice.
actual_prediction	Vector of length nrow(X) containing the model's predictions at the actual value of the predictors in the order of Xice.
xlab	String with the predictor name corresponding to predictor. If predictor is a column number, xlab is set to colnames(X)[, predictor].
nominal_axis	If TRUE, length(gridpts) is 5 or fewer; otherwise FALSE. When TRUE the plot function treats the x-axis as if x is nominal.
range_y	If y was passed, the range of the response. Otherwise it defaults to be max(ice_curves) - min(ice_curves) and a message is printed to the console.
sd_y	If y was passed, the standard deviation of the response. Otherwise it is defaults to sd(actual_prediction) and a message is printed to the console.
Xice	A matrix containing the subset of X for which ICE curves are estimated. Observations are ordered to be increasing in predictor. This ordering is the same one as in ice_curves, xj and actual_prediction, meaning for all these objects the i-th element refers to the same observation in X.
pdp	A vector of size length(gridpts) which is a numerical approximation to the partial dependence function (PDP) corresponding to the estimated ICE curves. See Goldstein et al (2013) for a discussion of how the PDP is a form of post-processing. See Friedman (2001) for a description of PDPs.
predictor	Same as the argument, see argument description.
logodds	Same as the argument, see argument description.
indices_to_build	Same as the argument, see argument description.
frac_to_build	Same as the argument, see argument description.
predictfcn	Same as the argument, see argument description.

References

Jerome Friedman. Greedy Function Approximation: A Gradient Boosting Machine. *The Annals of Statistics*, 29(5): 1189-1232, 2001.

Goldstein, A., Kapelner, A., Bleich, J., and Pitkin, E., Peeking Inside the Black Box: Visualizing Statistical Learning With Plots of Individual Conditional Expectation. (2014) *Journal of Computational and Graphical Statistics*, in press

See Also

[plot.ice](#), [print.ice](#), [summary.ice](#)

Examples

```

## Not run:
require(ICEbox)
require(randomForest)
require(MASS) #has Boston Housing data, Pima

##### regression example
data(Boston) #Boston Housing data
X = Boston
y = X$medv
X$medv = NULL

## build a RF:
bhd_rf_mod = randomForest(X, y)

## Create an 'ice' object for the predictor "age":
bhd.ice = ice(object = bhd_rf_mod, X = X, y = y, predictor = "age", frac_to_build = .1)

## End(Not run)

```

melt_ice_curves_cpp *Melt Matrix to Long Format Vector*

Description

Efficiently converts a matrix to a long-format vector (row-major order) for plotting.

Usage

```
melt_ice_curves_cpp(x, n_cores = 1L)
```

Arguments

x	Numeric Matrix
n_cores	Number of cores to use

plot.dice *Create a plot of a dice object.*

Description

Plotting of dice objects.

Usage

```
## S3 method for class 'dice'
plot(
  x,
  plot_margin = 0.05,
  frac_to_plot = 1,
  plot_sd = TRUE,
  plot_orig_pts_deriv = TRUE,
  pts_preds_size = 1.5,
  colorvec,
  color_by = NULL,
  x_quantile = TRUE,
  plot_dpd = TRUE,
  rug_quantile = seq(from = 0, to = 1, by = 0.1),
  verbose = TRUE,
  ...
)
```

Arguments

<code>x</code>	Object of class <code>dice</code> to plot.
<code>plot_margin</code>	Extra margin to pass to <code>ylim</code> as a fraction of the range of <code>x\$d_ice_curves</code> .
<code>frac_to_plot</code>	If <code>frac_to_plot</code> is less than 1, randomly plot <code>frac_to_plot</code> fraction of the curves in <code>x\$d_ice_curves</code> .
<code>plot_sd</code>	If TRUE, plot the cross-observation sd of partial derivatives below the derivative plots.
<code>plot_orig_pts_deriv</code>	If TRUE, marks each curve at the location of the derivative estimate at the location of predictor actually occurring in the data. If FALSE no mark is drawn.
<code>pts_preds_size</code>	Size of points to make if <code>plot_orig_pts_deriv</code> is TRUE.
<code>colorvec</code>	Optional vector of colors to use for each curve.
<code>color_by</code>	Optional variable name (or column number) in <code>Xice</code> to color curves by. If the <code>color_by</code> variable has 10 or fewer unique values, a discrete set of colors is used for each value and a legend is printed and returned. If there are more values, curves are colored from light to dark corresponding to low to high values of the variable specified by <code>color_by</code> .
<code>x_quantile</code>	If TRUE, the plot is drawn with the x-axis taken to be <code>quantile(gridpts)</code> . If FALSE, the predictor's original scale is used.
<code>plot_dpd</code>	If TRUE, the estimated derivative of the PDP is plotted and highlighted in yellow.
<code>rug_quantile</code>	If not null, tick marks are drawn on the x-axis corresponding to the vector of quantiles specified by this parameter. Forced to NULL when <code>x_quantile</code> is set to TRUE.
<code>verbose</code>	If TRUE, prints the color legend to the console.
<code>...</code>	Additional plotting arguments.

Value

A list with the following elements.

<code>plot_points_indices</code>	Row numbers of <code>Xice</code> of those observations presented in the plot.
<code>legend_text</code>	If the <code>color_by</code> argument was used, a legend describing the map between the <code>color_by</code> predictor and curve colors.
<code>plot</code>	The <code>ggplot</code> object used for plotting.

See Also

[dice](#)

Examples

```
## Not run:
require(ICEbox)
require(randomForest)
require(MASS) #has Boston Housing data, Pima

data(Boston) #Boston Housing data
X = Boston
y = X$medv
X$medv = NULL

## build a RF:
bhd_rf_mod = randomForest(X, y)

## Create an 'ice' object for the predictor "age":
bhd.ice = ice(object = bhd_rf_mod, X = X, y = y, predictor = "age", frac_to_build = .1)

# estimate derivatives, then plot.
bhd.dice = dice(bhd.ice)
plot(bhd.dice)

## End(Not run)
```

`plot.ice`

Plotting of ice objects.

Description

Plotting of ice objects.

Usage

```
## S3 method for class 'ice'
plot(
  x,
  plot_margin = 0.05,
  frac_to_plot = 1,
  plot_points_indices = NULL,
  plot_orig_pts_preds = TRUE,
  pts_preds_size = 1.5,
  colorvec,
  color_by = NULL,
  x_quantile = TRUE,
  plot_pdp = TRUE,
  centered = FALSE,
  prop_range_y = TRUE,
  rug_quantile = seq(from = 0, to = 1, by = 0.1),
  centered_percentile = 0,
  point_labels = NULL,
  point_labels_size = NULL,
  prop_type = "sd",
  verbose = TRUE,
  num_cores = 1,
  ...
)
```

Arguments

- x** Object of class ice to plot.
- plot_margin** Extra margin to pass to ylim as a fraction of the range of x\$ice_curves.
- frac_to_plot** If frac_to_plot is less than 1, randomly plot frac_to_plot fraction of the curves in x\$ice_curves.
- plot_points_indices** If not NULL, this plots only the indices of interest. If not NULL, frac_to_plot must be 1 otherwise an error is thrown. Default is NULL.
- plot_orig_pts_preds** If TRUE, marks each curve at the location of the observation's actual fitted value. If FALSE, no mark is drawn.
- pts_preds_size** Size of points to make if plot_origin_pts_preds is TRUE.
- colorvec** Optional vector of colors to use for each curve.
- color_by** Optional variable name in Xice, column number in Xice, or data vector of the correct length to color curves by. If the color_by variable has 10 or fewer unique values, a discrete set of colors is used for each value and a legend is printed and returned. If there are more values, curves are colored from light to dark corresponding to low to high values of the variable specified by color_by.
- x_quantile** If TRUE, the plot is drawn with the x-axis taken to be quantile(gridpts). If FALSE, the predictor's original scale is used.

<code>plot_pdp</code>	If TRUE, the PDP is plotted and highlighted in yellow.
<code>centered</code>	If TRUE, all curves are re-centered to be 0 at the quantile given by <code>centered_percentile</code> . See Goldstein et al (2013) for details and examples. If FALSE, the original <code>ice_curves</code> are plotted.
<code>prop_range_y</code>	When TRUE and <code>centered</code> =TRUE as well, the range of the right vertical axis displays the centered values as a fraction of the sd of the fitted values on actual observations if <code>prop_type</code> is missing or set to "sd". If <code>prop_type</code> is set to "range", the right axis displays the centered values as a fraction of the range of the fitted values over the actual observations.
<code>rug_quantile</code>	If not NULL, tick marks are drawn on the x-axis corresponding to the vector of quantiles specified by this parameter. Forced to NULL when <code>x_quantile</code> is set to TRUE.
<code>centered_percentile</code>	The percentile of predictor for which all <code>ice_curves</code> are "pinched together" and set to be 0. Default is 0.
<code>point_labels</code>	If not NULL, labels to plot next to each point. Default is NULL.
<code>point_labels_size</code>	If not NULL, size of labels to plot next to each point. Default is NULL which means it's the size of <code>pts_preds_size</code> .
<code>prop_type</code>	Scaling factor for the right vertical axis in centered plots if <code>prop_range_y</code> is TRUE. Can be one of "sd" (default) or "range". Ignored if <code>centered</code> and <code>prop_range_y</code> are not both TRUE.
<code>verbose</code>	If TRUE, prints the color legend to the console.
<code>num_cores</code>	Used for parallel plotting speedup. Default is 1.
...	Other arguments to be passed to the <code>plot</code> function.

Value

A list with the following elements.

`plot_points_indices`

Row numbers of `Xice` of those observations presented in the plot.

`legend_text`

If the `color_by` argument was used, a legend describing the map between the `color_by` predictor and curve colors.

See Also

[ice](#)

Examples

```
## Not run:
require(ICEbox)
require(randomForest)
require(MASS) #has Boston Housing data, Pima
data(Boston) #Boston Housing data
```

```

X = Boston
y = X$medv
X$medv = NULL

## build a RF:
bhd_rf_mod = randomForest(X, y)

## Create an 'ice' object for the predictor "age":
bhd.ice = ice(object = bhd_rf_mod, X = X, y = y, predictor = "age",
               frac_to_build = .1)

## plot
plot(bhd.ice, x_quantile = TRUE, plot_pdp = TRUE, frac_to_plot = 1)

## centered plot
plot(bhd.ice, x_quantile = TRUE, plot_pdp = TRUE, frac_to_plot = 1,
      centered = TRUE)

## color the curves by high and low values of 'rm'.
# First create an indicator variable which is 1 if the number of
# rooms is greater than the median:
median_rm = median(X$rm)
bhd.ice$Xice$I_rm = ifelse(bhd.ice$Xice$rm > median_rm, 1, 0)

plot(bhd.ice, frac_to_plot = 1, centered = TRUE, prop_range_y = TRUE,
      x_quantile = T, plot_orig_pts_preds = T, color_by = "I_rm")
bhd.ice = ice(object = bhd_rf_mod, X = X, y = y, predictor = "age",
               frac_to_build = 1)
plot(bhd.ice, frac_to_plot = 1, centered = TRUE, prop_range_y = TRUE,
      x_quantile = T, plot_orig_pts_preds = T, color_by = y)

## End(Not run)

```

print.dice

Print method for dice objects.

Description

Prints a summary of a dice object.

Usage

```

## S3 method for class 'dice'
print(x, ...)

```

Arguments

x	Object of class dice.
...	Ignored for now.

print.ice	<i>Print method for ice objects.</i>
-----------	--------------------------------------

Description

Prints a summary of an ice object.

Usage

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

Arguments

x	Object of class ice.
...	Ignored for now.

rowCenter_cpp	<i>Row-wise Centering</i>
---------------	---------------------------

Description

Centers each row of a matrix by subtracting the row mean.

Usage

```
rowCenter_cpp(x, n_cores = 1L)
```

Arguments

x	Numeric Matrix
n_cores	Number of cores to use

`sg_smooth_cpp`*Savitzky-Golay Filter for Matrix (Row-wise)*

Description

Smooths each row of a matrix using a Savitzky-Golay filter.

Usage

```
sg_smooth_cpp(x, window_size, order, deriv, n_cores = 1L)
```

Arguments

<code>x</code>	Matrix to smooth row-wise
<code>window_size</code>	Size of the filter window (must be odd)
<code>order</code>	Polynomial order
<code>deriv</code>	Derivative order (0=smooth, 1=first deriv, etc.)
<code>n_cores</code>	Number of cores to use

`summary.dice`*Summary function for dice objects.*

Description

Alias of `print` method.

Usage

```
## S3 method for class 'dice'  
summary(object, ...)
```

Arguments

<code>object</code>	Object of class <code>dice</code> .
<code>...</code>	Ignored for now.

summary.ice*Summary function for ice objects.*

Description

Alias of `print` method.

Usage

```
## S3 method for class 'ice'  
summary(object, ...)
```

Arguments

object	Object of class <code>ice</code> .
...	Ignored for now.

transform_ice_curves_cpp*Probability Transformation*

Description

Efficiently applies logodds or probit transformation to a matrix.

Usage

```
transform_ice_curves_cpp(x, method, n_cores = 1L)
```

Arguments

x	Numeric Matrix (probabilities)
method	1 for centered logodds, 2 for probit
n_cores	Number of cores to use

WhiteWine

Data concerning white wine.

Description

The WhiteWine data frame has 4898 rows and 12 columns and concerns white wines from a region in Portugal. The response variable, quality, is a wine quality metric, taken to be the median preference score of three blind tasters on a scale of 1-10. The 11 covariates are physicochemical metrics of wine quality such as citric acid content, sulphates, etc.

Usage

```
data(WhiteWine)
```

Format

A data frame of 4898 cases on 12 variables.

Source

K Bache and M Lichman. UCI machine learning repository, 2013. <http://archive.ics.uci.edu/ml>

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