

# Package ‘mdatools’

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**Author** Sergey Kucheryavskiy (<<https://orcid.org/0000-0002-3145-7244>>)

**Maintainer** Sergey Kucheryavskiy <svkucheryavski@gmail.com>

**Description** Projection based methods for preprocessing,  
exploring and analysis of multivariate data used in chemometrics.  
S. Kucheryavskiy (2020) <[doi:10.1016/j.chemolab.2020.103937](https://doi.org/10.1016/j.chemolab.2020.103937)>.

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**Imports** methods, graphics, grDevices, stats, Matrix

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## R topics documented:

as.matrix.classres . . . . .	10
as.matrix.ldecomp . . . . .	10
as.matrix.plsdares . . . . .	11
as.matrix.plsres . . . . .	11
as.matrix.regcoeffs . . . . .	12
as.matrix.regres . . . . .	12
as.matrix.simcamres . . . . .	13
as.matrix.simcares . . . . .	13
capitalize . . . . .	14
carbs . . . . .	14
categorize . . . . .	15

categorize.pca . . . . .	15
categorize.pls . . . . .	16
chisq.crit . . . . .	17
chisq.prob . . . . .	17
classify.plsda . . . . .	18
classify.simca . . . . .	18
classmodel.processRefValues . . . . .	19
classres . . . . .	19
classres.getPerformance . . . . .	20
confint.regcoeffs . . . . .	21
constraint . . . . .	22
constraintAngle . . . . .	22
constraintClosure . . . . .	23
constraintNonNegativity . . . . .	23
constraintNorm . . . . .	24
constraints.list . . . . .	24
constraintUnimod . . . . .	24
crossval . . . . .	25
crossval.getParams . . . . .	25
crossval.regmodel . . . . .	26
crossval.simca . . . . .	26
crossval.str . . . . .	27
dd.crit . . . . .	27
ddmoments.param . . . . .	28
ddrobust.param . . . . .	28
ellipse . . . . .	29
employ.constraint . . . . .	29
employ.prep . . . . .	30
eye . . . . .	30
fprintf . . . . .	31
getCalibrationData . . . . .	31
getCalibrationData.pca . . . . .	32
getCalibrationData.simcam . . . . .	32
getConfidenceEllipse . . . . .	33
getConfusionMatrix . . . . .	33
getConfusionMatrix.classres . . . . .	34
getConvexHull . . . . .	34
getDataLabels . . . . .	35
getImplementedConstraints . . . . .	35
getImplementedPrepMethods . . . . .	35
getLabelsAsIndices . . . . .	36
getLabelsAsValues . . . . .	36
getMainTitle . . . . .	36
getPlotColors . . . . .	37
getProbabilities . . . . .	37
getProbabilities.pca . . . . .	38
getProbabilities.simca . . . . .	38
getPureVariables . . . . .	39

getR	39
getRegcoeffs	40
getRegcoeffs.regmodel	40
getRes	41
getSelectedComponents	42
getSelectivityRatio	42
getSelectivityRatio.pls	43
getVariance.mcr	43
getVIPScores	44
getVIPScores.pls	44
hotelling.crit	45
hotelling.prob	45
imshow	46
ipls	46
ipls.backward	49
ipls.forward	49
jm.crit	50
jm.prob	50
ldecomp	51
ldecomp.getDistances	52
ldecomp.getLimitsCoordinates	52
ldecomp.getLimParams	53
ldecomp.getQLimits	53
ldecomp.getT2Limits	54
ldecomp.getVariances	54
ldecomp.plotResiduals	55
mcr	56
mcrals	57
mcrals.cal	61
mcrals.fcnnls	63
mcrals.nnls	63
mcrals.ols	64
mcrpure	65
mda.cbind	68
mda.data2im	68
mda.df2mat	69
mda.exclcols	69
mda.exclrows	70
mda.getattr	70
mda.getexclind	71
mda.im2data	71
mda.inclcols	72
mda.inclrows	72
mda.purge	73
mda.purgeCols	73
mda.purgeRows	73
mda.rbind	74
mda.setattr	74

mda.setimbg . . . . .	75
mda.show . . . . .	75
mda.subset . . . . .	76
mda.t . . . . .	76
mdaplot . . . . .	77
mdaplot.areColors . . . . .	80
mdaplot.formatValues . . . . .	80
mdaplot.getColors . . . . .	81
mdaplot.getXAxisLim . . . . .	82
mdaplot.getXTickLabels . . . . .	82
mdaplot.getXTicks . . . . .	83
mdaplot.getYAxisLim . . . . .	83
mdaplot.getYTickLabels . . . . .	84
mdaplot.getYTicks . . . . .	84
mdaplot.plotAxes . . . . .	85
mdaplot.prepareColors . . . . .	86
mdaplot.showColorbar . . . . .	86
mdaplot.showLines . . . . .	87
mdaplotg . . . . .	87
mdaplotg.getLegend . . . . .	90
mdaplotg.getXLim . . . . .	91
mdaplotg.getYLim . . . . .	91
mdaplotg.prepareData . . . . .	92
mdaplotg.processParam . . . . .	93
mdaplotg.showLegend . . . . .	93
mdaplotyy . . . . .	94
mdatools . . . . .	96
pca . . . . .	97
pca.cal . . . . .	101
pca.getB . . . . .	102
pca.mvreplace . . . . .	102
pca.nipals . . . . .	104
pca.run . . . . .	104
pca.svd . . . . .	105
pcares . . . . .	105
pcv . . . . .	107
pellets . . . . .	108
people . . . . .	109
pinv . . . . .	109
plot.classres . . . . .	110
plot.ipls . . . . .	110
plot.mcr . . . . .	111
plot.pca . . . . .	111
plot.pcares . . . . .	112
plot.pls . . . . .	112
plot.plsda . . . . .	113
plot.plsdares . . . . .	113
plot.plsres . . . . .	114

plot.randtest . . . . .	115
plot.regcoeffs . . . . .	115
plot.regres . . . . .	116
plot.simca . . . . .	117
plot.simcam . . . . .	117
plot.simcamres . . . . .	118
plotBars . . . . .	118
plotBiplot . . . . .	119
plotBiplot.pca . . . . .	119
plotConfidenceEllipse . . . . .	120
plotContributions . . . . .	121
plotContributions.mcr . . . . .	121
plotConvexHull . . . . .	122
plotCooman . . . . .	123
plotCooman.simcam . . . . .	123
plotCooman.simcamres . . . . .	124
plotCorr . . . . .	125
plotCorr.randtest . . . . .	126
plotCumVariance . . . . .	126
plotCumVariance.ldecomp . . . . .	127
plotCumVariance.mcr . . . . .	127
plotCumVariance.pca . . . . .	128
plotDensity . . . . .	128
plotDiscriminationPower . . . . .	129
plotDiscriminationPower.simcam . . . . .	129
plotDistDoF . . . . .	130
plotErrorbars . . . . .	131
plotExtreme . . . . .	131
plotExtreme.pca . . . . .	132
plotHist . . . . .	133
plotHist.randtest . . . . .	133
plotHotellingEllipse . . . . .	134
plotLines . . . . .	135
plotLoadings . . . . .	136
plotLoadings.pca . . . . .	136
plotMisclassified . . . . .	137
plotMisclassified.classmodel . . . . .	138
plotMisclassified.classres . . . . .	138
plotModelDistance . . . . .	139
plotModelDistance.simcam . . . . .	139
plotModellingPower . . . . .	141
plotPerformance . . . . .	141
plotPerformance.classmodel . . . . .	142
plotPerformance.classres . . . . .	143
plotPointsShape . . . . .	144
plotPredictions . . . . .	144
plotPredictions.classmodel . . . . .	145
plotPredictions.classres . . . . .	145

plotPredictions.regmodel . . . . .	146
plotPredictions.regres . . . . .	147
plotPredictions.simcam . . . . .	148
plotPredictions.simcamres . . . . .	149
plotProbabilities . . . . .	149
plotProbabilities.classres . . . . .	150
plotPurity . . . . .	150
plotPurity.mcrpure . . . . .	151
plotPuritySpectra . . . . .	151
plotPuritySpectra.mcrpure . . . . .	152
plotQDoF . . . . .	153
plotRegcoeffs . . . . .	153
plotRegcoeffs.regmodel . . . . .	154
plotRegressionLine . . . . .	154
plotResiduals . . . . .	155
plotResiduals.ldecomp . . . . .	155
plotResiduals.pca . . . . .	156
plotResiduals.regres . . . . .	157
plotRMSE . . . . .	158
plotRMSE.ipls . . . . .	159
plotRMSE.regmodel . . . . .	160
plotRMSE.regres . . . . .	160
plotScatter . . . . .	161
plotScores . . . . .	162
plotScores.ldecomp . . . . .	163
plotScores.pca . . . . .	163
plotSelection . . . . .	164
plotSelection.ipls . . . . .	165
plotSelectivityRatio . . . . .	166
plotSelectivityRatio.pls . . . . .	166
plotSensitivity . . . . .	167
plotSensitivity.classmodel . . . . .	167
plotSensitivity.classres . . . . .	168
plotseries . . . . .	168
plotSpecificity . . . . .	169
plotSpecificity.classmodel . . . . .	170
plotSpecificity.classres . . . . .	170
plotSpectra . . . . .	171
plotSpectra.mcr . . . . .	171
plotT2DoF . . . . .	172
plotVariance . . . . .	172
plotVariance.ldecomp . . . . .	173
plotVariance.mcr . . . . .	174
plotVariance.pca . . . . .	174
plotVariance.pls . . . . .	175
plotVariance.plsres . . . . .	176
plotVIPScores . . . . .	177
plotVIPScores.pls . . . . .	177

plotWeights	178
plotWeights.pls	178
plotXCumVariance	179
plotXCumVariance.pls	179
plotXCumVariance.plsres	180
plotXLoadings	180
plotXLoadings.pls	181
plotXResiduals	181
plotXResiduals.pls	182
plotXResiduals.plsres	183
plotXScores	184
plotXScores.pls	184
plotXScores.plsres	185
plotXVariance	186
plotXVariance.pls	186
plotXVariance.plsres	187
plotXYLoadings	187
plotXYLoadings.pls	188
plotXYResiduals	188
plotXYResiduals.pls	189
plotXYResiduals.plsres	190
plotXYScores	191
plotXYScores.pls	191
plotXYScores.plsres	192
plotYCumVariance	193
plotYCumVariance.pls	193
plotYCumVariance.plsres	194
plotYResiduals	194
plotYResiduals.plsres	195
plotYResiduals.regmodel	195
plotYVariance	196
plotYVariance.pls	196
plotYVariance.plsres	197
pls	198
pls.cal	203
pls.getLimitsCoordinates	204
pls.getZLimits	204
pls.run	205
pls.simpls	205
plsda	206
plsdares	209
plsres	212
predict.mcrals	215
predict.mcrpure	215
predict.pca	216
predict.pls	216
predict.plsda	217
predict.simca	218

predict.simcam . . . . .	218
prep . . . . .	219
prep.alsbasecorr . . . . .	220
prep.autoscale . . . . .	221
prep.generic . . . . .	221
prep.list . . . . .	222
prep.msc . . . . .	222
prep.norm . . . . .	223
prep.ref2km . . . . .	223
prep.savgol . . . . .	224
prep.snv . . . . .	225
prep.transform . . . . .	225
prep.varsel . . . . .	226
preparePlotData . . . . .	227
prepCalData . . . . .	227
print.classres . . . . .	228
print.ipls . . . . .	228
print.ldecomp . . . . .	229
print.mcrals . . . . .	229
print.mcrpure . . . . .	230
print.pca . . . . .	230
print.pcares . . . . .	231
print.pls . . . . .	231
print.plsda . . . . .	232
print.plsdares . . . . .	232
print.plsres . . . . .	233
print.randtest . . . . .	233
print.regcoeffs . . . . .	234
print.regmodel . . . . .	234
print.regres . . . . .	235
print.simca . . . . .	235
print.simcam . . . . .	236
print.simcamres . . . . .	236
print.simcares . . . . .	237
randtest . . . . .	237
regcoeffs . . . . .	239
regcoeffs.getStats . . . . .	240
regres . . . . .	241
regres.bias . . . . .	241
regres.err . . . . .	242
regres.r2 . . . . .	242
regres.rmse . . . . .	243
regres.slope . . . . .	243
regress.addattrs . . . . .	243
repmat . . . . .	244
rotationMatrixToX1 . . . . .	244
selectCompNum . . . . .	245
selectCompNum.pca . . . . .	245



selectCompNum.pls . . . . .	246
selratio . . . . .	247
setDistanceLimits . . . . .	247
setDistanceLimits.pca . . . . .	248
setDistanceLimits.pls . . . . .	249
showDistanceLimits . . . . .	250
showLabels . . . . .	250
showPredictions . . . . .	251
showPredictions.classres . . . . .	251
simca . . . . .	252
simcam . . . . .	255
simcam.getPerformanceStats . . . . .	257
simcamres . . . . .	257
simcares . . . . .	259
simdata . . . . .	261
splitExcludedData . . . . .	261
splitPlotData . . . . .	262
summary.classres . . . . .	262
summary.ipls . . . . .	263
summary.ldecomp . . . . .	263
summary.mcrals . . . . .	264
summary.mcrpure . . . . .	264
summary.pca . . . . .	265
summary.pcares . . . . .	265
summary.pls . . . . .	266
summary.plsda . . . . .	266
summary.plsdares . . . . .	267
summary.plsres . . . . .	267
summary.randtest . . . . .	268
summary.regcoeffs . . . . .	268
summary.regmodel . . . . .	269
summary.regres . . . . .	269
summary.simca . . . . .	270
summary.simcam . . . . .	270
summary.simcamres . . . . .	271
summary.simcares . . . . .	271
unmix.mcrpure . . . . .	272
vipscores . . . . .	272

---

as.matrix.classres      *as.matrix method for classification results*

---

### Description

Generic as.matrix function for classification results. Returns matrix with performance values for specific class.

### Usage

```
## S3 method for class 'classres'
as.matrix(x, ncomp = NULL, nc = 1, ...)
```

### Arguments

x	classification results (object of class plsdares, simcamres, etc.).
ncomp	model complexity (number of components) to show the parameters for.
nc	if there are several classes, which class to show the parameters for.
...	other arguments

---

as.matrix.ldecomp      *as.matrix method for ldecomp object*

---

### Description

Generic as.matrix function for linear decomposition. Returns a matrix with information about the decomposition.

### Usage

```
## S3 method for class 'ldecomp'
as.matrix(x, ncomp = NULL, ...)
```

### Arguments

x	object of class ldecomp
ncomp	number of components to get the result for (if NULL will return for each available)
...	other arguments

---

as.matrix.plsdares      *as.matrix method for PLS-DA results*

---

### Description

Returns a matrix with model performance statistics for PLS-DA results

### Usage

```
## S3 method for class 'plsdares'  
as.matrix(x, ncomp = NULL, nc = 1, ...)
```

### Arguments

x	PLS-DA results (object of class plsdares)
ncomp	number of components to calculate the statistics for (if NULL gets for all components)
nc	for which class to calculate the statistics for
...	other arguments

---

as.matrix.plsres      *as.matrix method for PLS results*

---

### Description

Returns a matrix with model performance statistics for PLS results

### Usage

```
## S3 method for class 'plsres'  
as.matrix(x, ncomp = NULL, ny = 1, ...)
```

### Arguments

x	PLS results (object of class plsres)
ncomp	number of components to calculate the statistics for
ny	for which response variable calculate the statistics for
...	other arguments

---

as.matrix.regcoeffs     *as.matrix method for regression coefficients class*

---

### Description

returns matrix with regression coefficients for given response number and amount of components

### Usage

```
## S3 method for class 'regcoeffs'
as.matrix(x, ncomp = 1, ny = 1, ...)
```

### Arguments

x	regression coefficients object (class regcoeffs)
ncomp	number of components to return the coefficients for
ny	number of response variable to return the coefficients for
...	other arguments

---

as.matrix.regres     *as.matrix method for regression results*

---

### Description

Returns a matrix with model performance statistics for regression results

### Usage

```
## S3 method for class 'regres'
as.matrix(x, ncomp = NULL, ny = 1, ...)
```

### Arguments

x	regression results (object of class regres)
ncomp	model complexity (number of components) to calculate the statistics for (can be a vector)
ny	for which response variable calculate the statistics for
...	other arguments

---

as.matrix.simcamres     *as.matrix method for SIMCAM results*

---

### Description

Generic as.matrix function for SIMCAM results. Returns matrix with performance values for specific class.

### Usage

```
## S3 method for class 'simcamres'  
as.matrix(x, nc = seq_len(x$nclasses), ...)
```

### Arguments

x	classification results (object of class plsdares, simcamres, etc.).
nc	vector with classes to use.
...	other arguments

---

as.matrix.simcares     *as.matrix method for SIMCA classification results*

---

### Description

Generic as.matrix function for classification results. Returns matrix with performance values for specific class.

### Usage

```
## S3 method for class 'simcares'  
as.matrix(x, ncomp = NULL, ...)
```

### Arguments

x	classification results (object of class plsdares, simcamres, etc.).
ncomp	model complexity (number of components) to show the parameters for.
...	other arguments

---

capitalize	<i>Capitalize text or vector with text values</i>
------------	---

---

**Description**

Capitalize text or vector with text values

**Usage**

```
capitalize(str)
```

**Arguments**

str	text of vector with text values
-----	---------------------------------

---

carbs	<i>Raman spectra of carbonhydrates</i>
-------	--

---

**Description**

The dataset consists of Raman spectra of fructose, lactose, and ribose as well as spectra of their mixtures.

**Usage**

```
data(simdata)
```

**Format**

The data is a list (carbs) with the following fields:

- \$D a matrix (21x1401) with spectral values for the mixtures.
- \$S a matrix (1401x3) with spectral values for the pure components.
- \$C a matrix (21x3) with concentration of the pure components.

**Details**

The dataset consists of Raman spectra of fructose, lactose, and ribose as well as spectra of their mixtures. The original spectra were downloaded from publicly available SPECARB library [1], created by S.B. Engelsen. The spectra were truncated to the range from 200 to 1600 cm<sup>-1</sup>.

The spectra of mixtures were created by linear combinations of the original spectra:

$$D = CS' + E$$

Concentrations of the components, C, follow a simplex lattice design with four levels. Some noise calculated as a random number uniformly distributed between 0% and 3% of maximum initial

intensity (E) was added to each spectrum of the dataset, D, individually.

## References

1. Engelsen S.B., Database on Raman spectra of carbohydrates. Available at: <http://www.models.life.ku.dk/~specarb/specarb> [visited 31.05.2020]

---

categorize	<i>Categorize PCA results</i>
------------	-------------------------------

---

## Description

Categorize PCA results

## Usage

```
categorize(obj, ...)
```

## Arguments

obj	object with PCA model
...	other parameters

---

categorize.pca	<i>Categorize PCA results based on orthogonal and score distances.</i>
----------------	--

---

## Description

The method compares score and orthogonal distances of PCA results from res with critical limits computed for the PCA model and categorizes the corresponding objects as "regular", "extreme" or "outlier".

## Usage

```
## S3 method for class 'pca'
categorize(obj, res = obj$res$cal, ncomp = obj$ncomp.selected, ...)
```

## Arguments

obj	object with PCA model
res	object with PCA results
ncomp	number of components to use for the categorization
...	other parameters

**Details**

The method does not categorize hidden values if any.

**Value**

vector (factor) with results of categorization.

---

categorize.pls	<i>Categorize data rows based on PLS results and critical limits for total distance.</i>
----------------	--

---

**Description**

The method uses full distance for decomposition of X-data and squared Y-residuals of PLS results from res with critical limits computed for the PLS model and categorizes the corresponding objects as "regular", "extreme" or "outlier".

**Usage**

```
## S3 method for class 'pls'  
categorize(obj, res = obj$res$cal, ncomp = obj$ncomp.selected, ...)
```

**Arguments**

obj	object with PCA model
res	object with PCA results
ncomp	number of components to use for the categorization
...	other parameters

**Details**

The method does not categorize hidden values if any. It is based on the approach described in [1] and works only if data driven approach is used for computing critical limits.

**Value**

vector (factor) with results of categorization.

**References**

1. Rodionova O. Ye., Pomerantsev A. L. Detection of Outliers in Projection-Based Modeling. Analytical Chemistry (2020, in publish). doi: 10.1021/acs.analchem.9b04611



---

chisq.crit	<i>Calculates critical limits for distance values using Chi-square distribution</i>
------------	---

---

**Description**

The method is based on Chi-squared distribution with  $DF = 2 * (m(u)/s(u)^2)$

**Usage**

```
chisq.crit(param, alpha = 0.05, gamma = 0.01)
```

**Arguments**

param	matrix with distribution parameters
alpha	significance level for extreme objects
gamma	significance level for outliers

---

chisq.prob	<i>Calculate probabilities for distance values using Chi-square distribution</i>
------------	--

---

**Description**

Calculate probabilities for distance values using Chi-square distribution

**Usage**

```
chisq.prob(u, param)
```

**Arguments**

u	vector with distances
param	vector with distribution parameters

---

classify.plsda	<i>PLS-DA classification</i>
----------------	------------------------------

---

**Description**

Converts PLS predictions of y values to predictions of classes

**Usage**

```
classify.plsda(model, y)
```

**Arguments**

model	a PLS-DA model (object of class plsda)
y	a matrix with predicted y values

**Details**

This is a service function for PLS-DA class, do not use it manually.

**Value**

Classification results (an object of class classes)

---

classify.simca	<i>SIMCA classification</i>
----------------	-----------------------------

---

**Description**

Make classification based on calculated T2 and Q values and corresponding limits

**Usage**

```
classify.simca(obj, pca.res, c.ref = NULL)
```

**Arguments**

obj	a SIMCA model (object of class simca)
pca.res	results of projection data to PCA space
c.ref	vector with class reference values

**Details**

This is a service function for SIMCA class, do not use it manually.

**Value**

vector with predicted class values (c.pred)

---

```
classmodel.processRefValues
```

*Check reference class values and convert it to a factor if necessary*

---

### Description

Check reference class values and convert it to a factor if necessary

### Usage

```
classmodel.processRefValues(c.ref, classnames = NULL)
```

### Arguments

c.ref	class reference values provided by user
classnames	text with class name in case of logical reference values

---

classes	<i>Results of classification</i>
---------	----------------------------------

---

### Description

classes is used to store results classification for one or multiple classes.

### Usage

```
classes(c.pred, c.ref = NULL, p.pred = NULL, ncomp.selected = 1)
```

### Arguments

c.pred	matrix with predicted values (+1 or -1) for each class.
c.ref	matrix with reference values for each class.
p.pred	matrix with probability values for each class.
ncomp.selected	vector with selected number of components for each class.

### Details

There is no need to create a classes object manually, it is created automatically when build a classification model (e.g. using [simca](#) or [plsda](#)) or apply the model to new data. For any classification method from `mdatools`, a class using to represent results of classification (e.g. [simcares](#)) inherits fields and methods of `classes`.

**Value**

c.pred	predicted class values (+1 or -1).
p.pred	predicted class probabilities.
c.ref	reference (true) class values if provided.
The following fields are available only if reference values were provided.	
tp	number of true positives.
tn	number of true negatives.
fp	number of false positives.
fn	number of false negatives.
specificity	specificity of predictions.
sensitivity	sensitivity of predictions.
misclassified	ratio of misclassified objects.

**See Also**

Methods classes class:

<a href="#">showPredictions.classes</a>	shows table with predicted values.
<a href="#">plotPredictions.classes</a>	makes plot with predicted values.
<a href="#">plotSensitivity.classes</a>	makes sn plot.
<a href="#">plotSpecificity.classes</a>	makes specificity plot.
<a href="#">plotMisclassified.classes</a>	makes ms ratio plot.
<a href="#">plotPerformance.classes</a>	makes plot with misclassified ratio, specificity and sensitivity values.

---

classes.getPerformance

*Calculation of classification performance parameters*

---

**Description**

Calculates and returns performance parameters for classification result (e.g. number of false negatives, false positives, sn, specificity, etc.).

**Usage**

```
classes.getPerformance(c.ref, c.pred)
```

**Arguments**

c.ref	reference class values for objects (vector with numeric or text values)
c.pred	predicted class values for objects (array nobj x ncomponents x nclasses)

**Details**

The function is called automatically when a classification result with reference values is created, for example when applying a plsda or simca models.

**Value**

Returns a list with following fields:

\$fn	number of false negatives (nclasses x ncomponents)
\$fp	number of false positives (nclasses x ncomponents)
\$tp	number of true positives (nclasses x ncomponents)
\$sensitivity	sn values (nclasses x ncomponents)
\$specificity	specificity values (nclasses x ncomponents)
\$specificity	ms ratio values (nclasses x ncomponents)

---

confint.regcoeffs	<i>Confidence intervals for regression coefficients</i>
-------------------	---

---

**Description**

returns matrix with confidence intervals for regression coefficients for given response number and number of components.

**Usage**

```
## S3 method for class 'regcoeffs'
confint(object, parm = NULL, level = 0.95, ncomp = 1, ny = 1, ...)
```

**Arguments**

object	regression coefficients object (class regcoeffs)
parm	not used, needed for compatibility with general method
level	confidence level
ncomp	number of components (one value)
ny	index of response variable (one value)
...	other arguments

---

constraint	<i>Class for MCR-ALS constraint</i>
------------	-------------------------------------

---

**Description**

Class for MCR-ALS constraint

**Usage**

```
constraint(name, params = NULL, method = NULL)
```

**Arguments**

name	short text with name for the constraint
params	a list with parameters for the constraint method (if NULL - default parameters will be used)
method	method to call when applying the constraint, provide it only for user defined constraints

**Details**

Use this class to create constraints and add them to a list for MCR-ALS curve resolution (see [mcraals](#)). Either provide name and parameters to one of the existing constraint implementations or make your own. See the list of implemented constraints by running `constraints()`

For your own constraint you need to create a method, which takes matrix with values (either spectra or contributions being resolved) as the first argument, does something and then return a matrix with the same dimension as the result. The method can have any number of optional parameters.

See help for [mcraals](#) or Bookdown tutorial for details.

---

constraintAngle	<i>Method for angle constraint</i>
-----------------	------------------------------------

---

**Description**

Adds a small portion of mean to contributions or spectra to increase contrast

**Usage**

```
constraintAngle(x, d, weight = 0.05)
```

**Arguments**

x	data matrix (spectra or contributions)
d	matrix with the original spectral values
weight	how many percent of mean to add (between 0 and 1)

---

constraintClosure      *Method for closure constraint*

---

**Description**

Force rows of data sum up to given value

**Usage**

```
constraintClosure(x, d, sum = 1)
```

**Arguments**

x	data matrix (spectra or contributions)
d	matrix with the original spectral values
sum	which value the spectra or contributions should sum up to

---

constraintNonNegativity  
*Method for non-negativity constraint*

---

**Description**

Set all negative values in the matrix to 0

**Usage**

```
constraintNonNegativity(x, d)
```

**Arguments**

x	data matrix (spectra or contributions)
d	matrix with the original spectral values

---

constraintNorm	<i>Method for normalization constraint</i>
----------------	--

---

**Description**

Normalize rows of matrix to unit length or area

**Usage**

```
constraintNorm(x, d, type = "length")
```

**Arguments**

x	data matrix (spectra or contributions)
d	matrix with the original spectral values
type	type of normalization ("area", "length" or "sum")

---

constraints.list	<i>Shows information about all implemented constraints</i>
------------------	--

---

**Description**

Shows information about all implemented constraints

**Usage**

```
constraints.list()
```

---

constraintUnimod	<i>Method for unimodality constraint</i>
------------------	--

---

**Description**

forces column of matrix to have one maximum each

**Usage**

```
constraintUnimod(x, d, tol = 0)
```

**Arguments**

x	data matrix (spectra or contributions)
d	matrix with the original spectral values
tol	tolerance (value between 0 and 1) to take make method stable to small fluctuations



---

crossval	<i>Generate sequence of indices for cross-validation</i>
----------	--

---

**Description**

Generates and returns sequence of object indices for each segment in random segmented cross-validation

**Usage**

```
crossval(cv = 1, nobj = NULL, resp = NULL)
```

**Arguments**

cv	cross-validation settings, can be a number or a list. If cv is a number, it will be used as a number of segments for random cross-validation (if cv = 1, full cross-validation will be performed), if it is a list, the following syntax can be used: cv = list('rand', nseg, nrep) for random repeated cross-validation with nseg segments and nrep repetitions or cv = list('ven', nseg) for systematic splits to nseg segments ('venetian blinds').
nobj	number of objects in a dataset
resp	vector with response values to use in case of venetian blinds

**Value**

matrix with object indices for each segment

---

crossval.getParams	<i>Define parameters based on 'cv' value</i>
--------------------	--

---

**Description**

Define parameters based on 'cv' value

**Usage**

```
crossval.getParams(cv, nobj)
```

**Arguments**

cv	settings for cross-validation provided by user
nobj	number of objects in calibration set

---

crossval.regmodel      *Cross-validation of a regression model*

---

**Description**

Does cross-validation of a regression model

**Usage**

```
crossval.regmodel(obj, x, y, cv, cal.fun)
```

**Arguments**

obj	a regression model (object of class regmodel)
x	a matrix with x values (predictors from calibration set)
y	a matrix with y values (responses from calibration set)
cv	number of segments (if cv = 1, full cross-validation will be used)
cal.fun	reference to function for model calibration

**Value**

object of class plsres with results of cross-validation

---

crossval.simca      *Cross-validation of a SIMCA model*

---

**Description**

Does the cross-validation of a SIMCA model

**Usage**

```
crossval.simca(obj, x, cv)
```

**Arguments**

obj	a SIMCA model (object of class simca)
x	a matrix with x values (predictors from calibration set)
cv	number of segments (if cv = 1, full cross-validation will be used)

**Value**

object of class simcares with results of cross-validation

---

crossval.str	<i>String with description of cross-validation method</i>
--------------	---

---

**Description**

String with description of cross-validation method

**Usage**

```
crossval.str(cv)
```

**Arguments**

cv	a list with cross-validation settings
----	---------------------------------------

**Value**

a string with the description text

---

dd.crit	<i>Calculates critical limits for distance values using Data Driven moments approach</i>
---------	--

---

**Description**

Calculates critical limits for distance values using Data Driven moments approach

**Usage**

```
dd.crit(paramQ, paramT2, alpha = 0.05, gamma = 0.01)
```

**Arguments**

paramQ	matrix with parameters for distribution of Q distances
paramT2	matrix with parameters for distribution of T2 distances
alpha	significance level for extreme objects
gamma	significance level for outliers

---

<code>ddmoments.param</code>	<i>Calculates critical limits for distance values using Data Driven moments approach</i>
------------------------------	--

---

**Description**

Calculates critical limits for distance values using Data Driven moments approach

**Usage**

`ddmoments.param(U)`

**Arguments**

`U` matrix or vector with distance values

---

<code>ddrobust.param</code>	<i>Calculates critical limits for distance values using Data Driven robust approach</i>
-----------------------------	---

---

**Description**

Calculates critical limits for distance values using Data Driven robust approach

**Usage**

`ddrobust.param(U, ncomp, alpha, gamma)`

**Arguments**

`U` matrix or vector with distance values  
`ncomp` number of components  
`alpha` significance level for extreme objects  
`gamma` significance level for outliers

---

ellipse                      *Create ellipse on the current plot*

---

**Description**

Create ellipse on the current plot

**Usage**

```
ellipse(xc = 0, yc = 0, a, b, col = "black", lty = 1, ...)
```

**Arguments**

xc	coordinate of center (x)
yc	coordinate of center (y)
a	major axis
b	minor axis
col	color of the ellipse line
lty	type of the ellipse line
...	any argument suitable for lines function

---

employ.constraint            *Applies constraint to a dataset*

---

**Description**

Applies constraint to a dataset

**Usage**

```
employ.constraint(obj, x, d, ...)
```

**Arguments**

obj	object with constraint
x	matrix with pure spectra or contributions
d	matrix with original spectral values
...	other arguments

---

`employ.prep`*Applies a list with preprocessing methods to a dataset*

---

**Description**

Applies a list with preprocessing methods to a dataset

**Usage**

```
employ.prep(obj, x, ...)
```

**Arguments**

<code>obj</code>	list with preprocessing methods (created using prep function).
<code>x</code>	matrix with dataset
<code>...</code>	other arguments

---

`eye`*Create the identity matrix*

---

**Description**

Create the identity matrix

**Usage**

```
eye(n)
```

**Arguments**

<code>n</code>	Size of the matrix
----------------	--------------------

**Value**

The identity matrix (n x n)

---

`fprintf`                    *Imitation of fprintf() function*

---

**Description**

Imitation of fprintf() function

**Usage**

`fprintf(...)`

**Arguments**

...                    arguments for sprintf function

---

`getCalibrationData`        *Calibration data*

---

**Description**

Calibration data

**Usage**

`getCalibrationData(obj, ...)`

**Arguments**

`obj`                    a model object  
...                    other arguments

**Details**

Generic function getting calibration data from a linear decomposition model (e.g. PCA)

getCalibrationData.pca

*Returns matrix with original calibration data*

---

### Description

Returns matrix with original calibration data

### Usage

```
## S3 method for class 'pca'  
getCalibrationData(obj)
```

### Arguments

obj                    object with PCA model

---

getCalibrationData.simcam

*Get calibration data*

---

### Description

Get data, used for calibration of the SIMCAM individual models and combine to one dataset.

### Usage

```
## S3 method for class 'simcam'  
getCalibrationData(obj, ...)
```

### Arguments

obj                    SIMCAM model (object of class simcam)  
...                    other arguments

### Details

See examples in help for [simcam](#) function.



---

getConfidenceEllipse *Compute confidence ellipse for a set of points*

---

**Description**

Compute confidence ellipse for a set of points

**Usage**

```
getConfidenceEllipse(points, conf.level = 0.95, n = 100)
```

**Arguments**

points	matrix of data frame with coordinates of the points
conf.level	confidence level for the ellipse
n	number of points in the ellipse coordinates

**Value**

matrix with coordinates of the ellipse points (x and y)

---

getConfusionMatrix *Confusion matrix for classification results*

---

**Description**

Confusion matrix for classification results

**Usage**

```
getConfusionMatrix(obj, ...)
```

**Arguments**

obj	classification results (object of class simcares, simcamres, etc)
...	other parameters.

**Details**

Returns confusion matrix for classification results represented by the object.

---

```
getConfusionMatrix.classres
```

*Confusion matrix for classification results*

---

**Description**

The columns of the matrix correspond to classification results, rows - to the real classes. In case of soft classification with multiple classes (e.g. SIMCAM) sum of values for every row will not correspond to the total number of class members as the same object can be classified as a member of several classes or non of them.

**Usage**

```
## S3 method for class 'classres'
getConfusionMatrix(obj, ncomp = obj$ncomp.selected, ...)
```

**Arguments**

obj	classification results (object of class simcares, simcamres, etc)
ncomp	number of components to make the matrix for (NULL - use selected for a model).
...	other arguments

**Details**

Returns confusion matrix for classification results represented by the object.

---

```
getConvexHull
```

*Compute coordinates of a closed convex hull for data points*

---

**Description**

Compute coordinates of a closed convex hull for data points

**Usage**

```
getConvexHull(points)
```

**Arguments**

points	matrix of data frame with coordinates of the points
--------	---

---

getDataLabels	<i>Create a vector with labels for plot series</i>
---------------	--

---

**Description**

For scatter plots labels correspond to rows of the data (names, values, indices, etc.). For non-scatter plots labels correspond to the columns (names, indices or max value for each column)

**Usage**

```
getDataLabels(ps, labels = NULL)
```

**Arguments**

ps	'plotseries' object
labels	vector with user defined labels or type of labels to show ("values", "names", "indices")

---

getImplementedConstraints	<i>Shows a list with implemented constraints</i>
---------------------------	--

---

**Description**

Shows a list with implemented constraints

**Usage**

```
getImplementedConstraints()
```

---

getImplementedPrepMethods	<i>Shows a list with implemented preprocessing methods</i>
---------------------------	--

---

**Description**

Shows a list with implemented preprocessing methods

**Usage**

```
getImplementedPrepMethods()
```

---

getLabelsAsIndices      *Create labels as column or row indices*

---

**Description**

Create labels as column or row indices

**Usage**

getLabelsAsIndices(ps)

**Arguments**

ps                      'plotseries' object

---

getLabelsAsValues      *Create labels from data values*

---

**Description**

Create labels from data values

**Usage**

getLabelsAsValues(ps)

**Arguments**

ps                      'plotseries' object

---

getMainTitle            *Get main title*

---

**Description**

returns main title for a plot depending on a user choice

**Usage**

getMainTitle(main, ncomp, default)

**Arguments**

main	main title of a plot, provided by user
ncomp	number of components to select, provided by user
default	default title for the plot

**Details**

Depedning on a user choice it returns main title for a plot

---

getPlotColors	<i>Define colors for plot series</i>
---------------	--------------------------------------

---

**Description**

Define colors for plot series

**Usage**

```
getPlotColors(ps, col, opacity, cgroup, colmap)
```

**Arguments**

ps	'plotseries' object
col	color specified by user (if any)
opacity	opacity for the color
cgroup	vector for color grouping (if any)
colmap	name or values for colormap

---

getProbabilities	<i>Get class belonging probability</i>
------------------	--

---

**Description**

Compute class belonging probabilities for classification results.

**Usage**

```
getProbabilities(obj, ...)
```

**Arguments**

obj	an object with classification results (e.g. SIMCA)
...	other parameters

---

getProbabilities.pca *Probabilities for residual distances*

---

### Description

Probabilities for residual distances

### Usage

```
## S3 method for class 'pca'  
getProbabilities(obj, ncomp, q, h, ...)
```

### Arguments

obj	object with PCA model
ncomp	number of components to compute the probability for
q	vector with squared orthogonal distances for given number of components
h	vector with score distances for given number of components
...	other parameters

### Details

Computes p-value for every object being from the same populaion as calibration set based on its orthogonal and score distances.

---

getProbabilities.simca  
*Probabilities of class belonging for PCA/SIMCA results*

---

### Description

Probabilities of class belonging for PCA/SIMCA results

### Usage

```
## S3 method for class 'simca'  
getProbabilities(obj, ncomp, q, h, ...)
```

### Arguments

obj	object with PCA model
ncomp	number of components to compute the probability for
q	vector with squared orthogonal distances for given number of components
h	vector with score distances for given number of components
...	other parameters

**Details**

Computes p-value for every object being from the same populaion as calibration set based on its orthogonal and score distances.

---

getPureVariables	<i>Identifies pure variables</i>
------------------	----------------------------------

---

**Description**

The method identifies indices of pure variables using the SIMPLISMA algorithm.

**Usage**

```
getPureVariables(D, ncomp, purevars, offset)
```

**Arguments**

D	matrix with the spectra
ncomp	number of pure components
purevars	user provided values gor pure variables (no calculation will be run in this case)
offset	offset (between 0 and 1) for calculation of parameter alpha

**Value**

The function returns a list with with following fields:

ncomp	number of pure components.
purvars	vector with indices for pure variables.
purityspec	matrix with purity values for each resolved components.
purity	vector with purity values for resolved components.

---

getR	<i>Creates rotation matrix to map a set vectors base1 to a set of vectors base2.</i>
------	--

---

**Description**

In both sets vectors should be orthonormal.

**Usage**

```
getR(base1, base2)
```

**Arguments**

base1            Matrix (JxA) with A orthonormal vectors as columns to be rotated ( $A \leq J$ )  
 base2            Matrix (JxA) with A orthonormal vectors as columns, base1 should be aligned with

**Value**

Rotation matrix (JxJ)

---

getRegcoeffs            *Get regression coefficients*

---

**Description**

Generic function for getting regression coefficients from PLS model

**Usage**

```
getRegcoeffs(obj, ...)
```

**Arguments**

obj                a PLS model  
 ...                other parameters

---

getRegcoeffs.regmodel    *Regression coefficients for PLS model'*

---

**Description**

Returns a matrix with regression coefficients for the PLS model which can be applied to a data directly

**Usage**

```
## S3 method for class 'regmodel'
getRegcoeffs(
  obj,
  ncomp = obj$ncomp.selected,
  ny = 1,
  full = FALSE,
  alpha = 0.05,
  ...
)
```



**Arguments**

obj	a PLS model (object of class pls)
ncomp	number of components to return the coefficients for
ny	if y is multivariate which variables you want to see the coefficients for
full	if TRUE the method also shows p-values and t-values as well as confidence intervals for the coefficients (if available)
alpha	significance level for confidence intervals (a number between 0 and 1, e.g. 0.05)
...	other parameters

**Details**

The method recalculates the regression coefficients found by the PLS algorithm taking into account centering and scaling of predictors and responses, so the matrix with coefficients can be applied directly to original data ( $yp = Xb$ ).

If number of components is not specified, the optimal number, selected by user or identified by a model will be used.

If Jack-knifing method was used to get statistics for the coefficient the method returns all statistics as well (p-value, t-value, confidence interval). In this case user has to specified a number of y-variable (if there are many) to get the statistics and the coefficients for. The confidence interval is computed for unstandardized coefficients.

**Value**

A matrix with regression coefficients and (optionally) statistics.

---

getRes	<i>Return list with valid results</i>
--------	---------------------------------------

---

**Description**

Return list with valid results

**Usage**

```
getRes(res, classname = "ldecomp")
```

**Arguments**

res	list with results
classname	name of class (for result object) to look for

---

getSelectedComponents *Get selected components*

---

**Description**

returns number of components depending on a user choice

**Usage**

```
getSelectedComponents(obj, ncomp = NULL)
```

**Arguments**

obj	an MDA model or result object (e.g. pca, pls, simca, etc)
ncomp	number of components to select, provided by user

**Details**

Depedning on a user choice it returns optimal number of component for the model (if use did not provide any value) or check the user choice for correctness and returns it back

---

getSelectivityRatio *Selectivity ratio*

---

**Description**

Generic function for returning selectivity ratio values for regression model (PCR, PLS, etc)

**Usage**

```
getSelectivityRatio(obj, ...)
```

**Arguments**

obj	a regression model
...	other parameters

---

getSelectivityRatio.pls  
*Selectivity ratio for PLS model*

---

**Description**

Returns vector with Selectivity ratio values. This function is a proxy for [selratio](#) and will be removed in future releases.

**Usage**

```
## S3 method for class 'pls'  
getSelectivityRatio(obj, ncomp = obj$ncomp.selected, ...)
```

**Arguments**

obj	a PLS model (object of class pls)
ncomp	number of components to get the values for (if NULL user selected as optimal will be used)
...	other parameters

**Value**

vector with selectivity ratio values

**References**

[1] Tarja Rajalahti et al. Chemometrics and Laboratory Systems, 95 (2009), pp. 35-48.

---

getVariance.mcr      *Compute explained variance for MCR case*

---

**Description**

Compute explained variance for MCR case

**Usage**

```
getVariance.mcr(obj, x)
```

**Arguments**

obj	object of class mcr
x	original spectral data

getVIPScores

*VIP scores*

---

**Description**

Generic function for returning VIP scores values for regression model (PCR, PLS, etc)

**Usage**

```
getVIPScores(obj, ...)
```

**Arguments**

obj	a regression model
...	other parameters

---

getVIPScores.pls

*VIP scores for PLS model*

---

**Description**

Returns vector with VIP scores values. This function is a proxy for [vipscores](#) and will be removed in future releases.

**Usage**

```
## S3 method for class 'pls'  
getVIPScores(obj, ncomp = obj$ncomp.selected, ...)
```

**Arguments**

obj	a PLS model (object of class pls)
ncomp	number of components to count
...	other parameters

**Value**

matrix nvar x 1 with VIP score values

---

hotelling.crit	<i>Calculate critical limits for distance values using Hotelling T2 distribution</i>
----------------	--

---

**Description**

Calculate critical limits for distance values using Hotelling T2 distribution

**Usage**

```
hotelling.crit(nobj, ncomp, alpha = 0.05, gamma = 0.01)
```

**Arguments**

nobj	number of objects in calibration set
ncomp	number of components
alpha	significance level for extreme objects
gamma	significance level for outliers

**Value**

vector with four values: critical limits for given alpha and gamma, mean distance and DoF.

---

hotelling.prob	<i>Calculate probabilities for distance values and given parameters using Hotelling T2 distribution</i>
----------------	---

---

**Description**

Calculate probabilities for distance values and given parameters using Hotelling T2 distribution

**Usage**

```
hotelling.prob(u, ncomp, nobj)
```

**Arguments**

u	vector with distances
ncomp	number of components
nobj	number of objects in calibration set

---

imshow	<i>show image data as an image</i>
--------	------------------------------------

---

### Description

show image data as an image

### Usage

```
imshow(
  data,
  channels = 1,
  show.excluded = FALSE,
  main = paste0(" ", colnames(data)[channels]),
  colmap = "jet"
)
```

### Arguments

data	data with image
channels	indices for one or three columns to show as image channels
show.excluded	logical, if TRUE the method also shows the excluded (hidden) pixels
main	main title for the image
colmap	colormap using to show the intensity levels

---

ipls	<i>Variable selection with interval PLS</i>
------	---

---

### Description

Applies iPLS algorithm to find variable intervals most important for prediction

### Usage

```
ipls(
  x,
  y,
  glob.ncomp = 10,
  center = TRUE,
  scale = FALSE,
  cv = list("ven", 10),
  exclcols = NULL,
  exclrows = NULL,
  int.ncomp = glob.ncomp,
```

```

int.num = NULL,
int.width = NULL,
int.limits = NULL,
int.niter = NULL,
ncomp.selcrit = "min",
method = "forward",
x.test = NULL,
y.test = NULL,
silent = FALSE
)

```

### Arguments

x	a matrix with predictor values
y	a vector with response values
glob.ncomp	maximum number of components for a global PLS model
center	logical, center or not the data values
scale	logical, standardize or not the data values
cv	cross-validation settings (see details)
exclcols	columns of x to be excluded from calculations (numbers, names or vector with logical values)
exclrows	rows to be excluded from calculations (numbers, names or vector with logical values)
int.ncomp	maximum number of components for interval PLS models
int.num	number of intervals
int.width	width of intervals
int.limits	a two column matrix with manual intervals specification
int.niter	maximum number of iterations (if NULL it will be the same as number of intervals)
ncomp.selcrit	criterion for selecting optimal number of components ('min' for minimum of RMSECV)
method	iPLS method ('forward' or 'backward')
x.test	matrix with predictors for test set (by default is NULL, if specified, is used instead of cv).
y.test	matrix with responses for test set.
silent	logical, show or not information about selection process

### Details

The algorithm splits the predictors into several intervals and tries to find a combination of the intervals, which gives best prediction performance. There are two selection methods: "forward" when the intervals are successively included, and "backward" when the intervals are successively excluded from a model. On the first step the algorithm finds the best (forward) or the worst (backward) individual interval. Then it tests the others to find the one which gives the best model in a

combination with the already selected/excluded one. The procedure continues until the maximum number of iteration is reached.

There are several ways to specify the intervals. First of all either number of intervals (`int.num`) or width of the intervals (`int.width`) can be provided. Alternatively one can specify the limits (first and last variable number) of the intervals manually with `int.limits`.

Cross-validation settings, `cv`, can be a number or a list. If `cv` is a number, it will be used as a number of segments for random cross-validation (if `cv = 1`, full cross-validation will be performed). If it is a list, the following syntax can be used: `cv = list('rand', nseg, nrep)` for random repeated cross-validation with `nseg` segments and `nrep` repetitions or `cv = list('ven', nseg)` for systematic splits to `nseg` segments ('venetian blinds').

### Value

object of 'ipls' class with several fields, including:

<code>var.selected</code>	a vector with indices of selected variables
<code>int.selected</code>	a vector with indices of selected intervals
<code>int.num</code>	total number of intervals
<code>int.width</code>	width of the intervals
<code>int.limits</code>	a matrix with limits for each interval
<code>int.stat</code>	a data frame with statistics for the selection algorithm
<code>glob.stat</code>	a data frame with statistics for the first step (individual intervals)
<code>gm</code>	global PLS model with all variables included
<code>om</code>	optimized PLS model with selected variables

### References

[1] Lars Noergaard et al. Interval partial least-squares regression (iPLS): a comparative chemometric study with an example from near-infrared spectroscopy. *Appl.Spec.* 2000; 54: 413-419

### Examples

```
library(mdatools)

## forward selection for simdata

data(simdata)
Xc = simdata$spectra.c
yc = simdata$conc.c[, 3, drop = FALSE]

# run iPLS and show results
im = ipls(Xc, yc, int.ncomp = 5, int.num = 10, cv = 4, method = "forward")
summary(im)
plot(im)

# show "developing" of RMSECV during the algorithm execution
plotRMSE(im)
```



```
# plot predictions before and after selection
par(mfrow = c(1, 2))
plotPredictions(im$gm)
plotPredictions(im$om)

# show selected intervals on spectral plot
ind = im$var.selected
mspectrum = apply(Xc, 2, mean)
plot(simdata$wavelength, mspectrum, type = 'l', col = 'lightblue')
points(simdata$wavelength[ind], mspectrum[ind], pch = 16, col = 'blue')
```

---

ipls.backward	<i>Runs the backward iPLS algorithm</i>
---------------	---

---

### Description

Runs the backward iPLS algorithm

### Usage

```
ipls.backward(x, y, obj, int.stat, glob.stat)
```

### Arguments

x	a matrix with predictor values
y	a vector with response values
obj	object with initial settings for iPLS algorithm
int.stat	data frame with initial interval statistics
glob.stat	data frame with initial global statistics

---

ipls.forward	<i>Runs the forward iPLS algorithm</i>
--------------	--

---

### Description

Runs the forward iPLS algorithm

### Usage

```
ipls.forward(x, y, obj, int.stat, glob.stat)
```

**Arguments**

x	a matrix with predictor values
y	a vector with response values
obj	object with initial settings for iPLS algorithm
int.stat	data frame with initial interval statistics
glob.stat	data frame with initial global statistics

---

jm.crit	<i>Calculate critical limits for distance values using Jackson-Mudholkar approach</i>
---------	---

---

**Description**

Calculate critical limits for distance values using Jackson-Mudholkar approach

**Usage**

```
jm.crit(residuals, eigenvals, alpha = 0.05, gamma = 0.01)
```

**Arguments**

residuals	matrix with PCA residuals
eigenvals	vector with eigenvalues for PCA components
alpha	significance level for extreme objects
gamma	significance level for outliers

**Value**

vector with four values: critical limits for given alpha and gamma, mean distance and DoF.

---

jm.prob	<i>Calculate probabilities for distance values and given parameters using Hotelling T2 distribution</i>
---------	---

---

**Description**

Calculate probabilities for distance values and given parameters using Hotelling T2 distribution

**Usage**

```
jm.prob(u, eigenvals, ncomp)
```

**Arguments**

u	vector with distances
eigenvals	vector with eigenvalues for PCA components
ncomp	number of components

---

ldecomp	<i>Class for storing and visualising linear decomposition of dataset (<math>X = TP' + E</math>)</i>
---------	---

---

**Description**

Creates an object of ldecomp class.

**Usage**

```
ldecomp(scores, loadings, residuals, eigenvals, ncomp.selected = ncol(scores))
```

**Arguments**

scores	matrix with score values (I x A).
loadings	matrix with loading values (J x A).
residuals	matrix with data residuals (I x J)
eigenvals	vector with eigenvalues for the loadings
ncomp.selected	number of selected components

**Details**

ldecomp is a general class for storing results of decomposition of dataset in form  $X = TP' + E$ . Here, X is a data matrix, T - matrix with scores, P - matrix with loadings and E - matrix with residuals. It is used, for example, for PCA results ([pcares](#)), in PLS and other methods. The class also includes methods for calculation of residual distances and explained variance.

There is no need to use the ldecomp manually. For example, when build PCA model with [pca](#) or apply it to a new data, the results will automatically inherit all methods of ldecomp.

**Value**

Returns an object (list) of ldecomp class with following fields:

scores	matrix with score values (I x A).
residuals	matrix with data residuals (I x J).
T2	matrix with score distances (I x A).
Q	matrix with orthogonal distances (I x A).
ncomp.selected	selected number of components.
expvar	explained variance for each component.
cumexpvar	cumulative explained variance.

---

ldecomp.getDistances *Compute score and residual distances*

---

### Description

Compute orthogonal Euclidean distance from object to PC space (Q, q) and Mahalanobis squared distance between projection of the object to the space and its origin (T2, h).

### Usage

```
ldecomp.getDistances(scores, loadings, residuals, eigenvals)
```

### Arguments

scores	matrix with scores (T).
loadings	matrix with loadings (P).
residuals	matrix with residuals (E).
eigenvals	vector with eigenvalues for the components

### Details

The distances are calculated for every 1:n components, where n goes from 1 to ncomp (number of columns in scores and loadings).

### Value

Returns a list with Q, T2 and tnorm values for each component.

---

ldecomp.getLimitsCoordinates  
*Compute coordinates of lines or curves with critical limits*

---

### Description

Compute coordinates of lines or curves with critical limits

### Usage

```
ldecomp.getLimitsCoordinates(  
  Qlim,  
  T2lim,  
  ncomp,  
  norm,  
  log,  
  show.limits = c(TRUE, TRUE)  
)
```

**Arguments**

Qlim	matrix with critical limits for orthogonal distances
T2lim	matrix with critical limits for score distances
ncomp	number of components for computing the coordinates
norm	logical, shall distance values be normalized or not
log	logical, shall log transformation be applied or not
show.limits	vector with two logical values defining if limits for extreme and/or outliers must be shown

**Value**

list with two matrices (x and y coordinates of corresponding limits)

---

`ldecomp.getLimParams` *Compute parameters for critical limits based on calibration results*

---

**Description**

Compute parameters for critical limits based on calibration results

**Usage**

```
ldecomp.getLimParams(U)
```

**Arguments**

U	matrix with residual distances
---	--------------------------------

---

`ldecomp.getQLimits` *Compute critical limits for orthogonal distances (Q)*

---

**Description**

Compute critical limits for orthogonal distances (Q)

**Usage**

```
ldecomp.getQLimits(lim.type, alpha, gamma, params, residuals, eigenvals)
```

**Arguments**

lim.type	which method to use for calculation of critical limits for residuals
alpha	significance level for extreme limits.
gamma	significance level for outlier limits.
params	distribution parameters returned by Idecomp.getLimParams
residuals	matrix with residuals (E)
eigenvals	egenvalues for the components used to decompose the data

---

Idecomp.getT2Limits    *Compute critical limits for score distances (T2)*

---

**Description**

Compute critical limits for score distances (T2)

**Usage**

```
Idecomp.getT2Limits(lim.type, alpha, gamma, params)
```

**Arguments**

lim.type	which method to use for calculation ("chisq", "ddmoments", "ddrobust")
alpha	significance level for extreme limits.
gamma	significance level for outlier limits.
params	distribution parameters returned by Idecomp.getLimParams

---

Idecomp.getVariances    *Compute explained variance*

---

**Description**

Computes explained variance and cumulative explained variance for data decomposition.

**Usage**

```
Idecomp.getVariances(scores, loadings, residuals, Q)
```

**Arguments**

scores	matrix with scores (T).
loadings	matrix with loadings (P).
residuals	matrix with residuals (E).
Q	matrix with squared orthogonal distances.

**Value**

Returns a list with two vectors.

---

ldecomp.plotResiduals *Residuals distance plot for a set of ldecomp objects*

---

**Description**

Shows a plot with score (T2, h) vs orthogonal (Q, q) distances and corresponding critical limits for given number of components.

**Usage**

```
ldecomp.plotResiduals(
  res,
  Qlim,
  T2lim,
  ncomp,
  log = FALSE,
  norm = FALSE,
  cgroup = NULL,
  xlim = NULL,
  ylim = NULL,
  show.limits = c(TRUE, TRUE),
  lim.col = c("darkgray", "darkgray"),
  lim.lwd = c(1, 1),
  lim.lty = c(2, 3),
  show.legend = TRUE,
  legend.position = "topright",
  show.excluded = FALSE,
  ...
)
```

**Arguments**

res	list with result objects to show the plot for
Qlim	matrix with critical limits for orthogonal distance
T2lim	matrix with critical limits for score distance
ncomp	how many components to use (by default optimal value selected for the model will be used)
log	logical, apply log transformation to the distances or not (see details)
norm	logical, normalize distance values or not (see details)
cgroup	color grouping of plot points (works only if one result object is available)
xlim	limits for x-axis (if NULL will be computed automatically)

<code>ylim</code>	limits for y-axis (if NULL will be computed automatically)
<code>show.limits</code>	vector with two logical values defining if limits for extreme and/or outliers must be shown
<code>lim.col</code>	vector with two values - line color for extreme and outlier limits
<code>lim.lwd</code>	vector with two values - line width for extreme and outlier limits
<code>lim.lty</code>	vector with two values - line type for extreme and outlier limits
<code>show.legend</code>	logical, show or not legend on the plot (if more than one result object)
<code>legend.position</code>	if legend must be shown, where it should be
<code>show.excluded</code>	logical, show or hide rows marked as excluded (attribute 'exclrows').
<code>...</code>	other plot parameters (see <code>mdaplotg</code> for details)

### Details

The function is a bit more advanced version of `plotResiduals.ldecomp`. It allows to show distance values for several result objects (e.g. calibration and test set or calibration and new prediction set) as well as display the corresponding critical limits in form of lines or curves.

Depending on how many result objects your model has or how many you specified manually, using the `res` parameter, the plot behaves in a bit different way.

If only one result object is provided, then it allows to colorise the points using `cgroup` parameter. If two or more result objects are provided, then the function show distances in groups, and adds corresponding legend.

The function can show distance values normalised ( $h/h_0$  and  $q/q_0$ ) as well as with log transformation ( $\log(1 + h/h_0)$ ,  $\log(1 + q/q_0)$ ). The latter is useful if distribution of the points is skewed and most of them are densely located around bottom left corner.

---

mcr

*General class for Multivariate Curve Resolution model*

---

### Description

`mcr` is used to store and visualise general MCR data and results.

### Usage

```
mcr(x, ncomp, method, exclrows = NULL, exclcols = NULL, info = "", ...)
```

### Arguments

<code>x</code>	spectra of mixtures (as matrix or data frame)
<code>ncomp</code>	number of pure components to resolve
<code>method</code>	function for computing spectra of pure components



exclrows	rows to be excluded from calculations (numbers, names or vector with logical values)
exclcols	columns to be excluded from calculations (numbers, names or vector with logical values)
info	text with information about the MCR model
...	other parameters related to specific method

---

mcrals

*Multivariate curve resolution using Alternating Least Squares*


---

## Description

mcralls allows to resolve spectroscopic data to linear combination of individual spectra and contributions using the alternating least squares (ALS) algorithm with constraints.

## Usage

```
mcrals(
  x,
  ncomp,
  cont.constraints = list(),
  spec.constraints = list(),
  spec.ini = matrix(runif(ncol(x) * ncomp), ncol(x), ncomp),
  cont.forced = matrix(NA, nrow(x), ncomp),
  spec.forced = matrix(NA, ncol(x), ncomp),
  cont.solver = mcrals.nnl,
  spec.solver = mcrals.nnl,
  exclrows = NULL,
  exclcols = NULL,
  verbose = FALSE,
  max.niter = 100,
  tol = 10^-6,
  info = ""
)
```

## Arguments

x	spectra of mixtures (matrix or data frame).
ncomp	number of components to calculate.
cont.constraints	a list with constraints to be applied to contributions (see details).
spec.constraints	a list with constraints to be applied to spectra (see details).
spec.ini	a matrix with initial estimation of the pure components spectra.
cont.forced	a matrix which allows to force some of the concentration values (see details).

<code>spec.forced</code>	a matrix which allows to force some of the spectra values (see details).
<code>cont.solver</code>	which function to use as a solver for resolving of pure components contributions (see details).
<code>spec.solver</code>	which function to use as a solver for resolving of pure components spectra (see details).
<code>exclrows</code>	rows to be excluded from calculations (numbers, names or vector with logical values).
<code>exclcols</code>	columns to be excluded from calculations (numbers, names or vector with logical values).
<code>verbose</code>	logical, if TRUE information about every iteration will be shown.
<code>max.niter</code>	maximum number of iterations.
<code>tol</code>	tolerance, when explained variance change is smaller than this value, iterations stop.
<code>info</code>	a short text with description of the case (optional).

## Details

The method implements the iterative ALS algorithm, where, at each iteration, spectra and contributions of each chemical component are estimated and then a set of constraints is applied to each. The method is well described in [1, 2].

The method assumes that the spectra (D) is a linear combination of pure components spectra (S) and pure component concentrations (C):

$$D = CS' + E$$

So the task is to get C and S by knowing D. In order to do that you need to provide:

1. Constraints for spectra and contributions. The constraints should be provided as a list with name of the constraint and all necessary parameters. You can see which constraints and parameters are currently supported by running `constraintList()`. See the code examples below or a Bookdown tutorial for more details.
2. Initial estimation of the pure components spectra, S. By default method uses a matrix with random numbers but you can provide a better guess (for example by running `mcrpure`) as a first step.
3. Which solver to use for resolving spectra and concentrations. There are two built in solvers: `mcrals.nls` (default) and `mcrals.ols`. The first implements non-negative least squares method which gives non-negative (thus physically meaningful) solutions. The second is ordinary least squares and if you want to get non-negative spectra and/or contributions in this case you need to provide a non-negativity constraint.

The algorithm iteratively resolves C and S and checks how well  $CS'$  is to D. The iterations stop either when number exceeds value in `max.niter` or when improvements (difference between explained variance on current and previous steps) is smaller than `tol` value.

Parameters `cont.force` and `spec.force` allows you to force some parts of the contributions or the spectra to be equal to particular pre-defined values. In this case you need to provide the parameters (or just one of them) in form of a matrix. For example `cont.force` should have as many rows as many you have in the original spectral data  $x$  and as many columns as many pure components you want to resolve. Feel all values of this matrix with NA and the values you want to force with

real numbers. For example if you know that in the first measurement concentration of 2 and 3 components was zero, set the corresponding values of `cont.force` to zero. See also the last case in the examples section.

### Value

Returns an object of `mcrpure` class with the following fields:

<code>resspec</code>	matrix with resolved spectra.
<code>rescont</code>	matrix with resolved contributions.
<code>cont.constraints</code>	list with contribution constraints provided by user.
<code>spec.constraints</code>	list with spectra constraints provided by user.
<code>expvar</code>	vector with explained variance for each component (in percent).
<code>cumexpvar</code>	vector with cumulative explained variance for each component (in percent).
<code>ncomp</code>	number of resolved components
<code>max.niter</code>	maximum number of iterations
<code>info</code>	information about the model, provided by user when build the model.

More details and examples can be found in the Bookdown tutorial.

### Author(s)

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

### References

1. J. Jaumot, R. Gargallo, A. de Juan, and R. Tauler, "A graphical user-friendly interface for MCR-ALS: a new tool for multivariate curve resolution in MATLAB", *Chemometrics and Intelligent # Laboratory Systems* 76, 101-110 (2005).

### See Also

Methods for `mcrals` objects:

<code>summary.mcrals</code>	shows some statistics for the case.
<code>predict.mcrals</code>	computes contributions by projection of new spectra to the resolved ones.

Plotting methods for `mcrals` objects:

<code>plotSpectra.mcr</code>	shows plot with resolved spectra.
<code>plotContributions.mcr</code>	shows plot with resolved contributions.
<code>plotVariance.mcr</code>	shows plot with explained variance.
<code>plotCumVariance.mcr</code>	shows plot with cumulative explained variance.

## Examples

```
library(mdatools)

# resolve mixture of carbohydrates Raman spectra

data(carbs)

# define constraints for contributions
cc <- list(
  constraint("nonneg")
)

# define constraints for spectra
cs <- list(
  constraint("nonneg"),
  constraint("norm", params = list(type = "area"))
)

# because by default initial approximation is made by using random numbers
# we need to seed the generator in order to get reproducible results
set.seed(6)

# run ALS
m <- mcrals(carbs$D, ncomp = 3, cont.constraints = cc, spec.constraints = cs)
summary(m)

# plot cumulative and individual explained variance
par(mfrow = c(1, 2))
plotVariance(m)
plotCumVariance(m)

# plot resolved spectra (all of them or individually)
par(mfrow = c(2, 1))
plotSpectra(m)
plotSpectra(m, comp = 2:3)

# plot resolved contributions (all of them or individually)
par(mfrow = c(2, 1))
plotContributions(m)
plotContributions(m, comp = 2:3)

# of course you can do this manually as well, e.g. show original
# and resolved spectra
par(mfrow = c(1, 1))
mdaplotg(
  list(
    "original" = prep.norm(carbs$D, "area"),
    "resolved" = prep.norm(mda.subset(mda.t(m$resspec), 1), "area")
  ), col = c("gray", "red"), type = "l"
```

```

)

# in case if you have reference spectra of components you can compare them with
# the resolved ones:
par(mfrow = c(3, 1))
for (i in 1:3) {
  mdaplotg(
    list(
      "pure" = prep.norm(mda.subset(mda.t(carbs$S), 1), "area"),
      "resolved" = prep.norm(mda.subset(mda.t(m$respec), 1), "area")
    ), col = c("gray", "red"), type = "l", lwd = c(3, 1)
  )
}

# This example shows how to force some of the contribution values
# First of all we combine the matrix with mixtures and the pure spectra, so the pure
# spectra are on top of the combined matrix
Dplus <- mda.rbind(mda.t(carbs$S), carbs$D)

# since we know that concentration of C2 and C3 is zero in the first row (it is a pure
# spectrum of first component), we can force them to be zero in the optimization procedure.
# Similarly we can do this for second and third rows.

cont.forced <- matrix(NA, nrow(Dplus), 3)
cont.forced[1, ] <- c(NA, 0, 0)
cont.forced[2, ] <- c(0, NA, 0)
cont.forced[3, ] <- c(0, 0, NA)

m <- mcrals(Dplus, 3, cont.forced = cont.forced, cont.constraints = cc, spec.constraints = cs)
plot(m)

# See bookdown tutorial for more details.

```

---

mcrals.cal

*Identifies pure variables*


---

## Description

The method identifies indices of pure variables using the SIMPLISMA algorithm.

## Usage

```

mcrals.cal(
  D,
  ncomp,
  cont.constraints,
  spec.constraints,

```

```

spec.ini,
cont.forced,
spec.forced,
cont.solver,
spec.solver,
max.niter,
tol,
verbose
)

```

### Arguments

D	matrix with the spectra
ncomp	number of pure components
cont.constraints	a list with constraints to be applied to contributions (see details).
spec.constraints	a list with constraints to be applied to spectra (see details).
spec.ini	a matrix with initial estimation of the pure components spectra.
cont.forced	a matrix which allows to force some of the concentration values (see details).
spec.forced	a matrix which allows to force some of the spectra values (see details).
cont.solver	which function to use as a solver for resolving of pure components contributions (see details).
spec.solver	which function to use as a solver for resolving of pure components spectra (see details).
max.niter	maximum number of iterations.
tol	tolerance, when explained variance change is smaller than this value, iterations stop.
verbose	logical, if TRUE information about every iteration will be shown.

### Value

The function returns a list with with following fields:

ncomp	number of pure components.
resspec	matrix with resolved spectra.
rescont	matrix with resolved contributions.
cont.constraints	list with contribution constraints provided by user.
spec.constraints	list with spectra constraints provided by user.
max.niter	maximum number of iterations

---

mcrals.fcnnls                      *Fast combinatorial non-negative least squares*

---

### Description

Fast combinatorial non-negative least squares

### Usage

```
mcrals.fcnnls(
  D,
  A,
  tol = 10 * .Machine$double.eps * as.numeric(sqrt(crossprod(A[, 1]))) * nrow(A)
)
```

### Arguments

D	a matrix
A	a matrix
tol	tolerance parameter for algorithm convergence

### Details

Computes Fast combinatorial NNLS solution for B:  $D = AB'$  subject to  $B \geq 0$ . Implements the method described in [1].

### References

1. Van Benthem, M.H. and Keenan, M.R. (2004), Fast algorithm for the solution of large scale non-negativity-constrained least squares problems. J. Chemometrics, 18: 441-450. doi:10.1002/cem.889

---

mcrals.nnls                      *Non-negative least squares*

---

### Description

Non-negative least squares

### Usage

```
mcrals.nnls(
  D,
  A,
  tol = 10 * .Machine$double.eps * as.numeric(sqrt(crossprod(A[, 1]))) * nrow(A)
)
```

**Arguments**

D	a matrix
A	a matrix
tol	tolerance parameter for algorithm convergence

**Details**

Computes NNLS solution for  $B: D = AB'$  subject to  $B \geq 0$ . Implements the active-set based algorithm proposed by Lawson and Hanson [1].

**References**

1. Lawson, Charles L.; Hanson, Richard J. (1995). Solving Least Squares Problems. SIAM.

---

mcrals.ols

*Ordinary least squares*

---

**Description**

Ordinary least squares

**Usage**

mcrals.ols(D, A)

**Arguments**

D	a matrix
A	a matrix

**Details**

Computes OLS solution for  $D = AB'$  (or  $D' = AB'$ ), where D, A are known



---

mcrpure

*Multivariate curve resolution based on pure variables*


---

### Description

mcrpure allows to resolve spectroscopic data to linear combination of individual spectra and contributions using the pure variables approach.

### Usage

```
mcrpure(
  x,
  ncomp,
  purevars = NULL,
  offset = 0.05,
  exclrows = NULL,
  exclcols = NULL,
  info = ""
)
```

### Arguments

x	spectra of mixtures (matrix or data frame).
ncomp	maximum number of components to calculate.
purevars	vector with indices for pure variables (optional, if you want to provide the variables directly).
offset	offset for correcting noise in computing maximum angles (should be value within [0, 1]).
exclrows	rows to be excluded from calculations (numbers, names or vector with logical values).
exclcols	columns to be excluded from calculations (numbers, names or vector with logical values).
info	a short text with description of the case (optional).

### Details

The method estimates purity of each variable and then uses the purest ones to decompose the spectral data into spectra ('resspec') and contributions ('rescont') of individual chemical components by ordinary least squares.

The pure variables are identified using stepwise maximum angle calculations and described in detail in [1]. So the purity of a spectral variable (wavelength, wavenumber) is actually an angle (measured in degrees) between the variable and vector of ones for the first component; and between the variable and space formed by previously found pure variables for the other components.

**Value**

Returns an object of `mcrpure` class with the following fields:

<code>resspec</code>	matrix with resolved spectra.
<code>rescont</code>	matrix with resolved contributions.
<code>purevars</code>	indices of the selected pure variables.
<code>purevals</code>	purity values for the selected pure variables.
<code>puritiespec</code>	purity spectra (matrix with purity values for each variable and component).
<code>expvar</code>	vector with explained variance for each component (in percent).
<code>cumexpvar</code>	vector with cumulative explained variance for each component (in percent).
<code>offset</code>	offset value used to compute the purity
<code>ncomp</code>	number of resolved components
<code>info</code>	information about the model, provided by user when build the model.

More details and examples can be found in the Bookdown tutorial.

**Author(s)**

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

**References**

1. Willem Windig, Neal B. Gallagher, Jeremy M. Shaver, Barry M. Wise. A new approach for interactive self-modeling mixture analysis. *Chemometrics and Intelligent Laboratory Systems*, 77 (2005) 85–96. DOI: 10.1016/j.chemolab.2004.06.009

**See Also**

Methods for `mcrpure` objects:

<code>summary.mcrpure</code>	shows some statistics for the case.
<code>unmix.mcrpure</code>	makes unmixing of new set of spectra.
<code>predict.mcrpure</code>	computes contributions by projection of new spectra to the resolved ones.

Plotting methods for `mcrpure` objects:

<code>plotPurity.mcrpure</code>	shows plot with maximum purity of each component.
<code>plotPuritySpectra.mcrpure</code>	shows plot with purity spectra.
<code>plotSpectra.mcr</code>	shows plot with resolved spectra.
<code>plotContributions.mcr</code>	shows plot with resolved contributions.
<code>plotVariance.mcr</code>	shows plot with explained variance.
<code>plotCumVariance.mcr</code>	shows plot with cumulative explained variance.

**Examples**

```

library(mdatools)

# resolve mixture of carbohydrates Raman spectra

data(carbs)
m = mcrpure(carbs$D, ncomp = 3)

# examples for purity spectra plot (you can select which components to show)
par(mfrow = c(2, 1))
plotPuritySpectra(m)
plotPuritySpectra(m, comp = 2:3)

# you can do it manually and combine e.g. with original spectra
par(mfrow = c(1, 1))
mdaplotg(
  list(
    "spectra" = prep.norm(carbs$D, "area"),
    "purity" = prep.norm(mda.subset(mda.t(m$respec), 1), "area")
  ), col = c("gray", "red"), type = "l"
)

# show the maximum purity for each component
par(mfrow = c(1, 1))
plotPurity(m)

# plot cumulative and individual explained variance
par(mfrow = c(1, 2))
plotVariance(m)
plotCumVariance(m)

# plot resolved spectra (all of them or individually)
par(mfrow = c(2, 1))
plotSpectra(m)
plotSpectra(m, comp = 2:3)

# plot resolved contributions (all of them or individually)
par(mfrow = c(2, 1))
plotContributions(m)
plotContributions(m, comp = 2:3)

# of course you can do this manually as well, e.g. show original
# and resolved spectra
par(mfrow = c(1, 1))
mdaplotg(
  list(
    "original" = prep.norm(carbs$D, "area"),
    "resolved" = prep.norm(mda.subset(mda.t(m$respec), 1), "area")
  ), col = c("gray", "red"), type = "l"
)

# in case if you have reference spectra of components you can compare them with

```

```

# the resolved ones:
par(mfrow = c(3, 1))
for (i in 1:3) {
  mdaplotg(
    list(
      "pure" = prep.norm(mda.subset(mda.t(carbs$S), 1), "area"),
      "resolved" = prep.norm(mda.subset(mda.t(m$respec), 1), "area")
    ), col = c("gray", "red"), type = "l", lwd = c(3, 1)
  )
}

# See bookdown tutorial for more details.

```

---

mda.cbind	<i>A wrapper for cbind() method with proper set of attributes</i>
-----------	---

---

**Description**

A wrapper for cbind() method with proper set of attributes

**Usage**

```
mda.cbind(...)
```

**Arguments**

... datasets (data frames or matrices) to bind

**Value**

the merged datasets

---

mda.data2im	<i>Convert data matrix to an image</i>
-------------	--

---

**Description**

Convert data matrix to an image

**Usage**

```
mda.data2im(data)
```

**Arguments**

data data matrix

---

mda.df2mat	<i>Convert data frame to a matrix</i>
------------	---------------------------------------

---

**Description**

The function converts data frame to a numeric matrix.

**Usage**

```
mda.df2mat(x, full = FALSE)
```

**Arguments**

x	a data frame
full	logical, if TRUE number of dummy variables for a factor will be the same as number of levels, otherwise by one smaller

**Details**

If one or several columns of the data frame are factors they will be converted to a set of dummy variables. If any columns/rows were hidden in the data frame they will remain hidden in the matrix. If there are factors among the hidden columns, the corresponding dummy variables will be hidden as well.

All other attributes (names, axis names, etc.) will be inherited.

**Value**

a numeric matrix

---

mda.exclcols	<i>Exclude/hide columns in a dataset</i>
--------------	--

---

**Description**

Exclude/hide columns in a dataset

**Usage**

```
mda.exclcols(x, ind)
```

**Arguments**

x	dataset (data frame or matrix).
ind	indices of columns to exclude (numbers, names or logical values)

**Details**

The method assign attribute 'exclcols', which contains number of columns, which should be excluded/hidden from calculations and plots (without removing them physically). The argument ind should contain column numbers (excluding already hidden), names or logical values.

**Value**

dataset with excluded columns

---

mda.exclrows	<i>Exclude/hide rows in a dataset</i>
--------------	---------------------------------------

---

**Description**

Exclude/hide rows in a dataset

**Usage**

```
mda.exclrows(x, ind)
```

**Arguments**

x	dataset (data frame or matrix).
ind	indices of rows to exclude (numbers, names or logical values)

**Details**

The method assign attribute 'exclrows', which contains number of rows, which should be excluded/hidden from calculations and plots (without removing them physically). The argument ind should contain rows numbers (excluding already hidden), names or logical values.

**Value**

dataset with excluded rows

---

mda.getattr	<i>Get data attributes</i>
-------------	----------------------------

---

**Description**

Returns a list with important data attributes (name, xvalues, excluded rows and columns, etc.)

**Usage**

```
mda.getattr(x)
```

**Arguments**

x	a dataset
---	-----------

---

mda.getexclind	<i>Get indices of excluded rows or columns</i>
----------------	--

---

**Description**

Get indices of excluded rows or columns

**Usage**

```
mda.getexclind(excl, names, n)
```

**Arguments**

excl	vector with excluded values (logical, text or numbers)
names	vector with names for rows or columns
n	number of rows or columns

---

mda.im2data	<i>Convert image to data matrix</i>
-------------	-------------------------------------

---

**Description**

Convert image to data matrix

**Usage**

```
mda.im2data(img)
```

**Arguments**

img	an image (3-way array)
-----	------------------------

---

mda.inclcols	<i>Include/unhide the excluded columns</i>
--------------	--

---

**Description**

include columns specified by user (earlier excluded using mda.exclcols)

**Usage**

```
mda.inclcols(x, ind)
```

**Arguments**

x	dataset (data frame or matrix).
ind	number of excluded columns to include

**Value**

dataset with included columns.

---

mda.inclrows	<i>include/unhide the excluded rows</i>
--------------	---

---

**Description**

include rows specified by user (earlier excluded using mda.exclrows)

**Usage**

```
mda.inclrows(x, ind)
```

**Arguments**

x	dataset (data frame or matrix).
ind	number of excluded rows to include

**Value**

dataset with included rows



---

mda.purge	<i>Removes excluded (hidden) rows and colmns from data</i>
-----------	--

---

**Description**

Removes excluded (hidden) rows and colmns from data

**Usage**

```
mda.purge(data)
```

**Arguments**

data	data frame or matrix with data
------	--------------------------------

---

mda.purgeCols	<i>Removes excluded (hidden) colmns from data</i>
---------------	---

---

**Description**

Removes excluded (hidden) colmns from data

**Usage**

```
mda.purgeCols(data)
```

**Arguments**

data	data frame or matrix with data
------	--------------------------------

---

mda.purgeRows	<i>Removes excluded (hidden) rows from data</i>
---------------	---

---

**Description**

Removes excluded (hidden) rows from data

**Usage**

```
mda.purgeRows(data)
```

**Arguments**

data	data frame or matrix with data
------	--------------------------------

---

mda.rbind	<i>A wrapper for rbind() method with proper set of attributes</i>
-----------	---

---

**Description**

A wrapper for rbind() method with proper set of attributes

**Usage**

```
mda.rbind(...)
```

**Arguments**

... datasets (data frames or matrices) to bind

**Value**

the merged datasets

---

mda.setattr	<i>Set data attributes</i>
-------------	----------------------------

---

**Description**

Set most important data attributes (name, xvalues, excluded rows and columns, etc.) to a dataset

**Usage**

```
mda.setattr(x, attrs, type = "all")
```

**Arguments**

x	a dataset
attrs	list with attributes
type	a text variable telling which attributes to set ('all', 'row', 'col')

---

mda.setimbg	<i>Remove background pixels from image data</i>
-------------	---

---

**Description**

Remove background pixels from image data

**Usage**

```
mda.setimbg(data, bgpixels)
```

**Arguments**

data	a matrix with image data
bgpixels	vector with indices or logical values corresponding to background pixels

---

mda.show	<i>Wrapper for show() method</i>
----------	----------------------------------

---

**Description**

Wrapper for show() method

**Usage**

```
mda.show(x, n = 50)
```

**Arguments**

x	data set
n	number of rows to show

---

mda.subset	<i>A wrapper for subset() method with proper set of attributed</i>
------------	--

---

**Description**

A wrapper for subset() method with proper set of attributed

**Usage**

```
mda.subset(x, subset = NULL, select = NULL)
```

**Arguments**

x	dataset (data frame or matrix)
subset	which rows to keep (indices, names or logical values)
select	which columns to select (indices, names or logical values)

**Details**

The method works similar to the standard subset() method, with minor differences. First of all it keeps (and correct, if necessary) all important attributes. If only columns are selected, it keeps all excluded rows as excluded. If only rows are selected, it keeps all excluded columns. If both rows and columns are selected it removed all excluded elements first and then makes the subset.

The parameters subset and select may each be a vector with numbers or nanes without excluded elements, or a logical expression.

**Value**

a data with the subset

---

mda.t	<i>A wrapper for t() method with proper set of attributes</i>
-------	---

---

**Description**

A wrapper for t() method with proper set of attributes

**Usage**

```
mda.t(x)
```

**Arguments**

x	dataset (data frames or matrices) to transpose
---	--

**Value**

the transposed dataset

---

mdaplot

*Plotting function for a single set of objects*

---

**Description**

mdaplot is used to make different kinds of plot for one set of data objects.

**Usage**

```
mdaplot(  
  data = NULL,  
  ps = NULL,  
  type = "p",  
  pch = 16,  
  col = NULL,  
  bg = par("bg"),  
  bwd = 0.8,  
  border = NA,  
  lty = 1,  
  lwd = 1,  
  cex = 1,  
  cgroup = NULL,  
  xlim = NULL,  
  ylim = NULL,  
  colmap = "default",  
  labels = NULL,  
  main = NULL,  
  xlab = NULL,  
  ylab = NULL,  
  show.labels = FALSE,  
  show.colorbar = !is.null(cgroup),  
  show.lines = FALSE,  
  show.grid = TRUE,  
  grid.lwd = 0.5,  
  grid.col = "lightgray",  
  show.axes = TRUE,  
  xticks = NULL,  
  yticks = NULL,  
  xticklabels = NULL,  
  yticklabels = NULL,  
  xlas = 0,  
  ylas = 0,  
  lab.col = "darkgray",
```

```

lab.cex = 0.65,
show.excluded = FALSE,
col.excluded = "#C0C0C0",
nbins = 60,
force.x.values = NA,
opacity = 1,
pch.colinv = FALSE,
...
)

```

### Arguments

<code>data</code>	a vector, matrix or a data.frame with data values.
<code>ps</code>	'plotseries' object, if NULL will be created based on the provided data values
<code>type</code>	type of the plot ("p", "d", "l", "b", "h", "e").
<code>pch</code>	a character for markers (same as plot parameter).
<code>col</code>	a color for markers or lines (same as plot parameter).
<code>bg</code>	background color for scatter plots with 'pch=21:25'.
<code>bwd</code>	a width of a bar as a percent of a maximum space available for each bar.
<code>border</code>	color for border of bars (if barplot is used)
<code>lty</code>	line type
<code>lwd</code>	line width
<code>cex</code>	scale factor for the marker
<code>cgroup</code>	a vector with values to use for make color groups.
<code>xlim</code>	limits for the x axis (if NULL, will be calculated automatically).
<code>ylim</code>	limits for the y axis (if NULL, will be calculated automatically).
<code>colmap</code>	a colormap to use for coloring the plot items.
<code>labels</code>	a vector with text labels for data points or one of the following: "names", "indices", "values".
<code>main</code>	an overall title for the plot (same as plot parameter).
<code>xlab</code>	a title for the x axis (same as plot parameter).
<code>ylab</code>	a title for the y axis (same as plot parameter).
<code>show.labels</code>	logical, show or not labels for the data objects.
<code>show.colorbar</code>	logical, show or not colorbar legend if color grouping is on.
<code>show.lines</code>	vector with two coordinates (x, y) to show horizontal and vertical line cross the point.
<code>show.grid</code>	logical, show or not a grid for the plot.
<code>grid.lwd</code>	line thickness (width) for the grid.
<code>grid.col</code>	line color for the grid.
<code>show.axes</code>	logical, make a normal plot or show only elements (markers, lines, bars) without axes.

<code>xticks</code>	values for x ticks.
<code>yticks</code>	values for y ticks.
<code>xticklabels</code>	labels for x ticks.
<code>yticklabels</code>	labels for y ticks.
<code>xlas</code>	orientation of xticklabels.
<code>ylas</code>	orientation of yticklabels.
<code>lab.col</code>	color for data point labels.
<code>lab.cex</code>	size for data point labels.
<code>show.excluded</code>	logical, show or hide rows marked as excluded (attribute 'exclrows').
<code>col.excluded</code>	color for the excluded objects (rows).
<code>nbins</code>	if scatter density plot is shown, number of segments to split the plot area into. (see also <code>?smoothScatter</code> )
<code>force.x.values</code>	vector with corrected x-values for a bar plot (do not specify this manually).
<code>opacity</code>	opacity for plot colors (value between 0 and 1).
<code>pch.colinv</code>	allows to swap values for 'col' and 'bg' for scatter plots with 'pch' values from 21 to 25.
<code>...</code>	other plotting arguments.

## Details

Most of the parameters are similar to what are used with standard `plot` function. The differences are described below.

The function makes a plot of one set of objects. It can be a set of points (scatter plot), bars, lines, scatter-lines, errorbars or an image. The data is organized as a data frame, matrix or vector. For scatter and only first two columns will be used, for bar plot only values from the first row. It is recommended to use `mda.subset` method if plot should be made only for a subset of the data, especially if you have any excluded rows or columns or other special attributed, described in the Bookdown tutorial.

If data is a data frame and contains one or more factors, they will be converted to a dummy variables (using function `mda.df2mat`) and appears at the end (last columns) if line or bar plot is selected.

The function allows to colorize lines and points according to values of a parameter `cgroup`. The parameter must be a vector with the same elements as number of objects (rows) in the data. The values are divided into up to eight intervals and for each interval a particular color from a selected color scheme is assigned. Parameter `show.colorbar` allows to turn off and on a color bar legend for this option.

The used color scheme is defined by the `colmap` parameter. The default scheme is based on color brewer ([colorbrewer2.org](http://colorbrewer2.org)) diverging scheme with eight colors. There is also a gray scheme (`colmap = "gray"`) and user can define its own just by specifying the needed sequence of colors (e.g. `colmap = c("red", "yellow", "green")`, two colors is minimum). The scheme will then be generated automatically as a gradient among the colors.

Besides that the function allows to change tick values and corresponding tick labels for x and y axis, see Bookdown tutorial for more details.

**Author(s)**

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

**See Also**

[mdaplotg](#) - to make plots for several sets of data objects (groups of objects).

**Examples**

```
# See all examples in the tutorial.
```

---

mdaplot.areColors	<i>Check color values</i>
-------------------	---------------------------

---

**Description**

Checks if elements of argument are valid color values

**Usage**

```
mdaplot.areColors(palette)
```

**Arguments**

palette	vector with possibly color values (names, RGB, etc.)
---------	--

---

mdaplot.formatValues	<i>Format vector with numeric values</i>
----------------------	--

---

**Description**

Format vector with values, so only significant decimal numbers are left.

**Usage**

```
mdaplot.formatValues(data, round.only = F, digits = 3)
```

**Arguments**

data	vector or matrix with values
round.only	logical, do formatting or only round the values
digits	how many significant digits take into account



**Details**

Function takes into account difference between values and the values themselves.

**Value**

matrix with formatted values

---

mdaplot.getColors	<i>Color values for plot elements</i>
-------------------	---------------------------------------

---

**Description**

Generate vector with color values for plot objects (lines, points, bars), depending on number of groups for the objects.

**Usage**

```
mdaplot.getColors(  
  ngroups = NULL,  
  cgroup = NULL,  
  colmap = "default",  
  opacity = 1,  
  maxsplits = 64  
)
```

**Arguments**

ngroups	number of colors to create.
cgroup	vector of values, used for color grouping of plot points or lines.
colmap	which colormap to use ('default', 'gray', 'old', or user defined in form c('col1', 'col2', ...)).
opacity	opacity for colors (between 0 and 1)
maxsplits	if continuous values are used for color grouping - how many groups to create?

**Value**

Returns vector with generated color values

---

mdaplot.getXAxisLim     *Calculate limits for x-axis.*

---

### Description

Calculates limits for x-axis depending on data values that have to be plotted, extra plot elements that have to be shown and margins.

### Usage

```
mdaplot.getXAxisLim(
  ps,
  xlim,
  show.labels = FALSE,
  show.lines = FALSE,
  show.excluded = FALSE,
  bwd = 0.8
)
```

### Arguments

ps	'plotseries' object.
xlim	limits provided by user
show.labels	logical, will data labels be shown on the plot
show.lines	logical or numeric with line coordinates to be shown on the plot.
show.excluded	logical, will excluded values be shown on the plot
bwd	if limits are computed for bar plot, this is a bar width (otherwise NULL)

### Value

Returns a vector with two limits.

---

mdaplot.getXTickLabels  
                           *Prepare xticklabels for plot*

---

### Description

Prepare xticklabels for plot

### Usage

```
mdaplot.getXTickLabels(xticklabels, xticks, excluded_cols)
```

**Arguments**

xticklabels	xticklables provided by user (if any)
xticks	xticks (provided or computed)
excluded_cols	columns excluded from plot data (if any)

---

mdaplot.getXTicks      *Prepare xticks for plot*

---

**Description**

Prepare xticks for plot

**Usage**

```
mdaplot.getXTicks(xticks, xlim, x_values = NULL, type = NULL)
```

**Arguments**

xticks	xticks provided by user (if any)
xlim	limits for x axis
x_values	x values for the plot data object
type	type of the plot

---

mdaplot.getYAxisLim      *Calculate limits for y-axis.*

---

**Description**

Calculates limits for y-axis depending on data values that have to be plotted, extra plot elements that have to be shown and margins.

**Usage**

```
mdaplot.getYAxisLim(
  ps,
  ylim,
  show.lines = FALSE,
  show.excluded = FALSE,
  show.labels = FALSE,
  show.colorbar = FALSE
)
```

**Arguments**

ps	'plotseries' object.
ylim	limits provided by user
show.lines	logical or numeric with line coordinates to be shown on the plot.
show.excluded	logical, will excluded values be shown on the plot
show.labels	logical, will data labels be shown on the plot
show.colorbar	logical, will colorbar be shown on the plot

**Value**

Returns a vector with two limits.

---

mdaplot.getYTickLabels  
*Prepare yticklabels for plot*

---

**Description**

Prepare yticklabels for plot

**Usage**

```
mdaplot.getYTickLabels(yticklabels, yticks, excluded_rows)
```

**Arguments**

yticklabels	yticklables provided by user (if any)
yticks	yticks (provided or computed)
excluded_rows	rows excluded from plot data (if any)

---

mdaplot.getYTicks      *Prepare yticks for plot*

---

**Description**

Prepare yticks for plot

**Usage**

```
mdaplot.getYTicks(yticks, ylim, y_values = NULL, type = NULL)
```

**Arguments**

yticks	yticks provided by user (if any)
ylim	limits for y axis
y_values	y values for the plot data object
type	type of the plot

---

mdaplot.plotAxes	<i>Create axes plane</i>
------------------	--------------------------

---

**Description**

Creates an empty axes plane for given parameters

**Usage**

```
mdaplot.plotAxes(
  xticklabels = NULL,
  yticklabels = NULL,
  xlim = xlim,
  ylim = ylim,
  xticks = NULL,
  yticks = NULL,
  main = NULL,
  xlab = NULL,
  ylab = NULL,
  xlas = 0,
  ylas = 0,
  show.grid = TRUE,
  grid.lwd = 0.5,
  grid.col = "lightgray"
)
```

**Arguments**

xticklabels	labels for x ticks
yticklabels	labels for y ticks
xlim	vector with limits for x axis
ylim	vector with limits for y axis
xticks	values for x ticks
yticks	values for y ticks
main	main title for the plot
xlab	label for x axis
ylab	label for y axis

xlas	orientation of xticklabels
ylas	orientation of yticklabels
show.grid	logical, show or not axes grid
grid.lwd	line thinckness (width) for the grid
grid.col	line color for the grid

---

mdaplot.prepareColors *Prepare colors based on palette and opacity value*

---

### Description

Prepare colors based on palette and opacity value

### Usage

```
mdaplot.prepareColors(palette, ncolors, opacity)
```

### Arguments

palette	vector with main colors for current palette
ncolors	number of colors to generate
opacity	opacity for the colors (one value or individual for each color)

### Value

vector with colors

---

mdaplot.showColorbar *Plot colorbar*

---

### Description

Shows a colorbar if plot has color grouping of elements (points or lines).

### Usage

```
mdaplot.showColorbar(  
  cgroup,  
  colmap = "default",  
  lab.col = "darkgray",  
  lab.cex = 0.65  
)
```

**Arguments**

cgroup	a vector with values used to make color grouping of the elements
colmap	a colormap to be used for color generation
lab.col	color for legend labels
lab.cex	size for legend labels

---

mdaplot.showLines      *Plot lines*

---

**Description**

Shows horisontal and vertical lines on a plot.

**Usage**

```
mdaplot.showLines(point, lty = 2, lwd = 0.75, col = rgb(0.2, 0.2, 0.2))
```

**Arguments**

point	vector with two values: x coordinate for vertical point y for horizontal
lty	line type
lwd	line width
col	color of lines

**Details**

If it is needed to show only one line, the other coordinate shall be set to NA.

---

mdaplotg      *Plotting function for several plot series*

---

**Description**

mdaplotg is used to make different kinds of plots or their combination for several sets of objects.

**Usage**

```

mdaplotg(
  data,
  groupby = NULL,
  type = "p",
  pch = 16,
  lty = 1,
  lwd = 1,
  cex = 1,
  col = NULL,
  bwd = 0.8,
  legend = NULL,
  xlab = NULL,
  ylab = NULL,
  main = NULL,
  labels = NULL,
  ylim = NULL,
  xlim = NULL,
  colmap = "default",
  legend.position = "topright",
  show.legend = TRUE,
  show.labels = FALSE,
  show.lines = FALSE,
  show.grid = TRUE,
  grid.lwd = 0.5,
  grid.col = "lightgray",
  xticks = NULL,
  xticklabels = NULL,
  yticks = NULL,
  yticklabels = NULL,
  show.excluded = FALSE,
  lab.col = "darkgray",
  lab.cex = 0.65,
  xlas = 1,
  ylas = 1,
  opacity = 1,
  ...
)

```

**Arguments**

<code>data</code>	a matrix, data frame or a list with data values (see details below).
<code>groupby</code>	one or several factors used to create groups of data matrix rows (works if data is a matrix)
<code>type</code>	type of the plot ('p', 'l', 'b', 'h', 'e').
<code>pch</code>	a character for markers (same as plot parameter).
<code>lty</code>	the line type (same as plot parameter).



lwd	the line width (thickness) (same as plot parameter).
cex	the cex factor for the markers (same as plot parameter).
col	colors for the plot series
bwd	a width of a bar as a percent of a maximum space available for each bar.
legend	a vector with legend elements (if NULL, no legend will be shown).
xlab	a title for the x axis (same as plot parameter).
ylab	a title for the y axis (same as plot parameter).
main	an overall title for the plot (same as plot parameter).
labels	what to use as labels ('names' - row names, 'indices' - row indices, 'values' - values).
ylim	limits for the y axis (if NULL, will be calculated automatically).
xlim	limits for the x axis (if NULL, will be calculated automatically).
colmap	a colormap to generate colors if col is not provided
legend.position	position of the legend ('topleft', 'topright', 'top', 'bottomleft', 'bottomright', 'bottom').
show.legend	logical, show or not legend for the data objects.
show.labels	logical, show or not labels for the data objects.
show.lines	vector with two coordinates (x, y) to show horizontal and vertical line cross the point.
show.grid	logical, show or not a grid for the plot.
grid.lwd	line thinckness (width) for the grid
grid.col	line color for the grid
xticks	tick values for x axis.
xticklabels	labels for x ticks.
yticks	tick values for y axis.
yticklabels	labels for y ticks.
show.excluded	logical, show or hide rows marked as excluded (attribute 'exclrows')
lab.col	color for data point labels.
lab.cex	size for data point labels.
xlas	orientation of xticklabels
ylas	orientation of yticklabels
opacity	opacity for plot colors (value between 0 and 1)
...	other plotting arguments.

## Details

The `mdaplotg` function is used to make a plot with several sets of objects. Simply speaking, use it when you need a plot with legend. For example to show line plot with spectra from calibration and test set, scatter plot with height and weight values for women and men, and so on.

Most of the parameters are similar to `mdaplot`, the difference is described below.

The data should be organized as a list, every item is a matrix (or data frame) with data for one set of objects. Alternatively you can provide data as a matrix and use parameter `groupby` to create the groups. See tutorial for more details.

There is no color grouping option, because color is used to separate the sets. Marker symbol, line style and type, etc. can be defined as a single value (one for all sets) and as a vector with one value for each set.

## Author(s)

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

---

`mdaplotg.getLegend`      *Create and return vector with legend values*

---

## Description

Create and return vector with legend values

## Usage

```
mdaplotg.getLegend(ps, data.names, legend = NULL)
```

## Arguments

<code>ps</code>	list with plot series
<code>data.names</code>	names of the data sets
<code>legend</code>	legend values provided by user

## Value

vector of text values for the legend

---

mdaplotg.getXLim      *Compute x-axis limits for mdaplotg*

---

**Description**

Compute x-axis limits for mdaplotg

**Usage**

```
mdaplotg.getXLim(  
    ps,  
    xlim,  
    show.excluded,  
    show.legend,  
    show.labels,  
    legend.position,  
    bwd = NULL  
)
```

**Arguments**

ps	list with plotseries
xlim	limits provided by user
show.excluded	logical, will excluded values also be shown
show.legend	will legend be shown on the plot
show.labels	will labels be shown on the plot
legend.position	position of legend on the plot (if shown)
bwd	size of bar for bar plot

**Value**

vector with two values

---

mdaplotg.getYLim      *Compute y-axis limits for mdaplotg*

---

**Description**

Compute y-axis limits for mdaplotg

**Usage**

```
mdaplotg.getYLim(
  ps,
  ylim,
  show.excluded,
  show.legend,
  legend.position,
  show.labels
)
```

**Arguments**

ps	list with plotseries
ylim	limits provided by user
show.excluded	logical, will excluded values also be shown
show.legend	will legend be shown on the plot
legend.position	position of legend on the plot (if shown)
show.labels	logical, will data point labels also be shown

**Value**

vector with two values

---

mdaplotg.prepareData *Prepare data for mdaplotg*

---

**Description**

Prepare data for mdaplotg

**Usage**

```
mdaplotg.prepareData(data, type, groupby)
```

**Arguments**

data	datasets (in form of list, matrix or data frame)
type	vector with type for dataset
groupby	factor or data frame with factors - used to split data matrix into groups

**Value**

list of datasets

The method should prepare data as a list of datasets (matrices or data frames). One list element will be used to create one plot series.

If 'data' is matrix or data frame and not 'groupby' parameter is provided, then every row will be taken as separate set. This option is available only for line or bar plots.

---

mdaplotg.processParam *Check mdaplotg parameters and replicate them if necessary*

---

**Description**

Check mdaplotg parameters and replicate them if necessary

**Usage**

```
mdaplotg.processParam(param, name, is.type, ngroups)
```

**Arguments**

param	A parameter to check
name	name of the parameter (needed for error message)
is.type	function to use for checking parameter type
ngroups	number of groups (plot series)

---

mdaplotg.showLegend *Show legend for mdaplotg*

---

**Description**

Shows a legend for plot elements or their groups.

**Usage**

```
mdaplotg.showLegend(
  legend,
  col,
  pt.bg = NA,
  pch = NULL,
  lty = NULL,
  lwd = NULL,
  cex = 1,
  bty = "o",
  position = "topright",
  plot = TRUE,
  ...
)
```

**Arguments**

legend	vector with text elements for the legend items
col	vector with color values for the legend items
pt.bg	vector with background colors for the legend items (e.g. for pch = 21:25)
pch	vector with marker symbols for the legend items
lty	vector with line types for the legend items
lwd	vector with line width values for the legend items
cex	vector with cex factor for the points
bty	border type for the legend
position	legend position ("topright", "topleft", "bottomright", "bottomleft", "top", "bottom")
plot	logical, show legend or just calculate and return its size
...	other parameters

---

mdaplotyy

*Create line plot with double y-axis*


---

**Description**

mdaplotyy create line plot for two plot series and uses separate y-axis for each.

**Usage**

```
mdaplotyy(
  data,
  type = "l",
  col = mdaplot.getColors(2),
  lty = c(1, 1),
  lwd = c(1, 1),
  pch = (if (type == "b") c(16, 16) else c(NA, NA)),
  cex = 1,
  xlim = NULL,
  ylim = NULL,
  main = attr(data, "name"),
  xlab = attr(data, "xaxis.name"),
  ylab = rownames(data),
  labels = "values",
  show.labels = FALSE,
  lab.cex = 0.65,
  lab.col = "darkgray",
  show.grid = TRUE,
  grid.lwd = 0.5,
  grid.col = "lightgray",
```

```

    xticks = NULL,
    xticklabels = NULL,
    xlas = 0,
    ylas = 0,
    show.legend = TRUE,
    legend.position = "topright",
    legend = ylab,
    ...
)

```

### Arguments

<code>data</code>	a matrix or a data.frame with two rows of values.
<code>type</code>	type of the plot ("l" or "b").
<code>col</code>	a color for markers or lines (same as plot parameter) for each series.
<code>lty</code>	line type for each series (two values)
<code>lwd</code>	line width for each series (two values)
<code>pch</code>	a character for markers (same as plot parameter) for each series (two values).
<code>cex</code>	scale factor for the markers
<code>xlim</code>	limits for the x axis (if NULL, will be calculated automatically).
<code>ylim</code>	limits for the y axis, either list with two vectors (one for each series) or NULL.
<code>main</code>	an overall title for the plot (same as plot parameter).
<code>xlab</code>	a title for the x axis (same as plot parameter).
<code>ylab</code>	a title for each of the two y axis (as a vector of two text values).
<code>labels</code>	a vector with text labels for data points or one of the following: "names", "indices", "values".
<code>show.labels</code>	logical, show or not labels for the data objects.
<code>lab.cex</code>	size for data point labels.
<code>lab.col</code>	color for data point labels.
<code>show.grid</code>	logical, show or not a grid for the plot.
<code>grid.lwd</code>	line thickness (width) for the grid.
<code>grid.col</code>	line color for the grid.
<code>xticks</code>	values for x ticks.
<code>xticklabels</code>	labels for x ticks.
<code>xlas</code>	orientation of xticklabels.
<code>ylas</code>	orientation of yticklabels (will be applied to both y axes).
<code>show.legend</code>	logical show legend with name of each plot series or not
<code>legend.position</code>	position of legend if it must be shown
<code>legend</code>	values for the legend
<code>...</code>	other plotting arguments.

**Details**

This plot has properties both `mdaplot` and `mdaplotg`, so when you specify color, line properties etc. you have to do it for both plot series.

**Author(s)**

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

**See Also**

[mdaplotg](#) - to make plots for several sets of data objects (groups of objects).

**Examples**

```
# See all examples in the tutorial.
```

---

mdatools

*Package for Multivariate Data Analysis (Chemometrics)*

---

**Description**

This package contains classes and functions for most common methods used in Chemometrics. For a complete list of functions, use `library(help = 'mdatools')`.

**Details**

The project is hosted on GitHub (<https://svkucheryavski.github.io/mdatools/>), there you can also find a Bookdown user tutorial explaining most important features of the package. There is also a dedicated YouTube channel (<https://www.youtube.com/channel/UCox0H4utfMq4Flu2kymuyTA>) with introductory Chemometric course with examples based on mdatools functionality.

Every method is represented by two classes: a model class for keeping all parameters and information about the model, and a class for keeping and visualising results of applying the model to particular data values.

Every model class, e.g. `pls`, has all needed functionality implemented as class methods, including model calibration, validation (test set and cross-validation), visualisation of the calibration and validation results with various plots and summary statistics.

So far the following modelling and validation methods are implemented:

<code>pca</code> , <code>pcares</code>	Principal Component Analysis (PCA).
<code>pls</code> , <code>plsres</code>	Partial Least Squares regression (PLS).
<code>simca</code> , <code>simcares</code>	Soft Independent Modelling of Class Analogues (SIMCA)
<code>simcam</code> , <code>simcamres</code>	SIMCA for multiple classes case (SIMCA)
<code>plsda</code> , <code>plsdares</code>	Partial Least Squares Dscriminant Analysis (PLS-DA).
<code>randtest</code>	Randomization test for PLS-regression.
<code>ipls</code>	Interval PLS variable.



<a href="#">mcrals</a>	Multivariate Curve Resolution with Alternating Least Squares.
<a href="#">mcrpure</a>	Multivariate Curve Resolution with Purity approach.
<a href="#">pcv</a>	Procrustes Cross Validation.

Methods for data preprocessing:

<a href="#">prep.autoscale</a>	data mean centering and/or standardization.
<a href="#">prep.savgol</a>	Savitzky-Golay transformation.
<a href="#">prep.snv</a>	Standard normal variate.
<a href="#">prep.msc</a>	Multiplicative scatter correction.
<a href="#">prep.norm</a>	Spectra normalization.
<a href="#">prep.alsbasecorr</a>	Baseline correction with Asymmetric Least Squares.

All plotting methods are based on two functions, [mdaplot](#) and [mdaplotg](#). The functions extend the basic functionality of R plots and allow to make automatic legend and color grouping of data points or lines with colorbar legend, automatically adjust axes limits when several data groups are plotted and so on.

#### Author(s)

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

---

pca

*Principal Component Analysis*

---

#### Description

pca is used to build and explore a principal component analysis (PCA) model.

#### Usage

```
pca(
  x,
  ncomp = min(nrow(x) - 1, ncol(x), 20),
  center = TRUE,
  scale = FALSE,
  exclrows = NULL,
  exclcols = NULL,
  x.test = NULL,
  method = "svd",
  rand = NULL,
  lim.type = "ddmoments",
  alpha = 0.05,
  gamma = 0.01,
  info = ""
)
```

**Arguments**

<code>x</code>	calibration data (matrix or data frame).
<code>ncomp</code>	maximum number of components to calculate.
<code>center</code>	logical, do mean centering of data or not.
<code>scale</code>	logical, do standardization of data or not.
<code>exclrows</code>	rows to be excluded from calculations (numbers, names or vector with logical values)
<code>exclcols</code>	columns to be excluded from calculations (numbers, names or vector with logical values)
<code>x.test</code>	test data (matrix or data frame).
<code>method</code>	method to compute principal components ("svd", "nipals").
<code>rand</code>	vector with parameters for randomized PCA methods (if NULL, conventional PCA is used instead)
<code>lim.type</code>	which method to use for calculation of critical limits for residual distances (see details)
<code>alpha</code>	significance level for extreme limits for T2 and Q distances.
<code>gamma</code>	significance level for outlier limits for T2 and Q distances.
<code>info</code>	a short text with model description.

**Details**

Note, that from v. 0.10.0 cross-validation is no more supported in PCA.

If number of components is not specified, a minimum of number of objects - 1 and number of variables in calibration set is used. One can also specified an optimal number of component, once model is calibrated (`ncomp.selected`). The optimal number of components is used to build a residuals distance plot, as well as for SIMCA classification.

If some of rows of calibration set should be excluded from calculations (e.g. because they are outliers) you can provide row numbers, names, or logical values as parameter `exclrows`. In this case they will be completely ignored we model is calibrated. However, score and residuls distances will be computed for these rows as well and then hidden. You can show them on corresponding plots by using parameter `show.excluded = TRUE`.

It is also possible to exclude selected columns from calculations by provideing parameter `exclcols` in form of column numbers, names or logical values. In this case loading matrix will have zeros for these columns. This allows to compute PCA models for selected variables without removing them physically from a dataset.

Take into account that if you see other packages to make plots (e.g. `ggplot2`) you will not be able to distinguish between hidden and normal objects.

By default loadings are computed for the original dataset using either SVD or NIPALS algorithm. However, for datasets with large number of rows (e.g. hyperspectral images), there is a possibility to run algorithms based on random permutations [1, 2]. In this case you have to define parameter `rand` as a vector with two values: `p` - oversampling parameter and `k` - number of iterations. Usually `rand = c(15, 0)` or `rand = c(5, 1)` are good options, which give quite almost precise solution but much faster.

There are several ways to calculate critical limits for orthogonal (Q, q) and score (T2, h) distances. In `mdatools` you can specify one of the following methods via parameter `lim.type`: "jm" Jackson-Mudholkar approach [3], "chisq" - method based on chi-square distribution [4], "ddmoments" and "ddrobust" - related to data driven method proposed in [5]. The "ddmoments" is based on method of moments for estimation of distribution parameters (also known as "classical" approach) while "ddrobust" is based in robust estimation.

If `lim.type="chisq"` or `lim.type="jm"` is used, only limits for Q-distances are computed based on corresponding approach, limits for T2-distances are computed using Hotelling's T-squared distribution. The methods utilizing the data driven approach calculate limits for combination of the distances bases on chi-square distribution and parameters estimated from the calibration data.

The critical limits are calculated for a significance level defined by parameter 'alpha'. You can also specify another parameter, 'gamma', which is used to calculate acceptance limit for outliers (shown as dashed line on residual distance plot).

You can also recalculate the limits for existent model by using different values for alpha and gamme, without recomputing the model itself. In this case use the following code (it is assumed that you current PCA/SIMCA model is stored in variable `m`): `m = setDistanceLimits(m, lim.type, alpha, gamma)`.

In case of PCA the critical limits are just shown on residual plot as lines and can be used for detection of extreme objects (solid line) and outliers (dashed line). When PCA model is used for classification in SIMCA (see [simca](#)) the limits are also employed for classification of objects.

## Value

Returns an object of `pca` class with following fields:

<code>ncomp</code>	number of components included to the model.
<code>ncomp.selected</code>	selected (optimal) number of components.
<code>loadings</code>	matrix with loading values ( <code>nvar</code> x <code>ncomp</code> ).
<code>eigenvals</code>	vector with eigenvalues for all existent components.
<code>expvar</code>	vector with explained variance for each component (in percent).
<code>cumexpvar</code>	vector with cumulative explained variance for each component (in percent).
<code>T2lim</code>	statistical limit for T2 distance.
<code>Qlim</code>	statistical limit for Q residuals.
<code>info</code>	information about the model, provided by user when build the model.
<code>calres</code>	an object of class <code>pcares</code> with PCA results for a calibration data.
<code>testres</code>	an object of class <code>pcares</code> with PCA results for a test data, if it was provided.

More details and examples can be found in the Bookdown tutorial.

## Author(s)

Sergey Kucheryavskiy ([svkucheryavski@gmail.com](mailto:svkucheryavski@gmail.com))

## References

1. N. Halko, P.G. Martinsson, J.A. Tropp. Finding structure with randomness: probabilistic algorithms for constructing approximate matrix decompositions. *SIAM Review*, 53 (2010) pp. 217-288.
2. S. Kucheryavskiy, Blessing of randomness against the curse of dimensionality, *Journal of Chemometrics*, 32 (2018).
3. J.E. Jackson, *A User's Guide to Principal Components*, John Wiley & Sons, New York, NY (1991).
4. A.L. Pomerantsev, Acceptance areas for multivariate classification derived by projection methods, *Journal of Chemometrics*, 22 (2008) pp. 601-609.
5. A.L. Pomerantsev, O.Ye. Rodionova, Concept and role of extreme objects in PCA/SIMCA, *Journal of Chemometrics*, 28 (2014) pp. 429-438.

## See Also

Methods for `pca` objects:

<code>plot.pca</code>	makes an overview of PCA model with four plots.
<code>summary.pca</code>	shows some statistics for the model.
<code>categorize.pca</code>	categorize data rows as "normal", "extreme" or "outliers".
<code>selectCompNum.pca</code>	set number of optimal components in the model
<code>setDistanceLimits.pca</code>	set critical limits for residuals
<code>predict.pca</code>	applies PCA model to a new data.

Plotting methods for `pca` objects:

<code>plotScores.pca</code>	shows scores plot.
<code>plotLoadings.pca</code>	shows loadings plot.
<code>plotVariance.pca</code>	shows explained variance plot.
<code>plotCumVariance.pca</code>	shows cumulative explained variance plot.
<code>plotResiduals.pca</code>	shows plot for residual distances (Q vs. T2).
<code>plotBiplot.pca</code>	shows bi-plot.
<code>plotExtreme.pca</code>	shows extreme plot.
<code>plotT2DoF</code>	plot with degrees of freedom for score distance.
<code>plotQDoF</code>	plot with degrees of freedom for orthogonal distance.
<code>plotDistDoF</code>	plot with degrees of freedom for both distances.

Most of the methods for plotting data are also available for PCA results (`pcarets`) objects. Also check `pca.mvreplace`, which replaces missing values in a data matrix with approximated using iterative PCA decomposition.

## Examples

```
library(mdatools)
### Examples for PCA class
```

```
## 1. Make PCA model for People data with autoscaling

data(people)
model = pca(people, scale = TRUE, info = "Simple PCA model")
model = selectCompNum(model, 4)
summary(model)
plot(model, show.labels = TRUE)

## 2. Show scores and loadings plots for the model

par(mfrow = c(2, 2))
plotScores(model, comp = c(1, 3), show.labels = TRUE)
plotScores(model, comp = 2, type = "h", show.labels = TRUE)
plotLoadings(model, comp = c(1, 3), show.labels = TRUE)
plotLoadings(model, comp = c(1, 2), type = "h", show.labels = TRUE)
par(mfrow = c(1, 1))

## 3. Show residual distance and variance plots for the model
par(mfrow = c(2, 2))
plotVariance(model, type = "h")
plotCumVariance(model, show.labels = TRUE, legend.position = "bottomright")
plotResiduals(model, show.labels = TRUE)
plotResiduals(model, ncomp = 2, show.labels = TRUE)
par(mfrow = c(1, 1))
```

---

pca.cal

*PCA model calibration*

---

## Description

Calibrates (builds) a PCA model for given data and parameters

## Usage

```
pca.cal(x, ncomp, center, scale, method, rand = NULL)
```

## Arguments

x	matrix with data values
ncomp	number of principal components to calculate
center	logical, do mean centering or not
scale	logical, do standardization or not
method	algorithm for computing PC space (only 'svd' and 'nipals' are supported so far)
rand	vector with parameters for randomized PCA methods (if NULL, conventional PCA is used instead)

**Value**

an object with calibrated PCA model

---

pca.getB	<i>Low-dimensional approximation of data matrix X</i>
----------	---

---

**Description**

Low-dimensional approximation of data matrix X

**Usage**

```
pca.getB(X, k = NULL, rand = NULL, dist = "unif")
```

**Arguments**

X	data matrix
k	rank of X (number of components)
rand	a vector with two values - number of iterations (q) and oversampling parameter (p)
dist	distribution for generating random numbers, 'unif' or 'norm'

---

pca.mvreplace	<i>Replace missing values in data</i>
---------------	---------------------------------------

---

**Description**

pca.mvreplace is used to replace missing values in a data matrix with approximated by iterative PCA decomposition.

**Usage**

```
pca.mvreplace(  
  x,  
  center = TRUE,  
  scale = FALSE,  
  maxncomp = 10,  
  expvarlim = 0.95,  
  covlim = 10^-6,  
  maxiter = 100  
)
```

**Arguments**

x	a matrix with data, containing missing values.
center	logical, do centering of data values or not.
scale	logical, do standardization of data values or not.
maxncomp	maximum number of components in PCA model.
expvarlim	minimum amount of variance, explained by chosen components (used for selection of optimal number of components in PCA models).
covlim	convergence criterion.
maxiter	maximum number of iterations if convergence criterion is not met.

**Details**

The function uses iterative PCA modeling of the data to approximate and impute missing values. The result is most optimal for data sets with low or moderate level of noise and with number of missing values less than 10% for small dataset and up to 20% for large data.

**Value**

Returns the same matrix x where missing values are replaced with approximated.

**Author(s)**

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

**References**

Philip R.C. Nelson, Paul A. Taylor, John F. MacGregor. Missing data methods in PCA and PLS: Score calculations with incomplete observations. *Chemometrics and Intelligent Laboratory Systems*, 35 (1), 1996.

**Examples**

```
library(mdatools)

## A very simple example of imputing missing values in a data with no noise

# generate a matrix with values
s = 1:6
odata = cbind(s, 2*s, 4*s)

# make a matrix with missing values
mdata = odata
mdata[5, 2] = mdata[2, 3] = NA

# replace missing values with approximated
rdata = pca.mvreplace(mdata, scale = TRUE)

# show all matrices together
show(cbind(odata, mdata, round(rdata, 2)))
```

---

pca.nipals                      *NIPALS based PCA algorithm*

---

**Description**

Calculates principal component space using non-linear iterative partial least squares algorithm (NIPALS)

**Usage**

```
pca.nipals(x, ncomp = min(ncol(x), nrow(x) - 1), tol = 10^-10)
```

**Arguments**

x	a matrix with data values (preprocessed)
ncomp	number of components to calculate
tol	tolerance (if difference in eigenvalues is smaller - convergence achieved)

**Value**

a list with scores, loadings and eigenvalues for the components

**References**

Geladi, Paul; Kowalski, Bruce (1986), "Partial Least Squares Regression:A Tutorial", *Analytica Chimica Acta* 185: 1-17

---

pca.run                              *Runs one of the selected PCA methods*

---

**Description**

Runs one of the selected PCA methods

**Usage**

```
pca.run(x, ncomp, method, rand = NULL)
```

**Arguments**

x	data matrix
ncomp	number of components
method	name of PCA methods ('svd', 'nipals')
rand	parameters for randomized algorithm (if not NULL)



---

pca.svd                      *Singular Values Decomposition based PCA algorithm*

---

**Description**

Computes principal component space using Singular Values Decomposition

**Usage**

```
pca.svd(x, ncomp = min(ncol(x), nrow(x) - 1))
```

**Arguments**

x                      a matrix with data values (preprocessed)  
ncomp                  number of components to calculate

**Value**

a list with scores, loadings and eigenvalues for the components

---

pcares                      *Results of PCA decomposition*

---

**Description**

pcares is used to store and visualise results for PCA decomposition.

**Usage**

```
pcares(...)
```

**Arguments**

...                      all arguments supported by ldecomp.

**Details**

In fact pcares is a wrapper for [ldecomp](#) - general class for storing results for linear decomposition  $X = TP' + E$ . So, most of the methods, arguments and returned values are inherited from ldecomp.

There is no need to create a pcares object manually, it is created automatically when build a PCA model (see [pca](#)) or apply the model to a new data (see [predict.pca](#)). The object can be used to show summary and plots for the results.

It is assumed that data is a matrix or data frame with I rows and J columns.

**Value**

Returns an object (list) of class `pcares` and `ldecomp` with following fields:

<code>scores</code>	matrix with score values (I x A).
<code>residuals</code>	matrix with data residuals (I x J).
<code>T2</code>	matrix with score distances (I x A).
<code>Q</code>	matrix with orthogonal distances (I x A).
<code>ncomp.selected</code>	selected number of components.
<code>expvar</code>	explained variance for each component.
<code>cumexpvar</code>	cumulative explained variance.

**See Also**

Methods for `pcares` objects:

<code>print.pcares</code>	shows information about the object.
<code>summary.pcares</code>	shows statistics for the PCA results.

Methods, inherited from `ldecomp` class:

<code>plotScores.ldecomp</code>	makes scores plot.
<code>plotVariance.ldecomp</code>	makes explained variance plot.
<code>plotCumVariance.ldecomp</code>	makes cumulative explained variance plot.
<code>plotResiduals.ldecomp</code>	makes Q vs. T2 distance plot.

Check also [pca](#) and [ldecomp](#).

**Examples**

```
### Examples for PCA results class

library(mdatools)

## 1. Make a model for every odd row of People data
## and apply it to the objects from every even row

data(people)
x = people[seq(1, 32, 2), ]
x.new = people[seq(1, 32, 2), ]

model = pca(people, scale = TRUE, info = "Simple PCA model")
model = selectCompNum(model, 4)

res = predict(model, x.new)
summary(res)
```

```

plot(res)

## 1. Make PCA model for People data with autoscaling
## and full cross-validation and get calibration results

data(people)
model = pca(people, scale = TRUE, info = "Simple PCA model")
model = selectCompNum(model, 4)

res = model$calres
summary(res)
plot(res)

## 2. Show scores plots for the results
par(mfrow = c(2, 2))
plotScores(res)
plotScores(res, cgroup = people[, "Beer"], show.labels = TRUE)
plotScores(res, comp = c(1, 3), show.labels = TRUE)
plotScores(res, comp = 2, type = "h", show.labels = TRUE)
par(mfrow = c(1, 1))

## 3. Show residuals and variance plots for the results
par(mfrow = c(2, 2))
plotVariance(res, type = "h")
plotCumVariance(res, show.labels = TRUE)
plotResiduals(res, show.labels = TRUE, cgroup = people[, "Sex"])
plotResiduals(res, ncomp = 2, show.labels = TRUE)
par(mfrow = c(1, 1))

```

---

pcv

---

*Compute matrix with pseudo-validation set*


---

## Description

The method computes pseudo-validation matrix  $X_{pv}$ , based on PCA decomposition of calibration set  $X$  and systematic (venetian blinds) cross-validation. It is assumed that data rows are ordered correctly, so systematic cross-validation can be applied.

All details can be found in [1]

## Usage

```

pcv(
  x,
  ncomp = min(round(nrow(x)/nseg) - 1, col(x), 20),
  nseg = 4,
  scale = FALSE
)

```

**Arguments**

x	matrix with calibration set (I×J)
ncomp	number of components for PCA decomposition
nseg	number of segments in cross-validation
scale	logical, standardize columns of X prior to decomposition or not

**Value**

Pseudo-validation matrix (I×J)

**References**

1. Kucheryavskiy, S., Zhilin, S., Rodionova, O., & Pomerantsev, A. Procrustes Cross-Validation—A Bridge between Cross-Validation and Independent Validation Sets. *Analytical Chemistry*, 92 (17), 2020. pp.11842–11850. DOI: 10.1021/acs.analchem.0c02175

---

pellets

*Image data*

---

**Description**

Dataset for showing how mdatools works with images. It is an RGB image represented as 3-way array.

**Usage**

```
data(people)
```

**Format**

a 3-way array (height x width x channels).

**Details**

This is an image with pellets of four different colours mixed in a glass volume.

---

people

*People data*

---

### Description

Dataset for exploratory analysis with 32 objects (male and female persons) and 12 variables.

### Usage

```
data(people)
```

### Format

a matrix with 32 observations (persons) and 12 variables.

- [ , 1] Height in cm.
- [ , 2] Weight in kg.
- [ , 3] Hair length (-1 for short, +1 for long).
- [ , 4] Shoe size (EU standard).
- [ , 5] Age, years.
- [ , 6] Income, euro per year.
- [ , 7] Beer consumption, liters per year.
- [ , 8] Wine consumption, liters per year.
- [ , 9] Sex (-1 for male, +1 for female).
- [ , 10] Swimming ability (index, based on 500 m swimming time).
- [ , 11] Region (-1 for Scandinavia, +1 for Mediterranean).
- [ , 12] IQ (European standardized test).

### Details

The data was taken from the book [1] and is in fact a small subset of a pan-European demographic survey. It includes information about 32 persons, 16 represent northern Europe (Scandinavians) and 16 are from the Mediterranean regions. In both groups there are 8 male and 8 female persons. The data includes both quantitative and qualitative variables and is particularly useful for benchmarking exploratory data analysis methods.

### Source

1. K. Esbensen. Multivariate Data Analysis in Practice. Camo, 2002.

---

pinv

*Pseudo-inverse matrix*

---

**Description**

Computes pseudo-inverse matrix using SVD

**Usage**

```
pinv(data)
```

**Arguments**

data                    a matrix with data values to compute inverse for

plot.classres            *Plot function for classification results*

**Description**

Generic plot function for classification results. Alias for [plotPredictions.classres](#).

**Usage**

```
## S3 method for class 'classres'
plot(x, ...)
```

**Arguments**

x                        classification results (object of class plsdare, simcamres, etc.).  
 ...                      other arguments for plotPredictions() method.

plot.ipls                *Overview plot for iPLS results*

**Description**

Shows a plot for iPLS results.

**Usage**

```
## S3 method for class 'ipls'
plot(x, ...)
```

**Arguments**

x                        a (object of class pca)  
 ...                      other arguments

**Details**

See details for [plotSelection.ipls](#).

---

plot.mcr	<i>Plot summary for MCR model</i>
----------	-----------------------------------

---

**Description**

Plot summary for MCR model

**Usage**

```
## S3 method for class 'mcr'
plot(x, ...)
```

**Arguments**

x	mcr model object
...	other parameters

---

plot.pca	<i>Model overview plot for PCA</i>
----------	------------------------------------

---

**Description**

Shows a set of plots (scores, loadings, residuals and explained variance) for PCA model.

**Usage**

```
## S3 method for class 'pca'
plot(
  x,
  comp = c(1, 2),
  ncomp = x$ncomp.selected,
  show.labels = FALSE,
  show.legend = TRUE,
  ...
)
```

**Arguments**

x	a PCA model (object of class pca)
comp	vector with two values - number of components to show the scores and loadings plots for
ncomp	number of components to show the residuals plot for
show.labels	logical, show or not labels for the plot objects
show.legend	logical, show or not a legend on the plot
...	other arguments

**Details**

See examples in help for [pca](#) function.

---

plot.pcares	<i>Plot method for PCA results object</i>
-------------	---

---

**Description**

Show several plots to give an overview about the PCA results

**Usage**

```
## S3 method for class 'pcares'
plot(x, comp = c(1, 2), ncomp = x$ncomp.selected, show.labels = TRUE, ...)
```

**Arguments**

x	PCA results (object of class pcares)
comp	which components to show the scores plot for (can be one value or vector with two values).
ncomp	how many components to use for showing the residual distance plot
show.labels	logical, show or not labels for the plot objects
...	other arguments

---

plot.pls	<i>Model overview plot for PLS</i>
----------	------------------------------------

---

**Description**

Shows a set of plots (x residuals, regression coefficients, RMSE and predictions) for PLS model.

**Usage**

```
## S3 method for class 'pls'
plot(x, ncomp = x$ncomp.selected, ny = 1, show.legend = TRUE, ...)
```

**Arguments**

x	a PLS model (object of class pls)
ncomp	how many components to use (if NULL - user selected optimal value will be used)
ny	which y variable to show the summary for (if NULL, will be shown for all)
show.legend	logical, show or not a legend on the plot
...	other arguments



**Details**

See examples in help for [pls](#) function.

---

plot.plsda	<i>Model overview plot for PLS-DA</i>
------------	---------------------------------------

---

**Description**

Shows a set of plots (x residuals, regression coefficients, misclassification ratio and predictions) for PLS-DA model.

**Usage**

```
## S3 method for class 'plsda'
plot(x, ncomp = x$ncomp.selected, nc = 1, show.legend = TRUE, ...)
```

**Arguments**

x	a PLS-DA model (object of class plsda)
ncomp	how many components to use (if NULL - user selected optimal value will be used)
nc	which class to show the plots
show.legend	logical, show or not a legend on the plot
...	other arguments

**Details**

See examples in help for [plsda](#) function.

---

plot.plsdares	<i>Overview plot for PLS-DA results</i>
---------------	---

---

**Description**

Shows a set of plots (x residuals, y variance, classification performance and predictions) for PLS-DA results.

**Usage**

```
## S3 method for class 'plsdares'
plot(x, nc = 1, ncomp = x$ncomp.selected, show.labels = FALSE, ...)
```

**Arguments**

x	PLS-DA results (object of class plsdares)
nc	which class to show the plot for
ncomp	how many components to use
show.labels	logical, show or not labels for the plot objects
...	other arguments

**Details**

See examples in help for [pls](#) function.

---

plot.plsres

*Overview plot for PLS results*

---

**Description**

Shows a set of plots for PLS results.

**Usage**

```
## S3 method for class 'plsres'
plot(x, ncomp = x$ncomp.selected, ny = 1, show.labels = FALSE, ...)
```

**Arguments**

x	PLS results (object of class plsres)
ncomp	how many components to use (if NULL - user selected optimal value will be used)
ny	which y variable to show the summary for (if NULL, will be shown for all)
show.labels	logical, show or not labels for the plot objects
...	other arguments

**Details**

See examples in help for [plsres](#) function.

---

plot.randtest	<i>Plot for randomization test results</i>
---------------	--

---

**Description**

Makes a bar plot with alpha values for each component.

**Usage**

```
## S3 method for class 'randtest'
plot(x, main = "Alpha", xlab = "Components", ylab = "", ...)
```

**Arguments**

x	results of randomization test (object of class 'randtest')
main	main title for the plot
xlab	label for x axis
ylab	label for y axis
...	other optional arguments

**Details**

See examples in help for [randtest](#) function.

---

plot.regcoeffs	<i>Regression coefficients plot</i>
----------------	-------------------------------------

---

**Description**

Shows plot with regression coefficient values for every predictor variable (x)

**Usage**

```
## S3 method for class 'regcoeffs'
plot(
  x,
  ncomp = 1,
  ny = 1,
  type = (if (x$nvar > 30) "l" else "h"),
  col = c(mdaplot.getColors(1), "lightgray"),
  show.lines = c(NA, 0),
  show.ci = FALSE,
  alpha = 0.05,
  ylab = paste0("Coefficients (", x$respnames[ny], ")"),
  ...
)
```

**Arguments**

x	regression coefficients object (class regcoeffs)
ncomp	number of components to use for creating the plot
ny	index of response variable to make the plot for
type	type of the plot
col	vector with two colors for the plot (one is used to show real coefficient and another one to show confidence intervals)
show.lines	allows to show horizontal line at c(NA, 0)
show.ci	logical, show or not confidence intervals if they are available
alpha	significance level for confidence intervals (a number between 0 and 1, e.g. for 95% alpha = 0.05)
ylab	label for y-axis
...	other arguments for plotting methods (e.g. main, xlab, etc)

---

plot.regres

*Plot method for regression results*


---

**Description**

Plot method for regression results

**Usage**

```
## S3 method for class 'regres'
plot(x, ...)
```

**Arguments**

x	regression results (object of class regres)
...	other arguments

**Details**

This is a shortcut for [plotPredictions.regres](#)

---

plot.simca                    *Model overview plot for SIMCA*

---

**Description**

Shows a set of plots for SIMCA model.

**Usage**

```
## S3 method for class 'simca'  
plot(x, comp = c(1, 2), ncomp = x$ncomp.selected, ...)
```

**Arguments**

x	a SIMCA model (object of class simca)
comp	which components to show on scores and loadings plot
ncomp	how many components to use for residuals plot
...	other arguments

**Details**

See examples in help for [simcam](#) function.

---

plot.simcam                    *Model overview plot for SIMCAM*

---

**Description**

Shows a set of plots for SIMCAM model.

**Usage**

```
## S3 method for class 'simcam'  
plot(x, nc = c(1, 2), ...)
```

**Arguments**

x	a SIMCAM model (object of class simcam)
nc	vector with two values - classes (SIMCA models) to show the plot for
...	other arguments

**Details**

See examples in help for [simcam](#) function.

---

plot.simcamres	<i>Model overview plot for SIMCAM results</i>
----------------	---

---

**Description**

Just shows a prediction plot for SIMCAM results.

**Usage**

```
## S3 method for class 'simcamres'
plot(x, ...)
```

**Arguments**

x	SIMCAM results (object of class simcamres)
...	other arguments

**Details**

See examples in help for [simcamres](#) function.

---

plotBars	<i>Show plot series as bars</i>
----------	---------------------------------

---

**Description**

First row of the data matrix is taken for creating the bar series. In case of barplot color grouping is made based on columns (not rows as for all other plots).

**Usage**

```
plotBars(ps, col = ps$col, bwd = 0.8, border = NA, force.x.values = NA)
```

**Arguments**

ps	'plotseries' object
col	colors of the bars
bwd	width of the bars (as a ratio for max width)
border	color of bar edges
force.x.values	vector with corrected x-values for a bar plot (needed for group plots, do not change manually).

---

`plotBiplot`*Biplot*

---

**Description**

Biplot

**Usage**`plotBiplot(obj, ...)`**Arguments**

<code>obj</code>	a model or result object
<code>...</code>	other arguments

**Details**

Generic function for biplot

---

`plotBiplot.pca`*PCA biplot*

---

**Description**

Shows a biplot for selected components.

**Usage**

```
## S3 method for class 'pca'
plotBiplot(
  obj,
  comp = c(1, 2),
  pch = c(16, NA),
  col = mdaplot.getColors(2),
  main = "Biplot",
  lty = 1,
  lwd = 1,
  show.labels = FALSE,
  show.axes = TRUE,
  show.excluded = FALSE,
  lab.col = adjustcolor(col, alpha.f = 0.5),
  ...
)
```

**Arguments**

obj	a PCA model (object of class pca)
comp	a value or vector with several values - number of components to show the plot for
pch	a vector with two values - markers for scores and loadings
col	a vector with two colors for scores and loadings
main	main title for the plot
lty	line type for loadings
lwd	line width for loadings
show.labels	logical, show or not labels for the plot objects
show.axes	logical, show or not a axes lines crossing origin (0,0)
show.excluded	logical, show or hide rows marked as excluded (attribute 'exclrows')
lab.col	a vector with two colors for scores and loadings labels
...	other plot parameters (see mdaplotg for details)

---

plotConfidenceEllipse *Add confidence ellipse for groups of points on scatter plot*

---

**Description**

The method shows confidence ellipse for groups of points on a scatter plot made using 'mdaplot()' function with 'cgroup' parameter. It will work only if 'cgroup' is a factor.

**Usage**

```
plotConfidenceEllipse(p, conf.level = 0.95, lwd = 1, lty = 1, opacity = 0)
```

**Arguments**

p	plot data returned by function 'mdaplot()'.
conf.level	confidence level to make the ellipse for (between 0 and 1).
lwd	thickness of line used to show the hull.
lty	type of line used to show the hull.
opacity	of opacity is 0 ellipse is transparent otherwise semi-transparent.



**Examples**

```
# adds 90% confidence ellipse with semi-transparent area over two clusters of points

library(mdatools)
data(people)
group <- factor(people[, "Sex"], labels = c("Male", "Female"))

# first make plot and then add confidence ellipse
p <- mdaplot(people, type = "p", cgroup = group)
plotConfidenceEllipse(p, conf.level = 0.90, opacity = 0.2)
```

---

plotContributions      *Plot resolved contributions*

---

**Description**

Plot resolved contributions

**Usage**

```
plotContributions(obj, ...)
```

**Arguments**

obj	object with mcr case
...	other parameters

---

plotContributions.mcr      *Show plot with resolved contributions*

---

**Description**

Show plot with resolved contributions

**Usage**

```
## S3 method for class 'mcr'
plotContributions(
  obj,
  comp = seq_len(obj$ncomp),
  type = "l",
  col = mdaplot.getColors(obj$ncomp),
  ...
)
```

**Arguments**

obj	object of class mcr
comp	vector with number of components to make the plot for
type	type of the plot
col	vector with colors for individual components
...	other parameters suitable for mdaplotg

---

plotConvexHull	<i>Add convex hull for groups of points on scatter plot</i>
----------------	---

---

**Description**

The method shows convex hull for groups of points on a scatter plot made using ‘mdaplot()’ function with ‘cgroup’ parameter. It will work only if ‘cgroup’ is a factor.

**Usage**

```
plotConvexHull(p, lwd = 1, lty = 1, opacity = 0)
```

**Arguments**

p	plot data returned by function ‘mdaplot()’.
lwd	thickness of line used to show the hull.
lty	type of line used to show the hull.
opacity	of opacity is larger than 0 a semi-transparent polygon is shown over points.

**Examples**

```
# adds convex hull with semi-transparent area over two clusters of points

library(mdatools)
data(people)
group <- factor(people[, "Sex"], labels = c("Male", "Female"))

p <- mdaplot(people, type = "p", cgroup = group)
plotConvexHull(p)
```

---

plotCooman

*Cooman's plot*

---

### Description

Cooman's plot

### Usage

```
plotCooman(obj, ...)
```

### Arguments

obj	classification model or result object
...	other arguments

### Details

Generic function for Cooman's plot

---

plotCooman.simcam

*Cooman's plot for SIMCAM model*

---

### Description

Shows a Cooman's plot for a pair of SIMCA models

### Usage

```
## S3 method for class 'simcam'
plotCooman(
  obj,
  nc = c(1, 2),
  res = list(cal = obj$res[["cal"]]),
  groupby = res[[1]]$c.ref,
  main = "Cooman's plot",
  show.limits = TRUE,
  ...
)
```

**Arguments**

obj	a SIMCAM model (object of class simcam)
nc	vector with two values - classes (SIMCA models) to show the plot for
res	list with results to show the plot for
groupby	factor to use for grouping points on the plot
main	title of the plot
show.limits	logical, show or not critical limits
...	other plot parameters (see mdaplotg for details)

**Details**

Cooman's plot shows squared orthogonal distance from data points to two selected SIMCA models as well as critical limits for the distance (optional). In case if critical limits must be shown they are computed using chi-square distribution regardless which type of limits is employed for classification.

If only one result object is provided (e.g. results for calibration set or new predictions), then the points can be color grouped using 'groupby' parameter (by default reference class values are used to make the groups). In case of multiple result objects, the points are color grouped according to the objects (e.g. calibration set and test set).

---

plotCooman.simcamres *Cooman's plot for SIMCAM results*

---

**Description**

Shows a Cooman's plot for a pair of SIMCA models

**Usage**

```
## S3 method for class 'simcamres'
plotCooman(
  obj,
  nc = c(1, 2),
  main = "Cooman's plot",
  cgroup = obj$c.ref,
  show.plot = TRUE,
  ...
)
```

**Arguments**

obj	SIMCAM results (object of class simcamres)
nc	vector with two values - classes (SIMCA models) to show the plot for
main	main plot title
cgroup	vector of values to use for color grouping of plot points
show.plot	logical, show plot or just return plot data
...	other plot parameters (see mdaplotg for details)

**Details**

The plot is similar to [plotCooman.simcam](#) but shows points only for this result object and does not show critical limits (which are part of a model).

---

plotCorr	<i>Correlation plot</i>
----------	-------------------------

---

**Description**

Correlation plot

**Usage**

```
plotCorr(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for correlation plot

---

plotCorr.randtest      *Correlation plot for randomization test results*

---

### Description

Makes a plot with statistic values vs. coefficient of determination between permuted and reference y-values.

### Usage

```
## S3 method for class 'randtest'
plotCorr(
  obj,
  ncomp = obj$ncomp.selected,
  ylim = NULL,
  xlab = expression(r^2),
  ylab = "Test statistic",
  ...
)
```

### Arguments

obj	results of randomization test (object of class 'randtest')
ncomp	number of component to make the plot for
ylim	limits for y axis
xlab	label for x-axis
ylab	label for y-axis
...	other optional arguments

### Details

See examples in help for [randtest](#) function.

---

plotCumVariance      *Variance plot*

---

### Description

Variance plot

### Usage

```
plotCumVariance(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting explained variance for data decomposition

---

```
plotCumVariance.ldecomp
```

*Cumulative explained variance plot*

---

**Description**

Shows a plot with cumulative explained variance vs. number of components.

**Usage**

```
## S3 method for class 'ldecomp'
plotCumVariance(obj, type = "b", labels = "values", show.plot = TRUE, ...)
```

**Arguments**

obj	object of ldecomp class.
type	type of the plot
labels	what to show as labels for plot objects
show.plot	logical, shall plot be created or just plot series object is needed
...	most of graphical parameters from <a href="#">mdaplot</a> function can be used.

---

```
plotCumVariance.mcr Show plot with cumulative explained variance
```

---

**Description**

Show plot with cumulative explained variance

**Usage**

```
## S3 method for class 'mcr'
plotCumVariance(
  obj,
  type = "b",
  labels = "values",
  main = "Cumulative variance",
  xticks = seq_len(obj$ncomp),
  ...
)
```

**Arguments**

obj	object of class mcr
type	type of the plot
labels	what to use as data labels
main	title of the plot
xticks	vector with ticks for x-axis
...	other parameters suitable for mdaplot

---

plotCumVariance.pca     *Cumulative explained variance plot for PCA model*

---

**Description**

Shows a plot with cumulative explained variance for components.

**Usage**

```
## S3 method for class 'pca'
plotCumVariance(obj, legend.position = "bottomright", ...)
```

**Arguments**

obj	a PCA model (object of class pca)
legend.position	position of the legend
...	other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pca](#) function.

---

plotDensity     *Show plot series as density plot (using hex binning)*

---

**Description**

Show plot series as density plot (using hex binning)

**Usage**

```
plotDensity(ps, nbins = 60, colmap = ps$colmap)
```



**Arguments**

ps	'plotseries' object
nbins	number of bins in one dimension
colmap	colormap name or values used to create color gradient

---

```
plotDiscriminationPower
      Discrimination power plot
```

---

**Description**

Discrimination power plot

**Usage**

```
plotDiscriminationPower(obj, ...)
```

**Arguments**

obj	a model object
...	other arguments

**Details**

Generic function for plotting discrimination power values for classification model

---

```
plotDiscriminationPower.simcam
      Discrimination power plot for SIMCAM model
```

---

**Description**

Shows a plot with discrimination power of predictors for a pair of SIMCA models

**Usage**

```
## S3 method for class 'simcam'
plotDiscriminationPower(
  obj,
  nc = c(1, 2),
  type = "h",
  main = paste0("Discrimination power: ", obj$classnames[nc[1]], " vs. ",
    obj$classname[nc[2]]),
  xlab = attr(obj$dispower, "xaxis.name"),
  ylab = "",
  ...
)
```

**Arguments**

obj	a SIMCAM model (object of class simcam)
nc	vector with two values - classes (SIMCA models) to show the plot for
type	type of the plot
main	main plot title
xlab	label for x axis
ylab	label for y axis
...	other plot parameters (see mdaplotg for details)

**Details**

Discrimination power shows an ability of variables to separate classes. The power is computed similar to model distance, using variance of residuals. However in this case instead of sum the variance across all variables, we take the ratio separately for individual variables.

Discrimination power equal or above 3 is considered as high.

---

plotDistDoF	<i>Degrees of freedom plot for both distances</i>
-------------	---

---

**Description**

Shows a plot with degrees of freedom computed for score and orthogonal distances at given number of components using data driven approach ("ddmoments" or "ddrobust").

**Usage**

```
plotDistDoF(
  obj,
  type = "b",
  labels = "values",
  xticks = seq_len(obj$ncomp),
  ...
)
```

**Arguments**

obj	a PCA model (object of class pca)
type	type of the plot ("b", "l", "h")
labels	what to show as data points labels
xticks	vector with tick values for x-axis
...	other plot parameters (see mdaplotg for details)

**Details**

Work only if parameter `lim.type` equal to "ddmoments" or "ddrobust".

---

plotErrorbars	<i>Show plot series as error bars</i>
---------------	---------------------------------------

---

**Description**

It is assumed that first row of dataset contains the y-coordinates of points, second rows contains size of lower error bar and third - size for upper error bar. If only two rows are provided it is assumed that error bars are symmetric.

**Usage**

```
plotErrorbars(ps, col = ps$col, pch = 16, lwd = 1, cex = 1, ...)
```

**Arguments**

ps	'plotseries' object
col	color for the error bars
pch	marker symbol for the plot
lwd	line width for the error bars
cex	scale factor for the marker
...	other arguments for function 'points()'.

---

plotExtreme	<i>Shows extreme plot for SIMCA model</i>
-------------	---

---

**Description**

Generic function for creating extreme plot for SIMCA model

**Usage**

```
plotExtreme(obj, ...)
```

**Arguments**

obj	a SIMCA model
...	other parameters

---

plotExtreme.pca      *Extreme plot*

---

### Description

Shows a plot with number of expected vs. number of observed extreme objects for different significance levels (alpha values)

### Usage

```
## S3 method for class 'pca'
plotExtreme(
  obj,
  res = obj$res[["cal"]],
  comp = obj$ncomp.selected,
  main = "Extreme plot",
  xlab = "Expected",
  ylab = "Observed",
  pch = rep(21, length(comp)),
  bg = mdaplot.getColors(length(comp)),
  col = rep("white", length(comp)),
  lwd = ifelse(pch %in% 21:25, 0.25, 1),
  cex = rep(1.2, length(comp)),
  ellipse.col = "#cceedf",
  legend.position = "bottomright",
  ...
)
```

### Arguments

obj	a PCA model (object of class pca)
res	object with PCA results to show the plot for (e.g. calibration, test, etc)
comp	vector, number of components to show the plot for
main	plot title
xlab	label for x-axis
ylab	label for y-axis
pch	vector with values for pch parameter for each number of components
bg	vector with background color values for series of points (if pch=21:25)
col	vector with color values for series of points
lwd	line width for point symbols
cex	scale factor for data points
ellipse.col	color for tolerance ellipse
legend.position	position of the legend
...	other arguments

---

plotHist	<i>Statistic histogram</i>
----------	----------------------------

---

**Description**

Statistic histogram

**Usage**

```
plotHist(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting statistic histogram plot

---

plotHist.randtest	<i>Histogram plot for randomization test results</i>
-------------------	--

---

**Description**

Makes a histogram for statistic values distribution for particular component, also show critical value as a vertical line.

**Usage**

```
## S3 method for class 'randtest'  
plotHist(obj, ncomp = obj$ncomp.selected, bwd = 0.9, ...)
```

**Arguments**

obj	results of randomization test (object of class 'randtest')
ncomp	number of component to make the plot for
bwd	width of bars (between 0 and 1)
...	other optional arguments

**Details**

See examples in help for [randtest](#) function.

---

plotHotellingEllipse *Hotelling ellipse*

---

### Description

Add Hotelling ellipse to a scatter plot

### Usage

```
plotHotellingEllipse(p, conf.lim = 0.95, col = "#a0a0a0", lty = 3, ...)
```

### Arguments

p	plot series (e.g. from PCA scores plot)
conf.lim	confidence limit
col	color of the ellipse line
lty	line type (e.g. 1 for solid, 2 for dashed, etc.)
...	any argument suitable for lines function

### Details

The method is created to be used with PCA and PLS scores plots, so it shows the statistical limits computed using Hotelling  $T^2$  distribution in form of ellipse. The function works similar to [plotConvexHull](#) and [plotConfidenceEllipse](#) but does not require grouping of data points. Can be used together with functions [plotScores.pca](#), [plotScores.ldecomp](#), [plotXScores.pls](#), [plotXScores.plsres](#).

See examples for more details.

### Examples

```
# create PCA model for People data
data(people)
m <- pca(people, 4, scale = TRUE)

# make scores plot and show Hotelling ellipse with default settings
p <- plotScores(m, xlim = c(-8, 8), ylim = c(-8, 8))
plotHotellingEllipse(p)

# make scores plot and show Hotelling ellipse with manual settings
p <- plotScores(m, xlim = c(-8, 8), ylim = c(-8, 8))
plotHotellingEllipse(p, conf.lim = 0.99, col = "red")

# in case if you have both calibration and test set, 'plotScores()' returns
# plot series data for both, so you have to subset it and take the first series
# (calibration set) as shown below.
ind <- seq(1, 32, by = 4)
```

```
xc <- people[-ind, , drop = FALSE]
xt <- people[ind, , drop = FALSE]
m <- pca(xc, 4, scale = TRUE, x.test = xt)

p <- plotScores(m, xlim = c(-8, 8), ylim = c(-8, 8))
plotHotellingEllipse(p[[1]])
```

---

plotLines                      *Show plot series as set of lines*

---

### Description

Show plot series as set of lines

### Usage

```
plotLines(
  ps,
  col = ps$col,
  lty = 1,
  lwd = 1,
  cex = 1,
  col.excluded = "darkgray",
  show.excluded = FALSE,
  ...
)
```

### Arguments

ps	‘plotseries’ object
col	a color for markers or lines (same as plot parameter).
lty	line type
lwd	line width
cex	scale factor for the marker
col.excluded	color for the excluded lines.
show.excluded	logical, show or not the excluded data points
...	other arguments for function ‘lines()’.

plotLoadings      *Loadings plot*

---

**Description**

Loadings plot

**Usage**

```
plotLoadings(obj, ...)
```

**Arguments**

obj                  a model or result object  
...                  other arguments

**Details**

Generic function for plotting loadings values for data decomposition

---

plotLoadings.pca      *Loadings plot for PCA model*

---

**Description**

Shows a loadings plot for selected components.

**Usage**

```
## S3 method for class 'pca'  
plotLoadings(  
  obj,  
  comp = c(1, 2),  
  type = (if (length(comp) == 2) "p" else "l"),  
  show.legend = TRUE,  
  show.axes = TRUE,  
  ...  
)
```



**Arguments**

obj	a PCA model (object of class pca)
comp	a value or vector with several values - number of components to show the plot for
type	type of the plot ('b', 'l', 'h')
show.legend	logical, show or not a legend on the plot
show.axes	logical, show or not a axes lines crossing origin (0,0)
...	other plot parameters (see <code>mdaplotg</code> for details)

**Details**

See examples in help for [pca](#) function.

---

plotMisclassified      *Misclassification ratio plot*

---

**Description**

Misclassification ratio plot

**Usage**

```
plotMisclassified(obj, ...)
```

**Arguments**

obj	a model or a result object
...	other arguments

**Details**

Generic function for plotting missclassification values for classification model or results

plotMisclassified.classmodel

*Misclassified ratio plot for classification model*

---

### Description

Makes a plot with misclassified ratio values vs. model complexity (e.g. number of components)

### Usage

```
## S3 method for class 'classmodel'  
plotMisclassified(obj, ...)
```

### Arguments

obj                    classification model (object of class `plsda`, `simca`, etc.).  
...                    parameters for `plotPerformance.classmodel` function.

### Details

See examples in description of `plsda`, `simca` or `simcam`.

---

plotMisclassified.classres

*Misclassified ratio plot for classification results*

---

### Description

Makes a plot with ms ratio values vs. model complexity (e.g. number of components) for classification results.

### Usage

```
## S3 method for class 'classres'  
plotMisclassified(obj, ...)
```

### Arguments

obj                    classification results (object of class `plsdares`, `simcamres`, etc.).  
...                    other parameters for `plotPerformance.classres`

### Details

See examples in description of `plsdares`, `simcamres`, etc.

---

plotModelDistance      *Model distance plot*

---

**Description**

Model distance plot

**Usage**

```
plotModelDistance(obj, ...)
```

**Arguments**

obj	a model object
...	other arguments

**Details**

Generic function for plotting distance from object to a multivariate model

---

plotModelDistance.simcam  
*Model distance plot for SIMCAM model*

---

**Description**

Shows a plot with distance between one SIMCA model to others.

**Usage**

```
## S3 method for class 'simcam'  
plotModelDistance(  
  obj,  
  nc = 1,  
  type = "h",  
  xticks = seq_len(obj$nclasses),  
  xticklabels = obj$classnames,  
  main = paste0("Model distance (", obj$classnames[nc], ")"),  
  xlab = "Models",  
  ylab = "",  
  ...  
)
```

**Arguments**

obj	a SIMCAM model (object of class simcam)
nc	one value - number of class (SIMCA model) to show the plot for
type	type of the plot ("h", "l" or "b")
xticks	vector with tick values for x-axis
xticklabels	vector with tick labels for x-axis
main	main plot title
xlab	label for x axis
ylab	label for y axis
...	other plot parameters (see mdaPlotg for details)

**Details**

The plot shows similarity between a selected model and the others as a ratio of residual variance using the following algorithm. Let's take two SIMCA/PCA models, m1 and m2, which have optimal number of components A1 and A2. The models have been calibrated using calibration sets X1 and X2 with number of rows n1 and n2. Then we do the following:

1. Project X2 to model m1 and compute residuals, E12
2. Compute variance of the residuals as  $s12 = \text{sum}(E12^2) / n1$
3. Project X1 to model m2 and compute residuals, E21
4. Compute variance of the residuals as  $s21 = \text{sum}(E21^2) / n2$
5. Compute variance of residuals for m1 as  $s1 = \text{sum}(E1^2) / (n1 - A1 - 1)$
6. Compute variance of residuals for m2 as  $s2 = \text{sum}(E2^2) / (n2 - A2 - 1)$

The model distance then can be computed as:  $d = \sqrt{(s12 + s21) / (s1 + s2)}$

As one can see, if the two models and corresponding calibration sets are identical, then the distance will be  $\sqrt{(n - A - 1) / n}$ . For example, if  $n = 25$  and  $A = 2$ , then the distance between the model and itself is  $\sqrt{(22/25)} = \sqrt{0.88} = 0.938$ . This case is demonstrated in the example section.

In general, if distance between models is below one classes are overlapping. If it is above 3 the classes are well separated.

**Examples**

```
# create two calibration sets with n = 25 objects in each
data(iris)
x1 <- iris[1:25, 1:4]
x2 <- iris[51:75, 1:4]

# create two SIMCA models with A = 2
m1 <- simca(x1, 'setosa', ncomp = 2)
m2 <- simca(x2, 'versicolor', ncomp = 2)

# combine the models into SIMCAM class
m <- simcam(list(m1, m2))
```

```
# show the model distance plot with distance values as labels
# note, that distance between setosa and setosa is 0.938
plotModelDistance(m, show.labels = TRUE, labels = "values")
```

---

plotModellingPower      *Modelling power plot*

---

**Description**

Modelling power plot

**Usage**

```
plotModellingPower(obj, ...)
```

**Arguments**

obj	a model object
...	other arguments

**Details**

Generic function for plotting modelling power values for classification model

---

plotPerformance      *Classification performance plot*

---

**Description**

Classification performance plot

**Usage**

```
plotPerformance(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting classification performance for model or results

---

`plotPerformance.classmodel`*Performance plot for classification model*

---

## Description

Makes a plot with sensitivity values vs. model complexity (e.g. number of components)

## Usage

```
## S3 method for class 'classmodel'
plotPerformance(
  obj,
  nc = 1,
  param = "misclassified",
  type = "b",
  labels = "values",
  ylab = "",
  ylim = c(0, 1.15),
  xticks = seq_len(dim(obj$res$cal$c.pred)[2]),
  res = obj$res,
  ...
)
```

## Arguments

<code>obj</code>	classification model (object of class <code>plsda</code> , <code>simca</code> , etc.).
<code>nc</code>	class number to make the plot for.
<code>param</code>	which parameter to make the plot for ("specificity", "sensitivity", or "misclassified")
<code>type</code>	type of the plot
<code>labels</code>	what to show as labels for plot objects.
<code>ylab</code>	label for y axis
<code>ylim</code>	vector with two values - limits for y axis
<code>xticks</code>	vector with tick values for x-axis
<code>res</code>	list with result objects to show the plot for
<code>...</code>	most of the graphical parameters from <code>mdaplotg</code> function can be used.

---

plotPerformance.classres

*Performance plot for classification results*


---

### Description

Makes a plot with classification performance parameters vs. model complexity (e.g. number of components) for classification results.

### Usage

```
## S3 method for class 'classres'
plotPerformance(
  obj,
  nc = 1,
  type = "b",
  param = c("sensitivity", "specificity", "misclassified"),
  labels = "values",
  ylab = "",
  ylim = c(0, 1.1),
  xticks = seq_len(obj$ncomp),
  show.plot = TRUE,
  ...
)
```

### Arguments

obj	classification results (object of class <code>plsdare</code> s, <code>simcamres</code> , etc.).
nc	if there are several classes, which class to make the plot for.
type	type of the plot
param	which performance parameter to make the plot for (can be a vector with several values).
labels	what to show as labels for plot objects.
ylab	label for y axis
ylim	vector with two values - limits for y axis
xticks	vector with x-axis tick values
show.plot	logical, shall plot be created or just plot series object is needed
...	most of the graphical parameters from <code>mdaplot</code> function can be used.

### Details

See examples in description of [plsdare](#)s, [simcamres](#), etc.

---

plotPointsShape	<i>Add confidence ellipse or convex hull for group of points</i>
-----------------	--

---

**Description**

Add confidence ellipse or convex hull for group of points

**Usage**

```
plotPointsShape(p, lwd, lty, opacity, shape_function, ...)
```

**Arguments**

p	plot data returned by function 'mdaplot()'
lwd	thickness of line used to show the hull
lty	type of line used to show the hull
opacity	of opacity is larger than 0 a semi-transparent polygon is shown over points
shape_function	function which calculates and return coordinates of the shape
...	extra parameters for shape_function

---

plotPredictions	<i>Predictions plot</i>
-----------------	-------------------------

---

**Description**

Predictions plot

**Usage**

```
plotPredictions(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting predicted values for classification or regression model or results



---

```
plotPredictions.classmodel
```

*Predictions plot for classification model*

---

### Description

Makes a plot with class predictions for a classification model.

### Usage

```
## S3 method for class 'classmodel'
plotPredictions(
  obj,
  res.name = NULL,
  nc = seq_len(obj$nclasses),
  ncomp = NULL,
  main = NULL,
  ...
)
```

### Arguments

<code>obj</code>	a classification model (object of class <code>simca</code> , <code>plsda</code> , etc.). if <code>NULL</code> value is specified, the result will be selected automatically by checking the nearest available from test, cv and calibration results.
<code>res.name</code>	name of result object to make the plot for ("test", "cv" or "cal").
<code>nc</code>	vector with class numbers to make the plot for.
<code>ncomp</code>	what number of components to make the plot for.
<code>main</code>	title of the plot (if <code>NULL</code> will be set automatically)
<code>...</code>	most of the graphical parameters from <a href="#">mdaplotg</a> function can be used.

### Details

See examples in description of [plsda](#), [simca](#) or [simcam](#).

---

```
plotPredictions.classres
```

*Prediction plot for classification results*

---

### Description

Makes a plot with predicted class values for classification results.

**Usage**

```
## S3 method for class 'classres'
plotPredictions(
  obj,
  nc = seq_len(obj$nclasses),
  ncomp = obj$ncomp.selected,
  ylab = "",
  show.plot = TRUE,
  ...
)
```

**Arguments**

obj	classification results (object of class <code>plsdares</code> , <code>simcamres</code> , etc.).
nc	vector with classes to show predictions for.
ncomp	model complexity (number of components) to make the plot for.
ylab	label for y axis
show.plot	logical, shall plot be created or just plot series object is needed
...	most of the graphical parameters from <code>mdaplotg</code> or <code>mdaplot</code> function can be used.

**Details**

See examples in description of `plsdares`, `simcamres`, etc.

---

plotPredictions.regmodel

*Predictions plot for regression model*

---

**Description**

Shows plot with predicted vs. reference (measured) y values for selected components.

**Usage**

```
## S3 method for class 'regmodel'
plotPredictions(
  obj,
  ncomp = obj$ncomp.selected,
  ny = 1,
  legend.position = "topleft",
  show.line = TRUE,
  res = obj$res,
  ...
)
```

**Arguments**

obj	a regression model (object of class regmodel)
ncomp	how many components to use (if NULL - user selected optimal value will be used)
ny	number of response variable to make the plot for (if y is multivariate)
legend.position	position of legend on the plot (if shown)
show.line	logical, show or not line fit for the plot points
res	list with result objects
...	other plot parameters (see mdaPlotg for details)

---

 plotPredictions.regres

*Predictions plot for regression results*


---

**Description**

Shows plot with predicted y values.

**Usage**

```
## S3 method for class 'regres'
plotPredictions(
  obj,
  ny = 1,
  ncomp = obj$ncomp.selected,
  show.line = TRUE,
  show.stat = FALSE,
  stat.col = "#606060",
  stat.cex = 0.85,
  xlim = NULL,
  ylim = NULL,
  axes.equal = TRUE,
  show.plot = TRUE,
  ...
)
```

**Arguments**

obj	regression results (object of class regres)
ny	number of predictor to show the plot for (if y is multivariate)
ncomp	complexity of model (e.g. number of components) to show the plot for
show.line	logical, show or not line fit for the plot points
show.stat	logical, show or not legend with statistics on the plot

stat.col	color of text in legend with statistics
stat.cex	size of text in legend with statistics
xlim	limits for x-axis (if NULL will be computed automatically)
ylim	limits for y-axis (if NULL will be computed automatically)
axes.equal	logical, make limits for x and y axes equal or not
show.plot	logical, show plot or just return plot data
...	other plot parameters (see mdaplot for details)

### Details

If reference values are available, the function shows a scatter plot with predicted vs. reference values, otherwise predicted values are shown vs. object numbers.

---

plotPredictions.simcam

*Predictions plot for SIMCAM model*

---

### Description

Makes a plot with class predictions for calibration dataset.

### Usage

```
## S3 method for class 'simcam'
plotPredictions(
  obj,
  nc = seq_len(obj$nclasses),
  main = "SIMCAM Predictions (cal)",
  ...
)
```

### Arguments

obj	a SIMCAM model (object of class simcam)
nc	vector with class numbers to make the plot for.
main	plot title.
...	most of the graphical parameters from <a href="#">mdaplotg</a> function can be used.

### Details

See examples in description of [plsda](#), [simca](#) or [simcam](#).

---

`plotPredictions.simcamres`*Prediction plot for SIMCAM results*

---

**Description**

Makes a plot with predicted class values for classification results.

**Usage**

```
## S3 method for class 'simcamres'  
plotPredictions(obj, nc = seq_len(obj$nclasses), main = "Predictions", ...)
```

**Arguments**

<code>obj</code>	classification results (object of class <code>plsdares</code> , <code>simcamres</code> , etc.).
<code>nc</code>	vector with classes to show predictions for.
<code>main</code>	title of the plot
<code>...</code>	most of the graphical parameters from <code>mdaplotg</code> or <code>mdaplot</code> function can be used.

**Details**

See examples in description of [plsdares](#), [simcamres](#), etc.

---

`plotProbabilities`*Plot for class belonging probability*

---

**Description**

Makes a plot with class belonging probabilities for each object of the classification results. Works only with classification methods, which compute this probability (e.g. SIMCA).

**Usage**

```
plotProbabilities(obj, ...)
```

**Arguments**

<code>obj</code>	an object with classification results (e.g. SIMCA)
<code>...</code>	other parameters

---

```
plotProbabilities.classes
```

*Plot for class belonging probability*

---

### Description

Makes a plot with class belonging probabilities for each object of the classification results. Works only with classification methods, which compute this probability (e.g. SIMCA).

### Usage

```
## S3 method for class 'classes'
plotProbabilities(
  obj,
  ncomp = obj$ncomp.selected,
  nc = 1,
  type = "h",
  ylim = c(0, 1.1),
  show.lines = c(NA, 0.5),
  ...
)
```

### Arguments

obj	classification results (e.g. object of class simcamres).
ncomp	number of components to use the probabilities for.
nc	if there are several classes, which class to make the plot for.
type	type of the plot
ylim	vector with limits for y-axis
show.lines	shows a horizontal line at $p = 0.5$
...	most of the graphical parameters from <a href="#">mdaplot</a> function can be used.

---

```
plotPurity
```

*Plot purity values*

---

### Description

Plot purity values

### Usage

```
plotPurity(obj, ...)
```

**Arguments**

obj	object with mcr pure case
...	other parameters

---

plotPurity.mcrpure	<i>Purity values plot</i>
--------------------	---------------------------

---

**Description**

Purity values plot

**Usage**

```
## S3 method for class 'mcrpure'
plotPurity(
  obj,
  xticks = seq_len(obj$ncomp),
  type = "h",
  labels = "values",
  ...
)
```

**Arguments**

obj	mcrpure object
xticks	ticks for x axis
type	type of the plot
labels	what to use as data labels
...	other parameters suitable for mdaplot

The plot shows largest weighted purity value for each component graphically.

---

plotPuritySpectra	<i>Plot purity spectra</i>
-------------------	----------------------------

---

**Description**

Plot purity spectra

**Usage**

```
plotPuritySpectra(obj, ...)
```

**Arguments**

obj	object with mcr pure case
...	other parameters

---

```
plotPuritySpectra.mcrpure
      Purity spectra plot
```

---

## Description

Purity spectra plot

## Usage

```
## S3 method for class 'mcrpure'
plotPuritySpectra(
  obj,
  comp = seq_len(obj$ncomp),
  type = "l",
  col = mdaplot.getColors(obj$ncomp),
  show.lines = TRUE,
  lines.col = adjustcolor(col, alpha.f = 0.75),
  lines.lty = 3,
  lines.lwd = 1,
  ...
)
```

## Arguments

<code>obj</code>	mcrpure object
<code>comp</code>	vector of components to show the purity spectra for
<code>type</code>	type of the plot
<code>col</code>	colors for the plot (should be a vector with one value for each component in <code>obj</code> )
<code>show.lines</code>	if TRUE show the selected pure variables as vertical lines
<code>lines.col</code>	color for the selected pure variable lines (by default same as for plots but semi-transparent)
<code>lines.lty</code>	line type for the purity lines
<code>lines.lwd</code>	line width for the purity lines
<code>...</code>	other parameters suitable for <code>mdaplotg</code>

The plot shows weighted purity value of each variable separately for each specified component.



---

plotQDoF	<i>Degrees of freedom plot for orthogonal distance (Nh)</i>
----------	---

---

**Description**

Shows a plot with degrees of freedom computed for score distances at given number of components using data driven approach ("ddmoments" or "ddrobust").

**Usage**

```
plotQDoF(
  obj,
  type = "b",
  labels = "values",
  xticks = seq_len(obj$ncomp),
  ylab = "Nq",
  ...
)
```

**Arguments**

obj	a PCA model (object of class pca)
type	type of the plot ("b", "l", "h")
labels	what to show as data points labels
xticks	vector with tick values for x-axis
ylab	label for y-axis
...	other plot parameters (see mdaplotg for details)

**Details**

Work only if parameter `lim.type` equal to "ddmoments" or "ddrobust".

---

plotRegcoeffs	<i>Regression coefficients plot</i>
---------------	-------------------------------------

---

**Description**

Regression coefficients plot

**Usage**

```
plotRegcoeffs(obj, ...)
```

**Arguments**

obj            a model or result object  
 ...            other arguments

**Details**

Generic function for plotting regression coefficients values for a regression model

---

plotRegcoeffs.regmodel

*Regression coefficient plot for regression model*

---

**Description**

Shows plot with regression coefficient values. Is a proxy for `link{plot.regcoeffs}` method.

**Usage**

```
## S3 method for class 'regmodel'
plotRegcoeffs(obj, ncomp = obj$ncomp.selected, ...)
```

**Arguments**

obj            a regression model (object of class regmodel)  
 ncomp         number of components to show the plot for  
 ...            other plot parameters (see `link{plot.regcoeffs}` for details)

---

plotRegressionLine    *Add regression line for data points*

---

**Description**

Shows linear fit line for data points.

**Usage**

```
plotRegressionLine(p, col = p$col, ...)
```

**Arguments**

p              plot data returned by function 'mdaplot()'  
 col            color of line  
 ...            other parameters available for 'abline()' function

---

plotResiduals	<i>Residuals plot</i>
---------------	-----------------------

---

**Description**

Residuals plot

**Usage**

```
plotResiduals(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting residual values for data decomposition

---

plotResiduals.ldecomp	<i>Residual distance plot</i>
-----------------------	-------------------------------

---

**Description**

Shows a plot with orthogonal (Q, q) vs. score (T2, h) distances for data objects.

**Usage**

```
## S3 method for class 'ldecomp'  
plotResiduals(  
  obj,  
  ncomp = obj$ncomp.selected,  
  norm = FALSE,  
  log = FALSE,  
  show.labels = FALSE,  
  labels = "names",  
  show.plot = TRUE,  
  ...  
)
```

**Arguments**

obj	object of ldecomp class.
ncomp	number of components to show the plot for (if NULL, selected by model value will be used).
norm	logical, normalize distance values or not (see details)
log	logical, apply log transformation to the distances or not (see details)
show.labels	logical, show or not labels for the plot objects
labels	what to show as labels if necessary
show.plot	logical, shall plot be created or just plot series object is needed
...	most of graphical parameters from <code>mdaplot</code> function can be used.

---

plotResiduals.pca	<i>Residuals distance plot for PCA model</i>
-------------------	--

---

**Description**

Shows a plot with score (T2, h) vs orthogonal (Q, q) distances and corresponding critical limits for given number of components.

**Usage**

```
## S3 method for class 'pca'
plotResiduals(
  obj,
  ncomp = obj$ncomp.selected,
  log = FALSE,
  norm = TRUE,
  cgroup = NULL,
  xlim = NULL,
  ylim = NULL,
  show.limits = TRUE,
  lim.col = c("darkgray", "darkgray"),
  lim.lwd = c(1, 1),
  lim.lty = c(2, 3),
  res = obj$res,
  show.legend = TRUE,
  ...
)
```

**Arguments**

obj	a PCA model (object of class pca)
ncomp	how many components to use (by default optimal value selected for the model will be used)

log	logical, apply log transformation to the distances or not (see details)
norm	logical, normalize distance values or not (see details)
cgroup	color grouping of plot points (works only if one result object is available)
xlim	limits for x-axis
ylim	limits for y-axis
show.limits	logical, show or not lines/curves with critical limits for the distances
lim.col	vector with two values - line color for extreme and outlier limits
lim.lwd	vector with two values - line width for extreme and outlier limits
lim.lty	vector with two values - line type for extreme and outlier limits
res	list with result objects to show the plot for (by default, model results are used)
show.legend	logical, show or not a legend on the plot (needed if several result objects are available)
...	other plot parameters (see mdaPlotg for details)

### Details

The function is a bit more advanced version of [plotResiduals.ldecomp](#). It allows to show distance values for several result objects (e.g. calibration and test set or calibration and new prediction set) as well as display the corresponding critical limits in form of lines or curves.

Depending on how many result objects your model has or how many you specified manually, using the `res` parameter, the plot behaves in a bit different way.

If only one result object is provided, then it allows to colorise the points using `cgroup` parameter. If you specify `cgroup = "categories"` then it will show points as three groups: normal, extreme and outliers. If two or more result objects are provided, then the function show distances in groups, and adds corresponding legend.

The function can show distance values normalised ( $h/h_0$  and  $q/q_0$ ) as well as with log transformation ( $\log(1 + h/h_0)$ ,  $\log(1 + q/q_0)$ ). The latter is useful if distribution of the points is skewed and most of them are densely located around bottom left corner.

See examples in help for [pca](#) function.

---

plotResiduals.regres    *Residuals plot for regression results*

---

### Description

Shows plot with Y residuals (difference between predicted and reference values) for selected response variable and complexity (number of components).

**Usage**

```
## S3 method for class 'regres'
plotResiduals(
  obj,
  ny = 1,
  ncomp = obj$ncomp.selected,
  show.lines = c(NA, 0),
  show.plot = TRUE,
  ...
)
```

**Arguments**

obj	regression results (object of class regres)
ny	number of predictor to show the plot for (if y is multivariate)
ncomp	complexity of model (e.g. number of components) to show the plot for
show.lines	allows to show the horizontal line at $y = 0$
show.plot	logical, show plot or just return plot data
...	other plot parameters (see mdaplot for details)

---

plotRMSE

*RMSE plot*


---

**Description**

RMSE plot

**Usage**

```
plotRMSE(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting RMSE values vs. complexity of a regression model

---

plotRMSE.ipls	<i>RMSE development plot</i>
---------------	------------------------------

---

**Description**

Shows how RMSE develops for each iteration of iPLS selection algorithm

**Usage**

```
## S3 method for class 'ipls'
plotRMSE(
  obj,
  glob.ncomp = obj$gm$ncomp.selected,
  main = "RMSE development",
  xlab = "Iterations",
  ylab = if (is.null(obj$cv)) "RMSEP" else "RMSECV",
  xlim = NULL,
  ylim = NULL,
  ...
)
```

**Arguments**

obj	iPLS results (object of class ipls)
glob.ncomp	number of components for global PLS model with all intervals
main	main title for the plot
xlab	label for x-axis
ylab	label for y-axis
xlim	limits for x-axis
ylim	limits for y-axis
...	other arguments

**Details**

The plot shows RMSE values obtained at each iteration of the iPLS algorithm as bars. The first bar correspond to the global model with all variables included, second - to the model obtained at the first iteration and so on. Number at the bottom of each bar corresponds to the interval included or excluded at the particular iteration.

**See Also**

[summary.ipls](#), [plotSelection.ipls](#)

---

plotRMSE.regmodel      *RMSE plot for regression model*

---

### Description

Shows plot with root mean squared error values vs. number of components for PLS model.

### Usage

```
## S3 method for class 'regmodel'
plotRMSE(
  obj,
  ny = 1,
  type = "b",
  labels = "values",
  xticks = seq_len(obj$ncomp),
  res = obj$res,
  ylab = paste0("RMSE (", obj$res$cal$resnames[ny], ")"),
  ...
)
```

### Arguments

obj	a regression model (object of class regmodel)
ny	number of response variable to make the plot for (if y is multivariate)
type	type of the plot("b", "l" or "h")
labels	what to show as labels (vector or name, e.g. "names", "values", "indices")
xticks	vector with ticks for x-axis values
res	list with result objects
ylab	label for y-axis
...	other plot parameters (see mdaPlotg for details)

---

plotRMSE.regres      *RMSE plot for regression results*

---

### Description

Shows plot with RMSE values vs. model complexity (e.g. number of components).



**Usage**

```
## S3 method for class 'regres'
plotRMSE(
  obj,
  ny = 1,
  type = "b",
  xticks = seq_len(obj$ncomp),
  labels = "values",
  show.plot = TRUE,
  ylab = paste0("RMSE (", obj$responses[ny], ")"),
  ...
)
```

**Arguments**

obj	regression results (object of class regres)
ny	number of predictor to show the plot for (if y is multivariate)
type	type of the plot
xticks	vector with ticks for x-axis
labels	what to use as labels ("names", "values" or "indices")
show.plot	logical, show plot or just return plot data
ylab	label for y-axis
...	other plot parameters (see mdaPlot for details)

---

plotScatter

*Show plot series as set of points*


---

**Description**

Show plot series as set of points

**Usage**

```
plotScatter(
  ps,
  pch = 16,
  col = ps$col,
  bg = "white",
  lwd = 1,
  cex = 1,
  col.excluded = "lightgray",
  pch.colinv = FALSE,
  show.excluded = FALSE,
  ...
)
```

**Arguments**

ps	'plotseries' object
pch	size of point markers
col	color of the points
bg	background color of the points if 'pch=21:25'
lwd	line width for the error bars
cex	scale factor for the marker
col.excluded	color for excluded values (if must be shown)
pch.colinv	logical, should 'col' and 'bg' be switched if 'pch=21:25' and 'cgroup' is used to create colors.
show.excluded	logical, show or not the excluded data points
...	other arguments for function 'points()'.

---

plotScores

*Scores plot*


---

**Description**

Scores plot

**Usage**

```
plotScores(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for scores values for data decomposition

---

plotScores.ldecomp      *Scores plot*

---

### Description

Shows a plot with scores values for data objects.

### Usage

```
## S3 method for class 'ldecomp'  
plotScores(  
  obj,  
  comp = c(1, 2),  
  type = "p",  
  show.axes = TRUE,  
  show.plot = TRUE,  
  ...  
)
```

### Arguments

obj	object of ldecomp class.
comp	which components to show the plot for (can be one value or vector with two values).
type	type of the plot
show.axes	logical, show or not a axes lines crossing origin (0,0)
show.plot	logical, shall plot be created or just plot series object is needed
...	most of graphical parameters from <a href="#">mdaplot</a> function can be used.

---

plotScores.pca      *Scores plot for PCA model*

---

### Description

Shows a scores plot for selected components.

### Usage

```
## S3 method for class 'pca'  
plotScores(  
  obj,  
  comp = c(1, 2),  
  type = "p",  
  show.axes = TRUE,
```

```

    show.legend = TRUE,
    res = obj$res,
    ...
)

```

### Arguments

obj	a PCA model (object of class pca)
comp	a value or vector with several values - number of components to show the plot for
type	type of the plot ("p", "l", "b", "h")
show.axes	logical, show or not a axes lines crossing origin (0,0)
show.legend	logical, show or not a legend on the plot
res	list with result objects to show the variance for
...	other plot parameters (see mdaplotg for details)

### Details

If plot is created only for one result object (e.g. calibration set), then the behaviour and all settings for the scores plot are identical to [plotScores.ldecomp](#). In this case you can show scores as a scatter, line or bar plot for any number of components.

Otherwise (e.g. if model contains results for calibration and test set) the plot is a group plot created using [mdaplotg](#) method and only scatter plot can be used.

See examples in help for [pca](#) function.

---

plotSelection	<i>Selected intervals plot</i>
---------------	--------------------------------

---

### Description

Selected intervals plot

### Usage

```
plotSelection(obj, ...)
```

### Arguments

obj	a model or result object
...	other arguments

### Details

Generic function for plotting selected intervals or variables

---

plotSelection.ipls      *iPLS performance plot*

---

### Description

Shows PLS performance for each selected or excluded intervals at the first iteration

### Usage

```
## S3 method for class 'ipls'
plotSelection(
  obj,
  glob.ncomp = obj$gm$ncomp.selected,
  main = "iPLS results",
  xlab = obj$xaxis.name,
  ylab = if (is.null(obj$cv)) "RMSEP" else "RMSECV",
  xlim = NULL,
  ylim = NULL,
  ...
)
```

### Arguments

obj	iPLS results (object of class ipls)
glob.ncomp	number of components for global PLS model with all intervals
main	main title for the plot
xlab	label for x-axis
ylab	label for y-axis
xlim	limits for x-axis
ylim	limits for y-axis
...	other arguments

### Details

The plot shows intervals as bars, which height corresponds to RMSECV obtained when particular interval was selected (forward) or excluded (backward) from a model at the first iteration. The intervals found optimal after backward/forward iPLS selection are shown with green color while the other intervals are gray.

See examples in help for [ipls](#) function.

@seealso [summary.ipls](#), [plotRMSE.ipls](#)

---

plotSelectivityRatio *Selectivity ratio plot*

---

### Description

Generic function for plotting selectivity ratio values for regression model (PCR, PLS, etc)

### Usage

```
plotSelectivityRatio(obj, ...)
```

### Arguments

obj	a regression model
...	other parameters

---

plotSelectivityRatio.pls  
*Selectivity ratio plot for PLS model*

---

### Description

Computes and shows a plot for Selectivity ratio values for given number of components and response variable

### Usage

```
## S3 method for class 'pls'
plotSelectivityRatio(obj, ny = 1, ncomp = obj$ncomp.selected, type = "l", ...)
```

### Arguments

obj	a PLS model (object of class pls)
ny	which response to plot the values for (if y is multivariate), can be a vector.
ncomp	number of components to count
type	type of the plot
...	other plot parameters (see mdaplot for details)

### Details

See [vipcores](#) for more details.

---

plotSensitivity      *Sensitivity plot*

---

**Description**

Sensitivity plot

**Usage**

```
plotSensitivity(obj, ...)
```

**Arguments**

obj                  a model or result object  
...                  other arguments

**Details**

Generic function for plotting sensitivity values for classification model or results

---

plotSensitivity.classmodel  
*Sensitivity plot for classification model*

---

**Description**

Makes a plot with sensitivity values vs. model complexity (e.g. number of components)

**Usage**

```
## S3 method for class 'classmodel'  
plotSensitivity(obj, legend.position = "bottomright", ...)
```

**Arguments**

obj                  classification model (object of class `plsda`, `simca`, etc.).  
legend.position      position of the legend (as in `mdaplotg`).  
...                  parameters for `plotPerformance.classmodel` function.

**Details**

See examples in description of `plsda`, `simca` or `simcam`.

---

`plotSensitivity.classres`

*Sensitivity plot for classification results*

---

### Description

Makes a plot with sn values vs. model complexity (e.g. number of components) for classification results.

### Usage

```
## S3 method for class 'classres'
plotSensitivity(obj, legend.position = "bottomright", ...)
```

### Arguments

`obj` classification results (object of class `plsdare`s, `simcamres`, etc.).

`legend.position` position of the legend (as in `mdaplotg`).

`...` other parameters for `plotPerformance.classres`

### Details

See examples in description of `plsdare`s, `simcamres`, etc.

---

`plotseries`

*Create plot series object based on data, plot type and parameters*

---

### Description

The ‘`plotseries`’ object contains all necessary parameters to create main plots from data values, including values for x and y, correct handling of excluded rows and columns, color grouping (if any), limits and labels.

If both ‘`col`’ and ‘`cgroup`’ are specified, ‘`cgroup`’ will be ignored.

Labels can be either provided by user or generated automatically based on values, names or indices of data rows and columns. If series is made for scatter plot ‘`type="p"`’ then labels are required for each row of the original dataset. Otherwise (for line, bar and errorbar plot) labels correspond to data columns (variables).

The object has the following plotting methods once created: `plotScatter` `plotLines` `plotBars` `plotDensity` `plotErrorbars`



**Usage**

```
plotseries(
  data,
  type,
  cgroup = NULL,
  col = NULL,
  opacity = 1,
  colmap = "default",
  labels = NULL
)
```

**Arguments**

data	data to make the plot for (vector, matrix or data frame).
type	type of the plot.
cgroup	vector with values used to create a color grouping of the series instances.
col	color to show the series on plot with (user defined).
opacity	opacity of the colors (between 0 and 1).
colmap	colormap name to generate color/colors if they are not specified by user. See <a href="#">link{mdaplot.getColors}</a> for details.
labels	either vector with labels for the series instances or string ("names", "values", or "indices") if labels should be generated automatically.

---

plotSpecificity	<i>Specificity plot</i>
-----------------	-------------------------

---

**Description**

Specificity plot

**Usage**

```
plotSpecificity(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting specificity values for classification model or results

---

```
plotSpecificity.classmodel
    Specificity plot for classification model
```

---

**Description**

Makes a plot with specificity values vs. model complexity (e.g. number of components)

**Usage**

```
## S3 method for class 'classmodel'
plotSpecificity(obj, legend.position = "bottomright", ...)
```

**Arguments**

```
obj          classification model (object of class plsda, simca, etc.).
legend.position  position of the legend (as in mdaplotg).
...          parameters for plotPerformance.classmodel function.
```

**Details**

See examples in description of [plsda](#), [simca](#) or [simcam](#).

---

```
plotSpecificity.classres
    Specificity plot for classification results
```

---

**Description**

Makes a plot with specificity values vs. model complexity (e.g. number of components) for classification results.

**Usage**

```
## S3 method for class 'classres'
plotSpecificity(obj, legend.position = "bottomright", ...)
```

**Arguments**

```
obj          classification results (object of class plsdares, simcamres, etc.).
legend.position  position of the legend (as in mdaplotg).
...          other parameters for plotPerformance.classres
```

**Details**

See examples in description of [plsdare](#)s, [simcamres](#), etc.

---

plotSpectra	<i>Plot resolved spectra</i>
-------------	------------------------------

---

**Description**

Plot resolved spectra

**Usage**

```
plotSpectra(obj, ...)
```

**Arguments**

obj	object with mcr case
...	other parameters

---

plotSpectra.mcr	<i>Show plot with resolved spectra</i>
-----------------	--

---

**Description**

Show plot with resolved spectra

**Usage**

```
## S3 method for class 'mcr'
plotSpectra(
  obj,
  comp = seq_len(obj$ncomp),
  type = "l",
  col = mdaplot.getColors(obj$ncomp),
  ...
)
```

**Arguments**

obj	object of class mcr
comp	vector with number of components to make the plot for
type	type of the plot
col	vector with colors for individual components
...	other parameters suitable for mdaplotg

---

plotT2DoF	<i>Degrees of freedom plot for score distance (Nh)</i>
-----------	--

---

**Description**

Shows a plot with degrees of freedom computed for score distances at given number of components using data driven approach ("ddmoments" or "ddrobust").

**Usage**

```
plotT2DoF(
  obj,
  type = "b",
  labels = "values",
  xticks = seq_len(obj$ncomp),
  ylab = "Nh",
  ...
)
```

**Arguments**

obj	a PCA model (object of class pca)
type	type of the plot ("b", "l", "h")
labels	what to show as data points labels
xticks	vector with tick values for x-axis
ylab	label for y-axis
...	other plot parameters (see mdaplotg for details)

**Details**

Work only if parameter `lim.type` equal to "ddmoments" or "ddrobust".

---

plotVariance	<i>Variance plot</i>
--------------	----------------------

---

**Description**

Variance plot

**Usage**

```
plotVariance(obj, ...)
```

**Arguments**

obj            a model or result object  
 ...            other arguments

**Details**

Generic function for plotting explained variance for data decomposition

---

plotVariance.ldecomp    *Explained variance plot*

---

**Description**

Shows a plot with explained variance vs. number of components.

**Usage**

```
## S3 method for class 'ldecomp'
plotVariance(
  obj,
  type = "b",
  variance = "expvar",
  labels = "values",
  xticks = seq_len(obj$ncomp),
  show.plot = TRUE,
  ylab = "Explained variance, %",
  ...
)
```

**Arguments**

obj            object of ldecomp class.  
 type           type of the plot  
 variance       string, which variance to make the plot for ("expvar", "cumexpvar")  
 labels         what to show as labels for plot objects.  
 xticks         vector with ticks for x-axis  
 show.plot      logical, shall plot be created or just plot series object is needed  
 ylab           label for y-axis  
 ...            most of graphical parameters from [mdaplot](#) function can be used.

---

plotVariance.mcr      *Show plot with explained variance*

---

### Description

Show plot with explained variance

### Usage

```
## S3 method for class 'mcr'
plotVariance(
  obj,
  type = "h",
  labels = "values",
  main = "Variance",
  xticks = seq_len(obj$ncomp),
  ...
)
```

### Arguments

obj	object of class mcr
type	type of the plot
labels	what to use as data labels
main	title of the plot
xticks	vector with ticks for x-axis
...	other parameters suitable for mdaplot

---

plotVariance.pca      *Explained variance plot for PCA model*

---

### Description

Shows a plot with explained variance or cumulative explained variance for components.

### Usage

```
## S3 method for class 'pca'
plotVariance(
  obj,
  type = "b",
  labels = "values",
  variance = "expvar",
  xticks = seq_len(obj$ncomp),
```

```

    res = obj$res,
    ylab = "Explained variance, %",
    ...
)

```

### Arguments

obj	a PCA model (object of class <code>pca</code> )
type	type of the plot ("b", "l", "h")
labels	what to use as labels (if <code>show.labels = TRUE</code> )
variance	which variance to show
xticks	vector with ticks for x-axis
res	list with result objects to show the variance for
ylab	label for y-axis
...	other plot parameters (see <code>mdaplotg</code> for details)

### Details

See examples in help for [pca](#) function.

---

plotVariance.pls	<i>Variance plot for PLS</i>
------------------	------------------------------

---

### Description

Shows plot with variance values vs. number of components.

### Usage

```

## S3 method for class 'pls'
plotVariance(
  obj,
  decomp = "xdecomp",
  variance = "expvar",
  type = "b",
  labels = "values",
  res = obj$res,
  ylab = "Explained variance, %",
  ...
)

```

**Arguments**

obj	a PLS model (object of class pls)
decomp	which decomposition to use ("xdecomp" for x or "ydecomp" for y)
variance	which variance to use ("expvar", "cumexpvar")
type	type of the plot("b", "l" or "h")
labels	what to show as labels for plot objects.
res	list with result objects to show the plot for (by default, model results are used)
ylab	label for y-axis
...	other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pls](#) function.

---

plotVariance.plsres     *Explained X variance plot for PLS results*

---

**Description**

Shows plot with explained X variance vs. number of components.

**Usage**

```
## S3 method for class 'plsres'
plotVariance(obj, decomp = "xdecomp", variance = "expvar", ...)
```

**Arguments**

obj	PLS results (object of class plsres)
decomp	which dcomposition to use ("xdecomp" or "ydecomp")
variance	which variance to use ("expvar", "cumexpvar")
...	other plot parameters (see mdaplot for details)

**Details**

See examples in help for [plsres](#) function.



---

plotVIPScores	<i>VIP scores plot</i>
---------------	------------------------

---

**Description**

Generic function for plotting VIP scores values for regression model (PCR, PLS, etc)

**Usage**

```
plotVIPScores(obj, ...)
```

**Arguments**

obj	a regression model
...	other parameters

---

plotVIPScores.pls	<i>VIP scores plot for PLS model</i>
-------------------	--------------------------------------

---

**Description**

Shows a plot with VIP scores values for given number of components and response variable

**Usage**

```
## S3 method for class 'pls'
plotVIPScores(obj, ny = 1, ncomp = obj$ncomp.selected, type = "l", ...)
```

**Arguments**

obj	a PLS model (object of class pls)
ny	which response to plot the values for (if y is multivariate), can be a vector.
ncomp	number of components to count
type	type of the plot
...	other plot parameters (see mdaplot for details)

**Details**

See [vipcores](#) for more details.

---

plotWeights	<i>Plot for PLS weights</i>
-------------	-----------------------------

---

**Description**

Plot for PLS weights

**Usage**

```
plotWeights(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for weight plot

---

plotWeights.pls	<i>X loadings plot for PLS</i>
-----------------	--------------------------------

---

**Description**

Shows plot with X loading values for selected components.

**Usage**

```
## S3 method for class 'pls'
plotWeights(
  obj,
  comp = 1,
  type = (if (nrow(obj$weights) < 20) "h" else "l"),
  show.axes = TRUE,
  show.legend = TRUE,
  ...
)
```

**Arguments**

obj	a PLS model (object of class pls)
comp	which components to show the plot for (one or vector with several values)
type	type of the plot
show.axes	logical, show or not a axes lines crossing origin (0,0)
show.legend	logical, show or not a legend
...	other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pls](#) function.

---

plotXCumVariance      *X cumulative variance plot*

---

**Description**

X cumulative variance plot

**Usage**

```
plotXCumVariance(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting cumulative explained variance for decomposition of x data

---

plotXCumVariance.pls      *Cumulative explained X variance plot for PLS*

---

**Description**

Shows plot with cumulative explained X variance vs. number of components.

**Usage**

```
## S3 method for class 'pls'
plotXCumVariance(obj, type = "b", main = "Cumulative variance (X)", ...)
```

**Arguments**

obj	a PLS model (object of class pls)
type	type of the plot("b", "l" or "h")
main	title for the plot
...	other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pls](#) function.

plotXCumVariance.plsres

*Explained cumulative X variance plot for PLS results*

---

### Description

Shows plot with cumulative explained X variance vs. number of components.

### Usage

```
## S3 method for class 'plsres'  
plotXCumVariance(obj, main = "Cumulative variance (X)", ...)
```

### Arguments

obj	PLS results (object of class plsres)
main	main plot title
...	other plot parameters (see mdaplot for details)

### Details

See examples in help for [plsres](#) function.

---

plotXLoadings

*X loadings plot*

---

### Description

X loadings plot

### Usage

```
plotXLoadings(obj, ...)
```

### Arguments

obj	a model or result object
...	other arguments

### Details

Generic function for plotting loadings values for decomposition of x data

---

plotXLoadings.pls      *X loadings plot for PLS*

---

### Description

Shows plot with X loading values for selected components.

### Usage

```
## S3 method for class 'pls'  
plotXLoadings(  
  obj,  
  comp = c(1, 2),  
  type = "p",  
  show.axes = TRUE,  
  show.legend = TRUE,  
  ...  
)
```

### Arguments

obj	a PLS model (object of class pls)
comp	which components to show the plot for (one or vector with several values)
type	type of the plot
show.axes	logical, show or not a axes lines crossing origin (0,0)
show.legend	logical, show or not legend on the plot (when it is available)
...	other plot parameters (see mdaplotg for details)

### Details

See examples in help for [pls](#) function.

---

plotXResiduals      *X residuals plot*

---

### Description

X residuals plot

### Usage

```
plotXResiduals(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting x residuals for classification or regression model or results

---

plotXResiduals.pls      *Residual distance plot for decomposition of X data*

---

**Description**

Shows a plot with orthogonal distance vs score distance for PLS decomposition of X data.

**Usage**

```
## S3 method for class 'pls'
plotXResiduals(
  obj,
  ncomp = obj$ncomp.selected,
  norm = TRUE,
  log = FALSE,
  main = sprintf("X-distances (ncomp = %d)", ncomp),
  cgroup = NULL,
  xlim = NULL,
  ylim = NULL,
  show.limits = c(TRUE, TRUE),
  lim.col = c("darkgray", "darkgray"),
  lim.lwd = c(1, 1),
  lim.lty = c(2, 3),
  show.legend = TRUE,
  legend.position = "topright",
  res = obj$res,
  ...
)
```

**Arguments**

obj	a PLS model (object of class pls)
ncomp	how many components to use (by default optimal value selected for the model will be used)
norm	logical, normalize distance values or not (see details)
log	logical, apply log transformation to the distances or not (see details)
main	title for the plot

cgroup	color grouping of plot points (works only if one result object is available)
xlim	limits for x-axis
ylim	limits for y-axis
show.limits	vector with two logical values defining if limits for extreme and/or outliers must be shown
lim.col	vector with two values - line color for extreme and outlier limits
lim.lwd	vector with two values - line width for extreme and outlier limits
lim.lty	vector with two values - line type for extreme and outlier limits
show.legend	logical, show or not a legend on the plot (needed if several result objects are available)
legend.position	position of legend (if shown)
res	list with result objects to show the plot for (by default, model results are used)
...	other plot parameters (see <code>mdaplotg</code> for details)

### Details

The function is almost identical to [plotResiduals.pca](#).

---

`plotXResiduals.plsres` *X residuals plot for PLS results*

---

### Description

Shows a plot with Q residuals vs. Hotelling T2 values for PLS decomposition of x data.

### Usage

```
## S3 method for class 'plsres'
plotXResiduals(
  obj,
  ncomp = obj$ncomp.selected,
  norm = TRUE,
  log = FALSE,
  main = sprintf("X-distances (ncomp = %d)", ncomp),
  ...
)
```

### Arguments

obj	PLS results (object of class <code>plsres</code> )
ncomp	how many components to use (if NULL - user selected optimal value will be used)
norm	logical, normalize distance values or not (see details)

log                    logical, apply log transformation to the distances or not (see details)  
 main                    main title for the plot  
 ...                    other plot parameters (see mdaplot for details)

**Details**

See examples in help for [plsres](#) function.

---

plotXScores                    *X scores plot*

---

**Description**

X scores plot

**Usage**

```
plotXScores(obj, ...)
```

**Arguments**

obj                    a model or result object  
 ...                    other arguments

**Details**

Generic function for plotting scores values for decomposition of x data

---

plotXScores.pls                    *X scores plot for PLS*

---

**Description**

Shows plot with X scores values for selected components.

**Usage**

```
## S3 method for class 'pls'
plotXScores(
  obj,
  comp = c(1, 2),
  show.axes = T,
  main = "Scores (X)",
  res = obj$res,
  ...
)
```



**Arguments**

obj	a PLS model (object of class pls)
comp	which components to show the plot for (one or vector with several values)
show.axes	logical, show or not a axes lines crossing origin (0,0)
main	main plot title
res	list with result objects to show the plot for (by default, model results are used)
...	other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pls](#) function.

---

plotXScores.plsres      *X scores plot for PLS results*

---

**Description**

Shows plot with scores values for PLS decomposition of x data.

**Usage**

```
## S3 method for class 'plsres'
plotXScores(obj, comp = c(1, 2), main = "Scores (X)", ...)
```

**Arguments**

obj	PLS results (object of class plsres)
comp	which components to show the plot for (one or vector with several values)
main	main plot title
...	other plot parameters (see mdaplot for details)

**Details**

See examples in help for [plsres](#) function.

---

plotXVariance	<i>X variance plot</i>
---------------	------------------------

---

**Description**

X variance plot

**Usage**

```
plotXVariance(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting explained variance for decomposition of x data

---

plotXVariance.pls	<i>Explained X variance plot for PLS</i>
-------------------	--

---

**Description**

Shows plot with explained X variance vs. number of components.

**Usage**

```
## S3 method for class 'pls'  
plotXVariance(obj, type = "b", main = "Variance (X)", ...)
```

**Arguments**

obj	a PLS model (object of class pls)
type	type of the plot("b", "l" or "h")
main	title for the plot
...	other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pls](#) function.

---

plotXVariance.plsres    *Explained X variance plot for PLS results*

---

**Description**

Shows plot with explained X variance vs. number of components.

**Usage**

```
## S3 method for class 'plsres'  
plotXVariance(obj, main = "Variance (X)", ...)
```

**Arguments**

obj	PLS results (object of class plsres)
main	main plot title
...	other plot parameters (see mdaplot for details)

**Details**

See examples in help for [plsres](#) function.

---

plotXYLoadings    *X loadings plot*

---

**Description**

X loadings plot

**Usage**

```
plotXYLoadings(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting loadings values for decomposition of x and y data

---

plotXYLoadings.pls      *XY loadings plot for PLS*

---

**Description**

Shows plot with X and Y loading values for selected components.

**Usage**

```
## S3 method for class 'pls'  
plotXYLoadings(obj, comp = c(1, 2), show.axes = TRUE, ...)
```

**Arguments**

obj	a PLS model (object of class pls)
comp	which components to show the plot for (one or vector with several values)
show.axes	logical, show or not a axes lines crossing origin (0,0)
...	other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pls](#) function.

---

plotXYResiduals      *Plot for XY-residuals*

---

**Description**

Plot for XY-residuals

**Usage**

```
plotXYResiduals(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for XY-residuals plot

---

plotXYResiduals.pls     *Residual XY-distance plot*

---

### Description

Shows a plot with full X-distance (f) vs. orthogonal Y-distance (z) for PLS model results.

### Usage

```
## S3 method for class 'pls'
plotXYResiduals(
  obj,
  ncomp = obj$ncomp.selected,
  norm = TRUE,
  log = FALSE,
  main = sprintf("XY-distances (ncomp = %d)", ncomp),
  cgroup = NULL,
  xlim = NULL,
  ylim = NULL,
  show.limits = c(TRUE, TRUE),
  lim.col = c("darkgray", "darkgray"),
  lim.lwd = c(1, 1),
  lim.lty = c(2, 3),
  show.legend = TRUE,
  legend.position = "topright",
  res = obj$res,
  ...
)
```

### Arguments

obj	a PLS model (object of class pls)
ncomp	how many components to use (by default optimal value selected for the model will be used)
norm	logical, normalize distance values or not (see details)
log	logical, apply log tranformation to the distances or not (see details)
main	title for the plot
cgroup	color grouping of plot points (works only if one result object is available)
xlim	limits for x-axis
ylim	limits for y-axis
show.limits	vector with two logical values defining if limits for extreme and/or outliers must be shown
lim.col	vector with two values - line color for extreme and outlier limits
lim.lwd	vector with two values - line width for extreme and outlier limits

lim.lty	vector with two values - line type for extreme and outlier limits
show.legend	logical, show or not a legend on the plot (needed if several result objects are available)
legend.position	position of legend (if shown)
res	list with result objects to show the plot for (by default, model results are used)
...	other plot parameters (see mdaplotg for details)

### Details

The function presents a way to identify extreme objects and outliers based on both full distance for X-decomposition (known as  $f$ ) and squared residual distance for Y-decomposition ( $z$ ). The approach has been proposed in [1].

The plot is available only if data driven methods (classic or robust) have been used for computing of critical limits.

### References

1. Rodionova O. Ye., Pomerantsev A. L. Detection of Outliers in Projection-Based Modeling. Analytical Chemistry (2020, in publish). doi: 10.1021/acs.analchem.9b04611

---

plotXYResiduals.plsres

*Residual distance plot*

---

### Description

Shows a plot with orthogonal (Q, q) vs. score (T2, h) distances for data objects.

### Usage

```
## S3 method for class 'plsres'
plotXYResiduals(
  obj,
  ncomp = obj$ncomp.selected,
  norm = TRUE,
  log = FALSE,
  show.labels = FALSE,
  labels = "names",
  show.plot = TRUE,
  ...
)
```

**Arguments**

obj	object of ldecomp class.
ncomp	number of components to show the plot for (if NULL, selected by model value will be used).
norm	logical, normalize distance values or not (see details)
log	logical, apply log transformation to the distances or not (see details)
show.labels	logical, show or not labels for the plot objects
labels	what to show as labels if necessary
show.plot	logical, shall plot be created or just plot series object is needed
...	most of graphical parameters from <code>mdaplot</code> function can be used.

---

plotXYScores	<i>XY scores plot</i>
--------------	-----------------------

---

**Description**

XY scores plot

**Usage**

```
plotXYScores(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting scores values for decomposition of x and y data

---

plotXYScores.pls	<i>XY scores plot for PLS</i>
------------------	-------------------------------

---

**Description**

Shows plot with X vs. Y scores values for selected component.

**Usage**

```
## S3 method for class 'pls'
plotXYScores(obj, ncomp = 1, show.axes = T, res = obj$res, ...)
```

**Arguments**

obj	a PLS model (object of class pls)
ncomp	which component to show the plot for
show.axes	logical, show or not a axes lines crossing origin (0,0)
res	list with result objects to show the plot for (by default, model results are used)
...	other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pls](#) function.

---

plotXYScores.plsres    *XY scores plot for PLS results*

---

**Description**

Shows plot with X vs. Y scores values for PLS results.

**Usage**

```
## S3 method for class 'plsres'
plotXYScores(obj, ncomp = 1, show.plot = TRUE, ...)
```

**Arguments**

obj	PLS results (object of class plsres)
ncomp	which component to show the plot for
show.plot	logical, show plot or just return plot data
...	other plot parameters (see mdaplot for details)

**Details**

See examples in help for [plsres](#) function.



---

plotYCumVariance      *Y cumulative variance plot*

---

**Description**

Y cumulative variance plot

**Usage**

```
plotYCumVariance(obj, ...)
```

**Arguments**

obj                    a model or result object  
...                    other arguments

**Details**

Generic function for plotting cumulative explained variance for decomposition of y data

---

plotYCumVariance.pls      *Cumulative explained Y variance plot for PLS*

---

**Description**

Shows plot with cumulative explained Y variance vs. number of components.

**Usage**

```
## S3 method for class 'pls'  
plotYCumVariance(obj, type = "b", main = "Cumulative variance (Y)", ...)
```

**Arguments**

obj                    a PLS model (object of class pls)  
type                   type of the plot("b", "l" or "h")  
main                   title for the plot  
...                    other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pls](#) function.

---

```
plotYCumVariance.plsres
```

*Explained cumulative Y variance plot for PLS results*

---

### Description

Shows plot with cumulative explained Y variance vs. number of components.

### Usage

```
## S3 method for class 'plsres'
plotYCumVariance(obj, main = "Cumulative variance (Y)", ...)
```

### Arguments

obj	PLS results (object of class plsres)
main	main plot title
...	other plot parameters (see mdaplot for details)

### Details

See examples in help for [plsres](#) function.

---

```
plotYResiduals      Y residuals plot
```

---

### Description

Y residuals plot

### Usage

```
plotYResiduals(obj, ...)
```

### Arguments

obj	a model or result object
...	other arguments

### Details

Generic function for plotting y residuals for classification or regression model or results

---

plotYResiduals.plsres *Y residuals plot for PLS results*

---

**Description**

Shows a plot with Y residuals vs reference Y values for selected component.

**Usage**

```
## S3 method for class 'plsres'  
plotYResiduals(obj, ncomp = obj$ncomp.selected, ...)
```

**Arguments**

obj	PLS results (object of class plsres)
ncomp	how many components to use (if NULL - user selected optimal value will be used)
...	other plot parameters (see mdaplot for details)

**Details**

Proxy for [plotResiduals.regres](#) function.

---

plotYResiduals.regmodel  
*Y residuals plot for regression model*

---

**Description**

Shows plot with y residuals (predicted vs. reference values) for selected components.

**Usage**

```
## S3 method for class 'regmodel'  
plotYResiduals(  
  obj,  
  ncomp = obj$ncomp.selected,  
  ny = 1,  
  show.lines = c(NA, 0),  
  res = obj$res,  
  ...  
)
```

**Arguments**

obj	a regression model (object of class regmodel)
ncomp	how many components to use (if NULL - user selected optimal value will be used)
ny	number of response variable to make the plot for (if y is multivariate)
show.lines	allows to show the horizontal line at 0 level
res	list with result objects
...	other plot parameters (see mdaplotg for details)

---

plotYVariance	<i>Y variance plot</i>
---------------	------------------------

---

**Description**

Y variance plot

**Usage**

```
plotYVariance(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for plotting explained variance for decomposition of y data

---

plotYVariance.pls	<i>Explained Y variance plot for PLS</i>
-------------------	--

---

**Description**

Shows plot with explained Y variance vs. number of components.

**Usage**

```
## S3 method for class 'pls'
plotYVariance(obj, type = "b", main = "Variance (Y)", ...)
```

**Arguments**

obj	a PLS model (object of class pls)
type	type of the plot("b", "l" or "h")
main	title for the plot
...	other plot parameters (see mdaplotg for details)

**Details**

See examples in help for [pls](#) function.

---

plotYVariance.plsres *Explained Y variance plot for PLS results*

---

**Description**

Shows plot with explained Y variance vs. number of components.

**Usage**

```
## S3 method for class 'plsres'  
plotYVariance(obj, main = "Variance (Y)", ...)
```

**Arguments**

obj	PLS results (object of class plsres)
main	main plot title
...	other plot parameters (see mdaplot for details)

**Details**

See examples in help for [plsres](#) function.

pls

*Partial Least Squares regression***Description**

pls is used to calibrate, validate and use of partial least squares (PLS) regression model.

**Usage**

```
pls(
  x,
  y,
  ncomp = min(nrow(x) - 1, ncol(x), 20),
  center = TRUE,
  scale = FALSE,
  cv = NULL,
  exclcols = NULL,
  exclrows = NULL,
  x.test = NULL,
  y.test = NULL,
  method = "simpls",
  info = "",
  ncomp.selcrit = "min",
  lim.type = "ddmoments",
  alpha = 0.05,
  gamma = 0.01
)
```

**Arguments**

x	matrix with predictors.
y	matrix with responses.
ncomp	maximum number of components to calculate.
center	logical, center or not predictors and response values.
scale	logical, scale (standardize) or not predictors and response values.
cv	cross-validation settings (see details).
exclcols	columns of x to be excluded from calculations (numbers, names or vector with logical values)
exclrows	rows to be excluded from calculations (numbers, names or vector with logical values)
x.test	matrix with predictors for test set.
y.test	matrix with responses for test set.
method	algorithm for computing PLS model (only 'simpls' is supported so far)
info	short text with information about the model.

<code>ncomp.selcrit</code>	criterion for selecting optimal number of components ('min' for first local minimum of RMSECV and 'wold' for Wold's rule.)
<code>lim.type</code>	which method to use for calculation of critical limits for residual distances (see details)
<code>alpha</code>	significance level for extreme limits for T2 and Q distances.
<code>gamma</code>	significance level for outlier limits for T2 and Q distances.

### Details

So far only SIMPLS method [1] is available. Implementation works both with one and multiple response variables.

Like in `pca`, `pls` uses number of components (`ncomp`) as a minimum of number of objects - 1, number of x variables and the default or provided value. Regression coefficients, predictions and other results are calculated for each set of components from 1 to `ncomp`: 1, 1:2, 1:3, etc. The optimal number of components, (`ncomp.selected`), is found using first local minimum, but can be also forced to user defined value using function (`selectCompNum.pls`). The selected optimal number of components is used for all default operations - predictions, plots, etc.

Cross-validation settings, `cv`, can be a number or a list. If `cv` is a number, it will be used as a number of segments for random cross-validation (if `cv = 1`, full cross-validation will be performed). If it is a list, the following syntax can be used: `cv = list("rand", nseg, nrep)` for random repeated cross-validation with `nseg` segments and `nrep` repetitions or `cv = list("ven", nseg)` for systematic splits to `nseg` segments ('venetian blinds').

Calculation of confidence intervals and p-values for regression coefficients can be done based on Jack-Knifing resampling. This is done automatically if cross-validation is used. However it is recommended to use at least 10 segments for stable JK result. See help for `regcoeffs` objects for more details.

### Value

Returns an object of `pls` class with following fields:

<code>ncomp</code>	number of components included to the model.
<code>ncomp.selected</code>	selected (optimal) number of components.
<code>xcenter</code>	vector with values used to center the predictors (x).
<code>ycenter</code>	vector with values used to center the responses (y).
<code>xscale</code>	vector with values used to scale the predictors (x).
<code>yscale</code>	vector with values used to scale the responses (y).
<code>xloadings</code>	matrix with loading values for x decomposition.
<code>yloadings</code>	matrix with loading values for y decomposition.
<code>xeigenvals</code>	vector with eigenvalues of components (variance of x-scores).
<code>yeigenvals</code>	vector with eigenvalues of components (variance of y-scores).
<code>weights</code>	matrix with PLS weights.
<code>coeffs</code>	object of class <code>regcoeffs</code> with regression coefficients calculated for each component.

<code>info</code>	information about the model, provided by user when build the model.
<code>cv</code>	information cross-validation method used (if any).
<code>res</code>	a list with result objects (e.g. calibration, cv, etc.)

**Author(s)**

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

**References**

1. S. de Jong, Chemometrics and Intelligent Laboratory Systems 18 (1993) 251-263.
2. Tarja Rajalahti et al. Chemometrics and Laboratory Systems, 95 (2009), 35-48.
3. Il-Gyo Chong, Chi-Hyuck Jun. Chemometrics and Laboratory Systems, 78 (2005), 103-112.

**See Also**

Main methods for pls objects:

<code>print</code>	prints information about a pls object.
<code>summary.pls</code>	shows performance statistics for the model.
<code>plot.pls</code>	shows plot overview of the model.
<code>pls.simpls</code>	implementation of SIMPLS algorithm.
<code>predict.pls</code>	applies PLS model to a new data.
<code>selectCompNum.pls</code>	set number of optimal components in the model.
<code>setDistanceLimits.pls</code>	allows to change parameters for critical limits.
<code>categorize.pls</code>	categorize data rows similar to <code>categorize.pca</code> .
<code>selratio</code>	computes matrix with selectivity ratio values.
<code>vipscores</code>	computes matrix with VIP scores values.

Plotting methods for pls objects:

<code>plotXScores.pls</code>	shows scores plot for x decomposition.
<code>plotXYScores.pls</code>	shows scores plot for x and y decomposition.
<code>plotXLoadings.pls</code>	shows loadings plot for x decomposition.
<code>plotXYLoadings.pls</code>	shows loadings plot for x and y decomposition.
<code>plotXVariance.pls</code>	shows explained variance plot for x decomposition.
<code>plotYVariance.pls</code>	shows explained variance plot for y decomposition.
<code>plotXCumVariance.pls</code>	shows cumulative explained variance plot for y decomposition.
<code>plotYCumVariance.pls</code>	shows cumulative explained variance plot for y decomposition.
<code>plotXResiduals.pls</code>	shows distance/residuals plot for x decomposition.
<code>plotXYResiduals.pls</code>	shows joint distance plot for x and y decomposition.
<code>plotWeights.pls</code>	shows plot with weights.
<code>plotSelectivityRatio.pls</code>	shows plot with selectivity ratio values.
<code>plotVIPScores.pls</code>	shows plot with VIP scores values.

Methods inherited from `regmodel` object (parent class for `pls`):



<code>plotPredictions.regmodel</code>	shows predicted vs. measured plot.
<code>plotRMSE.regmodel</code>	shows RMSE plot.
<code>plotYResiduals.regmodel</code>	shows residuals plot for y values.
<code>getRegcoeffs.regmodel</code>	returns matrix with regression coefficients.

Most of the methods for plotting data (except loadings and regression coefficients) are also available for PLS results (`plsres`) objects. There is also a randomization test for PLS-regression (`randtest`) and implementation of interval PLS algorithm for variable selection (`ipls`)

## Examples

```
### Examples of using PLS model class
library(mdatools)

## 1. Make a PLS model for concentration of first component
## using full-cross validation and automatic detection of
## optimal number of components and show an overview

data(simdata)
x = simdata$spectra.c
y = simdata$conc.c[, 1]

model = pls(x, y, ncomp = 8, cv = 1)
summary(model)
plot(model)

## 2. Make a PLS model for concentration of first component
## using test set and 10 segment cross-validation and show overview

data(simdata)
x = simdata$spectra.c
y = simdata$conc.c[, 1]
x.t = simdata$spectra.t
y.t = simdata$conc.t[, 1]

model = pls(x, y, ncomp = 8, cv = 10, x.test = x.t, y.test = y.t)
model = selectCompNum(model, 2)
summary(model)
plot(model)

## 3. Make a PLS model for concentration of first component
## using only test set validation and show overview

data(simdata)
x = simdata$spectra.c
y = simdata$conc.c[, 1]
x.t = simdata$spectra.t
y.t = simdata$conc.t[, 1]

model = pls(x, y, ncomp = 6, x.test = x.t, y.test = y.t)
model = selectCompNum(model, 2)
```

```
summary(model)
plot(model)

## 4. Show variance and error plots for a PLS model
par(mfrow = c(2, 2))
plotXCumVariance(model, type = 'h')
plotYCumVariance(model, type = 'b', show.labels = TRUE, legend.position = 'bottomright')
plotRMSE(model)
plotRMSE(model, type = 'h', show.labels = TRUE)
par(mfrow = c(1, 1))

## 5. Show scores plots for a PLS model
par(mfrow = c(2, 2))
plotXScores(model)
plotXScores(model, comp = c(1, 3), show.labels = TRUE)
plotXYScores(model)
plotXYScores(model, comp = 2, show.labels = TRUE)
par(mfrow = c(1, 1))

## 6. Show loadings and coefficients plots for a PLS model
par(mfrow = c(2, 2))
plotXLoadings(model)
plotXLoadings(model, comp = c(1, 2), type = 'l')
plotXYLoadings(model, comp = c(1, 2), legend.position = 'topleft')
plotRegcoeffs(model)
par(mfrow = c(1, 1))

## 7. Show predictions and residuals plots for a PLS model
par(mfrow = c(2, 2))
plotXResiduals(model, show.label = TRUE)
plotYResiduals(model, show.label = TRUE)
plotPredictions(model)
plotPredictions(model, ncomp = 4, xlab = 'C, reference', ylab = 'C, predictions')
par(mfrow = c(1, 1))

## 8. Selectivity ratio and VIP scores plots
par(mfrow = c(2, 2))
plotSelectivityRatio(model)
plotSelectivityRatio(model, ncomp = 1)
par(mfrow = c(1, 1))

## 9. Variable selection with selectivity ratio
selratio = getSelectivityRatio(model)
selvar = !(selratio < 8)

xsel = x[, selvar]
modelsel = pls(xsel, y, ncomp = 6, cv = 1)
modelsel = selectCompNum(modelsel, 3)

summary(model)
summary(modelsel)

## 10. Calculate average spectrum and show the selected variables
```

```
i = 1:ncol(x)
ms = apply(x, 2, mean)

par(mfrow = c(2, 2))

plot(i, ms, type = 'p', pch = 16, col = 'red', main = 'Original variables')
plotPredictions(model)

plot(i, ms, type = 'p', pch = 16, col = 'lightgray', main = 'Selected variables')
points(i[selvar], ms[selvar], col = 'red', pch = 16)
plotPredictions(model$sel)

par(mfrow = c(1, 1))
```

---

pls.cal

*PLS model calibration*

---

### Description

Calibrates (builds) a PLS model for given data and parameters

### Usage

```
pls.cal(x, y, ncomp, center, scale, method = "simpls", cv = FALSE)
```

### Arguments

x	a matrix with x values (predictors)
y	a matrix with y values (responses)
ncomp	number of components to calculate
center	logical, do mean centering or not
scale	logical, do standardization or not
method	algorithm for computing PLS model (only 'simpls' is supported so far)
cv	logical, is model calibrated during cross-validation or not (or cv settings for calibration)

### Value

model an object with calibrated PLS model

---

pls.getLimitsCoordinates

*Compute coordinates of lines or curves with critical limits*

---

### Description

Compute coordinates of lines or curves with critical limits

### Usage

```
pls.getLimitsCoordinates(Qlim, T2lim, Zlim, nobj, ncomp, norm, log)
```

### Arguments

Qlim	matrix with critical limits for orthogonal distances (X)
T2lim	matrix with critical limits for score distances (X)
Zlim	matrix with critical limits for orthogonal distances (Y)
nobj	number of objects to compute the limits for
ncomp	number of components for computing the coordinates
norm	logical, shall distance values be normalized or not
log	logical, shall log transformation be applied or not

### Value

list with two matrices (x and y coordinates of corresponding limits)

---

pls.getZLimits

*Compute critical limits for orthogonal distances (Q)*

---

### Description

Compute critical limits for orthogonal distances (Q)

### Usage

```
pls.getZLimits(lim.type, alpha, gamma, params)
```

### Arguments

lim.type	which method to use for calculation of critical limits for residuals
alpha	significance level for extreme limits.
gamma	significance level for outlier limits.
params	distribution parameters returned by ldecomp.getLimParams

---

pls.run *Runs selected PLS algorithm*

---

**Description**

Runs selected PLS algorithm

**Usage**

```
pls.run(x, y, ncomp = min(nrow(x) - 1, ncol(x)), method = "simpls", cv = FALSE)
```

**Arguments**

x	a matrix with x values (predictors from calibration set)
y	a matrix with y values (responses from calibration set)
ncomp	how many components to compute
method	algorithm for computing PLS model
cv	logical, is this for CV or not

---

pls.simpls *SIMPLS algorithm*

---

**Description**

SIMPLS algorithm for calibration of PLS model

**Usage**

```
pls.simpls(x, y, ncomp, cv = FALSE)
```

**Arguments**

x	a matrix with x values (predictors)
y	a matrix with y values (responses)
ncomp	number of components to calculate
cv	logical, is model calibrated during cross-validation or not

**Value**

a list with computed regression coefficients, loadings and scores for x and y matrices, and weights.

**References**

[1]. S. de Jong. SIMPLS: An Alternative approach to partial least squares regression. *Chemometrics and Intelligent Laboratory Systems*, 18, 1993 (251-263).

---

plsda

*Partial Least Squares Discriminant Analysis*

---

### Description

plsda is used to calibrate, validate and use of partial least squares discrimination analysis (PLS-DA) model.

### Usage

```
plsda(  
  x,  
  c,  
  ncomp = min(nrow(x) - 1, ncol(x), 20),  
  center = TRUE,  
  scale = FALSE,  
  cv = NULL,  
  exclcols = NULL,  
  exclrows = NULL,  
  x.test = NULL,  
  c.test = NULL,  
  method = "simpls",  
  lim.type = "ddmoments",  
  alpha = 0.05,  
  gamma = 0.01,  
  info = "",  
  ncomp.selcrit = "min",  
  classname = NULL  
)
```

### Arguments

x	matrix with predictors.
c	vector with class membership (should be either a factor with class names/numbers in case of multiple classes or a vector with logical values in case of one class model).
ncomp	maximum number of components to calculate.
center	logical, center or not predictors and response values.
scale	logical, scale (standardize) or not predictors and response values.
cv	cross-validation settings (see details).
exclcols	columns of x to be excluded from calculations (numbers, names or vector with logical values)
exclrows	rows to be excluded from calculations (numbers, names or vector with logical values)
x.test	matrix with predictors for test set.

<code>c.test</code>	vector with reference class values for test set (same format as calibration values).
<code>method</code>	method for calculating PLS model.
<code>lim.type</code>	which method to use for calculation of critical limits for residual distances (see details)
<code>alpha</code>	significance level for extreme limits for T2 and Q distances.
<code>gamma</code>	significance level for outlier limits for T2 and Q distances.
<code>info</code>	short text with information about the model.
<code>ncomp.selcrit</code>	criterion for selecting optimal number of components ('min' for first local minimum of RMSECV and 'wold' for Wold's rule.)
<code>classname</code>	name (label) of class in case if PLS-DA is used for one-class discrimination model. In this case it is expected that parameter 'c' will be a vector with logical values.

### Details

The `plsda` class is based on `pls` with extra functions and plots covering classification functionality. All plots for `pls` can be used. E.g. if you want to see the real predicted values (y in PLS) instead of classes use `plotPredictions.pls(model)` instead of `plotPredictions(model)`.

Cross-validation settings, `cv`, can be a number or a list. If `cv` is a number, it will be used as a number of segments for random cross-validation (if `cv = 1`, full cross-validation will be performed). If it is a list, the following syntax can be used: `cv = list('rand', nseg, nrep)` for random repeated cross-validation with `nseg` segments and `nrep` repetitions or `cv = list('ven', nseg)` for systematic splits to `nseg` segments ('venetian blinds').

Calculation of confidence intervals and p-values for regression coefficients are available only by jack-knifing so far. See help for `regcoeffs` objects for details.

### Value

Returns an object of `plsda` class with following fields (most inherited from class `pls`):

<code>ncomp</code>	number of components included to the model.
<code>ncomp.selected</code>	selected (optimal) number of components.
<code>xloadings</code>	matrix with loading values for x decomposition.
<code>yloadings</code>	matrix with loading values for y (c) decomposition.
<code>weights</code>	matrix with PLS weights.
<code>coeffs</code>	matrix with regression coefficients calculated for each component.
<code>info</code>	information about the model, provided by user when build the model.
<code>calres</code>	an object of class <code>plsdare</code> s with PLS-DA results for a calibration data.
<code>testres</code>	an object of class <code>plsdare</code> s with PLS-DA results for a test data, if it was provided.
<code>cvres</code>	an object of class <code>plsdare</code> s with PLS-DA results for cross-validation, if this option was chosen.

**Author(s)**

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

**See Also**

Specific methods for plsda class:

<code>print.plsda</code>	prints information about a pls object.
<code>summary.plsda</code>	shows performance statistics for the model.
<code>plot.plsda</code>	shows plot overview of the model.
<code>predict.plsda</code>	applies PLS-DA model to a new data.

Methods, inherited from classmodel class:

<code>plotPredictions.classmodel</code>	shows plot with predicted values.
<code>plotSensitivity.classmodel</code>	shows sensitivity plot.
<code>plotSpecificity.classmodel</code>	shows specificity plot.
<code>plotMisclassified.classmodel</code>	shows misclassified ratio plot.

See also methods for class `pls`.

**Examples**

```
### Examples for PLS-DA model class

library(mdatools)

## 1. Make a PLS-DA model with full cross-validation and show model overview

# make a calibration set from iris data (3 classes)
# use names of classes as class vector
x.cal = iris[seq(1, nrow(iris), 2), 1:4]
c.cal = iris[seq(1, nrow(iris), 2), 5]

model = plsda(x.cal, c.cal, ncomp = 3, cv = 1, info = 'IRIS data example')
model = selectCompNum(model, 1)

# show summary and basic model plots
# misclassification will be shown only for first class
summary(model)
plot(model)

# summary and model plots for second class
summary(model, nc = 2)
plot(model, nc = 2)

# summary and model plot for specific class and number of components
summary(model, nc = 3, ncomp = 3)
```



```
plot(model, nc = 3, ncomp = 3)

## 2. Show performance plots for a model
par(mfrow = c(2, 2))
plotSpecificity(model)
plotSensitivity(model)
plotMisclassified(model)
plotMisclassified(model, nc = 2)
par(mfrow = c(1, 1))

## 3. Show both class and y values predictions
par(mfrow = c(2, 2))
plotPredictions(model)
plotPredictions(model, res = "cal", ncomp = 2, nc = 2)
plotPredictions(structure(model, class = "regmodel"))
plotPredictions(structure(model, class = "regmodel"), ncomp = 2, ny = 2)
par(mfrow = c(1, 1))

## 4. All plots from ordinary PLS can be used, e.g.:
par(mfrow = c(2, 2))
plotXYScores(model)
plotYVariance(model)
plotXResiduals(model)
plotRegcoeffs(model, ny = 2)
par(mfrow = c(1, 1))
```

---

plsdares

*PLS-DA results*

---

## Description

plsdares is used to store and visualize results of applying a PLS-DA model to a new data.

## Usage

```
plsdares(plsres, cres)
```

## Arguments

plsres	PLS results for the data.
cres	Classification results for the data.

## Details

Do not use plsdares manually, the object is created automatically when one applies a PLS-DA model to a new data set, e.g. when calibrate and validate a PLS-DA model (all calibration and validation results in PLS-DA model are stored as objects of plsdares class) or use function [predict.plsda](#).

The object gives access to all PLS-DA results as well as to the plotting methods for visualisation of the results. The `plsdares` class also inherits all properties and methods of `classres` and `plsres` classes.

If no reference values provided, classification statistics will not be calculated and performance plots will not be available.

### Value

Returns an object of `plsdares` class with fields, inherited from `classres` and `plsres`.

### See Also

Methods for `plsda` objects:

<code>print.plsda</code>	shows information about the object.
<code>summary.plsda</code>	shows statistics for results of classification.
<code>plot.plsda</code>	shows plots for overview of the results.

Methods, inherited from `classres` class:

<code>showPredictions.classres</code>	show table with predicted values.
<code>plotPredictions.classres</code>	makes plot with predicted values.
<code>plotSensitivity.classres</code>	makes plot with sensitivity vs. components values.
<code>plotSpecificity.classres</code>	makes plot with specificity vs. components values.
<code>plotPerformance.classres</code>	makes plot with both specificity and sensitivity values.

Methods for `plsres` objects:

<code>print</code>	prints information about a <code>plsres</code> object.
<code>summary.plsres</code>	shows performance statistics for the results.
<code>plot.plsres</code>	shows plot overview of the results.
<code>plotXScores.plsres</code>	shows scores plot for x decomposition.
<code>plotXYScores.plsres</code>	shows scores plot for x and y decomposition.
<code>plotXVariance.plsres</code>	shows explained variance plot for x decomposition.
<code>plotYVariance.plsres</code>	shows explained variance plot for y decomposition.
<code>plotXCumVariance.plsres</code>	shows cumulative explained variance plot for y decomposition.
<code>plotYCumVariance.plsres</code>	shows cumulative explained variance plot for y decomposition.
<code>plotXResiduals.plsres</code>	shows T2 vs. Q plot for x decomposition.
<code>plotYResiduals.plsres</code>	shows residuals plot for y values.

Methods inherited from `regres` class (parent class for `plsres`):

<code>plotPredictions.regres</code>	shows predicted vs. measured plot.
<code>plotRMSE.regres</code>	shows RMSE plot.

See also `plsda` - a class for PLS-DA models, `predict.plsda` applying PLS-DA model for a new dataset.

## Examples

```
### Examples for PLS-DA results class

library(mdatools)

## 1. Make a PLS-DA model with full cross-validation, get
## calibration results and show overview

# make a calibration set from iris data (3 classes)
# use names of classes as class vector
x.cal = iris[seq(1, nrow(iris), 2), 1:4]
c.cal = iris[seq(1, nrow(iris), 2), 5]

model = plsda(x.cal, c.cal, ncomp = 3, cv = 1, info = 'IRIS data example')
model = selectCompNum(model, 1)

res = model$calres

# show summary and basic plots for calibration results
summary(res)
plot(res)

## 2. Apply the calibrated PLS-DA model to a new dataset

# make a new data
x.new = iris[seq(2, nrow(iris), 2), 1:4]
c.new = iris[seq(2, nrow(iris), 2), 5]

res = predict(model, x.new, c.new)
summary(res)
plot(res)

## 3. Show performance plots for the results
par(mfrow = c(2, 2))
plotSpecificity(res)
plotSensitivity(res)
plotMisclassified(res)
plotMisclassified(res, nc = 2)
par(mfrow = c(1, 1))

## 3. Show both class and y values predictions
par(mfrow = c(2, 2))
plotPredictions(res)
plotPredictions(res, ncomp = 2, nc = 2)
plotPredictions(structure(res, class = "regres"))
plotPredictions(structure(res, class = "regres"), ncomp = 2, ny = 2)
par(mfrow = c(1, 1))

## 4. All plots from ordinary PLS results can be used, e.g.:
```

```

par(mfrow = c(2, 2))
plotXYScores(res)
plotYVariance(res, type = 'h')
plotXVariance(res, type = 'h')
plotXResiduals(res)
par(mfrow = c(1, 1))

```

---

plsres

*PLS results*


---

### Description

plsres is used to store and visualize results of applying a PLS model to a new data.

### Usage

```

plsres(
  y.pred,
  y.ref = NULL,
  ncomp.selected = dim(y.pred)[2],
  xdecomp = NULL,
  ydecomp = NULL,
  info = ""
)

```

### Arguments

y.pred	predicted y values.
y.ref	reference (measured) y values.
ncomp.selected	selected (optimal) number of components.
xdecomp	PLS decomposition of X data (object of class ldecomp).
ydecomp	PLS decomposition of Y data (object of class ldecomp).
info	information about the object.

### Details

Do not use plsres manually, the object is created automatically when one applies a PLS model to a new data set, e.g. when calibrate and validate a PLS model (all calibration and validation results in PLS model are stored as objects of plsres class) or use function [predict.pls](#).

The object gives access to all PLS results as well as to the plotting methods for visualisation of the results. The plsres class also inherits all properties and methods of regres - general class for regression results.

If no reference values provided, regression statistics will not be calculated and most of the plots not available. The class is also used for cross-validation results, in this case some of the values and methods are not available (e.g. scores and scores plot, etc.).

All plots are based on `mdaplot` function, so most of its options can be used (e.g. color grouping, etc.).

RPD is ratio of standard deviation of response values to standard error of prediction (SDy/SEP).

### Value

Returns an object of `plsres` class with following fields:

<code>ncomp</code>	number of components included to the model.
<code>ncomp.selected</code>	selected (optimal) number of components.
<code>y.ref</code>	a matrix with reference values for responses.
<code>y.pred</code>	a matrix with predicted values for responses.
<code>rmse</code>	a matrix with root mean squared error values for each response and component.
<code>slope</code>	a matrix with slope values for each response and component.
<code>r2</code>	a matrix with determination coefficients for each response and component.
<code>bias</code>	a matrix with bias values for each response and component.
<code>sep</code>	a matrix with standard error values for each response and component.
<code>rpd</code>	a matrix with RPD values for each response and component.
<code>xdecomp</code>	decomposition of predictors (object of class <code>ldecomp</code> ).
<code>ydecomp</code>	decomposition of responses (object of class <code>ldecomp</code> ).
<code>info</code>	information about the object.

### See Also

Methods for `plsres` objects:

<code>print</code>	prints information about a <code>plsres</code> object.
<code>summary.plsres</code>	shows performance statistics for the results.
<code>plot.plsres</code>	shows plot overview of the results.
<code>plotXScores.plsres</code>	shows scores plot for x decomposition.
<code>plotXYScores.plsres</code>	shows scores plot for x and y decomposition.
<code>plotXVariance.plsres</code>	shows explained variance plot for x decomposition.
<code>plotYVariance.plsres</code>	shows explained variance plot for y decomposition.
<code>plotXCumVariance.plsres</code>	shows cumulative explained variance plot for y decomposition.
<code>plotYCumVariance.plsres</code>	shows cumulative explained variance plot for y decomposition.
<code>plotXResiduals.plsres</code>	shows T2 vs. Q plot for x decomposition.
<code>plotYResiduals.plsres</code>	shows residuals plot for y values.

Methods inherited from `regres` class (parent class for `plsres`):

<code>plotPredictions.regres</code>	shows predicted vs. measured plot.
<code>plotRMSE.regres</code>	shows RMSE plot.

See also [pls](#) - a class for PLS models.

### Examples

```
### Examples of using PLS result class
library(mdatools)
## 1. Make a PLS model for concentration of first component
## using full-cross validation and get calibration results

data(simdata)
x = simdata$spectra.c
y = simdata$conc.c[, 1]

model = pls(x, y, ncomp = 8, cv = 1)
model = selectCompNum(model, 2)
res = model$calres

summary(res)
plot(res)

## 2. Make a PLS model for concentration of first component
## and apply model to a new dataset

data(simdata)
x = simdata$spectra.c
y = simdata$conc.c[, 1]

model = pls(x, y, ncomp = 6, cv = 1)
model = selectCompNum(model, 2)

x.new = simdata$spectra.t
y.new = simdata$conc.t[, 1]
res = predict(model, x.new, y.new)

summary(res)
plot(res)

## 3. Show variance and error plots for PLS results
par(mfrow = c(2, 2))
plotXCumVariance(res, type = 'h')
plotYCumVariance(res, type = 'b', show.labels = TRUE, legend.position = 'bottomright')
plotRMSE(res)
plotRMSE(res, type = 'h', show.labels = TRUE)
par(mfrow = c(1, 1))

## 4. Show scores plots for PLS results
## (for results plot we can use color grouping)
par(mfrow = c(2, 2))
plotXScores(res)
plotXScores(res, show.labels = TRUE, cgroup = y.new)
plotXYScores(res)
plotXYScores(res, comp = 2, show.labels = TRUE)
par(mfrow = c(1, 1))
```

```
## 5. Show predictions and residuals plots for PLS results
par(mfrow = c(2, 2))
plotXResiduals(res, show.label = TRUE, cgroup = y.new)
plotYResiduals(res, show.label = TRUE)
plotPredictions(res)
plotPredictions(res, ncomp = 4, xlab = 'C, reference', ylab = 'C, predictions')
par(mfrow = c(1, 1))
```

---

predict.mcrals	<i>MCR ALS predictions</i>
----------------	----------------------------

---

### Description

Applies MCR-ALS model to a new set of spectra and returns matrix with contributions.

### Usage

```
## S3 method for class 'mcrals'
predict(object, x, ...)
```

### Arguments

object	an MCR model (object of class mcr).
x	spectral values (matrix or data frame).
...	other arguments.

### Value

Matrix with contributions

---

predict.mcrpure	<i>MCR predictions</i>
-----------------	------------------------

---

### Description

Applies MCR model to a new set of spectra and returns matrix with contributions.

### Usage

```
## S3 method for class 'mcrpure'
predict(object, x, ...)
```

**Arguments**

object            an MCR model (object of class mcr).  
 x                spectral values (matrix or data frame).  
 ...              other arguments.

**Value**

Matrix with contributions

---

predict.pca	<i>PCA predictions</i>
-------------	------------------------

---

**Description**

Applies PCA model to a new data set.

**Usage**

```
## S3 method for class 'pca'
predict(object, x, ...)
```

**Arguments**

object            a PCA model (object of class pca).  
 x                data values (matrix or data frame).  
 ...              other arguments.

**Value**

PCA results (an object of class pcares)

---

predict.pls	<i>PLS predictions</i>
-------------	------------------------

---

**Description**

Applies PLS model to a new data set

**Usage**

```
## S3 method for class 'pls'
predict(object, x, y = NULL, cv = FALSE, ...)
```



**Arguments**

object	a PLS model (object of class pls)
x	a matrix with x values (predictors)
y	a matrix with reference y values (responses)
cv	logical, shall predictions be made for cross-validation procedure or not
...	other arguments

**Details**

See examples in help for [pls](#) function.

**Value**

PLS results (an object of class [plsres](#))

---

predict.plsda	<i>PLS-DA predictions</i>
---------------	---------------------------

---

**Description**

Applies PLS-DA model to a new data set

**Usage**

```
## S3 method for class 'plsda'
predict(object, x, c.ref = NULL, ...)
```

**Arguments**

object	a PLS-DA model (object of class plsda)
x	a matrix with x values (predictors)
c.ref	a vector with reference class values (should be a factor)
...	other arguments

**Details**

See examples in help for [plsda](#) function.

**Value**

PLS-DA results (an object of class [plsdares](#))

---

predict.simca            *SIMCA predictions*

---

**Description**

Applies SIMCA model to a new data set

**Usage**

```
## S3 method for class 'simca'
predict(object, x, c.ref = NULL, cal = FALSE, ...)
```

**Arguments**

object	a SIMCA model (object of class simca)
x	a matrix with x values (predictors)
c.ref	a vector with reference class names (same as class names for models)
cal	logical, are predictions for calibration set or not
...	other arguments

**Details**

See examples in help for [simca](#) function.

**Value**

SIMCA results (an object of class simcares)

---

predict.simcam            *SIMCA multiple classes predictions*

---

**Description**

Applies SIMCAM model (SIMCA for multiple classes) to a new data set

**Usage**

```
## S3 method for class 'simcam'
predict(object, x, c.ref = NULL, ...)
```

**Arguments**

object	a SIMCAM model (object of class simcam)
x	a matrix with x values (predictors)
c.ref	a vector with reference class names (same as class names in models)
...	other arguments

**Details**

See examples in help for [simcam](#) function.

**Value**

SIMCAM results (an object of class `simcamres`)

---

prep	<i>Class for preprocessing object</i>
------	---------------------------------------

---

**Description**

Class for preprocessing object

**Usage**

```
prep(name, params = NULL, method = NULL)
```

**Arguments**

name	short text with name for the preprocessing method.
params	a list with parameters for the method (if NULL - default parameters will be used).
method	method to call when applying the preprocessing, provide it only for user defined methods.

**Details**

Use this class to create a list with a sequence of preprocessing methods to keep them together in right order and with defined parameters. The list/object can be provided as an extra argument to any modelling function (e.g. `pca`, `pls`, etc), so the optimal model parameters and the optimal preprocessing will be stored together and can be applied to a raw data by using method `predict`.

For your own preprocessing method you need to create a function, which takes matrix with values (dataset) as the first argument, does something and then return a matrix with the same dimension and same attributes as the result. The method can have any number of optional parameters.

See Bookdown tutorial for details.

---

prep.alsbasecorr      *Baseline correction using assymetric least squares*

---

### Description

Baseline correction using assymetric least squares

### Usage

```
prep.alsbasecorr(data, plambda = 5, p = 0.1, max.niter = 10)
```

### Arguments

data	matrix with spectra (rows correspond to individual spectra)
plambda	power of the penalty parameter (e.g. if plambda = 5, lambda = 10 <sup>5</sup> )
p	assymetry ratio (should be between 0 and 1)
max.niter	maximum number of iterations

### Details

The function implements baseline correction algorithm based on Whittaker smoother. The method was first shown in [1]. The function has two main parameters - power of a penalty parameter (usually varies between 2 and 9) and the ratio of assymetry (usually between 0.1 and 0.001). The choice of the parameters depends on how broad the disturbances of the baseline are and how narrow the original spectral peaks are.

### Value

preprocessed spectra.

### Examples

```
# take spectra from carbs dataset
data(carbs)
spectra = mda.t(carbs$$S)

# apply the correction
pspectra = prep.alsbasecorr(spectra, plambda = 3, p = 0.01)

# show the original and the corrected spectra individually
par(mfrow = c(3, 1))
for (i in 1:3) {
  mdaplotg(list(
    original = mda.subset(spectra, i),
    corrected = mda.subset(pspectra, i)
  ), type = "l", col = c("black", "red"), lwd = c(2, 1), main = rownames(spectra)[i])
}
```

---

```
prep.autoscale
```

*Autoscale values*

---

**Description**

Autoscale (mean center and standardize) values in columns of data matrix.

The use of 'max.cov' allows to avoid overestimation of inert variables, which vary very little. Note, that the 'max.cov' value is already in percent, e.g. if 'max.cov = 0.1' it will compare the coefficient of variation of every variable with 0.1 want to use this option simply keep 'max.cov = 0'.

**Usage**

```
prep.autoscale(data, center = TRUE, scale = FALSE, max.cov = 0)
```

**Arguments**

data	a matrix with data values
center	a logical value or vector with numbers for centering
scale	a logical value or vector with numbers for weighting
max.cov	columns that have coefficient of variation (in percent) below or equal to 'max.cov' will not be scaled

**Value**

data matrix with processed values

---

```
prep.generic
```

*Generic function for preprocessing*

---

**Description**

Generic function for preprocessing

**Usage**

```
prep.generic(x, f, ...)
```

**Arguments**

x	data matrix to be preprocessed
f	function for preprocessing
...	arguments for the function f

---

prep.list	<i>Shows information about all implemented preprocessing methods.</i>
-----------	---

---

**Description**

Shows information about all implemented preprocessing methods.

**Usage**

```
prep.list()
```

---

prep.msc	<i>Multiplicative Scatter Correction transformation</i>
----------	---

---

**Description**

Applies Multiplicative Scatter Correction (MSC) transformation to data matrix (spectra)

**Usage**

```
prep.msc(data, mspectrum = NULL)
```

**Arguments**

data	a matrix with data values (spectra)
mspectrum	mean spectrum (if NULL will be calculated from spectra)

**Details**

MSC is used to remove scatter effects (baseline offset and slope) from spectral data, e.g. NIR spectra.

@examples

```
### Apply MSC to spectra from simdata
```

```
library(mdatools) data(simdata)
```

```
spectra = simdata$spectra.c cspectra = prep.msc(spectra)
```

```
par(mfrow = c(2, 1)) mdaplot(spectra, type = "l", main = "Before MSC") mdaplot(cspectra, type = "l", main = "After MSC")
```

**Value**

preprocessed spectra (calculated mean spectrum is assigned as attribut 'mspectrum')

---

```
prep.norm
```

*Normalization*

---

**Description**

Normalizes signals (rows of data matrix).

**Usage**

```
prep.norm(data, type = "area", col.ind = NULL)
```

**Arguments**

data	a matrix with data values
type	type of normalization "area", "length", "sum", "snv", or "is".
col.ind	indices of columns (can be either integer or logical values) for normalization to internal standard peak.

**Details**

The "area", "length", "sum" types do preprocessing to unit area (sum of absolute values), length or sum of all values in every row of data matrix. Type "snv" does the Standard Normal Variate normalization, similar to [prep.snv](#). Type "is" does the normalization to internal standard peak, whose position is defined by parameter 'col.ind'. If the position is a single value, the rows are normalized to the height of this peak. If 'col.ind' points on several adjacent values, the rows are normalized to the area under the peak - sum of the intensities.

**Value**

data matrix with normalized values

---

```
prep.ref2km
```

*Kubelka-Munk transformation*

---

**Description**

Applies Kubelka-Munk (km) transformation to data matrix (spectra)

**Usage**

```
prep.ref2km(data)
```

**Arguments**

data	a matrix with spectra values (absolute reflectance values)
------	--

**Details**

Kubelka-Munk is useful preprocessing method for diffuse reflection spectra (e.g. taken for powders or rough surface). It transforms the reflectance spectra  $R$  to  $K/M$  units as follows:  $(1 - R)^2 / 2R$

**Value**

preprocessed spectra.

---

prep.savgol

*Savitzky-Golay filter*

---

**Description**

Applies Savitzky-Golay filter to the rows of data matrix

**Usage**

```
prep.savgol(data, width = 3, porder = 1, dorder = 0)
```

**Arguments**

data	a matrix with data values
width	width of the filter window
porder	order of polynomial used for smoothing
dorder	order of derivative to take (0 - no derivative)

**Details**

The function implements algorithm described in [1] which handles the edge points correctly and does not require to cut the spectra.

**References**

1. Peter A. Gorry. General least-squares smoothing and differentiation by the convolution (Savitzky-Golay) method. Anal. Chem. 1990, 62, 6, 570–573, <https://doi.org/10.1021/ac00205a007>.



---

`prep.snv`*Standard Normal Variate transformation*

---

**Description**

Applies Standard Normal Variate (SNV) transformation to the rows of data matrix

**Usage**

```
prep.snv(data)
```

**Arguments**

`data` a matrix with data values

**Details**

SNV is a simple preprocessing to remove scatter effects (baseline offset and slope) from spectral data, e.g. NIR spectra.

@examples

```
### Apply SNV to spectra from simdata
```

```
library(mdatools) data(simdata)
```

```
spectra = simdata$spectra.c wavelength = simdata$wavelength
```

```
cspectra = prep.snv(spectra)
```

```
par(mfrow = c(2, 1)) mdaplot(cbind(wavelength, t(spectra)), type = 'l', main = 'Before SNV')
```

```
mdaplot(cbind(wavelength, t(cspectra)), type = 'l', main = 'After SNV')
```

**Value**

data matrix with processed values

---

`prep.transform`*Transformation*

---

**Description**

Transforms values from using any mathematical function (e.g. log).

**Usage**

```
prep.transform(data, fun, ...)
```

**Arguments**

data            a matrix with data values  
 fun            reference to a transformation function, e.g. 'log' or 'function(x) x^2'.  
 ...            optional parameters for the transformation function

**Value**

data matrix with transformed values

**Examples**

```
# generate a matrix with two columns
y <- cbind(rnorm(100, 10, 1), rnorm(100, 20, 2))

# apply log transformation
py1 = prep.transform(y, log)

# apply power transformation
py2 = prep.transform(y, function(x) x^-1.25)

# show distributions
par(mfrow = c(2, 3))
for (i in 1:2) {
  hist(y[, i], main = paste0("Original values, column #", i))
  hist(py1[, i], main = paste0("Log-transformed, column #", i))
  hist(py2[, i], main = paste0("Power-transformed, column #", i))
}
```

---

```
prep.varsel
```

```
Variable selection
```

---

**Description**

Returns dataset with selected variables

**Usage**

```
prep.varsel(data, var.ind)
```

**Arguments**

data            a matrix with data values  
 var.ind        indices of variables (columns) to select, can bet either numeric or logical

**Value**

data matrix with the selected variables (columns)

---

preparePlotData	<i>Take dataset and prepare them for plot</i>
-----------------	---

---

### Description

The function checks that 'data' contains correct numeric values, check for mandatory attributes (row and column names, x- and y-axis values and names, etc.) and add them if necessary.

Another things is to remove hidden columns and split the rest to visible and hidden values (if excluded rows are present).

### Usage

```
preparePlotData(data)
```

### Arguments

data	dataset (vector, matrix or data frame)
------	--

---

prepCalData	<i>Prepares calibration data</i>
-------------	----------------------------------

---

### Description

Prepares calibration data

### Usage

```
prepCalData(x, exclrows = NULL, exclcols = NULL, min.nrows = 1, min.ncols = 2)
```

### Arguments

x	matrix or data frame with values (calibration set)
exclrows	rows to be excluded from calculations (numbers, names or vector with logical values)
exclcols	columns to be excluded from calculations (numbers, names or vector with logical values)
min.nrows	smallest number of rows which must be in the dataset
min.ncols	smallest number of columns which must be in the dataset

---

print.classres	<i>Print information about classification result object</i>
----------------	---

---

**Description**

Generic print function for classification results. Prints information about major fields of the object.

**Usage**

```
## S3 method for class 'classres'  
print(x, str = "Classification results (class classres)\nMajor fields:", ...)
```

**Arguments**

x	classification results (object of class plsdares, simcamres, etc.).
str	User specified text (e.g. to be used for particular method, like PLS-DA, etc).
...	other arguments

---

print.ipls	<i>Print method for iPLS</i>
------------	------------------------------

---

**Description**

Prints information about the iPLS object structure

**Usage**

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

**Arguments**

x	a iPLS (object of class ipls)
...	other arguments

---

print.ldecomp	<i>Print method for linear decomposition</i>
---------------	--

---

**Description**

Generic print function for linear decomposition. Prints information about the ldecomp object.

**Usage**

```
## S3 method for class 'ldecomp'  
print(x, str = NULL, ...)
```

**Arguments**

x	object of class ldecomp
str	user specified text to show as a description of the object
...	other arguments

---

print.mcrals	<i>Print method for mcrpure object</i>
--------------	--

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	mcrpure object
...	other arguments

---

print.mcrpure	<i>Print method for mcrpure object</i>
---------------	--

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	mcrpure object
...	other arguments

---

print.pca	<i>Print method for PCA model object</i>
-----------	--

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	a PCA model (object of class pca)
...	other arguments

---

print.pcares	<i>Print method for PCA results object</i>
--------------	--

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	PCA results (object of class pcares)
...	other arguments

---

print.pls	<i>Print method for PLS model object</i>
-----------	--

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	a PLS model (object of class pls)
...	other arguments

---

print.plsda	<i>Print method for PLS-DA model object</i>
-------------	---

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	a PLS-DA model (object of class plsda)
...	other arguments

---

print.plsdares	<i>Print method for PLS-DA results object</i>
----------------	---

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	PLS-DA results (object of class plsdares)
...	other arguments



---

print.plsres	<i>print method for PLS results object</i>
--------------	--

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	PLS results (object of class plsres)
...	other arguments

---

print.randtest	<i>Print method for randtest object</i>
----------------	---

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	a randomization test results (object of class randtest)
...	other arguments

---

print.regcoeffs	<i>print method for regression coefficients class</i>
-----------------	---

---

**Description**

prints regression coefficient values for given response number and amount of components

**Usage**

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

**Arguments**

x	regression coefficients object (class regcoeffs)
...	other arguments

---

print.regmodel	<i>Print method for PLS model object</i>
----------------	--

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	a regression model (object of class regmodel)
...	other arguments

---

print.regres	<i>print method for regression results object</i>
--------------	---

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	regression results (object of class regres)
...	other arguments

---

print.simca	<i>Print method for SIMCA model object</i>
-------------	--

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	a SIMCA model (object of class simca)
...	other arguments

---

print.simcam	<i>Print method for SIMCAM model object</i>
--------------	---

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	a SIMCAM model (object of class simcam)
...	other arguments

---

print.simcamres	<i>Print method for SIMCAM results object</i>
-----------------	---

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	SIMCAM results (object of class simcamres)
...	other arguments

---

print.simcares	<i>Print method for SIMCA results object</i>
----------------	--

---

**Description**

Prints information about the object structure

**Usage**

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

**Arguments**

x	SIMCA results (object of class simcares)
...	other arguments

---

randtest	<i>Randomization test for PLS regression</i>
----------	--

---

**Description**

randtest is used to carry out randomization/permutation test for a PLS regression model

**Usage**

```
randtest(  
  x,  
  y,  
  ncomp = 15,  
  center = T,  
  scale = F,  
  nperm = 1000,  
  sig.level = 0.05,  
  silent = TRUE,  
  exclcols = NULL,  
  exclrows = NULL  
)
```

**Arguments**

x	matrix with predictors.
y	vector or one-column matrix with response.
ncomp	maximum number of components to test.
center	logical, center or not predictors and response values.
scale	logical, scale (standardize) or not predictors and response values.
nperm	number of permutations.
sig.level	significance level.
silent	logical, show or not test progress.
exclcols	columns of x to be excluded from calculations (numbers, names or vector with logical values)
exclrows	rows to be excluded from calculations (numbers, names or vector with logical values)

**Details**

The class implements a method for selection of optimal number of components in PLS1 regression based on the randomization test [1]. The basic idea is that for each component from 1 to ncomp a statistic T, which is a covariance between t-score (X score, derived from a PLS model) and the reference Y values, is calculated. By repeating this for randomly permuted Y-values a distribution of the statistic is obtained. A parameter alpha is computed to show how often the statistic T, calculated for permuted Y-values, is the same or higher than the same statistic, calculated for original data without permutations.

If a component is important, then the covariance for unpermuted data should be larger than the covariance for permuted data and therefore the value for alpha will be quite small (there is still a small chance to get similar covariance). This makes alpha very similar to p-value in a statistical test.

The randtest procedure calculates alpha for each component, the values can be observed using summary or plot functions. There are also several function, allowing e.g. to show distribution of statistics and the critical value for each component.

**Value**

Returns an object of randtest class with following fields:

nperm	number of permutations used for the test.
stat	statistic values calculated for each component.
alpha	alpha values calculated for each component.
statperm	matrix with statistic values for each permutation.
corrperm	matrix with correlation between predicted and reference y-values for each permutation.
ncomp.selected	suggested number of components.

## References

S. Wiklund et al. Journal of Chemometrics 21 (2007) 427-439.

## See Also

Methods for randtest objects:

<code>print.randtest</code>	prints information about a randtest object.
<code>summary.randtest</code>	shows summary statistics for the test.
<code>plot.randtest</code>	shows bar plot for alpha values.
<code>plotHist.randtest</code>	shows distribution of statistic plot.
<code>plotCorr.randtest</code>	shows determination coefficient plot.

## Examples

```
### Examples of using the test

## Get the spectral data from Simdata set and apply SNV transformation

data(simdata)

y = simdata$conc.c[, 3]
x = simdata$spectra.c
x = prep.snv(x)

## Run the test and show summary
## (normally use higher nperm values > 1000)
r = randtest(x, y, ncomp = 4, nperm = 200, silent = FALSE)
summary(r)

## Show plots

par( mfrow = c(3, 2))
plot(r)
plotHist(r, ncomp = 3)
plotHist(r, ncomp = 4)
plotCorr(r, 3)
plotCorr(r, 4)
par( mfrow = c(1, 1))
```

---

regcoeffs

*Regression coefficients*

---

## Description

class for storing and visualisation of regression coefficients for regression models

**Usage**

```
regcoeffs(coeffs, ci.coeffs = NULL, use.mean = TRUE)
```

**Arguments**

<code>coeffs</code>	array (npred x ncomp x nresp) with regression coefficients
<code>ci.coeffs</code>	array (npred x ncomp x nresp x cv) with regression coefficients for computing confidence intervals (e.g. from cross-validation) using Jack-Knifing method
<code>use.mean</code>	logical, tells how to compute standard error for regression coefficients. If TRUE mean values for <code>ci.coeffs</code> is computed first. If FALSE, values (coefficients computed for global model) are used as mean.

**Value**

a list (object of `regcoeffs` class) with fields, including:

<code>values</code>	an array (nvar x ncomp x ny) with regression coefficients
<code>se</code>	an array (nvar x ncomp x ny) with standard errors for the coefficients
<code>t.values</code>	an array (nvar x ncomp x ny) with t-values for the coefficients
<code>p.values</code>	an array (nvar x ncomp x ny) with p-values for coefficients

last three fields are available if parameter `ci.coeffs` was provided.

Check also [confint.regcoeffs](#), [summary.regcoeffs](#) and [plot.regcoeffs](#).

---

`regcoeffs.getStats`      *Distribution statistics for regression coefficients*

---

**Description**

calculates standard error, t-values and p-values for regression coefficients based on Jack-Knifing method.

**Usage**

```
regcoeffs.getStats(coeffs, ci.coeffs = NULL, use.mean = TRUE)
```

**Arguments**

<code>coeffs</code>	array (npred x ncomp x nresp) with regression coefficients
<code>ci.coeffs</code>	array (npred x ncomp x nresp x cv) with regression coefficients for computing confidence intervals (e.g. from cross-validation) using Jack-Knifing method
<code>use.mean</code>	logical, tells how to compute standard error for regression coefficients. If TRUE mean values for <code>ci.coeffs</code> is computed first. If FALSE, values (coefficients computed for global model) are used as mean.



**Value**

a list with statistics three arrays: standard error, t-values and p-values computed for each regression coefficient.

---

regres	<i>Regression results</i>
--------	---------------------------

---

**Description**

Class for storing and visualisation of regression predictions

**Usage**

```
regres(y.pred, y.ref = NULL, ncomp.selected = 1)
```

**Arguments**

`y.pred` vector or matrix with y predicted values  
`y.ref` vector with reference (measured) y values  
`ncomp.selected` if y.pred calculated for different components, which to use as default

**Value**

a list (object of regres class) with fields, including:

<code>y.pred</code>	a matrix with predicted values
<code>y.ref</code>	a vector with reference (measured) values
<code>ncomp.selected</code>	selected column/number of components for predictions
<code>rmse</code>	root mean squared error for predicted vs measured values
<code>slope</code>	slope for predicted vs measured values
<code>r2</code>	coefficient of determination for predicted vs measured values
<code>bias</code>	bias for predicted vs measured values
<code>rpd</code>	RPD values

---

regres.bias	<i>Prediction bias</i>
-------------	------------------------

---

**Description**

Calculates matrix with bias (average prediction error) for every response and components

**Usage**

```
regres.bias(err)
```

**Arguments**

err                    vector with difference between reference and predicted y-values

---

regres.err                    *Error of prediction*

---

**Description**

Calculates array of differences between predicted and reference values.

**Usage**

regres.err(y.pred, y.ref)

**Arguments**

y.pred                    matrix with predicted values  
y.ref                    vector with reference values

---

regres.r2                    *Determination coefficient*

---

**Description**

Calculates matrix with coefficient of determination for every response and components

**Usage**

regres.r2(err, ytot)

**Arguments**

err                    vector with difference between reference and predicted y-values  
ytot                    total variance for y-values

---

regres.rmse	<i>RMSE</i>
-------------	-------------

---

**Description**

Calculates matrix with root mean squared error of prediction for every response and components.

**Usage**

```
regres.rmse(err)
```

**Arguments**

err	vector with difference between reference and predicted y-values
-----	---

---

regres.slope	<i>Slope</i>
--------------	--------------

---

**Description**

Calculates matrix with slope of predicted and measured values for every response and components.

**Usage**

```
regres.slope(y.pred, y.ref)
```

**Arguments**

y.pred	matrix with predicted values
y.ref	vector with reference values

---

regress.addattrs	<i>Add names and attributes to matrix with statistics</i>
------------------	---

---

**Description**

Add names and attributes to matrix with statistics

**Usage**

```
regress.addattrs(stat, attrs, name)
```

**Arguments**

stat	matrix with statistics
attrs	attributes from error matrix
name	name of statistic

---

repmat	<i>Replicate matrix x</i>
--------	---------------------------

---

**Description**

Replicate matrix x

**Usage**

```
repmat(x, nrows, ncols = nrows)
```

**Arguments**

x	original matrix
nrows	number of times replicate matrix row wise
ncols	number of times replicate matrix columns wise

---

rotationMatrixToX1	<i>Creates a rotation matrix to map a vector x to [1 0 0 ... 0]</i>
--------------------	---

---

**Description**

Creates a rotation matrix to map a vector x to [1 0 0 ... 0]

**Usage**

```
rotationMatrixToX1(x)
```

**Arguments**

x	Vector (sequence with J coordinates)
---	--------------------------------------

**Value**

Rotation matrix (JxJ)

---

selectCompNum	<i>Select optimal number of components for a model</i>
---------------	--

---

**Description**

Generic function for selecting number of components for multivariate models (e.g. PCA, PLS, ...)

**Usage**

```
selectCompNum(obj, ncomp = NULL, ...)
```

**Arguments**

obj	a model object
ncomp	number of components to select
...	other arguments

---

selectCompNum.pca	<i>Select optimal number of components for PCA model</i>
-------------------	--

---

**Description**

Allows user to select optimal number of components for a PCA model

**Usage**

```
## S3 method for class 'pca'  
selectCompNum(obj, ncomp, ...)
```

**Arguments**

obj	PCA model (object of class pca)
ncomp	number of components to select
...	other parameters if any

**Value**

the same model with selected number of components

---

selectCompNum.pls      *Select optimal number of components for PLS model*

---

### Description

Allows user to select optimal number of components for PLS model

### Usage

```
## S3 method for class 'pls'
selectCompNum(obj, ncomp = NULL, selcrit = obj$ncomp.selcrit, ...)
```

### Arguments

obj	PLS model (object of class pls)
ncomp	number of components to select
selcrit	criterion for selecting optimal number of components ('min' for first local minimum of RMSECV and 'wold' for Wold's rule.)
...	other parameters if any

### Details

The method sets `ncomp.selected` parameter for the model and return it back. The parameter points out to the optimal number of components in the model. You can either specify it manually, as argument `ncomp`, or use one of the algorithms for automatic selection.

Automatic selection by default based on cross-validation statistics. If no cross-validation results are found in the model, the method will use test set validation results. If they are not available as well, the model will use calibration results and give a warning as in this case the selected number of components will lead to overfitted model.

There are two algorithms for automatic selection you can chose between: either first local minimum of RMSE ('selcrit="min"') or Wold's rule ('selcrit="wold"').

The first local minimum criterion finds at which component,  $A$ , error of prediction starts raising and selects  $(A - 1)$  as the optimal number. The Wold's criterion finds which component  $A$  does not make error smaller at least by 5 as the optimal number.

If model is PLS2 model (has several response variables) the method computes optimal number of components for each response and returns the smallest value. For example, if for the first response 2 components give the smallest error and for the second response this number is 3,  $A = 2$  will be selected as a final result.

It is not recommended to use automatic selection for real applications, always investigate your model (via RMSE, Y-variance plot, regression coefficients) to make correct decision.

See examples in help for [pls](#) function.

### Value

the same model with selected number of components

---

selratio	<i>Selectivity ratio calculation</i>
----------	--------------------------------------

---

**Description**

Calculates selectivity ratio for each component and response variable in the PLS model

**Usage**

```
selratio(obj, ncomp = obj$ncomp.selected)
```

**Arguments**

obj	a PLS model (object of class pls)
ncomp	number of components to count

**Value**

array nvar x ncomp x ny with selectivity ratio values

**References**

[1] Tarja Rajalahti et al. Chemometrics and Laboratory Systems, 95 (2009), pp. 35-48.

---

setDistanceLimits	<i>Set residual distance limits</i>
-------------------	-------------------------------------

---

**Description**

Calculates and set critical limits for residuals of PCA model

**Usage**

```
setDistanceLimits(obj, ...)
```

**Arguments**

obj	a model object
...	other parameters

---

setDistanceLimits.pca *Compute and set statistical limits for Q and T2 residual distances.*

---

### Description

Computes statistical limits for orthogonal and score distances (based on calibration set) and assign the calculated values as model properties.

### Usage

```
## S3 method for class 'pca'
setDistanceLimits(
  obj,
  lim.type = obj$lim.type,
  alpha = obj$alpha,
  gamma = obj$gamma,
  ...
)
```

### Arguments

obj	object with PCA model
lim.type	type of limits ("jm", "chisq", "ddmoments", "ddrobust")
alpha	significance level for detection of extreme objects
gamma	significance level for detection of outliers (for data driven approach)
...	other arguments

### Details

The limits can be accessed as fields of model objects: \$Qlim and \$T2lim. Each is a matrix with four rows and ncomp columns. First row contains critical limits for extremes, second row - for outliers, third row contains mean value for corresponding distance (or its robust estimate in case of lim.type = "ddrobust") and last row contains the degrees of freedom.

### Value

Object models with the two fields updated.



---

setDistanceLimits.pls *Compute and set statistical limits for residual distances.*

---

### Description

Computes statistical limits for orthogonal and score distances (x-decomposition) and orthogonal distance (y-decomposition) based on calibration set and assign the calculated values as model properties.

### Usage

```
## S3 method for class 'pls'
setDistanceLimits(
  obj,
  lim.type = obj$lim.type,
  alpha = obj$alpha,
  gamma = obj$gamma,
  ...
)
```

### Arguments

obj	object with PLS model
lim.type	type of limits ("jm", "chisq", "ddmoments", "ddrobust")
alpha	significance level for detection of extreme objects
gamma	significance level for detection of outliers (for data driven approach)
...	other arguments

### Details

The limits can be accessed as fields of model objects: \$Qlim, \$T2lim, and \$Zlim. Each is a matrix with four rows and ncomp columns. In case of limits for x-decomposition, first row contains critical limits for extremes, second row - for outliers, third row contains mean value for corresponding distances (or its robust estimate in case of lim.type = "ddrobust") and last row contains the degrees of freedom.

### Value

Object models with the three fields updated.

---

showDistanceLimits	<i>Show residual distance limits</i>
--------------------	--------------------------------------

---

**Description**

Calculates and set critical limits for residuals of PCA model

**Usage**

```
showDistanceLimits(obj, ...)
```

**Arguments**

obj	a model object
...	other parameters

---

showLabels	<i>Show labels on plot</i>
------------	----------------------------

---

**Description**

Show labels on plot

**Usage**

```
showLabels(
  ps,
  show.excluded = FALSE,
  pos = 3,
  cex = 0.65,
  col = "darkgray",
  force.x.values = NULL,
  bwd = 0.8
)
```

**Arguments**

ps	'plotseries' object
show.excluded	logical, are excluded rows also shown on the plot
pos	position of the labels relative to the data points
cex	size of the labels text
col	color of the labels text
force.x.values	vector with forced x-values (or NULL)
bwd	bar width in case of bar plot

---

showPredictions	<i>Predictions</i>
-----------------	--------------------

---

**Description**

Predictions

**Usage**

```
showPredictions(obj, ...)
```

**Arguments**

obj	a model or result object
...	other arguments

**Details**

Generic function for showing predicted values for classification or regression model or results

---

showPredictions.classres	<i>Show predicted class values</i>
--------------------------	------------------------------------

---

**Description**

Shows a table with predicted class values for classification result.

**Usage**

```
## S3 method for class 'classres'
showPredictions(obj, ncomp = obj$ncomp.selected, ...)
```

**Arguments**

obj	object with classification results (e.g. plsdares or simcamres).
ncomp	number of components to show the predictions for (NULL - use selected for a model).
...	other parameters

**Details**

The function prints a matrix where every column is a class and every row is an data object. The matrix has either -1 (does not belong to the class) or +1 (belongs to the class) values.

---

 simca

*SIMCA one-class classification*


---

### Description

simca is used to make SIMCA (Soft Independent Modelling of Class Analogies) model for one-class classification.

### Usage

```
simca(
  x,
  classname,
  ncomp = min(nrow(x) - 1, ncol(x) - 1, 20),
  x.test = NULL,
  c.test = NULL,
  cv = NULL,
  ...
)
```

### Arguments

x	a numerical matrix with data values.
classname	short text (up to 20 symbols) with class name.
ncomp	maximum number of components to calculate.
x.test	a numerical matrix with test data.
c.test	a vector with classes of test data objects (can be text with names of classes or logical).
cv	cross-validation settings (see details).
...	any other parameters suitable for <a href="#">pca</a> method.

### Details

SIMCA is in fact PCA model with additional functionality, so simca class inherits most of the functionality of [pca](#) class. It uses critical limits calculated for Q and T2 residuals calculated for PCA model for making classification decision.

Cross-validation settings, cv, can be a number or a list. If cv is a number, it will be used as a number of segments for random cross-validation (if cv = 1, full cross-validation will be preformed). If it is a list, the following syntax can be used: `cv = list('rand', nseg, nrep)` for random repeated cross-validation with nseg segments and nrep repetitions or `cv = list('ven', nseg)` for systematic splits to nseg segments ('venetian blinds').

**Value**

Returns an object of `simca` class with following fields:

<code>classname</code>	a short text with class name.
<code>calres</code>	an object of class <code>simcares</code> with classification results for a calibration data.
<code>testres</code>	an object of class <code>simcares</code> with classification results for a test data, if it was provided.
<code>cvres</code>	an object of class <code>simcares</code> with classification results for cross-validation, if this option was chosen.

Fields, inherited from `pca` class:

<code>ncomp</code>	number of components included to the model.
<code>ncomp.selected</code>	selected (optimal) number of components.
<code>loadings</code>	matrix with loading values (nvar x ncomp).
<code>eigenvals</code>	vector with eigenvalues for all existent components.
<code>expvar</code>	vector with explained variance for each component (in percent).
<code>cumexpvar</code>	vector with cumulative explained variance for each component (in percent).
<code>T2lim</code>	statistical limit for T2 distance.
<code>Qlim</code>	statistical limit for Q residuals.
<code>info</code>	information about the model, provided by user when build the model.

**Author(s)**

Sergey Kucheryavskiy (svkucheryavski@gmail.com)

**References**

S. Wold, M. Sjostrom. "SIMCA: A method for analyzing chemical data in terms of similarity and analogy" in B.R. Kowalski (ed.), Chemometrics Theory and Application, American Chemical Society Symposium Series 52, Wash., D.C., American Chemical Society, p. 243-282.

**See Also**

Methods for `simca` objects:

<code>print.simca</code>	shows information about the object.
<code>summary.simca</code>	shows summary statistics for the model.
<code>plot.simca</code>	makes an overview of SIMCA model with four plots.
<code>predict.simca</code>	applies SIMCA model to a new data.

Methods, inherited from `classmodel` class:

<code>plotPredictions.classmodel</code>	shows plot with predicted values.
---	-----------------------------------

<code>plotSensitivity.classmodel</code>	shows sensitivity plot.
<code>plotSpecificity.classmodel</code>	shows specificity plot.
<code>plotMisclassified.classmodel</code>	shows misclassified ratio plot.

Methods, inherited from `pca` class:

<code>selectCompNum.pca</code>	set number of optimal components in the model
<code>plotScores.pca</code>	shows scores plot.
<code>plotLoadings.pca</code>	shows loadings plot.
<code>plotVariance.pca</code>	shows explained variance plot.
<code>plotCumVariance.pca</code>	shows cumulative explained variance plot.
<code>plotResiduals.pca</code>	shows Q vs. T2 residuals plot.

## Examples

```
## make a SIMCA model for Iris setosa class with full cross-validation
library(mdatools)

data = iris[, 1:4]
class = iris[, 5]

# take first 20 objects of setosa as calibration set
se = data[1:20, ]

# make SIMCA model and apply to test set
model = simca(se, "setosa", cv = 1)
model = selectCompNum(model, 1)

# show information, summary and plot overview
print(model)
summary(model)
plot(model)

# show predictions
par(mfrow = c(2, 1))
plotPredictions(model, show.labels = TRUE)
plotPredictions(model, res = "cal", ncomp = 2, show.labels = TRUE)
par(mfrow = c(1, 1))

# show performance, modelling power and residuals for ncomp = 2
par(mfrow = c(2, 2))
plotSensitivity(model)
plotMisclassified(model)
plotLoadings(model, comp = c(1, 2), show.labels = TRUE)
plotResiduals(model, ncomp = 2)
par(mfrow = c(1, 1))
```

---

simcam	<i>SIMCA multiclass classification</i>
--------	--

---

### Description

simcam is used to combine several one-class SIMCA models for multiclass classification.

### Usage

```
simcam(models, info = "")
```

### Arguments

models	list with SIMCA models (simca objects).
info	optional text with information about the the object.

### Details

Besides the possibility for multiclass classification, SIMCAM also provides tools for investigation of relationship among individual models (classes), such as discrimination power of variables, Cooman's plot, model distance, etc.

When create simcam object, the calibration data from all individual SIMCA models is extracted and combined for making predictions and calculate performance of the multi-class model. The results are stored in \$calres field of the model object.

### Value

Returns an object of simcam class with following fields:

models	a list with provided SIMCA models.
dispower	an array with discrimination power of variables for each pair of individual models.
moddist	a matrix with distance between each each pair of individual models.
classnames	vector with names of individual classes.
nclasses	number of classes in the object.
info	information provided by user when create the object.
calres	an object of class <a href="#">simcamres</a> with classification results for a calibration data.

### See Also

Methods for simca objects:

<code>print.simcam</code>	shows information about the object.
<code>summary.simcam</code>	shows summary statistics for the models.
<code>plot.simcam</code>	makes an overview of SIMCAM model with two plots.
<code><a href="#">predict.simcam</a></code>	applies SIMCAM model to a new data.

<code>plotModelDistance.simcam</code>	shows plot with distance between individual models.
<code>plotDiscriminationPower.simcam</code>	shows plot with discrimination power.
<code>plotCooman.simcam</code>	shows Cooman's plot for calibration data.

Methods, inherited from `classmodel` class:

<code>plotPredictions.classmodel</code>	shows plot with predicted values.
<code>plotSensitivity.classmodel</code>	shows sensitivity plot.
<code>plotSpecificity.classmodel</code>	shows specificity plot.
<code>plotMisclassified.classmodel</code>	shows misclassified ratio plot.

Since SIMCAM objects and results are calculated only for optimal number of components, there is no sense to show such plots like sensitivity or specificity vs. number of components. However they are available as for any other classification model.

## Examples

```
## make a multiclass SIMCA model for Iris data
library(mdatools)

# split data
caldata = iris[seq(1, nrow(iris), 2), 1:4]
x.se = caldata[1:25, ]
x.ve = caldata[26:50, ]
x.vi = caldata[51:75, ]

x.test = iris[seq(2, nrow(iris), 2), 1:4]
c.test = iris[seq(2, nrow(iris), 2), 5]

# create individual models
m.se = simca(x.se, classname = "setosa")
m.se = selectCompNum(m.se, 1)

m.vi = simca(x.vi, classname = "virginica")
m.vi = selectCompNum(m.vi, 2)

m.ve = simca(x.ve, classname = "versicolor")
m.ve = selectCompNum(m.ve, 1)

# combine models into SIMCAM objects, show statistics and plots
m = simcam(list(m.se, m.vi, m.ve), info = "simcam model for Iris data")
summary(m)

# show predictions and residuals for calibration data
par(mfrow = c(2, 2))
plotPredictions(m)
plotCooman(m, nc = c(1, 2))
plotModelDistance(m, nc = 1)
plotDiscriminationPower(m, nc = c(1, 2))
```



```

par(mfrow = c(1, 1))

# apply the SIMCAM model to test set and show statistics and plots
res = predict(m, x.test, c.test)
summary(res)
plotPredictions(res)

```

---

```
simcam.getPerformanceStats
```

*Performance statistics for SIMCAM model*

---

### Description

Calculates discrimination power and distance between individual SIMCA models.

### Usage

```
simcam.getPerformanceStats(models, classnames)
```

### Arguments

models	list with SIMCA models (as provided to simcam class)
classnames	names of the classes for each model

---

```
simcamres
```

*Results of SIMCA multiclass classification*

---

### Description

simcamres is used to store results for SIMCA multiclass classification.

### Usage

```
simcamres(cres, pred.res)
```

### Arguments

cres	results of classification (class classres).
pred.res	list with prediction results from each model (pcars objects)

## Details

Class `simcamres` inherits all properties and methods of class `classres`, plus store values necessary to visualise prediction decisions (e.g. Cooman's plot or Residuals plot).

In contrast to `simcares` here only values for optimal (selected) number of components in each individual SIMCA models are presented.

There is no need to create a `simcamres` object manually, it is created automatically when make a SIMCAM model (see `simcam`) or apply the model to a new data (see `predict.simcam`). The object can be used to show summary and plots for the results.

## Value

Returns an object (list) of class `simcamres` with the same fields as `classres` plus extra fields for Q and T2 values and limits:

<code>c.pred</code>	predicted class values.
<code>c.ref</code>	reference (true) class values if provided.
<code>T2</code>	matrix with T2 values for each object and class.
<code>Q</code>	matrix with Q values for each object and class.
<code>T2lim</code>	vector with T2 statistical limits for each class.
<code>Qlim</code>	vector with Q statistical limits for each class.

The following fields are available only if reference values were provided.

<code>tp</code>	number of true positives.
<code>fp</code>	number of false positives.
<code>fn</code>	number of false negatives.
<code>specificity</code>	specificity of predictions.
<code>sensitivity</code>	sensitivity of predictions.

## See Also

Methods for `simcamres` objects:

<code>print.simcamres</code>	shows information about the object.
<code>summary.simcamres</code>	shows statistics for results of classification.
<code>plotCooman.simcamres</code>	makes Cooman's plot.

Methods, inherited from `classres` class:

<code>showPredictions.classres</code>	show table with predicted values.
<code>plotPredictions.classres</code>	makes plot with predicted values.

Check also `simcam`.

**Examples**

```
## see examples for simcam method.
```

---

simcares	<i>Results of SIMCA one-class classification</i>
----------	--

---

**Description**

@description simcares is used to store results for SIMCA one-class classification.

**Usage**

```
simcares(class.res, pca.res = NULL)
```

**Arguments**

class.res	results of classification (class <code>classres</code> ).
pca.res	results of PCA decomposition of data (class <code>pcares</code> ).

**Details**

Class `simcares` inherits all properties and methods of class `pcares`, and has additional properties and functions for representing of classification results, inherited from class `classres`.

There is no need to create a `simcares` object manually, it is created automatically when build a SIMCA model (see `simca`) or apply the model to a new data (see `predict.simca`). The object can be used to show summary and plots for the results.

**Value**

Returns an object (list) of class `simcares` with the same fields as `pcares` plus extra fields, inherited from `classres`:

c.pred	predicted class values (+1 or -1).
c.ref	reference (true) class values if provided.

The following fields are available only if reference values were provided.

tp	number of true positives.
fp	number of false positives.
fn	number of false negatives.
specificity	specificity of predictions.
sensitivity	sensitivity of predictions.

**See Also**

Methods for `simcares` objects:

```
print.simcares    shows information about the object.
summary.simcares  shows statistics for results of classification.
```

Methods, inherited from `classes` class:

```
showPredictions.classes  show table with predicted values.
plotPredictions.classes  predicted classes plot.
plotSensitivity.classes   sensitivity plot.
plotSpecificity.classes   specificity plot.
plotPerformance.classes   performance plot.
```

Methods, inherited from `ldecomp` class:

```
plotResiduals.ldecomp    makes Q2 vs. T2 residuals plot.
plotScores.ldecomp       makes scores plot.
plotVariance.ldecomp     makes explained variance plot.
plotCumVariance.ldecomp  makes cumulative explained variance plot.
```

Check also `simca` and `pcars`.

## Examples

```
## make a SIMCA model for Iris setosa class and show results for calibration set
library(mdatools)

data = iris[, 1:4]
class = iris[, 5]

# take first 30 objects of setosa as calibration set
se = data[1:30, ]

# make SIMCA model and apply to test set
model = simca(se, 'Se')
model = selectCompNum(model, 1)

# show information and summary
print(model$calres)
summary(model$calres)

# show plots
layout(matrix(c(1,1,2,3), 2, 2, byrow = TRUE))
plotPredictions(model$calres, show.labels = TRUE)
plotResiduals(model$calres, show.labels = TRUE)
plotPerformance(model$calres, show.labels = TRUE, legend.position = 'bottomright')
layout(1, 1, 1)

# show predictions table
```

```
showPredictions(model$calres)
```

---

simdata	<i>Spectral data of polyaromatic hydrocarbons mixing</i>
---------	--

---

### Description

Simdata contains training and test set with spectra and concentration values of polyaromatic hydrocarbons mixings.

### Usage

```
data(simdata)
```

### Format

The data is a list with following fields:

\$spectra.c	a matrix (100x150) with spectral values for the training set.
\$spectra.t	a matrix (100x150) with spectral values for the test set.
\$conc.c	a matrix (100x3) with concentration of components for the training set.
\$conc.t	a matrix (100x3) with concentration of components for the test set.
\$wavelength	a vector with spectra wavelength in nm.

### Details

This is a simulated data containing UV/Vis spectra of three component (polyaromatic hydrocarbons) mixings - C1, C2 and C3. The spectral range is between 210 and 360 nm. The spectra were simulated as a linear combination of pure component spectra plus 5% of random noise. The concentration range is (in moles): C1 [0, 1], C2 [0, 0.5], C3 [0, 0.1].

There are 100 mixings in a training set and 50 mixings in a test set. The data can be used for multivariate regression examples.

---

splitExcludedData	<i>Split the excluded part of data</i>
-------------------	--

---

### Description

Split the excluded part of data

### Usage

```
splitExcludedData(data, type)
```

**Arguments**

data	matrix with hidden data values
type	type of plot

---

splitPlotData	<i>Split dataset to x and y values depending on plot type</i>
---------------	---

---

**Description**

Split dataset to x and y values depending on plot type

**Usage**

```
splitPlotData(data, type)
```

**Arguments**

data	matrix with data values (visible or hidden)
type	type of plot

---

summary.classres	<i>Summary statistics about classification result object</i>
------------------	--

---

**Description**

Generic summary function for classification results. Prints performance values for the results.

**Usage**

```
## S3 method for class 'classres'
summary(
  object,
  ncomp = object$ncomp.selected,
  nc = seq_len(object$nclasses),
  ...
)
```

**Arguments**

object	classification results (object of class plsdares, simcamres, etc.).
ncomp	which number of components to make the plot for (use NULL to show results for all available).
nc	vector with class numbers to show the summary for.
...	other arguments

---

summary.ipls	<i>Summary for iPLS results</i>
--------------	---------------------------------

---

**Description**

Shows statistics and algorithm parameters for iPLS results.

**Usage**

```
## S3 method for class 'ipls'
summary(object, glob.ncomp = object$gm$ncomp.selected, ...)
```

**Arguments**

object	a iPLS (object of class ipls)
glob.ncomp	number of components for global PLS model with all intervals
...	other arguments

**Details**

The method shows information on the algorithm parameters as well as a table with selected or excluded interval. The table has the following columns: 'step' showing on which iteration an interval was selected or excluded, 'start' and 'end' show variable indices for the interval, 'nComp' is a number of components used in a model, 'RMSE' is RMSECV for the model and 'R2' is coefficient of determination for the same model.

---

summary.ldecomp	<i>Summary statistics for linear decomposition</i>
-----------------	--

---

**Description**

Generic summary function for linear decomposition. Prints statistic about the decomposition.

**Usage**

```
## S3 method for class 'ldecomp'
summary(object, str = NULL, ...)
```

**Arguments**

object	object of class ldecomp
str	user specified text to show as a description of the object
...	other arguments

summary.mcrals      *Summary method for mcrals object*

---

**Description**

Shows some statistics (explained variance, etc) for the case.

**Usage**

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

**Arguments**

object	mcrals object
...	other arguments

---

summary.mcrpure      *Summary method for mcrpure object*

---

**Description**

Shows some statistics (explained variance, etc) for the case.

**Usage**

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

**Arguments**

object	mcrpure object
...	other arguments



---

summary.pca	<i>Summary method for PCA model object</i>
-------------	--

---

**Description**

Shows some statistics (explained variance, eigenvalues) for the model.

**Usage**

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

**Arguments**

object	a PCA model (object of class pca)
...	other arguments

---

summary.pcares	<i>Summary method for PCA results object</i>
----------------	--

---

**Description**

Shows some statistics (explained variance, eigenvalues) about the results.

**Usage**

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

**Arguments**

object	PCA results (object of class pcares)
...	other arguments

---

`summary.pls`*Summary method for PLS model object*

---

**Description**

Shows performance statistics for the model.

**Usage**

```
## S3 method for class 'pls'  
summary(  
  object,  
  ncomp = object$ncomp.selected,  
  ny = seq_len(nrow(object$yloadings)),  
  ...  
)
```

**Arguments**

<code>object</code>	a PLS model (object of class <code>pls</code> )
<code>ncomp</code>	how many components to count.
<code>ny</code>	which y variables to show the summary for (can be a vector)
<code>...</code>	other arguments

---

`summary.plsda`*Summary method for PLS-DA model object*

---

**Description**

Shows some statistics for the model.

**Usage**

```
## S3 method for class 'plsda'  
summary(  
  object,  
  ncomp = object$ncomp.selected,  
  nc = seq_len(object$nclasses),  
  ...  
)
```

**Arguments**

object	a PLS-DA model (object of class plsda)
ncomp	how many components to use (if NULL - user selected optimal value will be used)
nc	which class to show the summary for (if NULL, will be shown for all)
...	other arguments

---

summary.plsdares	<i>Summary method for PLS-DA results object</i>
------------------	---

---

**Description**

Shows performance statistics for the results.

**Usage**

```
## S3 method for class 'plsdares'
summary(object, nc = seq_len(object$nclasses), ...)
```

**Arguments**

object	PLS-DA results (object of class plsdares)
nc	which class to show the summary for (if NULL, will be shown for all)
...	other arguments

---

summary.plsres	<i>summary method for PLS results object</i>
----------------	--

---

**Description**

Shows performance statistics for the results.

**Usage**

```
## S3 method for class 'plsres'
summary(object, ny = seq_len(object$nresp), ncomp = NULL, ...)
```

**Arguments**

object	PLS results (object of class plsres)
ny	for which response variable show the summary for
ncomp	how many components to use (if NULL - user selected optimal value will be used)
...	other arguments

---

summary.randtest	<i>Summary method for randtest object</i>
------------------	---

---

**Description**

Shows summary for randomization test results.

**Usage**

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

**Arguments**

object	randomization test results (object of class randtest)
...	other arguments

---

summary.regcoeffs	<i>Summary method for regcoeffs object</i>
-------------------	--

---

**Description**

Shows estimated coefficients and statistics (if available).

**Usage**

```
## S3 method for class 'regcoeffs'
summary(object, ncomp = 1, ny = 1, alpha = 0.05, ...)
```

**Arguments**

object	object of class regcoeffs
ncomp	how many components to use
ny	which y variable to show the summary for
alpha	significance level for confidence interval (if statistics available)
...	other arguments

**Details**

Statistics are shown if Jack-Knifing was used when model is calibrated.

---

summary.regmodel	<i>Summary method for regression model object</i>
------------------	---

---

**Description**

Shows performance statistics for the model.

**Usage**

```
## S3 method for class 'regmodel'
summary(
  object,
  ncomp = object$ncomp.selected,
  ny = seq_len(object$res$cal$nresp),
  res = object$res,
  ...
)
```

**Arguments**

object	a regression model (object of class regmodel)
ncomp	number of components to show summary for
ny	which y variables to show the summary for (can be a vector)
res	list of results to show summary for
...	other arguments

---

summary.regres	<i>summary method for regression results object</i>
----------------	---

---

**Description**

Shows performance statistics for the regression results.

**Usage**

```
## S3 method for class 'regres'
summary(object, ncomp = object$ncomp.selected, ny = seq_len(object$nresp), ...)
```

**Arguments**

object	regression results (object of class regres)
ncomp	model complexity to show the summary for (if NULL - shows for all available values)
ny	for which response variable show the summary for (one value or a vector)
...	other arguments

---

summary.simca	<i>Summary method for SIMCA model object</i>
---------------	--

---

**Description**

Shows performance statistics for the model.

**Usage**

```
## S3 method for class 'simca'  
summary(object, ncomp = object$ncomp.selected, res = object$res, ...)
```

**Arguments**

object	a SIMCA model (object of class simca)
ncomp	number of components to show summary for
res	list of result objects to show summary for
...	other arguments

---

summary.simcam	<i>Summary method for SIMCAM model object</i>
----------------	---

---

**Description**

Shows performance statistics for the model.

**Usage**

```
## S3 method for class 'simcam'  
summary(object, nc = seq_len(object$nclasses), ...)
```

**Arguments**

object	a SIMCAM model (object of class simcam)
nc	number of class to show summary for (can be vector)
...	other arguments

---

summary.simcamres      *Summary method for SIMCAM results object*

---

### Description

Shows performance statistics for the results.

### Usage

```
## S3 method for class 'simcamres'  
summary(object, nc = seq_len(object$nclasses), ...)
```

### Arguments

object	SIMCAM results (object of class simcamres)
nc	number of class to show summary for (can be vector)
...	other arguments

---

summary.simcares      *Summary method for SIMCA results object*

---

### Description

Shows performance statistics for the results.

### Usage

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

### Arguments

object	SIMCA results (object of class simcares)
...	other arguments

---

unmix.mcrpure	<i>Unmix spectral data using pure variables estimated before</i>
---------------	--

---

**Description**

Unmix spectral data using pure variables estimated before

**Usage**

```
unmix.mcrpure(obj, x)
```

**Arguments**

obj	mcrpure object
x	matrix with spectra

**Value**

Returns a list with resolved spectra and contributions (matrices).

---

vipcores	<i>VIP scores for PLS model</i>
----------	---------------------------------

---

**Description**

Calculates VIP (Variable Importance in Projection) scores for predictors for given number of components and response variable.

**Usage**

```
vipcores(obj, ncomp = obj$ncomp.selected)
```

**Arguments**

obj	a PLS model (object of class pls)
ncomp	number of components to count

**Details**

May take some time in case of large number of predictors Returns results as a column-vector, with all necessary attributes inherited (e.g. xaxis.values, excluded variables, etc.). If you want to make a plot use for example: `mdaplot(mda.t(v), type = "1")`, where v is a vector with computed VIP scores. Or just try [plotVIPScores.pls](#).



**Value**

matrix  $nvar \times ny$  with VIP score values (columns correspond to responses).

**References**

[1] Il-Gyo Chong, Chi-Hyuck Jun. *Chemometrics and Laboratory Systems*, 78 (2005), pp. 103-112.

# Index

## \* datasets

- carbs, 14
  - pellets, 108
  - people, 109
  - simdata, 261
- as.matrix.classres, 10
- as.matrix.ldecomp, 10
- as.matrix.plsdares, 11
- as.matrix.plsres, 11
- as.matrix.regcoeffs, 12
- as.matrix.regres, 12
- as.matrix.simcamres, 13
- as.matrix.simcares, 13
- capitalize, 14
- carbs, 14
- categorize, 15
- categorize.pca, 15, 200
- categorize.pls, 16, 200
- chisq.crit, 17
- chisq.prob, 17
- classify.plsda, 18
- classify.simca, 18
- classmodel.processRefValues, 19
- classres, 19, 210, 258–260
- classres.getPerformance, 20
- confint.regcoeffs, 21, 240
- constraint, 22
- constraintAngle, 22
- constraintClosure, 23
- constraintNonNegativity, 23
- constraintNorm, 24
- constraints.list, 24
- constraintUnimod, 24
- crossval, 25
- crossval.getParams, 25
- crossval.regmodel, 26
- crossval.simca, 26
- crossval.str, 27
- dd.crit, 27
- ddmoments.param, 28
- ddrobust.param, 28
- ellipse, 29
- employ.constraint, 29
- employ.prep, 30
- eye, 30
- fprintf, 31
- getCalibrationData, 31
- getCalibrationData.pca, 32
- getCalibrationData.simcam, 32
- getConfidenceEllipse, 33
- getConfusionMatrix, 33
- getConfusionMatrix.classres, 34
- getConvexHull, 34
- getDataLabels, 35
- getImplementedConstraints, 35
- getImplementedPrepMethods, 35
- getLabelsAsIndices, 36
- getLabelsAsValues, 36
- getMainTitle, 36
- getPlotColors, 37
- getProbabilities, 37
- getProbabilities.pca, 38
- getProbabilities.simca, 38
- getPureVariables, 39
- getR, 39
- getRegcoeffs, 40
- getRegcoeffs.regmodel, 40, 201
- getRes, 41
- getSelectedComponents, 42
- getSelectivityRatio, 42
- getSelectivityRatio.pls, 43
- getVariance.mcr, 43
- getVIPScores, 44
- getVIPScores.pls, 44
- hotelling.crit, 45

- hotelling.prob, 45
- imshow, 46
- ipls, 46, 96, 165, 201
- ipls.backward, 49
- ipls.forward, 49
- jm.crit, 50
- jm.prob, 50
- ldecomp, 51, 105, 106, 260
- ldecomp.getDistances, 52
- ldecomp.getLimitsCoordinates, 52
- ldecomp.getLimParams, 53
- ldecomp.getQLimits, 53
- ldecomp.getT2Limits, 54
- ldecomp.getVariances, 54
- ldecomp.plotResiduals, 55
- mcr, 56
- mcrals, 22, 57, 97
- mcrals.cal, 61
- mcrals.fcnnls, 63
- mcrals.nnls, 63
- mcrals.ols, 64
- mcrpure, 58, 59, 65, 66, 97
- mda.cbind, 68
- mda.data2im, 68
- mda.df2mat, 69, 79
- mda.exclcols, 69
- mda.exclrows, 70
- mda.getattr, 70
- mda.getexclind, 71
- mda.im2data, 71
- mda.inclcols, 72
- mda.inclrows, 72
- mda.purge, 73
- mda.purgeCols, 73
- mda.purgeRows, 73
- mda.rbind, 74
- mda.setattr, 74
- mda.setimbg, 75
- mda.show, 75
- mda.subset, 76, 79
- mda.t, 76
- mdaplot, 77, 90, 97, 127, 143, 146, 149, 150, 156, 163, 173, 191, 213
- mdaplot.areColors, 80
- mdaplot.formatValues, 80
- mdaplot.getColors, 81
- mdaplot.getXAxisLim, 82
- mdaplot.getXTickLabels, 82
- mdaplot.getXTicks, 83
- mdaplot.getYAxisLim, 83
- mdaplot.getYTickLabels, 84
- mdaplot.getYTicks, 84
- mdaplot.plotAxes, 85
- mdaplot.prepareColors, 86
- mdaplot.showColorbar, 86
- mdaplot.showLines, 87
- mdaplotg, 80, 87, 96, 97, 142, 145, 146, 148, 149, 164
- mdaplotg.getLegend, 90
- mdaplotg.getXLim, 91
- mdaplotg.getYLim, 91
- mdaplotg.prepareData, 92
- mdaplotg.processParam, 93
- mdaplotg.showLegend, 93
- mdaplotyy, 94
- mdatools, 96
- pca, 51, 96, 97, 105, 106, 112, 128, 137, 157, 164, 175, 199, 252–254
- pca.cal, 101
- pca.getB, 102
- pca.mvreplace, 100, 102
- pca.nipals, 104
- pca.run, 104
- pca.svd, 105
- pcares, 51, 96, 99, 100, 105, 259, 260
- pcv, 97, 107
- pellets, 108
- people, 109
- pinv, 109
- plot.classres, 110
- plot.ipls, 110
- plot.mcr, 111
- plot.pca, 111
- plot.pcares, 112
- plot.pls, 112, 200
- plot.plsda, 113
- plot.plsdares, 113
- plot.plsres, 114, 210, 213
- plot.randtest, 115, 239
- plot.regcoeffs, 115, 240
- plot.regres, 116
- plot.simca, 117
- plot.simcam, 117

- plot.simcamres, 118
- plotBars, 118, 168
- plotBiplot, 119
- plotBiplot.pca, 100, 119
- plotConfidenceEllipse, 120, 134
- plotContributions, 121
- plotContributions.mcr, 59, 66, 121
- plotConvexHull, 122, 134
- plotCooman, 123
- plotCooman.simcam, 123, 125, 256
- plotCooman.simcamres, 124, 258
- plotCorr, 125
- plotCorr.randtest, 126, 239
- plotCumVariance, 126
- plotCumVariance.ldecomp, 106, 127, 260
- plotCumVariance.mcr, 59, 66, 127
- plotCumVariance.pca, 100, 128, 254
- plotDensity, 128, 168
- plotDiscriminationPower, 129
- plotDiscriminationPower.simcam, 129, 256
- plotDistDoF, 100, 130
- plotErrorbars, 131, 168
- plotExtreme, 131
- plotExtreme.pca, 100, 132
- plotHist, 133
- plotHist.randtest, 133, 239
- plotHotellingEllipse, 134
- plotLines, 135, 168
- plotLoadings, 136
- plotLoadings.pca, 100, 136, 254
- plotMisclassified, 137
- plotMisclassified.classmodel, 138, 208, 254, 256
- plotMisclassified.classres, 20, 138
- plotModelDistance, 139
- plotModelDistance.simcam, 139, 256
- plotModellingPower, 141
- plotPerformance, 141
- plotPerformance.classmodel, 138, 142, 167, 170
- plotPerformance.classres, 20, 138, 143, 168, 170, 210, 260
- plotPointsShape, 144
- plotPredictions, 144
- plotPredictions.classmodel, 145, 208, 253, 256
- plotPredictions.classres, 20, 110, 145, 210, 258, 260
- plotPredictions.regmodel, 146, 201
- plotPredictions.regres, 116, 147, 210, 213
- plotPredictions.simcam, 148
- plotPredictions.simcamres, 149
- plotProbabilities, 149
- plotProbabilities.classres, 150
- plotPurity, 150
- plotPurity.mcrpure, 66, 151
- plotPuritySpectra, 151
- plotPuritySpectra.mcrpure, 66, 152
- plotQDoF, 100, 153
- plotRegcoeffs, 153
- plotRegcoeffs.regmodel, 154
- plotRegressionLine, 154
- plotResiduals, 155
- plotResiduals.ldecomp, 56, 106, 155, 157, 260
- plotResiduals.pca, 100, 156, 183, 254
- plotResiduals.regres, 157, 195
- plotRMSE, 158
- plotRMSE.ipls, 159, 165
- plotRMSE.regmodel, 160, 201
- plotRMSE.regres, 160, 210, 213
- plotScatter, 161, 168
- plotScores, 162
- plotScores.ldecomp, 106, 134, 163, 164, 260
- plotScores.pca, 100, 134, 163, 254
- plotSelection, 164
- plotSelection.ipls, 110, 159, 165
- plotSelectivityRatio, 166
- plotSelectivityRatio.pls, 166, 200
- plotSensitivity, 167
- plotSensitivity.classmodel, 167, 208, 254, 256
- plotSensitivity.classres, 20, 168, 210, 260
- plotseries, 168
- plotSpecificity, 169
- plotSpecificity.classmodel, 170, 208, 254, 256
- plotSpecificity.classres, 20, 170, 210, 260
- plotSpectra, 171
- plotSpectra.mcr, 59, 66, 171
- plotT2DoF, 100, 172

- plotVariance, 172
- plotVariance.ldecomp, 106, 173, 260
- plotVariance.mcr, 59, 66, 174
- plotVariance.pca, 100, 174, 254
- plotVariance.pls, 175
- plotVariance.plsres, 176
- plotVIPScores, 177
- plotVIPScores.pls, 177, 200, 272
- plotWeights, 178
- plotWeights.pls, 178, 200
- plotXCumVariance, 179
- plotXCumVariance.pls, 179, 200
- plotXCumVariance.plsres, 180, 210, 213
- plotXLoadings, 180
- plotXLoadings.pls, 181, 200
- plotXResiduals, 181
- plotXResiduals.pls, 182, 200
- plotXResiduals.plsres, 183, 210, 213
- plotXScores, 184
- plotXScores.pls, 134, 184, 200
- plotXScores.plsres, 134, 185, 210, 213
- plotXVariance, 186
- plotXVariance.pls, 186, 200
- plotXVariance.plsres, 187, 210, 213
- plotXYLoadings, 187
- plotXYLoadings.pls, 188, 200
- plotXYResiduals, 188
- plotXYResiduals.pls, 189, 200
- plotXYResiduals.plsres, 190
- plotXYScores, 191
- plotXYScores.pls, 191, 200
- plotXYScores.plsres, 192, 210, 213
- plotYCumVariance, 193
- plotYCumVariance.pls, 193, 200
- plotYCumVariance.plsres, 194, 210, 213
- plotYResiduals, 194
- plotYResiduals.plsres, 195, 210, 213
- plotYResiduals.regmodel, 195, 201
- plotYVariance, 196
- plotYVariance.pls, 196, 200
- plotYVariance.plsres, 197, 210, 213
- pls, 96, 113, 114, 176, 179, 181, 185, 186, 188, 192, 193, 197, 198, 208, 214, 217, 246
- pls.cal, 203
- pls.getLimitsCoordinates, 204
- pls.getZLimits, 204
- pls.run, 205
- pls.simpls, 200, 205
- plsda, 19, 96, 113, 138, 145, 148, 167, 170, 206, 211, 217
- plsdares, 96, 138, 143, 146, 149, 168, 171, 207, 209
- plsres, 96, 114, 176, 180, 184, 185, 187, 192, 194, 197, 201, 210, 212, 217
- predict.mcrals, 59, 215
- predict.mcrpure, 66, 215
- predict.pca, 100, 105, 216
- predict.pls, 200, 212, 216
- predict.plsda, 208, 209, 211, 217
- predict.simca, 218, 253, 259
- predict.simcam, 218, 255, 258
- prep, 219
- prep.alsbasecorr, 97, 220
- prep.autoscale, 97, 221
- prep.generic, 221
- prep.list, 222
- prep.msc, 97, 222
- prep.norm, 97, 223
- prep.ref2km, 223
- prep.savgol, 97, 224
- prep.snv, 97, 223, 225
- prep.transform, 225
- prep.varsels, 226
- preparePlotData, 227
- prepCalData, 227
- print.classres, 228
- print.ipls, 228
- print.ldecomp, 229
- print.mcrals, 229
- print.mcrpure, 230
- print.pca, 230
- print.pcares, 231
- print.pls, 231
- print.plsda, 232
- print.plsdares, 232
- print.plsres, 233
- print.randtest, 233
- print.regcoeffs, 234
- print.regmodel, 234
- print.regres, 235
- print.simca, 235
- print.simcam, 236
- print.simcamres, 236
- print.simcares, 237
- randtest, 96, 115, 126, 133, 201, 237

regcoeffs, [199](#), [207](#), [239](#)  
regcoeffs.getStats, [240](#)  
regres, [241](#)  
regres.bias, [241](#)  
regres.err, [242](#)  
regres.r2, [242](#)  
regres.rmse, [243](#)  
regres.slope, [243](#)  
regress.addattrs, [243](#)  
repmat, [244](#)  
rotationMatrixToX1, [244](#)

selectCompNum, [245](#)  
selectCompNum.pca, [100](#), [245](#), [254](#)  
selectCompNum.pls, [199](#), [200](#), [246](#)  
selratio, [43](#), [200](#), [247](#)  
setDistanceLimits, [247](#)  
setDistanceLimits.pca, [100](#), [248](#)  
setDistanceLimits.pls, [200](#), [249](#)  
showDistanceLimits, [250](#)  
showLabels, [250](#)  
showPredictions, [251](#)  
showPredictions.classres, [20](#), [210](#), [251](#),  
[258](#), [260](#)  
simca, [19](#), [96](#), [99](#), [138](#), [145](#), [148](#), [167](#), [170](#),  
[218](#), [252](#), [259](#), [260](#)  
simcam, [32](#), [96](#), [117](#), [138](#), [145](#), [148](#), [167](#), [170](#),  
[219](#), [255](#), [258](#)  
simcam.getPerformanceStats, [257](#)  
simcamres, [96](#), [118](#), [138](#), [143](#), [146](#), [149](#), [168](#),  
[171](#), [255](#), [257](#)  
simcares, [19](#), [96](#), [253](#), [259](#)  
simdata, [261](#)  
splitExcludedData, [261](#)  
splitPlotData, [262](#)  
summary.classres, [262](#)  
summary.ipls, [159](#), [165](#), [263](#)  
summary.ldecomp, [263](#)  
summary.mcrals, [264](#)  
summary.mcrpure, [264](#)  
summary.pca, [265](#)  
summary.pcares, [265](#)  
summary.pls, [200](#), [266](#)  
summary.plsda, [266](#)  
summary.plsdares, [267](#)  
summary.plsres, [210](#), [213](#), [267](#)  
summary.randtest, [239](#), [268](#)  
summary.regcoeffs, [240](#), [268](#)  
summary.regmodel, [269](#)  
summary.regres, [269](#)  
summary.simca, [270](#)  
summary.simcam, [270](#)  
summary.simcamres, [271](#)  
summary.simcares, [271](#)  
unmix.mcrpure, [66](#), [272](#)  
vipcores, [44](#), [166](#), [177](#), [200](#), [272](#)