

Package: pls4all (via r-universe)

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Type Package

Title Partial Least Squares and Chemometrics Engine

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Description A portable Partial Least Squares engine for chemometrics: the slim, PLS-focused distribution carved from the 'nirs4all-methods' library. It ships every method built on the shared PLS core (NIPALS, SIMPLS, SVD, kernel, wide-kernel, orthogonal-scores, power, randomized SVD, PCR): regression variants (sparse SIMPLS, CPPLS, weighted, robust, ridge, continuum, multi-block, GLM, MIR), adaptive AOM-PLS / POP-PLS operator selection, variable-selection methods (SPA, CARS, GA, random frog, stability selection, VIP), PLS diagnostics (Hotelling T2, Q residuals, DModX), and calibration transfer (PDS, DS). The spectroscopy-specific surface (spectral preprocessing, augmentation, sample filters, signal-type detection) lives in the full 'nirs4all-methods' distribution. The same C++17 numerical core powers both; here it is vendored and compiled from source at install time, with no external system libraries required.

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URL <https://github.com/GBeurier/nirs4all-methods>

BugReports <https://github.com/GBeurier/nirs4all-methods/issues>

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Contents

aom_pls	4
aom_preprocess	5
approximate_press	6
bagging_pls	6
bagging_pls_fit	7
bipls_select	8
boosting_pls	8
boosting_pls_fit	9
bve_select	10
cars_select	10
coef.n4m_fit	11
coefficient_select	11
continuum_regression	12
continuum_regression_fit	12
cppls	13
cppls_fit	14
di_pls	14
di_pls_fit	15
ds_fit	16
ecr	16
ecr_fit	17
emcuve_select	17
fused_sparse_pls_fit	18
ga_select	18
gpr_pls_fit	19
group_sparse_pls_fit	20
interval_select	21
ipw_select	21
irf_select	22
iriv_select	23
kernel_pls_fit	23
lw_pls_fit	24
mb_pls	24
mb_pls_fit	25
mir_pls	25
mir_pls_fit	26

missing_aware_nipals	27
missing_aware_nipals_fit	27
n_pls_fit	28
n4m_abi_version	28
n4m_fit	29
n4m_method	30
n4m_predict	31
n4m_version	31
o2pls	32
o2pls_fit	33
on_pls_fit	33
one_se_rule	34
opls	34
pds_fit	35
pls	35
pls_cox_fit	36
pls_diagnostics	37
pls_glm	37
pls_glm_fit	38
pls_lda_fit	39
pls_logistic_fit	39
pls_mdatools	40
pls_monitoring	41
pls_qda_fit	42
plsr	42
predict.n4m_fit	45
predict.n4m_method_fit	45
pso_select	46
random_frog_select	47
random_subspace_pls	48
random_subspace_pls_fit	49
randomization_select	49
recursive_pls	50
recursive_pls_fit	51
rep_select	51
ridge_pls	52
ridge_pls_fit	53
robust_pls	53
robust_pls_fit	54
rosa_fit	54
scars_select	55
selectivity_ratio_select	55
shaving_select	56
sipls_select	57
so_pls_fit	57
spa_select	58
sparse_pls	58
sparse_pls_da_fit	59

sparse_simpls_fit	60
st_select	60
stability_select	61
t2_select	61
uve_select	62
vip_select	62
vip_spa_select	63
vissa_select	63
weighted_pls	64
weighted_pls_fit	65
wvc_select	65
wvc_threshold_select	66

Index	67
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aom_pls	<i>AOM-PLS and POP-PLS adaptive operator selection</i>
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Description

aom_pls() (alias aompls()) runs global Adaptive Operator Mixture PLS selection: one preprocessing operator is selected for the whole PLS model. pop_pls() (alias poppls()) runs per-component operator selection, where each retained PLS component may use a different preprocessing operator.

Usage

```
aom_pls(X, Y, max_components = 3L, n_operators = 9L, cv = 3L)
```

```
aompls(X, Y, max_components = 3L, n_operators = 9L, cv = 3L)
```

```
pop_pls(X, Y, max_components = 3L, n_operators = 9L, cv = 3L)
```

```
poppls(X, Y, max_components = 3L, n_operators = 9L, cv = 3L)
```

Arguments

X	Numeric matrix of spectra, with rows as samples and columns as wavelengths.
Y	Numeric vector or matrix of responses, with one row per sample.
max_components	Maximum number of latent PLS components.
n_operators	Number of compact AOM bank operators to expose.
cv	Number of contiguous cross-validation folds.

Details

The operator bank mirrors the compact public nirs4a11 AOM bank: identity, Savitzky-Golay smoothers, Savitzky-Golay derivatives, polynomial detrending, and finite difference. The implementation is provided by the native n4m AOM selector ABI.

Value

A named list. Both functions return predictions, operator_kinds, component-selection diagnostics, cross-validation scores, and selected component metadata.

Examples

```
set.seed(1)
X <- matrix(rnorm(40 * 20), nrow = 40)
Y <- as.numeric(X[, 1] + rnorm(40, sd = 0.1))
fit <- aom_pls(X, Y, max_components = 2L)
dim(fit$predictions)
fit2 <- pop_pls(X, Y, max_components = 2L)
dim(fit2$predictions)
```

aom_preprocess

Adaptive Operator-Mixture preprocessing fit/transform.

Description

Fits the AOM preprocessing pipeline (operator bank + gating) over ‘X’ and returns a ‘n4m_method_fit’ object carrying the selected operators, the per-component gating weights, and the transformed spectra ready to feed into a downstream regression solver.

Usage

```
aom_preprocess(X, Y = NULL, n_operators = 1L, gating_mode = 0L)
```

Arguments

X	numeric matrix of spectra (rows = samples, cols = wavelengths).
Y	optional numeric vector of supervisory targets. When ‘NULL’, the unsupervised gating path is used (a zero target vector is substituted internally).
n_operators	number of operators in the AOM bank (default ‘1L’).
gating_mode	integer code selecting the gating strategy: ‘0L’ = hard-select per component, ‘1L’ = soft-mixture (default ‘0L’).

Value

A ‘n4m_method_fit’ object. Use ‘predict()’ for inference on new spectra and ‘coef()’ to extract the gating coefficients.

Examples

```
set.seed(1)
X <- matrix(rnorm(40 * 10), nrow = 40)
Y <- as.numeric(X[, 1] + rnorm(40, sd = 0.1))
fit <- aom_preprocess(X, Y)
class(fit)
```

approximate_press *Approximate-PRESS component selection.*

Description

For each component count k in $1..max_components$, fits SIMPLS and approximates PRESS via leverage-inflated in-sample residuals.

Usage

```
approximate_press(X, Y, max_components)
```

Arguments

X Numeric matrix of predictors (rows = samples, cols = features).
 Y Numeric matrix or vector of responses, with one row per sample.
max_components Integer; maximum number of components to scan.

Value

A list with 'press_per_component', 'rmse_per_component', 'selected_n_components' (the argmin of PRESS, 1-based as an integer length 1).

bagging_pls *Bagging PLS — formula entry point.*

Description

Bagging PLS — formula entry point.

Usage

```
bagging_pls(  
  formula,  
  data,  
  ncomp = 2L,  
  n_estimators = 50L,  
  seed = 0L,  
  na.action = stats::na.omit  
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
n_estimators	Integer >= 1. Number of bootstrap / boosting / random-subspace estimators.
seed	Integer. Random seed for reproducibility.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

bagging_pls_fit	<i>Bagging PLS (bootstrap aggregation of PLS regressors).</i>
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Description

Bagging PLS (bootstrap aggregation of PLS regressors).

Usage

```
bagging_pls_fit(X, Y, n_components, n_estimators = 50L, seed = 0L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_estimators	Integer >= 1. Number of bootstrap / boosting / random-subspace estimators.
seed	Integer. Random seed for reproducibility.

bipls_select	<i>biPLS — backward interval PLS.</i>
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Description

biPLS — backward interval PLS.

Usage

```
bipls_select(X, Y, n_components, interval_width = 10L, min_intervals = 1L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
interval_width	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
min_intervals	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

boosting_pls	<i>Boosting PLS — formula entry point.</i>
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Description

Boosting PLS — formula entry point.

Usage

```
boosting_pls(
  formula,
  data,
  ncomp = 2L,
  n_estimators = 50L,
  learning_rate = 0.1,
  na.action = stats::na.omit
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
n_estimators	Integer >= 1. Number of bootstrap / boosting / random-subspace estimators.
learning_rate	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

boosting_pls_fit	<i>Boosting PLS (stage-wise refit with learning_rate).</i>
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Description

Boosting PLS (stage-wise refit with learning_rate).

Usage

```
boosting_pls_fit(X, Y, n_components, n_estimators = 50L, learning_rate = 0.1)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_estimators	Integer >= 1. Number of bootstrap / boosting / random-subspace estimators.
learning_rate	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.

bve_select	<i>BVE-PLS.</i>
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Description

BVE-PLS.

Usage

```
bve_select(X, Y, n_components, n_steps = 10L, min_features = 5L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_steps	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
min_features	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

cars_select	<i>CARS — Competitive Adaptive Reweighted Sampling.</i>
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Description

Fit-data selector with a built-in 5-fold validation plan (default C-side fallback). For a custom plan, use the lower-level Python binding or extend `r_methods.c`.

Usage

```
cars_select(X, Y, n_components, n_iterations = 50L, min_features = 5L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_iterations	Number of CARS iterations (typical 50-100).
min_features	Lower bound on the final subset size.

Value

A list with ‘selected_indices’ (1-based) and ‘best_rmse’.

coef.n4m_fit	<i>Extract the regression coefficients of a [pls()]-fitted model.</i>
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Description

Returns the $(p \times q)$ coefficient matrix (p predictors by q targets) read from the fitted libn4m model via the `'n4m_model_get_array'` C ABI (tag `'N4M_MODEL_COEFFICIENTS'`). Rows are named after the predictors when their names are available from the model terms.

Usage

```
## S3 method for class 'n4m_fit'
coef(object, ...)
```

Arguments

object	A <code>'n4m_fit'</code> returned by <code>[pls()]</code> .
...	Ignored (for S3 generic compatibility).

Value

A numeric $p \times q$ matrix of regression coefficients.

coefficient_select	<i>Coefficient-magnitude ranker.</i>
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Description

Coefficient-magnitude ranker.

Usage

```
coefficient_select(model, X, top_k)
```

Arguments

model	Method-specific parameter. See the underlying <code>'*_fit()'</code> function for the exact semantics.
X	Numeric matrix used for the fit (re-passed for diagnostics).
top_k	Method-specific parameter. See the underlying <code>'*_fit()'</code> function for the exact semantics.

Value

A list with `'scores'` (lcoefl sums) and `'selected_indices'`.

continuum_regression *Continuum regression — formula entry point.*

Description

Continuum regression — formula entry point.

Usage

```
continuum_regression(
  formula,
  data,
  ncomp = 2L,
  tau = 0.5,
  na.action = stats::na.omit
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
tau	Numeric in [0, 1]. Continuum regression mixing parameter.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

continuum_regression_fit
Continuum regression (tau in [0, 1]).

Description

Continuum regression (tau in [0, 1]).

Usage

```
continuum_regression_fit(X, Y, n_components, tau = 0.5)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
tau	Numeric in [0, 1]. Continuum regression mixing parameter.

cppls *Canonical Powered PLS — formula entry point.*

Description

Canonical Powered PLS — formula entry point.

Usage

```
cppls(formula, data, ncomp = 2L, gamma = 0.5, na.action = stats::na.omit)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
gamma	Numeric. CPPLS / kernel band parameter.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

 cppls_fit

Canonical Powered PLS fit (Indahl 2005).

Description

Canonical Powered PLS fit (Indahl 2005).

Usage

```
cppls_fit(X, Y, n_components, gamma = 0.5)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
gamma	Power exponent in [0, 1]. 0 recovers SIMPLS, 1 is fully power-rescaled.

Value

A list with ‘coefficients’, ‘predictions’, ‘x_mean’, ‘y_mean’, ‘rmse’.

 di_pls

Domain-invariant PLS – formula entry point.

Description

Domain-invariant PLS – formula entry point.

Usage

```
di_pls(
  formula,
  data,
  ncomp = 2L,
  X_target,
  di_lambda = 1,
  na.action = stats::na.omit
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
X_target	Numeric matrix for the target domain.
di_lambda	Numeric DI-PLS penalty.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model and in-sample predictions.

di_pls_fit

Domain-Invariant PLS (Nikzad-Langerodi 2018).

Description

Domain-Invariant PLS (Nikzad-Langerodi 2018).

Usage

```
di_pls_fit(X_source, Y_source, n_components, X_target, di_lambda = 1)
```

Arguments

X_source	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
Y_source	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
n_components	Integer. Number of latent components.
X_target	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
di_lambda	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.

`ds_fit` *Direct Standardization (calibration transfer).*

Description

Direct Standardization (calibration transfer).

Usage

```
ds_fit(X_source, X_target)
```

Arguments

<code>X_source</code>	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.
<code>X_target</code>	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.

`ecr` *Elastic Component Regression — formula entry point.*

Description

Elastic Component Regression — formula entry point.

Usage

```
ecr(formula, data, ncomp = 2L, alpha = 0.5, na.action = stats::na.omit)
```

Arguments

<code>formula</code>	A two-sided formula (e.g. <code>'y ~ .'</code> or <code>'y ~ x1 + x2 + x3'</code>). Response on the left, predictors on the right.
<code>data</code>	A <code>'data.frame'</code> (or anything <code>'as.data.frame'</code> -coercible) containing the response and predictor columns referenced by <code>'formula'</code> .
<code>ncomp</code>	Integer. Number of latent components.
<code>alpha</code>	Numeric in $[0, 1]$. Elastic-net / penalty mixing parameter.
<code>na.action</code>	What to do with <code>'NA'</code> s. Default: <code>'na.omit'</code> .

Value

A `'n4m_method_fit'` object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use `'predict()'` for inference and `'coef()'` to extract regression coefficients.

ecr_fit *Elastic Component Regression (Liu 2009/2010).*

Description

Elastic Component Regression (Liu 2009/2010).

Usage

```
ecr_fit(X, Y, n_components, alpha = 0.5)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
alpha	Numeric in [0, 1]. Elastic-net / penalty mixing parameter.

emcuve_select *EMCUVE — ensemble Monte Carlo UVE.*

Description

EMCUVE — ensemble Monte Carlo UVE.

Usage

```
emcuve_select(
  X,
  Y,
  n_components,
  noise_features = NULL,
  noise_seed = 0L,
  n_ensembles = 5L,
  vote_threshold = 0.5
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
noise_features	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

noise_seed	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
n_ensembles	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
vote_threshold	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

fused_sparse_pls_fit *Fused-sparse PLS (L1 + adjacent-coef smoothing).*

Description

Fused-sparse PLS (L1 + adjacent-coef smoothing).

Usage

```
fused_sparse_pls_fit(
  X,
  Y,
  n_components,
  l1_lambda = 0.05,
  fusion_lambda = 0.05
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
l1_lambda	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
fusion_lambda	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

ga_select *GA-PLS — genetic algorithm variable selection.*

Description

GA-PLS — genetic algorithm variable selection.

Usage

```
ga_select(
  X,
  Y,
  n_components,
  n_generations = 50L,
  population_size = 50L,
  min_features = NULL,
  max_features = NULL,
  mutation_rate = 0.01,
  seed = 0L
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_generations	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
population_size	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
min_features	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
max_features	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
mutation_rate	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
seed	Integer. Random seed for reproducibility.

gpr_pls_fit

Gaussian Process Regression on PLS scores (single-target Y).

Description

Gaussian Process Regression on PLS scores (single-target Y).

Usage

```
gpr_pls_fit(
  X,
  Y,
  n_components,
```

```

length_scale = 1,
noise_level = 1e-04,
seed = 0L
)

```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
length_scale	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
noise_level	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
seed	Integer. Random seed for reproducibility.

group_sparse_pls_fit *Group-sparse PLS (group L1 across feature groups).*

Description

Group-sparse PLS (group L1 across feature groups).

Usage

```
group_sparse_pls_fit(X, Y, n_components, group_assignment, group_lambda = 0.05)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
group_assignment	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
group_lambda	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

interval_select	<i>Interval selector (iPLS).</i>
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Description

Interval selector (iPLS).

Usage

```
interval_select(X, Y, n_components, interval_width = 10L, step = 1L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
interval_width	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
step	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

ipw_select	<i>IPW-PLS.</i>
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Description

IPW-PLS.

Usage

```
ipw_select(
  X,
  Y,
  n_components,
  n_iterations = 10L,
  top_k = 10L,
  damping = 0.5,
  weight_floor = 1e-06
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_iterations	Integer ≥ 1 . Number of outer-loop iterations.
top_k	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.
damping	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.
weight_floor	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.

irf_select

IRF — Interval Random Frog.

Description

IRF — Interval Random Frog.

Usage

```
irf_select(
  X,
  Y,
  n_components,
  n_iterations = 100L,
  window_size = 10L,
  initial_intervals = 10L,
  top_k = 5L,
  seed = 0L
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_iterations	Integer ≥ 1 . Number of outer-loop iterations.
window_size	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.
initial_intervals	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.

top_k	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
seed	Integer. Random seed for reproducibility.

iriv_select	<i>IRIV — Iteratively Retains Informative Variables.</i>
-------------	--

Description

IRIV — Iteratively Retains Informative Variables.

Usage

```
iriv_select(X, Y, n_components, max_rounds = 20L, seed = 0L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
max_rounds	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
seed	Integer. Random seed for reproducibility.

kernel_pls_fit	<i>Non-linear kernel PLS (Rosipal & Trejo 2001).</i>
----------------	--

Description

Non-linear kernel PLS (Rosipal & Trejo 2001).

Usage

```
kernel_pls_fit(
  X,
  Y,
  n_components,
  kernel_type = 1L,
  gamma = 0,
  coef0 = 1,
  degree = 3L
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
kernel_type	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
gamma	Numeric. CPPLS / kernel band parameter.
coef0	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
degree	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

lw_pls_fit	<i>Locally-weighted PLS (Næs & Centner 1998).</i>
------------	---

Description

Locally-weighted PLS (Næs & Centner 1998).

Usage

```
lw_pls_fit(X, Y, n_components, n_neighbors = 30L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_neighbors	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

mb_pls	<i>Multi-block PLS — formula entry point.</i>
--------	---

Description

Multi-block PLS — formula entry point.

Usage

```
mb_pls(formula, data, ncomp = 2L, block_sizes, na.action = stats::na.omit)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
block_sizes	Integer vector summing to the number of predictors.
na.action	What to do with 'NA's. Default: 'na.omit'.

mb_pls_fit	<i>Multi-block PLS (block-weighted SIMPLS).</i>
------------	---

Description

Multi-block PLS (block-weighted SIMPLS).

Usage

```
mb_pls_fit(X, Y, n_components, block_sizes)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
block_sizes	Integer vector; must sum to ncol(X).

Value

A list with 'coefficients', 'predictions', 'x_mean', 'x_scale', 'intercept', 'block_weights', 'rmse'.

mir_pls	<i>MIR-PLS — formula entry point.</i>
---------	---------------------------------------

Description

MIR-PLS — formula entry point.

Usage

```
mir_pls(formula, data, ncomp = 2L, na.action = stats::na.omit)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

mir_pls_fit

MIR-PLS — Multivariate Inverse Regression PLS.

Description

MIR-PLS — Multivariate Inverse Regression PLS.

Usage

```
mir_pls_fit(X, Y, n_components)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.

Value

A list with 'coefficients', 'predictions', 'x_mean', 'y_mean', 'rmse'.

missing_aware_nipals *Missing-aware NIPALS — formula entry point.*

Description

Missing-aware NIPALS — formula entry point.

Usage

```
missing_aware_nipals(formula, data, ncomp = 2L, na.action = stats::na.pass)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

missing_aware_nipals_fit

Missing-aware NIPALS PLS (Nelson 1996).

Description

Missing-aware NIPALS PLS (Nelson 1996).

Usage

```
missing_aware_nipals_fit(X, Y, n_components)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.

n_pls_fit	<i>N-PLS (3-way tensor) regression. 'X_flat' is the flattened (n, mode_j*mode_k) matrix.</i>
-----------	--

Description

N-PLS (3-way tensor) regression. 'X_flat' is the flattened (n, mode_j*mode_k) matrix.

Usage

```
n_pls_fit(X_flat, Y, n_components, mode_j, mode_k)
```

Arguments

X_flat	Numeric matrix. The flattened 3-way design tensor (rows = samples, cols = mode_j * mode_k).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
mode_j	Integer. Size of the first non-sample tensor mode.
mode_k	Integer. Size of the second non-sample tensor mode.

n4m_abi_version	<i>Loaded ABI version as an integer vector (major, minor, patch).</i>
-----------------	---

Description

Loaded ABI version as an integer vector (major, minor, patch).

Usage

```
n4m_abi_version()
```

Value

An integer vector of length 3.

n4m_fit

Fit a PLS regression model via the libn4m C ABI.

Description

Accepts a numeric matrix X ($n \times p$) and a numeric vector or matrix Y ($n \times q$). Both are coerced to double precision and row-major contiguous before being passed across the C boundary.

Usage

```
n4m_fit(
  X,
  Y,
  algo,
  n_components,
  store_scores = FALSE,
  center_x = TRUE,
  scale_x = TRUE,
  center_y = TRUE,
  scale_y = TRUE
)
```

Arguments

X	Numeric matrix, $n \times p$.
Y	Numeric matrix or vector.
algo	Character. Solver name (see Details).
n_components	Integer ≥ 1 .
store_scores	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.
center_x	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.
scale_x	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.
center_y	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.
scale_y	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.

Details

`'algo'` selects the solver. Recognized values: "pls_nipals", "pls_orthogonal_scores", "pls_simpls", "pls_kernel_algorithm", "pls_wide_kernel", "pls_svd", "pls_power", "pls_randomized_svd", "pcr_svd", "opls_nipals".

Value

An external pointer wrapping the fitted model handle. Pass it to [n4m_predict()] to obtain predictions. The model is freed automatically when the external pointer is garbage-collected.

n4m_method	<i>Low-level n4m method dispatcher.</i>
------------	---

Description

Low-level n4m method dispatcher.

Usage

```
n4m_method(
  algo,
  X,
  Y,
  n_components,
  params = list(),
  center_x = TRUE,
  scale_x = FALSE,
  center_y = TRUE,
  scale_y = FALSE
)
```

Arguments

algo	Character; algorithm name (see Details).
X	Numeric matrix or NULL (for one_se_rule_compute).
Y	Numeric matrix or vector. Pass an n x 1 placeholder for classifier-style fits where labels go in 'params\$y_labels'.
n_components	Positive integer.
params	Named list of algorithm-specific parameters (sparsity_lambda, sample_weights, block_sizes, X_target, y_labels, alpha_thresholds, ...).
center_x, scale_x, center_y, scale_y	Boolean preprocessing flags (default centering, no scaling — matches the Python tier-1 defaults).

Details

Supported algorithm names:

MethodResult fits (33): "sparse_simpls" "cppls" "ecr" "di_pls" "weighted_pls" "robust_pls" "ridge_pls" "continuum_regression" "recursive_pls" "n_pls" "kernel_pls" "o2pls" "sparse_pls_da" "group_sparse_pls" "fused_sparse_pls" "so_pls" "on_pls" "rosa" "bagging_pls" "boosting_pls" "random_subspace_pls"

"gpr_pls" "pls_glm" "pls_qda" "pls_cox" "pds" "ds" "mir_pls" "missing_aware_nipals" "mb_pls"
 "lw_pls" "pls_lda" "pls_logistic"

Selectors (24): "spa_select" "cars_select" "interval_select" "stability_select" "uve_select" "random_frog_select" "scars_select" "ga_select" "pso_select" "vissa_select" "shaving_select" "bve_select" "t2_select" "wvc_select" "wvc_threshold_select" "emcuve_select" "randomization_select" "bipls_select" "sipls_select" "rep_select" "ipw_select" "st_select" "iriv_select" "irf_select" "vip_spa_select"

Diagnostics (2 via dispatcher; the other 2 stay in r_methods.c): "approximate_press_compute"
 "one_se_rule_compute"

Value

A named list with the MethodResult arrays + scalars. Index fields ('selected_indices', 'top_k_intervals', ...) are 1-based.

n4m_predict	<i>Predict with a fitted n4m model.</i>
-------------	---

Description

Predict with a fitted n4m model.

Usage

```
n4m_predict(model, X)
```

Arguments

model	External pointer returned by [n4m_fit()].
X	Numeric matrix, n_new x p.

Value

Numeric matrix, n_new x n_targets.

n4m_version	<i>Runtime version string of the loaded libn4m.</i>
-------------	---

Description

Runtime version string of the loaded libn4m.

Usage

```
n4m_version()
```

Value

A character scalar like "0.67.0+abi.1.1.0".

o2pls	<i>O2-PLS — formula entry point (uses n_predictive for component count).</i>
-------	--

Description

O2-PLS — formula entry point (uses n_predictive for component count).

Usage

```
o2pls(
  formula,
  data,
  n_predictive = 2L,
  n_x_orthogonal = 1L,
  n_y_orthogonal = 1L,
  na.action = stats::na.omit
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
n_predictive	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
n_x_orthogonal	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
n_y_orthogonal	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

o2pls_fit	<i>O2-PLS (bi-directional OPLS).</i>
-----------	--------------------------------------

Description

O2-PLS (bi-directional OPLS).

Usage

```
o2pls_fit(X, Y, n_predictive = 2L, n_x_orthogonal = 1L, n_y_orthogonal = 1L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_predictive	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
n_x_orthogonal	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
n_y_orthogonal	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

on_pls_fit	<i>OnPLS — Orthogonal multi-block PLS (joint + unique loadings).</i>
------------	--

Description

OnPLS — Orthogonal multi-block PLS (joint + unique loadings).

Usage

```
on_pls_fit(X, Y, n_joint, n_unique_per_block, block_sizes)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_joint	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
n_unique_per_block	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
block_sizes	Integer vector. Per-block feature counts for multi-block PLS.

one_se_rule	<i>One-SE rule from a (max_components × n_folds) fold RMSE matrix.</i>
-------------	--

Description

One-SE rule from a (max_components × n_folds) fold RMSE matrix.

Usage

```
one_se_rule(fold_rmse_matrix)
```

Arguments

fold_rmse_matrix	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
------------------	---

opls	<i>Formula-based OPLS regression wrapper around the n4m C ABI.</i>
------	--

Description

Uses the same formula/S3 return contract as [pls()], with the model API configured as ‘algo = "opls_nipals"’.

Usage

```
opls(formula, data, ncomp = 2L, na.action = stats::na.omit)
```

Arguments

formula	A two-sided formula (e.g. ‘y ~ .’ or ‘y ~ x1 + x2 + x3’). Response on the left, predictors on the right.
data	A ‘data.frame’ (or anything ‘as.data.frame’-coercible) containing the response and predictor columns referenced by ‘formula’.
ncomp	Integer. Number of latent components.
na.action	What to do with ‘NA’s. Default: ‘na.omit’.

Value

An object of class ‘c("opls_fit", "n4m_fit", "pls_fit)"’.

pds_fit *Piecewise Direct Standardization (calibration transfer).*

Description

Piecewise Direct Standardization (calibration transfer).

Usage

```
pds_fit(X_source, X_target, window_half_width = 2L)
```

Arguments

X_source	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
X_target	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
window_half_width	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

pls *Formula-based PLS regression wrapper around the n4m C ABI*

Description

Tier-2 idiomatic R interface for tier-1 [n4m_fit()] / [n4m_predict()]. Provides a formula entry point and S3 generics so the returned object plays well with ‘predict()’, ‘summary()’, ‘coef()’, ‘print()’, base R ‘cv.glmnet’-style workflows, and any caret model that supports a ‘predict(object, newdata)’ adapter.

When the first argument is not a formula, ‘pls(x, y, ...)’ dispatches to the matrix-oriented [pls_mdatools()] compatibility facade.

Usage

```
pls(
  formula,
  data,
  ncomp = 2L,
  algo = "pls_nipals",
  na.action = stats::na.omit,
  ...
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
algo	Character. One of "pls_nipals", "pls_simpls", "pls_svd", "pls_power", "pls_kernel_algorithm", "pls_wide_kernel", "pls_orthogonal_scores", "pls_randomized_svd", "pcr_svd", "opls_nipals". Defaults to "pls_nipals" to match the R 'pls' package's default.
na.action	What to do with 'NA's. Default: 'na.omit'.
...	For matrix-style calls, forwarded to [pls_mdatools()].

Value

An object of class 'c("n4m_fit", "pls_fit)" with components: * 'model' — external pointer to the libn4m model handle * 'formula' — the call's formula * 'terms' — the 'terms()' object describing the model * 'xlevels' — factor levels of the predictors, for newdata coercion * 'call' — the original call (for 'print' / 'summary') * 'ncomp' — components used * 'algo' — solver used * 'n_features_in' — predictor count * 'response_name' — left-hand side of the formula (string)

Examples

```
## Not run:
set.seed(0)
df <- data.frame(
  x1 = rnorm(100), x2 = rnorm(100), x3 = rnorm(100)
)
df$y <- 2 * df$x1 - df$x2 + rnorm(100, sd = 0.1)
fit <- pls(y ~ ., data = df, ncomp = 3)
summary(fit)
preds <- predict(fit, newdata = df)

## End(Not run)
```

pls_cox_fit

PLS-Cox proportional hazards.

Description

PLS-Cox proportional hazards.

Usage

```
pls_cox_fit(X, n_components, survival_times, event_indicators)
```

Arguments

<code>X</code>	Numeric matrix of predictors (rows = samples, cols = features).
<code>n_components</code>	Integer. Number of latent components.
<code>survival_times</code>	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
<code>event_indicators</code>	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

`pls_diagnostics` *PLS diagnostics: T², Q, DModX from a fitted model.*

Description

PLS diagnostics: T², Q, DModX from a fitted model.

Usage

```
pls_diagnostics(model, X)
```

Arguments

<code>model</code>	External pointer from ‘n4m_fit()’.
<code>X</code>	Numeric matrix to score (typically the training matrix).

Value

A list with ‘t2’, ‘q’, ‘dmodx’ — each is a 1-row matrix of length `nrow(X)`.

`pls_glm` *PLS-GLM — formula entry point. Default is Gaussian; set ‘family = "poisson"’ for Poisson IRLS.*

Description

PLS-GLM — formula entry point. Default is Gaussian; set ‘family = "poisson"’ for Poisson IRLS.

Usage

```
pls_glm(
  formula,
  data,
  ncomp = 2L,
  family = "gaussian",
  na.action = stats::na.omit
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
family	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

pls_glm_fit

PLS-GLM — Gaussian (default) or Poisson IRLS.

Description

PLS-GLM — Gaussian (default) or Poisson IRLS.

Usage

```
pls_glm_fit(X, Y, n_components, poisson = FALSE)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
poisson	Logical; TRUE selects the Poisson-link path.

Value

A list with 'coefficients', 'intercept', 'predictions', 'x_mean', 'rmse'.

pls_lda_fit *PLS-LDA — Linear Discriminant Analysis on PLS scores.*

Description

PLS-LDA — Linear Discriminant Analysis on PLS scores.

Usage

```
pls_lda_fit(X, y_labels, n_components, n_classes = NULL)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
y_labels	Integer vector. Class labels.
n_components	Integer. Number of latent components.
n_classes	Integer >= 2. Number of class labels.

pls_logistic_fit *Multinomial logistic regression on PLS scores.*

Description

Multinomial logistic regression on PLS scores.

Usage

```
pls_logistic_fit(X, y_labels, n_components, n_classes = NULL)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
y_labels	Integer vector. Class labels.
n_components	Integer. Number of latent components.
n_classes	Integer >= 2. Number of class labels.

pls_mdatools

*mdatools-style matrix PLS interface***Description**

Matrix-oriented PLS facade modelled after `mdatools::pls(x, y, ...)` for NIRS/chemometrics workflows. The top-level `pls(x, y, ...)` function dispatches here when its first argument is not a formula.

Usage

```
pls_mdatools(
  x,
  y,
  ncomp = min(nrow(x) - 1L, ncol(x), 20L),
  center = TRUE,
  scale = FALSE,
  cv = NULL,
  exclcols = NULL,
  exclrows = NULL,
  x.test = NULL,
  y.test = NULL,
  method = c("simpls", "cppls", "kernelpls",
             "widekernelpls", "oscorespls", "nipals", "pcr"),
  info = "",
  ncomp.selcrit = "min",
  lim.type = "ddmoments",
  alpha = 0.05,
  gamma = 0.5,
  cv.scope = "local",
  prep = NULL,
  fit_components = TRUE,
  ...
)
```

Arguments

<code>x</code>	Numeric predictor matrix.
<code>y</code>	Numeric response vector or matrix.
<code>ncomp</code>	Maximum number of components.
<code>center</code>	Logical; center predictors and response.
<code>scale</code>	Logical; standardize predictors and response.
<code>cv</code>	NULL, an integer number of folds, or a list of assessment row indices.
<code>exclcols</code>	Optional predictor columns to exclude.
<code>exclrows</code>	Optional rows to exclude before fitting.

x.test	Optional external test predictor matrix.
y.test	Optional external test response vector or matrix.
method	PLS/PCR algorithm selector.
info	Free-form model label.
ncomp.selcrit	Stored for compatibility.
lim.type	Stored for compatibility.
alpha	Stored for compatibility.
gamma	CPPLS gamma when method = "cppls".
cv.scope	Stored for compatibility.
prep	Optional preprocessing function applied to x and x.test.
fit_components	Logical; fit component prefixes 1:ncomp.
...	Reserved for compatibility.

Value

A `n4m_mdatools_pls` object with calibration results in `calres`; optional cross-validation and test results are stored in `cvres` and `testres`.

Examples

```
## Not run:
set.seed(1)
X <- matrix(rnorm(80), 20, 4)
y <- X[, 1] - X[, 2] + rnorm(20, sd = 0.05)
fit <- pls_mdatools(X, y, ncomp = 3, cv = 5)
predict(fit, x = X)
fit$calres$rmse
fit$cvres$rmse

## End(Not run)
```

pls_monitoring *PLS process monitoring (Hotelling T^2 + Q with alarms).*

Description

Computes phase-I thresholds on ‘X_reference’, then evaluates ‘X_monitor’ against those thresholds and reports per-row alarms.

Usage

```
pls_monitoring(model, X_reference, X_monitor, alpha = 0.95)
```

Arguments

model	External pointer from 'n4m_fit()'.
X_reference	Phase-1 numeric matrix (used to set thresholds).
X_monitor	Phase-2 numeric matrix (rows are evaluated).
alpha	Confidence level (e.g. 0.95). Thresholds correspond to the (1 - alpha) quantile.

Value

A list with 't2', 'q', 't2_alarms', 'q_alarms', 'any_alarms', 't2_threshold', 'q_threshold'.

pls_qda_fit	<i>PLS-QDA (Quadratic Discriminant Analysis on PLS scores).</i>
-------------	---

Description

PLS-QDA (Quadratic Discriminant Analysis on PLS scores).

Usage

```
pls_qda_fit(X, y_labels, n_components)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
y_labels	Integer vector. Class labels.
n_components	Integer. Number of latent components.

pls	<i>PLS package compatibility facade</i>
-----	---

Description

These functions provide the familiar formula-oriented surface of the R **pls** package without importing it. Computation is backed by **n4m**; the facade is intended for NIRS/chemometrics scripts that already use `pls()`, `pcr()`, `predict()`, `RMSEP()` and `selectNcomp()` patterns.

Usage

```
pls(
  formula,
  ncomp = 2L,
  data,
  subset,
  na.action = stats::na.omit,
  method = c("kernelpls", "widekernelpls", "simpls",
            "oscorespls", "cppls", "nipals"),
  scale = FALSE,
  validation = c("none", "CV"),
  segments = 10L,
  center = TRUE,
  fit_components = TRUE,
  gamma = 0.5,
  sparsity_lambda = 0,
  ...
)

pcr(
  formula,
  ncomp = 2L,
  data,
  subset,
  na.action = stats::na.omit,
  method = c("svdpc", "pcr"),
  scale = FALSE,
  validation = c("none", "CV"),
  segments = 10L,
  center = TRUE,
  fit_components = TRUE,
  ...
)

mvr(
  formula,
  ncomp = 2L,
  data,
  subset,
  na.action = stats::na.omit,
  method = c("kernelpls", "widekernelpls", "simpls",
            "oscorespls", "cppls", "nipals", "pcr"),
  scale = FALSE,
  validation = c("none", "CV"),
  segments = 10L,
  center = TRUE,
  model = c("pls", "pcr"),
  fit_components = TRUE,
```

```

    gamma = 0.5,
    sparsity_lambda = 0,
    ...
)

MSEP(object, ...)

RMSEP(object, ...)

R2(object, ...)

selectNcomp(object, method = c("min"), estimate = c("CV", "train"), ...)

```

Arguments

formula	A two-sided model formula.
ncomp	Number of latent components.
data	Data frame containing formula variables.
subset	Optional row subset.
na.action	Missing-value action.
method	PLS/PCR algorithm selector.
scale	Logical; standardize predictors and response.
validation	"none" or "CV".
segments	Number of contiguous CV segments, or a list of assessment row indices.
center	Logical; center predictors and response.
fit_components	Logical; fit component prefixes 1:ncomp for component-wise prediction and metrics.
gamma	CPPLS gamma.
sparsity_lambda	Sparse-SIMPLS soft threshold for the dense SIMPLS-compatible path.
model	"pls" or "pcr".
object	A <code>n4m_mvr</code> object.
estimate	Metric source, "train" or "CV".
...	Reserved for compatibility.

Value

`pls()`, `pcr()` and `mvr()` return a `n4m_mvr` object. `MSEP()`, `RMSEP()` and `R2()` return a small `mvrVal`-compatible list with a `val` array. `selectNcomp()` returns an integer component count.

Examples

```
## Not run:
set.seed(1)
df <- data.frame(x1 = rnorm(40), x2 = rnorm(40))
df$y <- 2 * df$x1 - df$x2 + rnorm(40, sd = 0.05)
fit <- pls(y ~ ., data = df, ncomp = 2, method = "simpls")
predict(fit, df, ncomp = 1:2)
RMSEP(fit)
selectNcomp(fit)

## End(Not run)
```

predict.n4m_fit	<i>Predict from a [pls()]-fitted model.</i>
-----------------	---

Description

Predict from a [pls()]-fitted model.

Usage

```
## S3 method for class 'n4m_fit'
predict(object, newdata = NULL, ...)
```

Arguments

object	A 'n4m_fit' returned by [pls()].
newdata	A 'data.frame' (or matrix). If 'NULL', predictions on the training matrix are returned.
...	Ignored (for S3 generic compatibility).

Value

Numeric vector (for single-target regression) or matrix (multi-target).

predict.n4m_method_fit	<i>Predict from a MethodResult-based n4m fit.</i>
------------------------	---

Description

Predict from a MethodResult-based n4m fit.

Usage

```
## S3 method for class 'n4m_method_fit'
predict(object, newdata = NULL, ...)
```

Arguments

object	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
newdata	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
...	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

pso_select	<i>PSO-PLS (Binary Particle Swarm Optimization).</i>
------------	--

Description

PSO-PLS (Binary Particle Swarm Optimization).

Usage

```
pso_select(
  X,
  Y,
  n_components,
  n_swarm = 30L,
  n_iterations = 50L,
  w = 0.729,
  c1 = 1.494,
  c2 = 1.494,
  v_max = 4,
  seed = 0L
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_swarm	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
n_iterations	Integer ≥ 1 . Number of outer-loop iterations.
w	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

c1	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
c2	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
v_max	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
seed	Integer. Random seed for reproducibility.

random_frog_select *Random Frog (Phase 5g).*

Description

Random Frog (Phase 5g).

Usage

```
random_frog_select(
  X,
  Y,
  n_components,
  n_iterations = 100L,
  initial_size = 30L,
  min_size = NULL,
  max_size = NULL,
  top_k = 10L,
  seed = 0L
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_iterations	Integer >= 1. Number of outer-loop iterations.
initial_size	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
min_size	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
max_size	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
top_k	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
seed	Integer. Random seed for reproducibility.

random_subspace_pls *Random-subspace PLS — formula entry point.*

Description

Random-subspace PLS — formula entry point.

Usage

```
random_subspace_pls(  
  formula,  
  data,  
  ncomp = 2L,  
  n_estimators = 50L,  
  features_per_subspace = 10L,  
  seed = 0L,  
  na.action = stats::na.omit  
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
n_estimators	Integer >= 1. Number of bootstrap / boosting / random-subspace estimators.
features_per_subspace	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
seed	Integer. Random seed for reproducibility.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

`random_subspace_pls_fit`*Random-subspace PLS (Ho 1998).*

Description

Random-subspace PLS (Ho 1998).

Usage

```
random_subspace_pls_fit(  
  X,  
  Y,  
  n_components,  
  n_estimators = 50L,  
  features_per_subspace = 10L,  
  seed = 0L  
)
```

Arguments

<code>X</code>	Numeric matrix of predictors (rows = samples, cols = features).
<code>Y</code>	Numeric matrix or vector of responses, with one row per sample.
<code>n_components</code>	Integer. Number of latent components.
<code>n_estimators</code>	Integer ≥ 1 . Number of bootstrap / boosting / random-subspace estimators.
<code>features_per_subspace</code>	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
<code>seed</code>	Integer. Random seed for reproducibility.

`randomization_select` *Randomization test selector.*

Description

Randomization test selector.

Usage

```
randomization_select(  
  X,  
  Y,  
  n_components,  
  n_permutations = 100L,  
  randomization_seed = 0L,  
  alpha = 0.05  
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_permutations	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
randomization_seed	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
alpha	Numeric in [0, 1]. Elastic-net / penalty mixing parameter.

recursive_pls	<i>Recursive PLS – formula entry point.</i>
---------------	---

Description

Recursive PLS stores in-sample predictions rather than a reusable coefficient model. ‘predict()’ therefore accepts only the fitted training design.

Usage

```
recursive_pls(
  formula,
  data,
  ncomp = 2L,
  window_size = 50L,
  na.action = stats::na.omit
)
```

Arguments

formula	A two-sided formula (e.g. ‘y ~ .’ or ‘y ~ x1 + x2 + x3’). Response on the left, predictors on the right.
data	A ‘data.frame’ (or anything ‘as.data.frame’-coercible) containing the response and predictor columns referenced by ‘formula’.
ncomp	Integer. Number of latent components.
window_size	Integer moving-window size.
na.action	What to do with ‘NA’s. Default: ‘na.omit’.

recursive_pls_fit	<i>Moving-window recursive PLS.</i>
-------------------	-------------------------------------

Description

Moving-window recursive PLS.

Usage

```
recursive_pls_fit(X, Y, n_components, window_size = 50L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
window_size	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

rep_select	<i>REP-PLS.</i>
------------	-----------------

Description

REP-PLS.

Usage

```
rep_select(
  X,
  Y,
  n_components,
  n_steps = 10L,
  min_features = 5L,
  remove_count = 1L
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_steps	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

min_features	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.
remove_count	Method-specific parameter. See the underlying <code>*_fit()</code> function for the exact semantics.

ridge_pls	<i>Ridge PLS — formula entry point.</i>
-----------	---

Description

Ridge PLS — formula entry point.

Usage

```
ridge_pls(
  formula,
  data,
  ncomp = 2L,
  ridge_lambda = 1,
  na.action = stats::na.omit
)
```

Arguments

formula	A two-sided formula (e.g. <code>'y ~ .'</code> or <code>'y ~ x1 + x2 + x3'</code>). Response on the left, predictors on the right.
data	A <code>'data.frame'</code> (or anything <code>'as.data.frame'</code> -coercible) containing the response and predictor columns referenced by <code>'formula'</code> .
ncomp	Integer. Number of latent components.
ridge_lambda	Numeric ≥ 0 . Ridge regularisation strength.
na.action	What to do with <code>'NA'</code> s. Default: <code>'na.omit'</code> .

Value

A `'n4m_method_fit'` object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use `'predict()'` for inference and `'coef()'` to extract regression coefficients.

ridge_pls_fit	<i>L2-augmented PLS regression.</i>
---------------	-------------------------------------

Description

L2-augmented PLS regression.

Usage

```
ridge_pls_fit(X, Y, n_components, ridge_lambda = 1)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
ridge_lambda	Numeric ≥ 0 . Ridge regularisation strength.

robust_pls	<i>Robust PLS — formula entry point.</i>
------------	--

Description

Robust PLS — formula entry point.

Usage

```
robust_pls(
  formula,
  data,
  ncomp = 2L,
  huber_k = 1.345,
  max_irls_iter = 20L,
  na.action = stats::na.omit
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
huber_k	Numeric ≥ 0 . Huber loss tuning constant.

max_irls_iter Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

na.action What to do with ‘NA’s. Default: ‘na.omit’.

Value

A ‘n4m_method_fit’ object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use ‘predict()’ for inference and ‘coef()’ to extract regression coefficients.

robust_pls_fit *Robust PLS via Huber IRLS.*

Description

Robust PLS via Huber IRLS.

Usage

```
robust_pls_fit(X, Y, n_components, huber_k = 1.345, max_irls_iter = 20L)
```

Arguments

X Numeric matrix of predictors (rows = samples, cols = features).

Y Numeric matrix or vector of responses, with one row per sample.

n_components Integer. Number of latent components.

huber_k Numeric ≥ 0 . Huber loss tuning constant.

max_irls_iter Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

rosa_fit *ROSA — Response-Oriented Sequential Alternation.*

Description

ROSA — Response-Oriented Sequential Alternation.

Usage

```
rosa_fit(X, Y, n_components, block_sizes)
```

Arguments

X Numeric matrix of predictors (rows = samples, cols = features).

Y Numeric matrix or vector of responses, with one row per sample.

n_components Integer. Number of latent components.

block_sizes Integer vector. Per-block feature counts for multi-block PLS.

scars_select *SCARS — Stability + CARS.*

Description

SCARS — Stability + CARS.

Usage

```
scars_select(
  X,
  Y,
  n_components,
  n_iterations = 50L,
  min_features = 5L,
  sample_fraction = 0.8,
  seed = 0L
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_iterations	Integer >= 1. Number of outer-loop iterations.
min_features	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
sample_fraction	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
seed	Integer. Random seed for reproducibility.

selectivity_ratio_select
 Selectivity-ratio ranker.

Description

Selectivity-ratio ranker.

Usage

```
selectivity_ratio_select(model, X, top_k)
```

Arguments

model	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
X	Numeric matrix used for the fit (re-passed for diagnostics).
top_k	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

Value

A list with ‘scores’ and ‘selected_indices’.

shaving_select	<i>Shaving selector.</i>
----------------	--------------------------

Description

Shaving selector.

Usage

```
shaving_select(
  X,
  Y,
  n_components,
  n_steps = 10L,
  min_features = 5L,
  shave_fraction = 0.1
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_steps	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
min_features	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
shave_fraction	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

sipls_select	<i>siPLS — synergistic interval PLS.</i>
--------------	--

Description

siPLS — synergistic interval PLS.

Usage

```
sipls_select(X, Y, n_components, interval_width = 10L, combination_size = 2L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
interval_width	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
combination_size	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

so_pls_fit	<i>Sequential & Orthogonalised multi-block PLS (Næs et al. 2011). ‘block_sizes’ integer vector summing to ncol(X); ‘n_components_per_block’ integer vector of same length.</i>
------------	--

Description

Sequential & Orthogonalised multi-block PLS (Næs et al. 2011). ‘block_sizes’ integer vector summing to ncol(X); ‘n_components_per_block’ integer vector of same length.

Usage

```
so_pls_fit(X, Y, block_sizes, n_components_per_block)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
block_sizes	Integer vector. Per-block feature counts for multi-block PLS.
n_components_per_block	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

spa_select	<i>SPA — Successive Projections Algorithm.</i>
------------	--

Description

Fit-data selector: pass (X, Y) and the desired number of features.

Usage

```
spa_select(X, Y, n_components, top_k)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
top_k	Number of features to select.

Value

A list with 'selected_indices' (1-based) and 'best_rmse'.

sparse_pls	<i>Sparse SIMPLS — formula entry point.</i>
------------	---

Description

Sparse SIMPLS — formula entry point.

Usage

```
sparse_pls(
  formula,
  data,
  ncomp = 2L,
  sparsity_lambda = 0.05,
  na.action = stats::na.omit
)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
sparsity_lambda	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.
na.action	What to do with 'NA's. Default: 'na.omit'.

Value

A 'n4m_method_fit' object carrying the fitted model, in-sample predictions, training RMSE, and method-specific metadata. Use 'predict()' for inference and 'coef()' to extract regression coefficients.

sparse_pls_da_fit *Sparse PLS-DA classifier ('y_labels' is an integer vector of class IDs).*

Description

Sparse PLS-DA classifier ('y_labels' is an integer vector of class IDs).

Usage

```
sparse_pls_da_fit(X, y_labels, n_components, sparsity_lambda = 0.05)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
y_labels	Integer vector. Class labels.
n_components	Integer. Number of latent components.
sparsity_lambda	Method-specific parameter. See the underlying '*_fit()' function for the exact semantics.

`sparse_simpls_fit` *Sparse SIMPLS fit.*

Description

Sparse SIMPLS fit.

Usage

```
sparse_simpls_fit(X, Y, n_components, sparsity_lambda = 0.05)
```

Arguments

<code>X</code>	Numeric matrix.
<code>Y</code>	Numeric matrix or vector.
<code>n_components</code>	Integer ≥ 1 .
<code>sparsity_lambda</code>	Soft-threshold magnitude per component (≥ 0).

Value

A list with ‘coefficients’, ‘predictions’, ‘x_mean’, ‘y_mean’, ‘weights_w’, ‘rmse’.

`st_select` *ST-PLS — score-threshold selector.*

Description

ST-PLS — score-threshold selector.

Usage

```
st_select(X, Y, n_components, thresholds, min_selected = NULL)
```

Arguments

<code>X</code>	Numeric matrix of predictors (rows = samples, cols = features).
<code>Y</code>	Numeric matrix or vector of responses, with one row per sample.
<code>n_components</code>	Integer. Number of latent components.
<code>thresholds</code>	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
<code>min_selected</code>	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

stability_select	<i>Stability selector (coefficient stability, MCUVE-style).</i>
------------------	---

Description

Stability selector (coefficient stability, MCUVE-style).

Usage

```
stability_select(X, Y, n_components, top_k = 10L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
top_k	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

t2_select	<i>T2-PLS - sweep over alpha thresholds.</i>
-----------	--

Description

T2-PLS - sweep over alpha thresholds.

Usage

```
t2_select(X, Y, n_components, alpha_thresholds, min_selected = NULL)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
alpha_thresholds	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
min_selected	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

uve_select *UVE — Uninformative Variable Elimination.*

Description

UVE — Uninformative Variable Elimination.

Usage

```
uve_select(X, Y, n_components, noise_features = NULL, noise_seed = 0L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
noise_features	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
noise_seed	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

vip_select *VIP (Variable Importance in Projection) ranker.*

Description

Operates on an already-fitted n4m model handle (from ‘n4m_fit’). Returns the top-‘top_k’ features by VIP score.

Usage

```
vip_select(model, X, top_k)
```

Arguments

model	External pointer from ‘n4m_fit()’.
X	Numeric matrix used for the fit (re-passed for diagnostics).
top_k	Integer; number of features to return.

Value

A list with ‘scores’ (length p VIP scores) and ‘selected_indices’ (1-based, length top_k).

vip_spa_select	<i>VIP-SPA hybrid selector.</i>
----------------	---------------------------------

Description

VIP-SPA hybrid selector.

Usage

```
vip_spa_select(X, Y, n_components, vip_threshold = 0.3, top_k = 10L)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
vip_threshold	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
top_k	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

vissa_select	<i>VISSA — Variable Iterative Space Shrinkage Approach.</i>
--------------	---

Description

VISSA — Variable Iterative Space Shrinkage Approach.

Usage

```
vissa_select(
  X,
  Y,
  n_components,
  n_iterations = 20L,
  n_submodels = 100L,
  ratio_kept = 0.1,
  threshold = 0.5,
  floor_probability = 0.01,
  seed = 0L
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
n_iterations	Integer ≥ 1 . Number of outer-loop iterations.
n_submodels	Integer ≥ 1 . Number of inner sub-models per VISSA-style iteration.
ratio_kept	Numeric in (0, 1]. Fraction of features kept per iteration.
threshold	Numeric. Convergence / pruning threshold.
floor_probability	Numeric in [0, 0.5). Lower bound on per-feature retention probability.
seed	Integer. Random seed for reproducibility.

 weighted_pls

Sample-weighted PLS — formula entry point.

Description

Sample-weighted PLS — formula entry point.

Usage

```
weighted_pls(formula, data, ncomp = 2L, weights, na.action = stats::na.omit)
```

Arguments

formula	A two-sided formula (e.g. 'y ~ .' or 'y ~ x1 + x2 + x3'). Response on the left, predictors on the right.
data	A 'data.frame' (or anything 'as.data.frame'-coercible) containing the response and predictor columns referenced by 'formula'.
ncomp	Integer. Number of latent components.
weights	Numeric vector of length nrow(data) with sample weights.
na.action	What to do with 'NA's. Default: 'na.omit'.

weighted_pls_fit	<i>Sample-weighted PLS (sqrt(w)-prescaled SIMPLS).</i>
------------------	--

Description

Sample-weighted PLS (sqrt(w)-prescaled SIMPLS).

Usage

```
weighted_pls_fit(X, Y, n_components, sample_weights)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
sample_weights	Numeric vector of length nrow(X), strictly positive finite weights.

Value

A list with ‘coefficients’, ‘predictions’, ‘x_mean’, ‘y_mean’, ‘rmse’.

wvc_select	<i>WVC-PLS — weighted vector correlation top-k selector.</i>
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Description

WVC-PLS — weighted vector correlation top-k selector.

Usage

```
wvc_select(X, Y, n_components, top_k = 10L, normalize = TRUE)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
top_k	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
normalize	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

wvc_threshold_select *WVC-threshold selector.*

Description

WVC-threshold selector.

Usage

```
wvc_threshold_select(  
  X,  
  Y,  
  n_components,  
  normalize = TRUE,  
  threshold = 0,  
  threshold_factor = 1,  
  min_selected = 1L  
)
```

Arguments

X	Numeric matrix of predictors (rows = samples, cols = features).
Y	Numeric matrix or vector of responses, with one row per sample.
n_components	Integer. Number of latent components.
normalize	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
threshold	Numeric. Convergence / pruning threshold.
threshold_factor	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.
min_selected	Method-specific parameter. See the underlying ‘*_fit()’ function for the exact semantics.

Index

* models

- aom_pls, 4
- aom_pls, 4
- aom_preprocess, 5
- aompls (aom_pls), 4
- approximate_press, 6
- bagging_pls, 6
- bagging_pls_fit, 7
- bipls_select, 8
- boosting_pls, 8
- boosting_pls_fit, 9
- bve_select, 10
- cars_select, 10
- coef.n4m_fit, 11
- coef.n4m_mvr (plsr), 42
- coefficient_select, 11
- continuum_regression, 12
- continuum_regression_fit, 12
- cppls, 13
- cppls_fit, 14
- di_pls, 14
- di_pls_fit, 15
- ds_fit, 16
- ecr, 16
- ecr_fit, 17
- emcuve_select, 17
- fused_sparse_pls_fit, 18
- ga_select, 18
- gpr_pls_fit, 19
- group_sparse_pls_fit, 20
- interval_select, 21
- ipw_select, 21
- irf_select, 22
- iriv_select, 23
- kernel_pls_fit, 23
- lw_pls_fit, 24
- mb_pls, 24
- mb_pls_fit, 25
- mir_pls, 25
- mir_pls_fit, 26
- missing_aware_nipals, 27
- missing_aware_nipals_fit, 27
- MSEP (plsr), 42
- mvr (plsr), 42
- n4m_abi_version, 28
- n4m_fit, 29
- n4m_method, 30
- n4m_predict, 31
- n4m_version, 31
- n_pls_fit, 28
- o2pls, 32
- o2pls_fit, 33
- on_pls_fit, 33
- one_se_rule, 34
- opls, 34
- pcr (plsr), 42
- pds_fit, 35
- pls, 35
- pls4all_method (n4m_method), 30
- pls_cox_fit, 36
- pls_diagnostics, 37
- pls_glm, 37
- pls_glm_fit, 38
- pls_lda_fit, 39
- pls_logistic_fit, 39
- pls_mdtools, 40
- pls_monitoring, 41
- pls_qda_fit, 42

pls, 42
pop_pls (aom_pls), 4
poppls (aom_pls), 4
predict.n4m_fit, 45
predict.n4m_mdatools_pls
 (pls_mdatools), 40
predict.n4m_method_fit, 45
predict.n4m_mvr (pls), 42
pso_select, 46

R2 (pls), 42
random_frog_select, 47
random_subspace_pls, 48
random_subspace_pls_fit, 49
randomization_select, 49
recursive_pls, 50
recursive_pls_fit, 51
rep_select, 51
ridge_pls, 52
ridge_pls_fit, 53
RMSEP (pls), 42
robust_pls, 53
robust_pls_fit, 54
rosa_fit, 54

scars_select, 55
selectivity_ratio_select, 55
selectNcomp (pls), 42
shaving_select, 56
sipls_select, 57
so_pls_fit, 57
spa_select, 58
sparse_pls, 58
sparse_pls_da_fit, 59
sparse_simpls_fit, 60
st_select, 60
stability_select, 61

t2_select, 61

uve_select, 62

vip_select, 62
vip_spa_select, 63
vissa_select, 63

weighted_pls, 64
weighted_pls_fit, 65
wvc_select, 65
wvc_threshold_select, 66