Package ‘NNS’

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Type Package

Title Nonlinear Nonparametric Statistics

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Description Nonlinear nonparametric statistics using partial moments. Partial moments are the elements of variance and asymptotically approximate the area of f(x). These robust statistics provide the basis for nonlinear analysis while retaining linear equivalences. NNS offers: Numerical integration, Numerical differentiation, Clustering, Correlation, Dependence, Causal analysis, ANOVA, Regression, Classification, Seasonality, Autoregressive modeling, Normalization and Stochastic dominance. All routines based on: Viole, F. and Nawrocki, D. (2013), Nonlinear Nonparametric Statistics: Using Partial Moments (ISBN: 1490523995).

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BugReports https://github.com/OVVO-Financial/NNS/issues

LazyData TRUE

RoxygenNote 7.1.1

Depends R (>= 3.3.0), doParallel

Imports data.table, dtw, meboot, Rfast, rgl, stringr, tdigest

Suggests knitr, rmarkdown

VignetteBuilder knitr

NeedsCompilation no

Repository CRAN

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R topics documented:

`Co.LPM` *Co-Lower Partial Moment (Lower Left Quadrant 4)*

Description

This function generates a co-lower partial moment for between two equal length variables for any degree or target.

Usage

```
Co.LPM(degree.x, degree.y, x, y, target.x = mean(x), target.y = mean(y))
```

Arguments

degree.x	integer; Degree for variable X. (degree.x = 0) is frequency, (degree.x = 1) is area.
degree.y	integer; Degree for variable Y. (degree.y = 0) is frequency, (degree.y = 1) is area.
x	a numeric vector.
y	a numeric vector of equal length to x.
target.x	numeric; Typically the mean of Variable X for classical statistics equivalences, but does not have to be. (Vectorized)
target.y	numeric; Typically the mean of Variable Y for classical statistics equivalences, but does not have to be. (Vectorized)

Value

Co-LPM of two variables

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
Co.LPM(0, 0, x, y, mean(x), mean(y))
```

Co.UPM *Co-Upper Partial Moment (Upper Right Quadrant 1)*

Description

This function generates a co-upper partial moment between two equal length variables for any degree or target.

Usage

```
Co.UPM(degree.x, degree.y, x, y, target.x = mean(x), target.y = mean(y))
```

Arguments

degree.x	integer; Degree for variable X. (degree.x = 0) is frequency, (degree.x = 1) is area.
degree.y	integer; Degree for variable Y. (degree.y = 0) is frequency, (degree.y = 1) is area.
x	a numeric vector.
y	a numeric vector of equal length to x.
target.x	numeric; Typically the mean of Variable X for classical statistics equivalences, but does not have to be. (Vectorized)
target.y	numeric; Typically the mean of Variable Y for classical statistics equivalences, but does not have to be. (Vectorized)

Value

Co-UPM of two variables

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
Co.UPM(0, 0, x, y, mean(x), mean(y))
```

D.LPM *Divergent-Lower Partial Moment (Lower Right Quadrant 3)*

Description

This function generates a divergent lower partial moment between two equal length variables for any degree or target.

Usage

```
D.LPM(degree.x, degree.y, x, y, target.x = mean(x), target.y = mean(y))
```

Arguments

`degree.x`	integer; Degree for variable X. (`degree.x = 0`) is frequency, (`degree.x = 1`) is area.
`degree.y`	integer; Degree for variable Y. (`degree.y = 0`) is frequency, (`degree.y = 1`) is area.
`x`	a numeric vector.
`y`	a numeric vector of equal length to `x`.
`target.x`	numeric; Typically the mean of Variable X for classical statistics equivalences, but does not have to be. (Vectorized)
`target.y`	numeric; Typically the mean of Variable Y for classical statistics equivalences, but does not have to be. (Vectorized)

Value

Divergent LPM of two variables

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
D.LPM(0, 0, x, y, mean(x), mean(y))
```

D.UPM — *Divergent-Upper Partial Moment (Upper Left Quadrant 2)*

Description

This function generates a divergent upper partial moment between two equal length variables for any degree or target.

Usage

```
D.UPM(degree.x, degree.y, x, y, target.x = mean(x), target.y = mean(y))
```

Arguments

`degree.x`	integer; Degree for variable X. `(degree.x = 0)` is frequency, `(degree.x = 1)` is area.
`degree.y`	integer; Degree for variable Y. `(degree.y = 0)` is frequency, `(degree.y = 1)` is area.
`x`	a numeric vector.
`y`	a numeric vector of equal length to `x`.
`target.x`	numeric; Typically the mean of Variable X for classical statistics equivalences, but does not have to be. (Vectorized)
`target.y`	numeric; Typically the mean of Variable Y for classical statistics equivalences, but does not have to be. (Vectorized)

Value

Divergent UPM of two variables

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
D.UPM(0, 0, x, y, mean(x), mean(y))
```

---

dy.dx *Partial Derivative dy/dx*

Description

Returns the numerical partial derivate of `y` wrt `x` for a point of interest.

Usage

```
dy.dx(x, y, eval.point = median(x), deriv.method = "FD")
```

Arguments

x	a numeric vector.
y	a numeric vector.
eval.point	numeric or ("overall"); x point to be evaluated. Defaults to (eval.point = median(x)). Set to (eval.point = "overall") to find an overall partial derivative estimate (1st derivative only).
deriv.method	method of derivative estimation, options: ("NNS", "FD"); Determines the partial derivative from the coefficient of the NNS.reg output when (deriv.method = "NNS") or generates a partial derivative using the finite difference method (deriv.method = "FD") (Defualt).

Value

Returns a list of both 1st and 2nd derivative:

- dy.dx(...)$First the 1st derivative.
- dy.dx(...)$Second the 2nd derivative.

Note

If a vector of derivatives is required, ensure (deriv.method = "FD").

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Vinod, H. and Viole, F. (2017) "Nonparametric Regression Using Clusters" https://link.springer.com/article/10.1007/s10614-017-9713-5

Examples

```
## Not run:
x <- seq(0, 2 * pi, pi / 100) ; y <-sin(x)
dy.dx(x, y, eval.point = 1.75)

# Vector of derivatives
dy.dx(x, y, eval.point = c(1.75, 2.5), deriv.method = "FD")
## End(Not run)
```

dy.d_ *Partial Derivative dy/d_[wrt]*

Description

Returns the numerical partial derivative of `y` with respect to [wrt] any regressor for a point of interest. Finite difference method is used with NNS.reg estimates as `f(x + h)` and `f(x -h)` values.

Usage

```
dy.d_(
  x,
  y,
  wrt,
  eval.points = "obs",
  mixed = FALSE,
  ncores = NULL,
  messages = TRUE
)
```

Arguments

`x` a numeric matrix or data frame.

`y` a numeric vector with compatible dimensions to `x`.

`wrt` integer; Selects the regressor to differentiate with respect to (vectorized).

`eval.points` numeric or options: ("obs", "apd", "mean", "median", "last"); Regressor points to be evaluated.

- Numeric values must be in matrix or data.frame form to be evaluated for each regressor, otherwise, a vector of points will evaluate only at the `wrt` regressor. See examples for use cases.
- Set to `(eval.points = "obs")` (defalut) to find the average partial derivative at every observation of the variable with respect to *for specific tuples of given observations.*
- Set to `(eval.points = "apd")` to find the average partial derivative at every observation of the variable with respect to *over the entire distribution of other regressors.*
- Set to `(eval.points = "mean")` to find the partial derivative at the mean of value of every variable.
- Set to `(eval.points = "median")` to find the partial derivative at the median value of every variable.
- Set to `(eval.points = "last")` to find the partial derivative at the last observation of every value (relevant for time-series data).

`mixed` logical; `FALSE` (default) If mixed derivative is to be evaluated, set `(mixed = TRUE)`.

ncores	integer; value specifying the number of cores to be used in the parallelized procedure. If NULL (default), the number of cores to be used is equal to the number of cores of the machine - 1.
messages	logical; TRUE (default) Prints status messages.

Value

Returns column-wise matrix of wrt regressors:

- dy.d_(...)[,wrt]$First the 1st derivative
- dy.d_(...)[,wrt]$Second the 2nd derivative
- dy.d_(...)[,wrt]$Mixed the mixed derivative (for two independent variables only).

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Vinod, H. and Viole, F. (2020) "Comparing Old and New Partial Derivative Estimates from Nonlinear Nonparametric Regressions" https://www.ssrn.com/abstract=3681104

Examples

```
## Not run:
set.seed(123) ; x_1 <- runif(100) ; x_2 <- runif(100) ; y <- x_1 ^ 2 * x_2 ^ 2
B <- cbind(x_1, x_2)

#' ## To find derivatives of y wrt 1st regressor for specific points of both regressors
dy.d_(B, y, wrt = c(1, 2), eval.points = t(c(.5, .5)))

## To find average partial derivative of y wrt 1st regressor,
only supply 1 value in [eval.points], or a vector of [eval.points]:
dy.d_(B, y, wrt = 1, eval.points = .5)

dy.d_(B, y, wrt = 1, eval.points = fivenum(B[,1]))


## To find average partial derivative of y wrt 1st regressor,
for every oberservation of 1st regressor:
apd <- dy.d_(B, y, wrt = 1, eval.points = "apd")
plot(B[,1], apd[,1]$First)

## 95% Confidence Interval to test if 0 is within
### Lower CI
LPM.VaR(.025, 0, apd[,1]$First)

### Upper CI
UPM.VaR(.025, 0, apd[,1]$First)
```

```
## End(Not run)
```

LPM *Lower Partial Moment*

Description

This function generates a univariate lower partial moment for any degree or target.

Usage

```
LPM(degree, target, variable)
```

Arguments

degree	integer; (degree = 0) is frequency, (degree = 1) is area.
target	numeric; Typically set to mean, but does not have to be. (Vectorized)
variable	a numeric vector.

Value

LPM of variable

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100)
LPM(0, mean(x), x)
```

LPM.ratio *Lower Partial Moment RATIO*

Description

This function generates a standardized univariate lower partial moment for any degree or target.

Usage

```
LPM.ratio(degree, target, variable)
```

Arguments

`degree`	integer; (`degree = 0`) is frequency, (`degree = 1`) is area.
`target`	numeric; Typically set to mean, but does not have to be. (Vectorized)
`variable`	a numeric vector.

Value

Standardized LPM of variable

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Viole, F. (2017) "Continuous CDFs and ANOVA with NNS" https://www.ssrn.com/abstract=3007373

Examples

```
set.seed(123)
x <- rnorm(100)
LPM.ratio(0, mean(x), x)

## Not run:
## Empirical CDF (degree = 0)
lpm_cdf <- LPM.ratio(0, sort(x), x)
plot(sort(x), lpm_cdf)

## Continuous CDF (degree = 1)
lpm_cdf_1 <- LPM.ratio(1, sort(x), x)
plot(sort(x), lpm_cdf_1)

## Joint CDF
```

```
x <- rnorm(5000) ; y <- rnorm(5000)
plot3d(x, y, Co.LPM(0, 0, sort(x), sort(y), x, y), col = "blue", xlab = "X", ylab = "Y",
zlab = "Probability", box = FALSE)

## End(Not run)
```

LPM.VaR *LPM VaR*

Description

Generates a value at risk (VaR) quantile based on the Lower Partial Moment ratio.

Usage

```
LPM.VaR(percentile, degree, x)
```

Arguments

percentile	numeric [0, 1]; The percentile for left-tail VaR (vectorized).
degree	integer; (degree = 0) for discrete distributions, (degree = 1) for continuous distributions.
x	a numeric vector.

Value

Returns a numeric value representing the point at which "percentile" of the area of x is below.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100)

## For 5% quantile, left-tail
LPM.VaR(0.05, 0, x)
```

NNS.ANOVA *NNS ANOVA*

Description

Analysis of variance (ANOVA) based on lower partial moment CDFs for multiple variables. Returns a degree of certainty the difference in sample means is zero, not a p-value.

Usage

```
NNS.ANOVA(
  control,
  treatment,
  confidence.interval = 0.95,
  tails = "Both",
  pairwise = FALSE,
  plot = TRUE,
  binary = TRUE
)
```

Arguments

`control` a numeric vector, matrix or data frame.

`treatment` `NULL` (default) a numeric vector, matrix or data frame.

`confidence.interval` numeric [0, 1]; The confidence interval surrounding the control mean when (`binary = TRUE`). Defaults to (`confidence.interval = 0.95`).

`tails` options: ("Left", "Right", "Both"). `tails = "Both"`(Default) Selects the tail of the distribution to determine effect size.

`pairwise` logical; `FALSE` (defualt) Returns pairwise certainty tests when set to `pairwise = TRUE`.

`plot` logical; `TRUE` (default) Returns the boxplot of all variables along with grand mean identification. When (`binary = TRUE`), returns the boxplot of both variables along with grand mean identification and confidence interval thereof.

`binary` logical; `TRUE` (default) Selects binary analysis between a control and treatment variable.

Value

For (`binary = FALSE`) returns the degree certainty the difference in sample means is zero [0, 1].

For (`binary = TRUE`) returns:

- `"Control Mean"`
- `"Treatment Mean"`
- `"Grand Mean"`

- "Control CDF"
- "Treatment CDF"
- "Certainty" the certainty of the same population statistic
- "Lower Bound Effect" and "Upper Bound Effect" the effect size of the treatment for the specified confidence interval

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Viole, F. (2017) "Continuous CDFs and ANOVA with NNS" https://www.ssrn.com/abstract=3007373

Examples

```
### Binary analysis and effect size
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.ANOVA(control = x, treatment = y)

### Two variable analysis with no control variable
A <- cbind(x, y)
NNS.ANOVA(A)

### Multiple variable analysis with no control variable
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100) ; z <- rnorm(100)
A <- cbind(x, y, z)
NNS.ANOVA(A)
```

NNS.ARMA *NNS ARMA*

Description

Autoregressive model incorporating nonlinear regressions of component series.

Usage

```
NNS.ARMA(
  variable,
  h = 1,
  training.set = NULL,
  seasonal.factor = TRUE,
```

```
    weights = NULL,
    best.periods = 2,
    negative.values = FALSE,
    method = "nonlin",
    dynamic = FALSE,
    plot = TRUE,
    seasonal.plot = TRUE,
    conf.intervals = NULL,
    ncores = NULL
)
```

Arguments

`variable` a numeric vector.

`h` integer; 1 (default) Number of periods to forecast.

`training.set` numeric; `NULL` (default) Sets the number of variable observations (`variable[1 : training.set]`) to monitor performance of forecast over in-sample range.

`seasonal.factor` logical or integer(s); `TRUE` (default) Automatically selects the best seasonal lag from the seasonality test. To use weighted average of all seasonal lags set to (`seasonal.factor = FALSE`). Otherwise, directly input known frequency integer lag to use, i.e. (`seasonal.factor = 12`) for monthly data. Multiple frequency integers can also be used, i.e. (`seasonal.factor = c(12,24,36)`)

`weights` numeric; `NULL` (default) sets the weights of the `seasonal.factor` vector when specified as integers. If (`weights = NULL`) each `seasonal.factor` is weighted on its NNS.seas result and number of observations it contains.

`best.periods` integer; [2] (default) used in conjunction with (`seasonal.factor = FALSE`), uses the `best.periods` number of detected seasonal lags instead of `ALL` lags when (`seasonal.factor = FALSE`).

`negative.values` logical; `FALSE` (default) If the variable can be negative, set to (`negative.values = TRUE`). If there are negative values within the variable, `negative.values` will automatically be detected.

`method` options: ("lin", "nonlin", "both"); `"nonlin"` (default) To select the regression type of the component series, select (`method = "both"`) where both linear and nonlinear estimates are generated. To use a nonlinear regression, set to (`method = "nonlin"`); to use a linear regression set to (`method = "lin"`).

`dynamic` logical; `FALSE` (default) To update the seasonal factor with each forecast point, set to (`dynamic = TRUE`). The default is (`dynamic = FALSE`) to retain the original seasonal factor from the inputted variable for all ensuing `h`.

`plot` logical; `TRUE` (default) Returns the plot of all periods exhibiting seasonality and the `variable` level reference in upper panel. Lower panel returns original data and forecast.

seasonal.plot logical; TRUE (default) Adds the seasonality plot above the forecast. Will be set to FALSE if no seasonality is detected or seasonal.factor is set to an integer value.

conf.intervals numeric [0, 1]; NULL (default) Plots and returns the associated confidence intervals for the final estimate. Constructed using the maximum entropy bootstrap meboot on the final estimates.

ncores integer; value specifying the number of cores to be used in the parallelized procedure. If NULL (default), the number of cores to be used is equal to half the number of cores of the machine - 1.

Value

Returns a vector of forecasts of length (h) if no conf.intervals specified. Else, returns a data.table with the forecasts as well as lower and upper confidence intervals per forecast point.

Note

For monthly data series, increased accuracy may be realized from forcing seasonal factors to multiples of 12. For example, if the best periods reported are: {37, 47, 71, 73} use (seasonal.factor = c(36,48,72)).

(seasonal.factor = FALSE) can be a very computationally expensive exercise due to the number of seasonal periods detected.

If error encountered when (seasonal.factor = TRUE):

"NaNs produced Error in seq.default(length(variable)+1,1,-lag[i]) : wrong sign in 'by' argument"

use the combination of (seasonal.factor = FALSE,best.periods = 1).

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Viole, F. (2019) "Forecasting Using NNS" https://www.ssrn.com/abstract=3382300

Examples

```
## Nonlinear NNS.ARMA using AirPassengers monthly data and 12 period lag
## Not run:
NNS.ARMA(AirPassengers, h = 45, training.set = 100, seasonal.factor = 12, method = "nonlin")

## Linear NNS.ARMA using AirPassengers monthly data and 12, 24, and 36 period lags
NNS.ARMA(AirPassengers, h = 45, training.set = 120, seasonal.factor = c(12, 24, 36), method = "lin")

## Nonlinear NNS.ARMA using AirPassengers monthly data and 2 best periods lag
```

```
NNS.ARMA(AirPassengers, h = 45, training.set = 120, seasonal.factor = FALSE, best.periods = 2)
## End(Not run)
```

NNS.ARMA.optim *NNS ARMA Optimizer*

Description

Wrapper function for optimizing any combination of a given `seasonal.factor` vector in NNS.ARMA. Minimum sum of squared errors (forecast-actual) is used to determine optimum across all NNS.ARMA methods.

Usage

```
NNS.ARMA.optim(
  variable,
  training.set,
  seasonal.factor,
  negative.values = FALSE,
  obj.fn = expression(sum((predicted - actual)^2)),
  objective = "min",
  linear.approximation = TRUE,
  print.trace = TRUE,
  ncores = NULL
)
```

Arguments

`variable` a numeric vector.

`training.set` numeric; Sets the number of variable observations as the training set. See `Note` below for recommended uses.

`seasonal.factor` integers; Multiple frequency integers considered for NNS.ARMA model, i.e. `(seasonal.factor = c(12,24,36))`

`negative.values` logical; `FALSE` (default) If the variable can be negative, set to `(negative.values = TRUE)`.

`obj.fn` expression; `expression(sum((predicted -actual)^2))` (default) Sum of squared errors is the default objective function. Any `expression()` using the specific terms `predicted` and `actual` can be used.

`objective` options: ("min", "max") `"min"` (default) Select whether to minimize or maximize the objective function `obj.fn`.

`linear.approximation` logical; `TRUE` (default) Uses the best linear output from `NNS.reg` to generate a nonlinear and mixture regression for comparison. `FALSE` is a more exhaustive search over the objective space.

`print.trace` logical; `TRUE` (defualt) Prints current iteration information. Suggested as backup in case of error, best parameters to that point still known and copyable!

`ncores` integer; value specifying the number of cores to be used in the parallelized procedure. If NULL (default), the number of cores to be used is equal to half the number of cores of the machine.

Value

Returns a list containing:

- `$period` a vector of optimal seasonal periods
- `$weights` the optimal weights of each seasonal period between an equal weight or NULL weighting
- `$obj.fn` the objective function value
- `$method` the method identifying which NNS.ARMA method was used.
- `$bias.shift` a numerical result of the overall bias of the optimum objective function result. To be added to the final result when using the NNS.ARMA with the derived parameters.

Note

- Typically, `(training.set = length(variable) -2 * length(forecast horizon))` is used for optimization. Smaller samples would use `(training.set = length(variable) -length(forecast horizon))` in order to preserve information.
- The number of combinations will grow prohibitively large, they should be kept as small as possible. `seasonal.factor` containing an element too large will result in an error. Please reduce the maximum `seasonal.factor`.
- If variable cannot logically assume negative values, then the `$bias.shift` must be limited to 0 via a `pmax(0,...)` call.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
## Nonlinear NNS.ARMA period optimization using 2 yearly lags on AirPassengers monthly data
## Not run:
nns.optims <- NNS.ARMA.optim(AirPassengers[1:132], training.set = 120,
seasonal.factor = seq(12, 24, 6))

## Then use optimal parameters in NNS.ARMA to predict 12 periods in-sample.
## Note the {$bias.shift} usage in the {NNS.ARMA} function:
```

```
nns.estimates <- NNS.ARMA(AirPassengers, h = 12, training.set = 132,
seasonal.factor = nns.optims$periods, method = nns.optims$method) + nns.optims$bias.shift

## If variable cannot logically assume negative values
nns.estimates <- pmax(0, nns.estimates)

## End(Not run)
```

NNS.boost *NNS Boost*

Description

Ensemble method for classification using the predictions of the NNS multivariate regression NNS.reg collected from uncorrelated feature combinations.

Usage

```
NNS.boost(
  IVs.train,
  DV.train,
  IVs.test = NULL,
  type = NULL,
  representative.sample = FALSE,
  depth = "max",
  n.best = NULL,
  learner.trials = 100,
  epochs = NULL,
  CV.size = 0.25,
  ts.test = NULL,
  folds = 5,
  threshold = NULL,
  obj.fn = expression(sum((predicted - actual)^2)),
  objective = "min",
  extreme = FALSE,
  feature.importance = TRUE,
  status = TRUE,
  ncores = NULL
)
```

Arguments

IVs.train	a matrix or data frame of variables of numeric or factor data types.
DV.train	a numeric or factor vector with compatible dimensions to (IVs.train).
IVs.test	a matrix or data frame of variables of numeric or factor data types with compatible dimensions to (IVs.train). If NULL, will use (IVs.train) as default.

type	NULL (default). To perform a classification of discrete integer classes from factor target variable (DV.train), set to (type = "CLASS"), else for continuous (DV.train) set to (type = NULL).
representative.sample	logical; FALSE (default) Reduces observations of IVs.train to a set of representative observations per regressor.
depth	options: (integer, NULL, "max"); Specifies the order parameter in the NNS.reg routine, assigning a number of splits in the regressors. (depth = "max")(default) will be significantly faster, but increase the variance of results, which is suggested for mixed continuous and discrete (unordered, ordered) data.
n.best	integer; NULL (default) Sets the number of nearest regression points to use in weighting for multivariate regression at sqrt(# of regressors). Analogous to k in a k Nearest Neighbors algorithm. If NULL, determines the optimal clusters via the NNS.stack procedure.
learner.trials	integer; NULL (default) Sets the number of trials to obtain an accuracy threshold level. (learner.trials = 100) is the default setting.
epochs	integer; 2*length(DV.train) (default) Total number of feature combinations to run.
CV.size	numeric [0, 1]; (CV.size = .25) (default) Sets the cross-validation size. Defaults to 0.25 for a 25 percent random sampling of the training set.
ts.test	integer; NULL (default) Sets the length of the test set for time-series data; typically 2*h parameter value from NNS.ARMA or double known periods to forecast.
folds	integer; 5 (default) Sets the number of folds in the NNS.stack procedure for optimal n.best parameter.
threshold	numeric; NULL (default) Sets the obj.fn threshold to keep feature combinations.
obj.fn	expression; expression(sum((predicted -actual)^2)) (default) Sum of squared errors is the default objective function. Any expression() using the specific terms predicted and actual can be used. Automatically selects an accuracy measure when (type = "CLASS").
objective	options: ("min", "max") "max" (default) Select whether to minimize or maximize the objective function obj.fn.
extreme	logical; FALSE (default) Uses the maximum (minimum) threshold obtained from the learner.trials, rather than the upper (lower) quintile level for maximization (minimization) objective.
feature.importance	logical; TRUE (default) Plots the frequency of features used in the final estimate.
status	logical; TRUE (default) Prints status update message in console.
ncores	integer; value specifying the number of cores to be used in the parallelized procedure. If NULL (default), the number of cores to be used is equal to the number of cores of the machine - 1.

Value

Returns a vector of fitted values for the dependent variable test set $results, and the final feature loadings $feature.weights.

Note

Like a logistic regression, the (type = "CLASS") setting is not necessary for target variable of two classes e.g. [0, 1].

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. (2016) "Classification Using NNS Clustering Analysis" https://www.ssrn.com/abstract=2864711

Examples

```
## Using 'iris' dataset where test set [IVs.test] is 'iris' rows 141:150.
## Not run:
a <- NNS.boost(iris[1:140, 1:4], iris[1:140, 5],
IVs.test = iris[141:150, 1:4],
epochs = 100, learner.trials = 100,
type = "CLASS")

## Test accuracy
mean(a$results == as.numeric(iris[141:150, 5]))

## End(Not run)
```

NNS.caus *NNS Causation*

Description

Returns the causality from observational data between two variables.

Usage

```
NNS.caus(x, y, factor.2.dummy = FALSE, tau = 0, plot = FALSE)
```

Arguments

x	a numeric vector, matrix or data frame.
y	NULL (default) or a numeric vector with compatible dimsensions to x.
factor.2.dummy	logical; FALSE (default) Automatically augments variable matrix with numerical dummy variables based on the levels of factors. Includes dependent variable y.

`tau` options: ("cs", "ts", integer); 0 (default) Number of lagged observations to consider (for time series data). Otherwise, set `(tau = "cs")` for cross-sectional data. `(tau = "ts")` automatically selects the lag of the time series data, while `(tau = [integer])` specifies a time series lag.

`plot` logical; `FALSE` (default) Plots the raw variables, tau normalized, and cross-normalized variables.

Value

Returns the directional causation (x —> y) or (y —> x) and net quantity of association. For causal matrix, directional causation is returned as ([column variable] —> [row variable]). Negative numbers represent causal direction attributed to [row variable].

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
## Not run:
## x causes y...
set.seed(123)
x <- rnorm(1000) ; y <- x ^ 2
NNS.caus(x, y, tau = "cs")

## Causal matrix without per factor causation
NNS.caus(iris, tau = 0)

## Causal matrix with per factor causation
NNS.caus(iris, factor.2.dummy = TRUE, tau = 0)

## End(Not run)
```

NNS.CDF *NNS CDF*

Description

This function generates an empirical CDF using partial moment ratios LPM.ratio, and resulting survival, hazard and cumulative hazard functions.

Usage

```
NNS.CDF(variable, degree = 0, target = NULL, type = "CDF", plot = TRUE)
```

Arguments

variable	a numeric vector or data.frame of 2 variables for joint CDF.
degree	integer; (degree = 0) (default) is frequency, (degree = 1) is area.
target	numeric; NULL (default) Must lie within support of each variable.
type	options("CDF", "survival", "hazard", "cumulative hazard"); "CDF" (default) Selects type of function to return for bi-variate analysis. Multivariate analysis is restricted to "CDF".
plot	logical; plots CDF.

Value

Returns:

- "Function" a data.table containing the observations and resulting CDF of the variable.
- "target.value" value from the target argument.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Viole, F. (2017) "Continuous CDFs and ANOVA with NNS" https://www.ssrn.com/abstract=3007373

Examples

```
set.seed(123)
x <- rnorm(100)
NNS.CDF(x)

## Not run:
## Empirical CDF (degree = 0)
NNS.CDF(x)

## Continuous CDF (degree = 1)
NNS.CDF(x, 1)

## Joint CDF
x <- rnorm(5000) ; y <- rnorm(5000)
A <- cbind(x,y)

NNS.CDF(A, 0)

## Joint CDF with target
NNS.CDF(A, 0, target = c(0,0))

## End(Not run)
```

NNS.dep *NNS Dependence*

Description

Returns the dependence and nonlinear correlation between two variables based on higher order partial moment matrices measured by frequency or area.

Usage

```
NNS.dep(x, y = NULL, asym = FALSE, print.map = FALSE, ncores = NULL)
```

Arguments

`x`	a numeric vector, matrix or data frame.
`y`	`NULL` (default) or a numeric vector with compatible dimsensions to `x`.
`asym`	logical; `FALSE` (default) Allows for asymmetrical dependencies.
`print.map`	logical; `FALSE` (default) Plots quadrant means.
`ncores`	integer; value specifying the number of cores to be used in the parallelized procedure. If NULL (default), the number of cores to be used is equal to the number of cores of the machine - 1.

Value

Returns the bi-variate `"Correlation"` and `"Dependence"` or correlation / dependence matrix for matrix input.

Note

p-values and confidence intervals can be obtained from sampling random permutations of `y_p` and running `NNS.dep(x,y_p)` to compare against a null hypothesis of 0 correlation or independence between `x,y`.

`NNS.cor` has been deprecated `(NNS >= 0.5.4)` and can be called via `NNS.dep`.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
## Not run:
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.dep(x, y)

## Correlation / Dependence Matrix
x <- rnorm(100) ; y <- rnorm(100) ; z <- rnorm(100)
B <- cbind(x, y, z)
NNS.dep(B)


## p-values for [NNS.dep]
x <- seq(-5, 5, .1); y <- x^2 + rnorm(length(x))


nns_cor_dep <- NNS.dep(x, y, print.map = TRUE)
nns_cor_dep

## Create permutations of y
y_p <- replicate(1000, sample.int(length(y)))

## Generate new correlation and dependence measures on each new permutation of y
nns.mc <- apply(y_p, 2, function(g) NNS.dep(x, y[g]))

## Store results
cors <- unlist(lapply(nns.mc, "[[", 1))
deps <- unlist(lapply(nns.mc, "[[", 2))

## View results
hist(cors)
hist(deps)

## Left tailed correlation p-value
cor_p_value <- LPM(0, nns_cor_dep$Correlation, cors)
cor_p_value

## Right tailed correlation p-value
cor_p_value <- UPM(0, nns_cor_dep$Correlation, cors)
cor_p_value

## Confidence Intervals
## For 95th percentile VaR (both-tails) see [LPM.VaR] and [UPM.VaR]
## Lower CI
LPM.VaR(.025, 0, cors)
## Upper CI
UPM.VaR(.025, 0, cors)

## Left tailed dependence p-value
dep_p_value <- LPM(0, nns_cor_dep$Dependence, deps)
dep_p_value
```

```
## Right tailed dependence p-value
dep_p_value <- UPM(0, nns_cor_dep$Dependence, deps)
dep_p_value

## End(Not run)
```

`NNS.dep.base` *NNS Dependence Base*

Description

Internal function for NNS dependence NNS.dep parallel instances.

Usage

```
NNS.dep.base(
  x,
  y = NULL,
  order = NULL,
  degree = NULL,
  type = NULL,
  print.map = FALSE,
  asym = FALSE
)
```

Arguments

`x`	from NNS.part
`y`	from NNS.part
`order`	from NNS.part
`degree`	from NNS.part
`type`	from NNS.part
`print.map`	from NNS.part
`asym`	for asymmetrical dependecies

Value

Returns NNS dependence.

NNS.dep.hd — NNS Co-Partial Moments Higher Dimension Dependence

Description

Determines higher dimension dependence coefficients based on degree 0 co-partial moments.

Usage

```
NNS.dep.hd(x, plot = FALSE, independence.overlay = FALSE)
```

Arguments

`x` a numeric matrix or data frame.

`plot` logical; `FALSE` (default) Generates a 3d scatter plot with regression points using plot3d.

`independence.overlay` logical; `FALSE` (default) Creates and overlays independent Co.LPM and Co.UPM regions to visually reference the difference in dependence from the data.frame of variables being analyzed. Under independence, the light green and red shaded areas would be occupied by green and red data points respectively.

Value

- $actual.observations Number of Co.LPM and Co.UPM observations.
- $independent.null Expected number of Co.LPM and Co.UPM observations under the null hypothesis of independence.
- $Dependence Multivariate nonlinear dependence coefficient [0,1]

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. (2016) "Beyond Correlation: Using the Elements of Variance for Conditional Means and Probabilities" https://www.ssrn.com/abstract=2745308.

Examples

```
set.seed(123)
x <- rnorm(1000) ; y <- rnorm(1000) ; z <- rnorm(1000)
A <- data.frame(x, y, z)
NNS.dep.hd(A, plot = TRUE, independence.overlay = TRUE)
```

NNS.diff *NNS Numerical Differentiation*

Description

Determines numerical derivative of a given function using projected secant lines on the y-axis. These projected points infer finite steps h, in the finite step method.

Usage

```
NNS.diff(f, point, h = 0.1, tol = 1e-10, print.trace = FALSE)
```

Arguments

f	an expression or call or a formula with no lhs.
point	numeric; Point to be evaluated for derivative of a given function f.
h	numeric [0, ...]; Initial step for secant projection. Defaults to (h = 0.1).
tol	numeric; Sets the tolerance for the stopping condition of the inferred h. Defualts to (tol = 1e-10).
print.trace	logical; FALSE (default) Displays each iteration, lower y-intercept, upper y-intercept and inferred h.

Value

Returns a matrix of values, intercepts, derivatives, inferred step sizes for multiple methods of estimation.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
f <- function(x) sin(x) / x
NNS.diff(f, 4.1)
```

NNS.distance *NNS Distance*

Description

Internal kernel function for NNS multivariate regression NNS.reg parallel instances.

Usage

```
NNS.distance(rpm, dist.estimate, type, k, n)
```

Arguments

`rpm`	REGRESSION.POINT.MATRIX from NNS.reg
`dist.estimate`	Vector to generate distances from.
`type`	"L1", "L2", "DTW" or "FACTOR"
`k`	`n.best` from NNS.reg
`n`	number of observations.

Value

Returns sum of weighted distances.

NNS.FSD *NNS FSD Test*

Description

Bi-directional test of first degree stochastic dominance using lower partial moments.

Usage

```
NNS.FSD(x, y, type = "discrete")
```

Arguments

`x`	a numeric vector.
`y`	a numeric vector.
`type`	options: ("discrete", "continuous"); `"discrete"` (default) selects the type of CDF.

Value

Returns one of the following FSD results: `"X FSD Y"`, `"Y FSD X"`, or `"NO FSD EXISTS"`.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2016) "LPM Density Functions for the Computation of the SD Efficient Set." Journal of Mathematical Finance, 6, 105-126. https://www.scirp.org/Journal/PaperInformation.aspx?PaperID=63817.

Viole, F. (2017) "A Note on Stochastic Dominance." https://www.ssrn.com/abstract=3002675.

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.FSD(x, y)
```

NNS.FSD.uni *NNS FSD Test uni-directional*

Description

Uni-directional test of first degree stochastic dominance using lower partial moments used in SD Efficient Set routine.

Usage

```
NNS.FSD.uni(x, y, type = "discrete")
```

Arguments

`x`	a numeric vector.
`y`	a numeric vector.
`type`	options: ("discrete", "continuous"); `"discrete"` (default) selects the type of CDF.

Value

Returns (1) if `"X FSD Y"`, else (0).

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2016) "LPM Density Functions for the Computation of the SD Efficient Set." Journal of Mathematical Finance, 6, 105-126. https://www.scirp.org/Journal/PaperInformation.aspx?PaperID=63817.

Viole, F. (2017) "A Note on Stochastic Dominance." https://www.ssrn.com/abstract=3002675.

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.FSD.uni(x, y)
```

NNS.meboot *NNS meboot*

Description

Adapted maximum entropy bootstrap routine from meboot https://cran.r-project.org/package=meboot.

Usage

```
NNS.meboot(
  x,
  reps = 999,
  setSpearman = NULL,
  drift = TRUE,
  trim = 0.1,
  xmin = NULL,
  xmax = NULL,
  reachbnd = TRUE,
  expand.sd = TRUE,
  force.clt = TRUE,
  scl.adjustment = FALSE,
  sym = FALSE,
  elaps = FALSE,
  colsubj,
  coldata,
  coltimes,
  ...
)
```

Arguments

x	vector of data, ts object or pdata.frame object.
reps	numeric; number of replicates to generate.

`setSpearman`	numeric [0,1]; The default setting `setSpearman = NULL` assumes that the user does not want to generate replicates that are perfectly dependent on original time series, `setSpearman=1` recovers the original `meboot(...)` settings. `setSpearman < 1` admits less perfect (more realistic for some purposes) dependence.
`drift`	logical; `TRUE` default preserves the drift of the original series.
`trim`	numeric [0,1]; The mean trimming proportion, defaults to `trim=0.1`.
`xmin`	numeric; the lower limit for the left tail.
`xmax`	numeric; the upper limit for the right tail.
`reachbnd`	logical; If `TRUE` potentially reached bounds (xmin = smallest value - trimmed mean and xmax = largest value + trimmed mean) are given when the random draw happens to be equal to 0 and 1, respectively.
`expand.sd`	logical; If `TRUE` the standard deviation in the ensemble is expanded. See `expand.sd` in `meboot::meboot`.
`force.clt`	logical; If `TRUE` the ensemble is forced to satisfy the central limit theorem. See `force.clt` in `meboot::meboot`.
`scl.adjustment`	logical; If `TRUE` scale adjustment is performed to ensure that the population variance of the transformed series equals the variance of the data.
`sym`	logical; If `TRUE` an adjustment is peformed to ensure that the ME density is symmetric.
`elaps`	logical; If `TRUE` elapsed time during computations is displayed.
`colsubj`	numeric; the column in `x` that contains the individual index. It is ignored if the input data `x` is not a `pdata.frame` object.
`coldata`	numeric; the column in `x` that contains the data of the variable to create the ensemble. It is ignored if the input data `x` is not a `pdata.frame` object.
`coltimes`	numeric; an optional argument indicating the column that contains the times at which the observations for each individual are observed. It is ignored if the input data `x` is not a `pdata.frame` object.
`...`	possible argument `fiv` to be passed to `expand.sd`.

Value

- x original data provided as input.
- replicates maximum entropy bootstrap replicates.
- ensemble average observation over all replicates.
- xx sorted order stats (xx[1] is minimum value).
- z class intervals limits.
- dv deviations of consecutive data values.
- dvtrim trimmed mean of dv.
- xmin data minimum for ensemble=xx[1]-dvtrim.
- xmax data x maximum for ensemble=xx[n]+dvtrim.
- desintxb desired interval means.
- ordxx ordered x values.
- kappa scale adjustment to the variance of ME density.
- elaps elapsed time.

References

- Vinod, H.D. and Viole, F. (2020) Arbitrary Spearman's Rank Correlations in Maximum Entropy Bootstrap and Improved Monte Carlo Simulations https://www.ssrn.com/abstract=3621614
- Vinod, H.D. (2013), Maximum Entropy Bootstrap Algorithm Enhancements. https://www.ssrn.com/abstract=2285041.
- Vinod, H.D. (2006), Maximum Entropy Ensembles for Time Series Inference in Economics, *Journal of Asian Economics*, **17**(6), pp. 955-978.
- Vinod, H.D. (2004), Ranking mutual funds using unconventional utility theory and stochastic dominance, *Journal of Empirical Finance*, **11**(3), pp. 353-377.

Examples

```
## Not run:
# To generate an orthogonal rank correlated time-series to AirPassengers
boots <- NNS.meboot(AirPassengers, reps=100, setSpearman = 0, xmin = 0)

# Verify correlation of replicates ensemble to original
cor(boots$ensemble, AirPassengers, method = "spearman")

# Plot all replicates
matplot(boots$replicates, type = 'l')

# Plot ensemble
lines(boots$ensemble, lwd = 3)

## End(Not run)
```

`NNS.norm` *NNS Normalization*

Description

Normalizes a matrix of variables based on nonlinear scaling normalization method.

Usage

```
NNS.norm(A, linear = FALSE, chart.type = NULL, location = "topleft")
```

Arguments

`A`	a numeric matrix or data frame.
`linear`	logical; `FALSE` (default) Performs a linear scaling normalization, resulting in equal means for all variables.
`chart.type`	options: ("l", "b"); `NULL` (default). Set `(chart.type = "l")` for line, `(chart.type = "b")` for boxplot.
`location`	Sets the legend location within the plot, per the x and y co-ordinates used in base graphics legend.

Value

Returns a data.frame of normalized values.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100) ; y<-rnorm(100)
A <- cbind(x, y)
NNS.norm(A)
```

NNS.part *NNS Partition Map*

Description

Creates partitions based on partial moment quadrant means, iteratively assigning identifications to observations based on those quadrants (unsupervised partitional and hierarchial clustering method). Basis for correlation, dependence NNS.dep, regression NNS.reg routines.

Usage

```
NNS.part(
  x,
  y,
  Voronoi = FALSE,
  type = NULL,
  order = NULL,
  obs.req = 8,
  min.obs.stop = TRUE,
  noise.reduction = "off"
)
```

Arguments

x	a numeric vector.
y	a numeric vector with compatible dimensions to x.
Voronoi	logical; FALSE (default) Displays a Voronoi type diagram using partial moment quadrants.

type
: NULL (default) Controls the partitioning basis. Set to (type = "XONLY") for X-axis based partitioning. Defaults to NULL for both X and Y-axis partitioning.

order
: integer; Number of partial moment quadrants to be generated. (order = "max") will institute a perfect fit.

obs.req
: integer; (8 default) Required observations per cluster where quadrants will not be further partitioned if observations are not greater than the entered value. Reduces minimum number of necessary observations in a quadrant to 1 when (obs.req = 1).

min.obs.stop
: logical; TRUE (default) Stopping condition where quadrants will not be further partitioned if a single cluster contains less than the entered value of obs.req.

noise.reduction
: the method of determining regression points options: ("mean", "median", "mode", "off"); (noise.reduction = "mean") uses means for partitions. (noise.reduction = "median") uses medians instead of means for partitions, while (noise.reduction = "mode") uses modes instead of means for partitions. Defaults to (noise.reduction = "off") where an overall central tendency measure is used.

Value

Returns:

- "dt" a data.table of x and y observations with their partition assignment "quadrant" in the 3rd column and their prior partition assignment "prior.quadrant" in the 4th column.
- "regression.points" the data.table of regression points for that given (order = ...).
- "order" the order of the final partition given "min.obs.stop" stopping condition.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.part(x, y)

## Data.table of observations and partitions
NNS.part(x, y, order = 1)$dt

## Regression points
NNS.part(x, y, order = 1)$regression.points

## Voronoi style plot
NNS.part(x, y, Voronoi = TRUE)
```

```
## Examine final counts by quadrant
DT <- NNS.part(x, y)$dt
DT[ , counts := .N, by = quadrant]
DT
```

NNS.PDF *NNS PDF*

Description

This function generates an empirical PDF using dy.dx on NNS.CDF.

Usage

```
NNS.PDF(variable, degree = 1, target = NULL, bins = NULL, plot = TRUE)
```

Arguments

variable	a numeric vector.
degree	integer; (degree = 0) is frequency, (degree = 1) (default) is area.
target	a numeric range of values [a,b] where a < b. NULL (default) uses the variable min and max observations respectively.
bins	integer; NULL Selects number of bins. Bin width defaults to density(x)$bw.
plot	logical; plots PDF.

Value

Returns a data.table containing the intervals used and resulting PDF of the variable.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100)
NNS.PDF(x)

## Custom target range
NNS.PDF(x, target = c(-5, 5))
```

`NNS.reg` *NNS Regression*

Description

Generates a nonlinear regression based on partial moment quadrant means.

Usage

```
NNS.reg(
  x,
  y,
  factor.2.dummy = TRUE,
  order = NULL,
  stn = 0.95,
  dim.red.method = NULL,
  tau = NULL,
  type = NULL,
  point.est = NULL,
  location = "top",
  return.values = TRUE,
  plot = TRUE,
  plot.regions = FALSE,
  residual.plot = TRUE,
  std.errors = FALSE,
  confidence.interval = NULL,
  threshold = 0,
  n.best = NULL,
  noise.reduction = "off",
  dist = "L2",
  ncores = NULL,
  multivariate.call = FALSE
)
```

Arguments

`x`	a vector, matrix or data frame of variables of numeric or factor data types.
`y`	a numeric or factor vector with compatible dimensions to `x`.
`factor.2.dummy`	logical; `TRUE` (default) Automatically augments variable matrix with numerical dummy variables based on the levels of factors.
`order`	integer; Controls the number of partial moment quadrant means. Users are encouraged to try different `(order = ...)` integer settings with `(noise.reduction = "off")`. `(order = "max")` will force a limit condition perfect fit.
`stn`	numeric [0, 1]; Signal to noise parameter, sets the threshold of `(NNS.dep)` which reduces `("order")` when `(order = NULL)`. Defaults to 0.95 to ensure high dependence for higher `("order")` and endpoint determination.

dim.red.method
: options: ("cor", "NNS.dep", "NNS.caus", "all", NULL) method for determining synthetic X* coefficients. Selection of a method automatically engages the dimension reduction regression. The default is `NULL` for full multivariate regression. `(dim.red.method = "NNS.dep")` uses NNS.dep for nonlinear dependence weights, while `(dim.red.method = "NNS.caus")` uses NNS.caus for causal weights. `(dim.red.method = "cor")` uses standard linear correlation for weights. `(dim.red.method = "all")` averages all methods for further feature engineering.

tau
: options("ts", NULL); `NULL`(default) To be used in conjunction with `(dim.red.method = "NNS.caus")` or `(dim.red.method = "all")`. If the regression is using time-series data, set `(tau = "ts")` for more accurate causal analysis.

type
: `NULL` (default). To perform a classification, set to `(type = "CLASS")`. Like a logistic regression, it is not necessary for target variable of two classes e.g. [0, 1].

point.est
: a numeric or factor vector with compatible dimensions to `x`. Returns the fitted value `y.hat` for any value of `x`.

location
: Sets the legend location within the plot, per the `x` and `y` co-ordinates used in base graphics legend.

return.values
: logical; `TRUE` (default), set to `FALSE` in order to only display a regression plot and call values as needed.

plot
: logical; `TRUE` (default) To plot regression.

plot.regions
: logical; `FALSE` (default). Generates 3d regions associated with each regression point for multivariate regressions. Note, adds significant time to routine.

residual.plot
: logical; `TRUE` (default) To plot `y.hat` and `Y`.

std.errors
: logical; `FALSE` (default) To provide standard errors of each linear segment in the `"Fitted.xy"` output.

confidence.interval
: numeric [0, 1]; `NULL` (default) Plots the associated confidence interval with the estimate and reports the standard error for each individual segment.

threshold
: numeric [0, 1]; `(threshold = 0)` (default) Sets the threshold for dimension reduction of independent variables when `(dim.red.method)` is not `NULL`.

n.best
: integer; `NULL` (default) Sets the number of nearest regression points to use in weighting for multivariate regression at `sqrt(# of regressors)`. `(n.best = "all")` will select and weight all generated regression points. Analogous to `k` in a `k Nearest Neighbors` algorithm. Different values of `n.best` are tested using cross-validation in NNS.stack.

noise.reduction
: the method of determining regression points options: ("mean", "median", "mode", "off"); In low signal:noise situations,`(noise.reduction = "mean")` uses means for NNS.dep restricted partitions, `(noise.reduction = "median")` uses medians instead of means for NNS.dep restricted partitions, while `(noise.reduction = "mode")` uses modes instead of means for NNS.dep restricted partitions. `(noise.reduction = "off")` uses an overall central tendency measure for partitions.

`dist` options:("L1", "L2", "DTW", "FACTOR") the method of distance calculation; Selects the distance calculation used. `dist = "L2"` (default) selects the Euclidean distance and `(dist = "L1")` seclects the Manhattan distance; `(dist = "DTW")` selects the dynamic time warping distance; `(dist = "FACTOR")` uses a frequency.

`ncores` integer; value specifying the number of cores to be used in the parallelized procedure. If NULL (default), the number of cores to be used is equal to the number of cores of the machine - 1.

`multivariate.call`
Internal parameter for multivariate regressions.

Value

UNIVARIATE REGRESSION RETURNS THE FOLLOWING VALUES:

- `"R2"` provides the goodness of fit;
- `"SE"` returns the overall standard error of the estimate between `y` and `y.hat`;
- `"Prediction.Accuracy"` returns the correct rounded `"Point.est"` used in classifications versus the categorical `y`;
- `"derivative"` for the coefficient of the `x` and its applicable range;
- `"Point.est"` for the predicted value generated;
- `"regression.points"` provides the points used in the regression equation for the given order of partitions;
- `"Fitted.xy"` returns a data.table of `x`, `y`, `y.hat`, `resid`, `NNS.ID`, `gradient`;

MULTIVARIATE REGRESSION RETURNS THE FOLLOWING VALUES:

- `"R2"` provides the goodness of fit;
- `"equation"` returns the numerator of the synthetic X* dimension reduction equation as a data.table consisting of regressor and its coefficient. Denominator is simply the length of all coefficients > 0, returned in last row of `equation` data.table.
- `"x.star"` returns the synthetic X* as a vector;
- `"rhs.partitions"` returns the partition points for each regressor `x`;
- `"RPM"` provides the Regression Point Matrix, the points for each `x` used in the regression equation for the given order of partitions;
- `"Point.est"` returns the predicted value generated;
- `"Fitted.xy"` returns a data.table of `x`,`y`, `y.hat`, `gradient`, and `NNS.ID`.

Note

Please ensure `point.est` is of compatible dimensions to `x`, error message will ensue if not compatible.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Vinod, H. and Viole, F. (2017) "Nonparametric Regression Using Clusters" https://link.springer.com/article/10.1007/s10614-017-9713-5

Vinod, H. and Viole, F. (2018) "Clustering and Curve Fitting by Line Segments" https://www.preprints.org/manuscript/201801.0090/v1

Examples

```
## Not run:
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.reg(x, y)

## Manual {order} selection
NNS.reg(x, y, order = 2)

## Maximum {order} selection
NNS.reg(x, y, order = "max")

## x-only paritioning (Univariate only)
NNS.reg(x, y, type = "XONLY")

## For Multiple Regression:
x <- cbind(rnorm(100), rnorm(100), rnorm(100)) ; y <- rnorm(100)
NNS.reg(x, y, point.est = c(.25, .5, .75))

## For Multiple Regression based on Synthetic X* (Dimension Reduction):
x <- cbind(rnorm(100), rnorm(100), rnorm(100)) ; y <- rnorm(100)
NNS.reg(x, y, point.est = c(.25, .5, .75), dim.red.method = "cor")

## IRIS dataset examples:
# Dimension Reduction:
NNS.reg(iris[,1:4], iris[,5], dim.red.method = "cor", order = 5)

# Dimension Reduction using causal weights:
NNS.reg(iris[,1:4], iris[,5], dim.red.method = "NNS.caus", order = 5)

# Multiple Regression:
NNS.reg(iris[,1:4], iris[,5], order = 2, noise.reduction = "off")

# Classification:
NNS.reg(iris[,1:4], iris[,5], point.est = iris[1:10, 1:4], type = "CLASS")$Point.est

## To call fitted values:
x <- rnorm(100) ; y <- rnorm(100)
NNS.reg(x, y)$Fitted

## To call partial derivative (univariate regression only):
NNS.reg(x, y)$derivative
```

```
## End(Not run)
```

`NNS.SD.efficient.set` *NNS SD Efficient Set*

Description

Determines the set of stochastic dominant variables for various degrees.

Usage

```
NNS.SD.efficient.set(x, degree, type = "discrete")
```

Arguments

`x`	a numeric matrix or data frame.
`degree`	numeric options: (1, 2, 3); Degree of stochastic dominance test from (1, 2 or 3).
`type`	options: ("discrete", "continuous"); `"discrete"` (default) selects the type of CDF.

Value

Returns set of stochastic dominant variable names.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2016) "LPM Density Functions for the Computation of the SD Efficient Set." Journal of Mathematical Finance, 6, 105-126. https://www.scirp.org/Journal/PaperInformation.aspx?PaperID=63817.

Viole, F. (2017) "A Note on Stochastic Dominance." https://www.ssrn.com/abstract=3002675.

Examples

```
set.seed(123)
x <- rnorm(100) ; y<-rnorm(100) ; z<-rnorm(100)
A <- cbind(x, y, z)
NNS.SD.efficient.set(A, 1)
```

NNS.seas *NNS Seasonality Test*

Description

Seasonality test based on the coefficient of variation for the variable and lagged component series. A result of 1 signifies no seasonality present.

Usage

```
NNS.seas(variable, modulo = NULL, mod.only = TRUE, plot = TRUE)
```

Arguments

variable	a numeric vector.
modulo	integer(s); NULL (default) Used to find the nearest multiple(s) in the reported seasonal period.
mod.only	logical; codeTRUE (default) Limits the number of seasonal periods returned to the specified modulo.
plot	logical; TRUE (default) Returns the plot of all periods exhibiting seasonality and the variable level reference.

Value

Returns a matrix of all periods exhibiting less coefficient of variation than the variable with "all.periods"; and the single period exhibiting the least coefficient of variation versus the variable with "best.period"; as well as a vector of "periods" for easy call into NNS.ARMA.optim. If no seasonality is detected, NNS.seas will return ("No Seasonality Detected").

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100)

## To call strongest period based on coefficient of variation:
NNS.seas(x, plot = FALSE)$best.period

## Using modulos for logical seasonal inference:
NNS.seas(x, modulo = c(2,3,5,7), plot = FALSE)
```

`NNS.SSD` *NNS SSD Test*

Description

Bi-directional test of second degree stochastic dominance using lower partial moments.

Usage

```
NNS.SSD(x, y)
```

Arguments

x	a numeric vector.
y	a numeric vector.

Value

Returns one of the following SSD results: "X SSD Y", "Y SSD X", or "NO SSD EXISTS".

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2016) "LPM Density Functions for the Computation of the SD Efficient Set." Journal of Mathematical Finance, 6, 105-126. https://www.scirp.org/Journal/PaperInformation.aspx?PaperID=63817.

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.SSD(x, y)
```

`NNS.SSD.uni` *NNS SSD Test uni-directional*

Description

Uni-directional test of second degree stochastic dominance using lower partial moments used in SD Efficient Set routine.

Usage

```
NNS.SSD.uni(x, y)
```

Arguments

x a numeric vector.

y a numeric vector.

Value

Returns (1) if "X SSD Y", else (0).

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2016) "LPM Density Functions for the Computation of the SD Efficient Set." Journal of Mathematical Finance, 6, 105-126. https://www.scirp.org/Journal/PaperInformation.aspx?PaperID=63817.

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.SSD.uni(x, y)
```

NNS.stack *NNS Stack*

Description

Prediction model using the predictions of the NNS base models NNS.reg as features (i.e. meta-features) for the stacked model.

Usage

```
NNS.stack(
  IVs.train,
  DV.train,
  IVs.test = NULL,
  type = NULL,
  obj.fn = expression(sum((predicted - actual)^2)),
  objective = "min",
  dist = "L2",
  CV.size = NULL,
  ts.test = NULL,
  folds = 5,
  order = NULL,
  norm = NULL,
```

```
  method = c(1, 2),
  stack = TRUE,
  dim.red.method = "cor",
  status = TRUE,
  ncores = NULL
)
```

Arguments

- `IVs.train` — a vector, matrix or data frame of variables of numeric or factor data types.
- `DV.train` — a numeric or factor vector with compatible dimensions to `(IVs.train)`.
- `IVs.test` — a vector, matrix or data frame of variables of numeric or factor data types with compatible dimensions to `(IVs.train)`. If NULL, will use `(IVs.train)` as default.
- `type` — `NULL` (default). To perform a classification of discrete integer classes from factor target variable `(DV.train)`, set to `(type = "CLASS")`, else for continuous `(DV.train)` set to `(type = NULL)`. Like a logistic regression, this setting is not necessary for target variable of two classes e.g. [0, 1].
- `obj.fn` — expression; `expression(sum((predicted -actual)^2))` (default) Sum of squared errors is the default objective function. Any `expression()` using the specific terms `predicted` and `actual` can be used.
- `objective` — options: ("min", "max") `"min"` (default) Select whether to minimize or maximize the objective function `obj.fn`.
- `dist` — options:("L1", "L2", "DTW", "FACTOR") the method of distance calculation; Selects the distance calculation used. `dist = "L2"` (default) selects the Euclidean distance and `(dist = "L1")` selects the Manhattan distance; `(dist = "DTW")` selects the dynamic time warping distance; `(dist = "FACTOR")` uses a frequency.
- `CV.size` — numeric [0, 1]; `NULL` (default) Sets the cross-validation size if `(IVs.test = NULL)`. Defaults to 0.25 for a 25 percent random sampling of the training set under `(CV.size = NULL)`.
- `ts.test` — integer; NULL (default) Sets the length of the test set for time-series data; typically `2*h` parameter value from NNS.ARMA or double known periods to forecast.
- `folds` — integer; `folds = 5` (default) Select the number of cross-validation folds.
- `order` — options: (integer, "max", NULL); `NULL` (default) Sets the order for NNS.reg, where `(order = "max")` is the k-nearest neighbors equivalent, which is suggested for mixed continuous and discrete (unordered, ordered) data.
- `norm` — options: ("std", "NNS", NULL); `NULL` (default) 3 settings offered: `NULL`, `"std"`, and `"NNS"`. Selects the `norm` parameter in NNS.reg.
- `method` — numeric options: (1, 2); Select the NNS method to include in stack. `(method = 1)` selects NNS.reg; `(method = 2)` selects NNS.reg dimension reduction regression. Defaults to `method = c(1,2)`, which will reduce the dimension first, then find the optimal `n.best`.

`stack` logical; `TRUE` (default) Uses dimension reduction output in `n.best` optimization, otherwise performs both analyses independently.

`dim.red.method` options: ("cor", "NNS.dep", "NNS.caus", "all") method for determining synthetic X* coefficients. `(dim.red.method = "cor")` (default) uses standard linear correlation for weights. `(dim.red.method = "NNS.dep")` uses NNS.dep for nonlinear dependence weights, while `(dim.red.method = "NNS.caus")` uses NNS.caus for causal weights. `(dim.red.method = "all")` averages all methods for further feature engineering.

`status` logical; `TRUE` (default) Prints status update message in console.

`ncores` integer; value specifying the number of cores to be used in the parallelized subroutine NNS.reg. If NULL (default), the number of cores to be used is equal to the number of cores of the machine - 1.

Value

Returns a vector of fitted values for the dependent variable test set for all models.

- `"NNS.reg.n.best"` returns the optimum `"n.best"` parameter for the NNS.reg multivariate regression. `"SSE.reg"` returns the SSE for the NNS.reg multivariate regression.
- `"OBJfn.reg"` returns the `obj.fn` for the NNS.reg regression.
- `"NNS.dim.red.threshold"` returns the optimum `"threshold"` from the NNS.reg dimension reduction regression.
- `"OBJfn.dim.red"` returns the `obj.fn` for the NNS.reg dimension reduction regression.
- `"reg"` returns NNS.reg output.
- `"dim.red"` returns NNS.reg dimension reduction regression output.
- `"stack"` returns the output of the stacked model.

Note

- Like a logistic regression, the `(type = "CLASS")` setting is not necessary for target variable of two classes e.g. [0, 1].
- Missing data should be handled prior as well using na.omit or complete.cases on the full dataset.

If error received:

`"Error in is.data.frame(x) : object 'RP' not found"`

reduce the `CV.size`.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. (2016) "Classification Using NNS Clustering Analysis" https://www.ssrn.com/abstract=2864711

Examples

```
## Using 'iris' dataset where test set [IVs.test] is 'iris' rows 141:150.
## Not run:
NNS.stack(iris[1:140, 1:4], iris[1:140, 5], IVs.test = iris[141:150, 1:4], type = "CLASS")

## Using 'iris' dataset to determine [n.best] and [threshold] with no test set.
NNS.stack(iris[ , 1:4], iris[ , 5], type = "CLASS")

## Selecting NNS.reg and dimension reduction techniques.
NNS.stack(iris[1:140, 1:4], iris[1:140, 5], iris[141:150, 1:4], method = c(1, 2), type = "CLASS")
## End(Not run)
```

`NNS.term.matrix` *NNS Term Matrix*

Description

Generates a term matrix for text classification use in NNS.reg.

Usage

```
NNS.term.matrix(x, oos = NULL, names = FALSE)
```

Arguments

`x`	Text A two column dataset should be used. Concatenate text from original sources to comply with format. Also note the possiblity of factors in `"DV"`, so `"as.numeric(as.character(...))"` is used to avoid issues.
`oos`	Out-of-sample text dataset to be classified.
`names`	Column names for `"IV"` and `"oos"`. Defaults to FALSE.

Value

Returns the text as independent variables `"IV"` and the classification as the dependent variable `"DV"`. Out-of-sample independent variables are returned with `"OOS"`.

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
x <- data.frame(cbind(c("sunny", "rainy"), c(1, -1)))
NNS.term.matrix(x)

### Concatenate Text with space seperator, cbind with "DV"
x <- data.frame(cbind(c("sunny", "rainy"), c("windy", "cloudy"), c(1, -1)))
x <- data.frame(cbind(paste(x[ , 1], x[ , 2], sep = " "), as.numeric(as.character(x[ , 3]))))
NNS.term.matrix(x)


### NYT Example
## Not run:
require(RTextTools)
data(NYTimes)

### Concatenate Columns 3 and 4 containing text, with column 5 as DV
NYT=data.frame(cbind(paste(NYTimes[ , 3], NYTimes[ , 4], sep = " "),
                     as.numeric(as.character(NYTimes[ , 5]))))
NNS.term.matrix(NYT)
## End(Not run)
```

NNS.TSD *NNS TSD Test*

Description

Bi-directional test of third degree stochastic dominance using lower partial moments.

Usage

```
NNS.TSD(x, y)
```

Arguments

x	a numeric vector.
y	a numeric vector.

Value

Returns one of the following TSD results: `"X TSD Y"`, `"Y TSD X"`, or `"NO TSD EXISTS"`.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2016) "LPM Density Functions for the Computation of the SD Efficient Set." Journal of Mathematical Finance, 6, 105-126. https://www.scirp.org/Journal/PaperInformation.aspx?PaperID=63817.

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.TSD(x, y)
```

NNS.TSD.uni *NNS TSD Test uni-directional*

Description

Uni-directional test of third degree stochastic dominance using lower partial moments used in SD Efficient Set routine.

Usage

```
NNS.TSD.uni(x, y)
```

Arguments

x	a numeric vector.
y	a numeric vector.

Value

Returns (1) if "X TSD Y", else (0).

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2016) "LPM Density Functions for the Computation of the SD Efficient Set." Journal of Mathematical Finance, 6, 105-126. https://www.scirp.org/Journal/PaperInformation.aspx?PaperID=63817.

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100)
NNS.TSD.uni(x, y)
```

NNS.VAR *NNS VAR*

Description

Nonparametric vector autoregressive model incorporating NNS.ARMA estimates of variables into NNS.reg for a multi-variate time-series forecast.

Usage

```
NNS.VAR(
  variables,
  h,
  tau = 1,
  dim.red.method = "cor",
  obj.fn = expression(sum((predicted - actual)^2)),
  objective = "min",
  status = TRUE,
  ncores = NULL
)
```

Arguments

`variables`	a numeric matrix or data.frame of contemporaneous time-series to forecast.
`h`	integer; 1 (default) Number of periods to forecast.
`tau`	positive integer [> 0]; 1 (default) Number of lagged observations to consider for the time-series data. Vector for single lag for each respective variable or list for multiple lags per each variable.
`dim.red.method`	options: ("cor", "NNS.dep", "NNS.caus", "all") method for reducing regressors via NNS.stack. `(dim.red.method = "cor")` (default) uses standard linear correlation for dimension reduction in the lagged variable matrix. `(dim.red.method = "NNS.dep")` uses NNS.dep for nonlinear dependence weights, while `(dim.red.method = "NNS.caus")` uses NNS.caus for causal weights. `(dim.red.method = "all")` averages all methods for further feature engineering.
`obj.fn`	expression; `expression(sum((predicted -actual)^2))` (default) Sum of squared errors is the default objective function. Any `expression()` using the specific terms `predicted` and `actual` can be used.
`objective`	options: ("min", "max") `"min"` (default) Select whether to minimize or maximize the objective function `obj.fn`.
`status`	logical; `TRUE` (default) Prints status update message in console.
`ncores`	integer; value specifying the number of cores to be used in the parallelized subroutine NNS.ARMA.optim. If NULL (default), the number of cores to be used is equal to the number of cores of the machine - 1.

Value

Returns the following matrices of forecasted variables:

- "interpolated_and_extrapolated" Returns a data.table of the linear interpolated and NNS.ARMA extrapolated values to replace NA values in the original variables argument. This is required for working with variables containing different frequencies, e.g. where NA would be reported for intra-quarterly data when indexed with monthly periods.
- "relevant_variables" Returns the relevant variables from the dimension reduction step.
- "univariate" Returns the univariate NNS.ARMA forecasts.
- "multivariate" Returns the multi-variate NNS.reg forecasts.
- "ensemble" Returns the ensemble of both "univariate" and "multivariate" forecasts.

Note

- dim.red.method = "cor" is significantly faster than the other methods, but comes at the expense of ignoring possible nonlinear relationships between lagged variables.
- Not recommended for factor variables, even after transformed to numeric. NNS.reg is better suited for factor or binary regressor extrapolation.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Viole, F. (2019) "Multi-variate Time-Series Forecasting: Nonparametric Vector Autoregression Using NNS" https://www.ssrn.com/abstract=3489550

Viole, F. (2020) "NOWCASTING with NNS" https://www.ssrn.com/abstract=3586658

Viole, F. (2019) "Forecasting Using NNS" https://www.ssrn.com/abstract=3382300

Vinod, H. and Viole, F. (2017) "Nonparametric Regression Using Clusters" https://link.springer.com/article/10.1007/s10614-017-9713-5

Vinod, H. and Viole, F. (2018) "Clustering and Curve Fitting by Line Segments" https://www.preprints.org/manuscript/201801.0090/v1

Examples

```
## Not run:
####################################################
### Standard Nonparametric Vector Autoregression ###
####################################################

set.seed(123)
x <- rnorm(100) ; y <- rnorm(100) ; z <- rnorm(100)
A <- cbind(x = x, y = y, z = z)
```

```
### Using lags 1:4 for each variable
NNS.VAR(A, h = 12, tau = 4, status = TRUE)

### Using lag 1 for variable 1, lag 3 for variable 2 and lag 3 for variable 3
NNS.VAR(A, h = 12, tau = c(1,3,3), status = TRUE)

### Using lags c(1,2,3) for variables 1 and 3, while using lags c(4,5,6) for variable 2
NNS.VAR(A, h = 12, tau = list(c(1,2,3), c(4,5,6), c(1,2,3)), status = TRUE)


#########################################
### NOWCASTING with Mixed Frequencies ###
#########################################

library(Quandl)
econ_variables <- Quandl(c("FRED/GDPC1", "FRED/UNRATE", "FRED/CPIAUCSL"),type = 'ts',
                         order = "asc", collapse = "monthly", start_date="2000-01-01")

### Note the missing values that need to be imputed
head(econ_variables)
tail(econ_variables)


NNS.VAR(econ_variables, h = 12, tau = 12, status = TRUE)

## End(Not run)
```

PM.matrix *Partial Moment Matrix*

Description

This function generates a co-partial moment matrix for the specified co-partial moment.

Usage

```
PM.matrix(LPM.degree, UPM.degree, target = "mean", variable, pop.adj = FALSE)
```

Arguments

`LPM.degree`	integer; Degree for `variable` below `target` deviations. `(degree = 0)` is frequency, `(degree = 1)` is area.
`UPM.degree`	integer; Degree for `variable` above `target` deviations. `(degree = 0)` is frequency, `(degree = 1)` is area.
`target`	numeric; Typically the mean of Variable X for classical statistics equivalences, but does not have to be. (Vectorized) `(target = "mean")` (default) will set the target as the mean of every variable.

`variable` a numeric matrix or data.frame.

`pop.adj` logical; `FALSE` (default) Adjusts the sample co-partial moment matrices for population statistics.

Value

Matrix of partial moment quadrant values (CUPM, DUPM, DLPM, CLPM), and overall covariance matrix. Uncalled quadrants will return a matrix of zeros.

Note

For divergent asymmetical `"D.LPM"` and `"D.UPM"` matrices, matrix is `D.LPM(column,row,...)`.

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Viole, F. (2017) "Bayes' Theorem From Partial Moments" https://www.ssrn.com/abstract=3457377

Examples

```
set.seed(123)
x <- rnorm(100) ; y <- rnorm(100) ; z <- rnorm(100)
A <- cbind(x,y,z)
PM.matrix(LPM.degree = 1, UPM.degree = 1, target = "mean", variable = A)

## Use of vectorized numeric targets (target_x, target_y, target_z)
PM.matrix(LPM.degree = 1, UPM.degree = 1, target = c(0, 0.15, .25), variable = A)

## Calling Individual Partial Moment Quadrants
cov.mtx <- PM.matrix(LPM.degree = 1, UPM.degree = 1, target = "mean", variable = A)
cov.mtx$cupm

## Full covariance matrix
cov.mtx$cov.matrix
```

`UPM` *Upper Partial Moment*

Description

This function generates a univariate upper partial moment for any degree or target.

Usage

```
UPM(degree, target, variable)
```

Arguments

degree	integer; (degree = 0) is frequency, (degree = 1) is area.
target	numeric; Typically set to mean, but does not have to be. (Vectorized)
variable	a numeric vector.

Value

UPM of variable

Author(s)

Fred Viole, OVVO Financial Systems

References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100)
UPM(0, mean(x), x)
```

UPM.ratio *Upper Partial Moment RATIO*

Description

This function generates a standardized univariate upper partial moment for any degree or target.

Usage

```
UPM.ratio(degree, target, variable)
```

Arguments

degree	integer; (degree = 0) is frequency, (degree = 1) is area.
target	numeric; Typically set to mean, but does not have to be. (Vectorized)
variable	a numeric vector.

Value

Standardized UPM of variable

Author(s)

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References

Viole, F. and Nawrocki, D. (2013) "Nonlinear Nonparametric Statistics: Using Partial Moments" https://www.amazon.com/dp/1490523995/ref=cm_sw_su_dp

Examples

```
set.seed(123)
x <- rnorm(100)
UPM.ratio(0, mean(x), x)

## Joint Upper CDF
## Not run:
x <- rnorm(5000) ; y <- rnorm(5000)
plot3d(x, y, Co.UPM(0, 0, sort(x), sort(y), x, y), col = "blue", xlab = "X", ylab = "Y",
zlab = "Probability", box = FALSE)

## End(Not run)
```

UPM.VaR *UPM VaR*

Description

Generates an upside value at risk (VaR) quantile based on the Upper Partial Moment ratio

Usage

```
UPM.VaR(percentile, degree, x)
```

Arguments

percentile	numeric [0, 1]; The percentile for right-tail VaR (vectorized).
degree	integer; `(degree = 0)` for discrete distributions, `(degree = 1)` for continuous distributions.
x	a numeric vector.

Value

Returns a numeric value representing the point at which `"percentile"` of the area of `x` is above.

Examples

```
set.seed(123)
x <- rnorm(100)

## For 5% quantile, right-tail
UPM.VaR(0.05, 0, x)
```

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