

Package ‘PredictorSelect’

July 21, 2025

Type Package

Title Out-of-Sample Predictability in Predictive Regressions with Many Predictor Candidates

Version 0.1.0

Description Consider a linear predictive regression setting with a potentially large set of candidate predictors. This work is concerned with detecting the presence of out of sample predictability based on out of sample mean squared error comparisons given in Gonzalo and Pitarakis (2023) <[doi:10.1016/j.ijforecast.2023.10.005](https://doi.org/10.1016/j.ijforecast.2023.10.005)>.

License GPL

Encoding UTF-8

RoxygenNote 7.3.1

NeedsCompilation no

Author Rong Peng [aut, cre]

Maintainer Rong Peng <r.peng@soton.ac.uk>

Repository CRAN

Date/Publication 2024-06-10 17:00:11 UTC

Contents

DMBAR_Test	1
recursive_hstep_fast	2

Index	4
--------------	----------

DMBAR_Test	<i>Constructs the DMBAR Test statistic in GP2023</i>
------------	--

Description

Consider a linear predictive regression setting with a potentially large set of candidate predictors. This work is concerned with detecting the presence of out of sample predictability based on out of sample MSE comparisons. For details of the test, please refer to Gonzalo and Pitarakis (2023).

Usage

```
DMBAR_Test(ehat0, ehatj, mu0, pvcutoffk)
```

Arguments

ehat0	n by 1 vector of out of sample forecast errors from benchmark model with only intercept.
ehatj	n by j vector of out of sample forecast errors from models (j=1,...,p) estimated with one predictor (j) per time.
mu0	sample split parameter (must be different from 0.5).
pvcutoffk	pvalue cutoff used to decide whether the global null is rejected when identifying the key player conditional on rejecting the global null.

Value

A list of Test statistic, pvalue and key player across 4 alternative formulations of the test statistics (Irvr under 0 vs 1; power enhanced vs non-power enhanced (notation: 0, 1, 0adj, 1adj).

References

Gonzalo, J., & Pitarakis, J. Y. (2023). Out-of-sample predictability in predictive regressions with many predictor candidates. *International Journal of Forecasting*, 1166-1178.

Examples

```
ehat0<- rnorm(15);
ehatj<- rnorm(15);
temp <- DMBAR_Test(ehat0,ehatj,mu0=0.4,pvcutoffk=0.1);
```

recursive_hstep_fast *Forecasting h-steps ahead using Recursive Least Squares Fast*

Description

Consider the following LS-fitted Model with intercept: $y_{(t+h)} = \beta_0 + x_{(jt)} * \beta + u_{(t+h)}$ which is used to generate out-of-sample forecasts of y, h-steps ahead (h=1,2,3,...). Notes: (1) first estimation window is (1,...,k0) and last window is (1,...,n-h) for $k0 = \text{round}(n*\pi_0)$. First forecast is $\hat{y}_{(k0+h|k0)}$ and last forecast is $\hat{y}_{(n|n-h)}$. There are a total of (n-h-k0+1) forecasts and corresponding forecast errors. (2) this fast version of the recursive least squares algorithm uses the Sherman-Morrison matrix formula to avoid matrix inversions at each recursion. (3) $x_{(jt)}$ is the jth predictor in x (jth column).

Usage

```
recursive_hstep_fast(y, x, pi0, h)
```

Arguments

<i>y</i>	an outcome series, which should be numeric and one dimensional.
<i>x</i>	a predictor matrix (intercept would be added automatically).
<i>pi0</i>	Fraction of the sample, which should be within 0 and 1.
<i>h</i>	Number of steps ahead to predict, which should be a positive integer.

Details

recursive_hstep_fast is the fast version that avoids the recursive calculation of inverse of the matrix using Sherman-Morrison formula.

Value

Series of residuals estimated

Examples

```
x<- rnorm(15);  
y<- x+rnorm(15);  
temp1 <- recursive_hstep_fast(y,x,pi0=0.5,h=1);
```

Index

DMBAR_Test, [1](#)

recursive_hstep_fast, [2](#)