Package: pEPA (via r-universe)

October 27, 2026

Type Package

Title Tests of Equal Predictive Accuracy for Panels of Forecasts

Version 1.2

Depends R (>= 2.10)

Imports Rcpp

LinkingTo Rcpp, RcppArmadillo

Date 2025-03-07

Author Krzysztof Drachal [aut, cre] (Faculty of Economic Sciences, University of Warsaw, Poland)

Maintainer Krzysztof Drachal <kdrachal@wne.uw.edu.pl>

Description Allows to perform the tests of equal predictive accuracy for panels of forecasts. Main references: Qu et al. (2024) <doi:10.1016/j.ijforecast.2023.08.001> and Akgun et al. (2024) <doi:10.1016/j.ijforecast.2023.02.001>.

License GPL-3

LazyData TRUE

URL https://CRAN.R-project.org/package=pEPA

Note Research funded by the grant of the National Science Centre, Poland, under the contract number DEC-2018/31/B/HS4/02021.

NeedsCompilation yes

Repository https://cran.r-universe.dev

Date/Publication 2025-03-07 20:30:01 UTC

RemoteUrl https://github.com/cran/pEPA

RemoteRef HEAD

RemoteSha bf5f8ab2e1c37fcabe5750965da3e6cfc65c129e

Contents

csc.C1.test . . . 2
csc.C3.test . . . 3
csc.test . . . 5
observed . . . 6
pool_av.S1.test . . . 7
pool_av.S3.test . . . 8
pool_av.test . . . 9
predicted . . . 11
tc.test . . . 12

Index **14**

`csc.C1.test` *Computes Test for Cross-Sectional Clusters.*

Description

This function computes test of the equal predictive accuracy for cross-sectional clusters. It corresponds to $C_{nT}^{(1)}$ statistic in the referenced paper by Akgun et al. (2024). The null hypothesis of this test is that a pair of forecasts have the same expected accuracy among cross-sectional clusters. However, their predictive accuracy can be different across the clusters, but the same among each cluster. The test is suitable for situations with cross-sectional independence.

Usage

```
csc.C1.test(evaluated1,evaluated2,realized,loss.type="SE",cl)
```

Arguments

`evaluated1`	same as in `pool_av.test`, but cross-sections are ordered rowwise
`evaluated2`	same as in `pool_av.test`, but cross-sections are ordered rowwise
`realized`	same as in `pool_av.test`, but cross-sections are ordered rowwise
`loss.type`	same as in `pool_av.test`
`cl`	`vector` of the beginning indices of rows for each pre-defined clusters – as a result always `cl[1]=1`

Value

class `htest` object, `list` of

`statistic`	test statistic
`parameter`	K, number of cross-sectional clusters
`alternative`	alternative hypothesis of the test
`p.value`	p-value
`method`	name of the test
`data.name`	names of the tested data

References

Akgun, O., Pirotte, A., Urga, G., Yang, Z. 2024. Equal predictive ability tests based on panel data with applications to OECD and IMF forecasts. *International Journal of Forecasting* **40**, 202–228.

See Also

pool_av.test, csc.C3.test

Examples

```
data(forecasts)
y <- t(observed)
# just to save time
y <- y[,1:40]
f.bsr <- matrix(NA,ncol=ncol(y),nrow=56)
f.dma <- f.bsr
# extract prices predicted by BSR rec and DMA methods
for (i in 1:56)
  {
    f.bsr[i,] <- predicted[[i]][1:40,1]
    f.dma[i,] <- predicted[[i]][1:40,9]
  }
# 2 cross-sectional clusters: energy commodities and non-energy commodities
cs.cl <- c(1,9)
t <- csc.C1.test(evaluated1=f.bsr,evaluated2=f.dma,realized=y,loss.type="SE",cl=cs.cl)
```

---

csc.C3.test *Computes Test for Cross-Sectional Clusters.*

Description

This function computes test of the equal predictive accuracy for cross-sectional clusters. It corresponds to $C_{nT}^{(3)}$ statistic in the referenced paper by Akgun et al. (2024). The null hypothesis of this test is that a pair of forecasts have the same expected accuracy among cross-sectional clusters. However, their predictive accuracy can be different across the clusters, but the same among each cluster. The test allows for strong cross-sectional dependence.

Usage

```
csc.C3.test(evaluated1,evaluated2,realized,loss.type="SE",cl)
```

Arguments

evaluated1	same as in pool_av.test, but cross-sections are ordered rowwise
evaluated2	same as in pool_av.test, but cross-sections are ordered rowwise
realized	same as in pool_av.test, but cross-sections are ordered rowwise

loss.type	same as in pool_av.test
cl	vector of the beginning indices of rows for each pre-defined clusters – as a result always cl[1]=1

Value

class htest object, list of

statistic	test statistic
parameter	K, number of cross-sectional clusters
alternative	alternative hypothesis of the test
p.value	p-value
method	name of the test
data.name	names of the tested data

References

Akgun, O., Pirotte, A., Urga, G., Yang, Z. 2024. Equal predictive ability tests based on panel data with applications to OECD and IMF forecasts. *International Journal of Forecasting* **40**, 202–228.

See Also

pool_av.test, csc.C1.test

Examples

```
data(forecasts)
y <- t(observed)
# just to reduce computation time restrict to energy commodities only
y <- y[1:8,]
f.bsr <- matrix(NA,ncol=ncol(y),nrow=8)
f.dma <- f.bsr
# extract prices predicted by BSR rec and DMA methods
for (i in 1:8)
  {
    f.bsr[i,] <- predicted[[i]][,1]
    f.dma[i,] <- predicted[[i]][,9]
  }
# 2 cross-sectional clusters: crude oil and other energy commodities
cs.cl <- c(1,4)
t <- csc.C3.test(evaluated1=f.bsr,evaluated2=f.dma,realized=y,loss.type="SE",cl=cs.cl)
```

csc.test *Computes Test for Cross-Sectional Clusters.*

Description

This function computes test of the equal predictive accuracy for cross-sectional clusters. The null hypothesis of this test is that a pair of forecasts have the same expected accuracy among cross-sectional clusters. However, their predictive accuracy can be different across the clusters, but the same among each cluster. The test is suitable if either: $K \geq 2$ and `significance level` ≤ 0.08326, or $2 \leq K \leq 14$ and `significance level` ≤ 0.1, or $K = \{2, 3\}$ and `significance level` ≤ 0.2, where K denotes the number of time clusters.

Usage

```
csc.test(evaluated1,evaluated2,realized,loss.type="SE",cl,dc=FALSE)
```

Arguments

`evaluated1`	same as in `pool_av.test`, but cross-sections are ordered rowwise
`evaluated2`	same as in `pool_av.test`, but cross-sections are ordered rowwise
`realized`	same as in `pool_av.test`, but cross-sections are ordered rowwise
`loss.type`	same as in `pool_av.test`
`cl`	`vector` of the beginning indices of rows for each pre-defined clusters – as a result always `cl[1]=1`
`dc`	`logical` indicating if apply decorrelating clusters, if not specified `dc=FALSE` is used

Value

class `htest` object, `list` of

`statistic`	test statistic
`parameter`	K, number of cross-sectional clusters
`alternative`	alternative hypothesis of the test
`p.value`	p-value
`method`	name of the test
`data.name`	names of the tested data

References

Qu, R., Timmermann, A., Zhu, Y. 2024. Comparing forecasting performance with panel data. *International Journal of Forecasting* **40**, 918–941.

See Also

`pool_av.test`

Examples

```
data(forecasts)
y <- t(observed)
f.bsr <- matrix(NA,ncol=ncol(y),nrow=56)
f.dma <- f.bsr
# extract prices predicted by BSR rec and DMA methods
for (i in 1:56)
  {
    f.bsr[i,] <- predicted[[i]][,1]
    f.dma[i,] <- predicted[[i]][,9]
  }
# 2 cross-sectional clusters: energy commodities and non-energy commodities
cs.cl <- c(1,9)
t <- csc.test(evaluated1=f.bsr,evaluated2=f.dma,realized=y,loss.type="SE",cl=cs.cl)
```

---

observed *Sample Panel of Commodities Spot Prices.*

Description

Observed spot prices of various commodities.

Usage

```
data(forecasts)
```

Format

observed is matrix object such that its columns correspond to spot prices of selected 56 commodities.

Details

They cover the period between 1996 and 2021, and are in monthly freqency. Variables names are the same as in the paper by Drachal and Pawłowski (2024). The observed prices were taken from The World Bank (2022).

References

Drachal, K., Pawłowski, M. 2024. Forecasting selected commodities' prices with the Bayesian symbolic regression. *International Journal of Financial Studies* **12**, 34, doi:10.3390/ijfs12020034

The World Bank. 2022. Commodity Markets. https://www.worldbank.org/en/research/commodity-markets

See Also

predicted

Examples

```
data(forecasts)
# WTI prices
t1 <- observed[,3]
```

`pool_av.S1.test` *Computes Test for Overall Equal Predictive Ability.*

Description

This function computes test of the equal predictive accuracy for the pooled average. It corresponds to $S_{nT}^{(1)}$ statistic in the referenced paper by Akgun et al. (2024). The null hypothesis of this test is that the pooled average loss is equal in expectation for a pair of forecasts from both considered methods. The alternative one is that the differences do not average out across the cross-sectional and time-series dimensions. The test is suitable for situations with cross-sectional independence.

Usage

```
pool_av.S1.test(evaluated1,evaluated2,realized,loss.type="SE")
```

Arguments

`evaluated1`	same as in `pool_av.test`
`evaluated2`	same as in `pool_av.test`
`realized`	same as in `pool_av.test`
`loss.type`	same as in `pool_av.test`

Value

class `htest` object, `list` of

`statistic`	test statistic
`alternative`	alternative hypothesis of the test
`p.value`	p-value
`method`	name of the test
`data.name`	names of the tested data

References

Akgun, O., Pirotte, A., Urga, G., Yang, Z. 2024. Equal predictive ability tests based on panel data with applications to OECD and IMF forecasts. *International Journal of Forecasting* **40**, 202–228.

See Also

`pool_av.test`, `pool_av.S3.test`

Examples

```
data(forecasts)
y <- t(observed)
f.bsr <- matrix(NA,ncol=ncol(y),nrow=56)
f.dma <- f.bsr
# extract prices predicted by BSR rec and DMA methods
for (i in 1:56)
  {
    f.bsr[i,] <- predicted[[i]][,1]
    f.dma[i,] <- predicted[[i]][,9]
  }
t <- pool_av.S1.test(evaluated1=f.bsr,evaluated2=f.dma,realized=y,loss.type="SE")
```

`pool_av.S3.test` *Computes Test for Overall Equal Predictive Ability.*

Description

This function computes test of the equal predictive accuracy for the pooled average. It corresponds to $S_{nT}^{(3)}$ statistic in the referenced paper by Akgun et al. (2024). The null hypothesis of this test is that the pooled average loss is equal in expectation for a pair of forecasts from both considered methods. The alternative one is that the differences do not average out across the cross-sectional and time-series dimensions. The test allows for strong cross-sectional dependence.

Usage

```
pool_av.S3.test(evaluated1,evaluated2,realized,loss.type="SE")
```

Arguments

`evaluated1`	same as in `pool_av.test`
`evaluated2`	same as in `pool_av.test`
`realized`	same as in `pool_av.test`
`loss.type`	same as in `pool_av.test`

Value

class `htest` object, `list` of

`statistic`	test statistic
`alternative`	alternative hypothesis of the test
`p.value`	p-value
`method`	name of the test
`data.name`	names of the tested data

References

Akgun, O., Pirotte, A., Urga, G., Yang, Z. 2024. Equal predictive ability tests based on panel data with applications to OECD and IMF forecasts. *International Journal of Forecasting* **40**, 202–228.

See Also

pool_av.test, pool_av.S1.test

Examples

```
data(forecasts)
y <- t(observed)
# just to reduce computation time shorten time-series
y <- y[,1:40]
f.bsr <- matrix(NA,ncol=ncol(y),nrow=56)
f.dma <- f.bsr
# extract prices predicted by BSR rec and DMA methods
for (i in 1:56)
  {
    f.bsr[i,] <- predicted[[i]][1:40,1]
    f.dma[i,] <- predicted[[i]][1:40,9]
  }
t <- pool_av.S3.test(evaluated1=f.bsr,evaluated2=f.dma,realized=y,loss.type="SE")
```

---

pool_av.test *Computes Test for the Pooled Average.*

Description

This function computes test of the equal predictive accuracy for the pooled average. The null hypothesis of this test is that the pooled average loss is equal in expectation for a pair of forecasts from both considered methods. The alternative hypothesis can be formulated as the differences do not average out across the cross-sectional and time-series dimensions.

Usage

```
pool_av.test(evaluated1,evaluated2,realized,loss.type="SE",J=NULL)
```

Arguments

evaluated1	matrix of forecasts from the first method, cross-sections are ordered by rows, and time by columns
evaluated2	matrix of forecasts from the second method, cross-sections are ordered by rows, and time by columns
realized	matrix of the observed values, cross-sections are ordered by rows, and time by columns

loss.type	a method to compute the loss function, loss.type="SE" applies squared errors, loss.type="AE" – absolute errors, loss.type="SPE" – squared proportional error (useful if errors are heteroskedastic), loss.type="ASE" – absolute scaled error, if loss.type is specified as some numeric, then the function of type exp(loss.type*errors)-1-loss.type*errors is applied (useful when it is more costly to underpredict realized than to overpredict), if not specified loss.type="SE" is used
J	numeric maximum lag length, if not specified J=round(T^(1/3)) is used, where T=ncol(realized)

Value

class htest object, list of

statistic	test statistic
parameter	J, maximum lag length
alternative	alternative hypothesis of the test
p.value	p-value
method	name of the test
data.name	names of the tested data

References

Hyndman, R.J., Koehler, A.B. 2006. Another look at measures of forecast accuracy. *International Journal of Forecasting* **22**, 679–688.

Qu, R., Timmermann, A., Zhu, Y. 2024. Comparing forecasting performance with panel data. *International Journal of Forecasting* **40**, 918–941.

Taylor, S. J., 2005. *Asset Price Dynamics, Volatility, and Prediction*, Princeton University Press.

Triacca, U., 2024. *Comparing Predictive Accuracy of Two Forecasts*, https://www.lem.sssup.it/phd/documents/Lesson19.pdf.

Examples

```
data(forecasts)
y <- t(observed)
f.bsr <- matrix(NA,ncol=ncol(y),nrow=56)
f.dma <- f.bsr
# extract prices predicted by BSR rec and DMA methods
for (i in 1:56)
  {
    f.bsr[i,] <- predicted[[i]][,1]
    f.dma[i,] <- predicted[[i]][,9]
  }
t <- pool_av.test(evaluated1=f.bsr,evaluated2=f.dma,realized=y,loss.type="SE")
```

predicted *Sample Panels of Commodities Spot Prices Forecasts.*

Description

Forecasts obtained from various methods applied to various commodities prices.

Usage

```
data(forecasts)
```

Format

`predicted` is `list` of forecasts of spot prices of selected 56 commodities. For each commodity `matrix` of forecasts generated by various methods is provided. Columns correspond to various methods.

Details

The forecasts were taken from Drachal and Pawłowski (2024). They cover the period between 1996 and 2021, and are in monthly freqency. Variables and methods names are the same as in that paper, where they are described in details.

References

Drachal, K., Pawłowski, M. 2024. Forecasting selected commodities' prices with the Bayesian symbolic regression. *International Journal of Financial Studies* **12**, 34, doi:10.3390/ijfs12020034

See Also

`observed`

Examples

```
data(forecasts)
# WTI prices predicted by BSR rec method
t2 <- predicted[[3]][,1]
```

`tc.test` *Computes Test for Time Clusters.*

Description

This function computes test of the equal predictive accuracy for time clusters. The null hypothesis of this test is that the equal predictive accuracy for the two methods holds within each of the time clusters. The test is suitable if either: $K \geq 2$ and `significance level` ≤ 0.08326, or $2 \leq K \leq 14$ and `significance level` ≤ 0.1, or $K = \{2, 3\}$ and `significance level` ≤ 0.2, where K denotes the number of time clusters.

Usage

```
tc.test(evaluated1,evaluated2,realized,loss.type="SE",cl)
```

Arguments

`evaluated1`	same as in `pool_av.test`
`evaluated2`	same as in `pool_av.test`
`realized`	same as in `pool_av.test`
`loss.type`	same as in `pool_av.test`
`cl`	`vector` of the beginning indices of each pre-defined blocks of time – as a result always `cl[1]=1`

Value

class `htest` object, `list` of

`statistic`	test statistic
`parameter`	K, number of time clusters
`alternative`	alternative hypothesis of the test
`p.value`	p-value
`method`	name of the test
`data.name`	names of the tested data

References

Qu, R., Timmermann, A., Zhu, Y. 2024. Comparing forecasting performance with panel data. *International Journal of Forecasting* **40**, 918–941.

See Also

`pool_av.test`

Examples

```
data(forecasts)
y <- t(observed)
f.bsr <- matrix(NA,ncol=ncol(y),nrow=56)
f.dma <- f.bsr
# extract prices predicted by BSR rec and DMA methods
for (i in 1:56)
  {
    f.bsr[i,] <- predicted[[i]][,1]
    f.dma[i,] <- predicted[[i]][,9]
  }
# 3 time clusters: Jun 1996 -- Nov 2007, Dec 2007 -- Jun 2009, Jul 2009 - Aug 2021
# rownames(observed)[1]
# rownames(observed)[139]
# rownames(observed)[158]
t.cl <- c(1,139,158)
t <- tc.test(evaluated1=f.bsr,evaluated2=f.dma,realized=y,loss.type="SE",cl=t.cl)
```

Index

csc.C1.test, 2, *4*
csc.C3.test, 3, *3*
csc.test, 5

list, *2*, *4*, *5*, *7*, *8*, *10–12*
logical, *5*

matrix, *6*, *9*, *11*

numeric, *10*

observed, 6, *11*

pool_av.S1.test, 7, *9*
pool_av.S3.test, *7*, 8
pool_av.test, *2–5*, *7*, *8*, 9, *9*, *12*
predicted, *6*, 11

tc.test, 12

vector, *2*, *4*, *5*, *12*