

Package ‘pEPA’

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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>.

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Contents

csc.C1.test	2
csc.C3.test	3
csc.test	4
observed	6
pool_av.S1.test	7
pool_av.S3.test	8

pool_av.test	9
predicted	11
tc.test	12

Index	14
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csc.C1.test	<i>Computes Test for Cross-Sectional Clusters.</i>
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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",c1)
```

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
c1	vector of the beginning indices of rows for each pre-defined clusters – as a result always <code>c1[1]=1</code>

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 <code>cl[1]=1</code>

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 <code>cl[1]=1</code>
dc	logical indicating if apply decorrelating clusters, if not specified <code>dc=FALSE</code> 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 frequency. 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](https://doi.org/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

<code>evaluated1</code>	same as in pool_av.test
<code>evaluated2</code>	same as in pool_av.test
<code>realized</code>	same as in pool_av.test
<code>loss.type</code>	same as in pool_av.test

Value

class `htest` object, [list](#) of

<code>statistic</code>	test statistic
<code>alternative</code>	alternative hypothesis of the test
<code>p.value</code>	p-value
<code>method</code>	name of the test
<code>data.name</code>	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 $\exp(\text{loss.type} \times \text{errors}) - 1 - \text{loss.type} \times \text{errors}$ 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 $J = \text{round}(T^{1/3})$ is used, where $T = \text{ncol}(\text{realized})$

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 frequency. 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](https://doi.org/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",c1)
```

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
c1	vector of the beginning indices of each pre-defined blocks of time – as a result always <code>c1[1]=1</code>

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, [4](#)

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](#), [7–9](#), [9](#), [12](#)

predicted, [6](#), [11](#)

tc.test, [12](#)

vector, [2](#), [3](#), [5](#), [12](#)