

# Package ‘gammadlm’

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**Type** Package

**Title** Maximum Likelihood Estimation of the Gamma Distributed-Lag Model with Multiple Explanatory Variables

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**Description** Implementation of the hill climbing algorithm for maximum likelihood estimation of the Gamma distributed-lag model with multiple explanatory variables. A panel structure can be taken into account.

**Depends** R (>= 3.5.0)

**Imports** graphics, stats

**License** GPL-2

**NeedsCompilation** no

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gammadlm-package	<i>Maximum Likelihood Estimation of the Gamma Distributed-Lag Model with Multiple Explanatory Variables</i>
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## Description

Implementation of the hill climbing algorithm for maximum likelihood estimation of the Gamma distributed-lag model with multiple explanatory variables. A panel structure can be taken into account.

## Details

Package: gammadlm  
 Type: Package  
 Version: 0.1.1  
 Date: 2022-02-02  
 License: GPL-2

Let  $Y$  be the response variable and  $X_1, \dots, X_J$  be  $J$  explanatory variables. Also, let  $y_t$  and  $x_{j,t}$  be, respectively, the value of  $Y$  and of  $X_j$  ( $j = 1, \dots, J$ ) observed at time  $t$ . Under the assumption that the time series of  $Y$  and of  $X_1, \dots, X_J$  are all weakly stationary (i.e., expected value and autocorrelation function independent of time), the Gamma distributed-lag model explaining  $Y$  from  $X_1, \dots, X_J$  is defined as:

$$y_t = \alpha + \sum_{j=1}^J \sum_{k=0}^{\infty} \beta_{j,k}(\theta_j, \delta_j, \lambda_j, \eta_j) x_{j,t-k} + \varepsilon_t$$

$$\beta_{j,k}(\theta_j, \delta_j, \lambda_j, \eta_j) = \theta_j w_{j,k}(\delta_j, \lambda_j, \eta_j)$$

$$w_{j,k}(\delta_j, \lambda_j, \eta_j) = \frac{(k+1-\eta_j)^{\frac{\delta_j}{1-\delta_j}} \lambda_j^{k-\eta_j}}{\sum_{l=0}^{\infty} (l+1-\eta_j)^{\frac{\delta_j}{1-\delta_j}} \lambda_j^{l-\eta_j}}$$

where:

- $\alpha$  is the intercept;
- $\beta_{j,k}$  is the dynamic coefficient for  $X_j$  at time lag  $k$ , equal to the scale parameter  $\theta_j$  times the weight  $w_{j,k}$ . The set  $\{w_{j,k} : k = 0, 1, \dots, \infty\}$  includes the weights for  $X_j$  and is defined by the shape parameters  $0 \leq \delta_j < 1$  and  $0 \leq \lambda_j < 1$  and the offset  $\eta_j$  (typically set to 0). The set  $\{\beta_{j,k} : k = 0, 1, \dots, \infty\}$  is called *lag distribution* of  $X_j$ ;
- $\varepsilon_t$  is the random error at time  $t$ .

In case of a panel structure, one intercept is specified for each unit of observation, while the lag distributions are the same for all units:

$$y_{i,t} = \alpha_i + \sum_{j=1}^J \sum_{k=0}^{\infty} \beta_{j,k}(\theta_j, \delta_j, \lambda_j, \eta_j) x_{i,j,t-k} + \varepsilon_{i,t}$$

where:

- $y_{i,t}$  and  $x_{i,j,t}$  are, respectively, the value of  $Y$  and of  $X_j$  ( $j = 1, \dots, J$ ) observed on unit  $i$  at time  $t$ ;
- $\alpha_i$  is the intercept for unit  $i$ ;
- $\varepsilon_{i,t}$  is the random error for unit  $i$  at time  $t$ .

The main functions of the package are:

- [unirootTest](#), to check stationarity;
- [preProcess](#), to apply the Box-Cox transformation and differencing in order to achieve stationarity;
- [gammadlm](#), to estimate the model through the hill climbing algorithm;
- [lagCoef](#), to see the estimated dynamic coefficients;
- [plot.gammadlm](#), to display the estimated lag distributions.

**Author(s)**

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**References**

A. Magrini (2022). A hill climbing algorithm for maximum likelihood estimation of the Gamma distributed-lag model with multiple explanatory variables. *Austrian Journal of Statistics*, 51(2): 40-46.

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gammadlm	<i>Estimation of a Gamma distributed-lag model</i>
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**Description**

Maximum likelihood estimation of a Gamma distributed-lag model with multiple explanatory variables using the hill climbing algorithm. A panel structure can be taken into account.

**Usage**

```
gammadlm(y.name, x.names, z.names=NULL, unit=NULL, time=NULL, data,
         offset=rep(0,length(x.names)), add.intercept=TRUE, control=list(nstart=NULL,
         delta.lim=NULL, lambda.lim=NULL, peak.lim=NULL, length.lim=NULL), quiet=FALSE)
```

**Arguments**

y.name	Character including the name of the response variable, that must be a quantitative variable. If a vector with length greater than 1 is provided, only the first element is considered.
x.names	Character vector of length 1 or greater including the names of the explanatory variables with lags, that must be quantitative variables. If the name of the response variable is indicated in x.names, then the autoregressive lag distribution will be estimated.
z.names	Character vector including the names of the explanatory variables without lags (optional). They may be either quantitative or qualitative variables. If NULL (the default), no explanatory variable without lags is included in the model.
unit	Character containing the name of the units of observation in case of a panel structure. If a vector with length greater than 1 is provided, only the first element is considered. If NULL (the default), a single unit of observation is assumed.
time	Character containing the name of the time variable, which must be in numeric or date format. If a vector with length greater than 1 is provided, only the first element is considered. If NULL (the default), data are assumed to be temporally ordered.
data	Object of class data.frame containing the variables in the model. Variables in y.name, x.names, z.names, unit and time cannot contain missing values.
offset	Numerical vector indicating, in order, the offset for each variable in x.names. Default is 0. If the name of the response variable is indicated in x.names and its offset is less than 1, then it will be set automatically to 1.
add.intercept	Logical value indicating whether intercepts should be included in the model. Default is TRUE. See 'Details'.

control	<p>A list including control options for the hill climbing algorithm.</p> <ul style="list-style-type: none"> <li>• <code>nstart</code>: positive integer value indicating the number of restarts. If equal to 1 or NULL (the default), shape parameters are initialized for one explanatory variable at a time based on OLS-grid search.</li> <li>• <code>delta.lim</code>: a named list with one component for each variable in <code>x.names</code>, that must be either a numerical vector of length 2 indicating the minimum and the maximum value of <math>\delta</math>, or a numerical value indicating the exact value of <math>\delta</math>. If there is no component in <code>delta.lim</code> for a certain variable in <code>x.names</code>, then the theoretical range <math>[0,1)</math> is assumed for <math>\delta</math>.</li> <li>• <code>lambda.lim</code>: the same as <code>delta.lim</code>, but it is about <math>\lambda</math> parameters.</li> <li>• <code>peak.lim</code>: the same as <code>delta.lim</code>, but it is about the peak of the lag distributions.</li> <li>• <code>length.lim</code>: the same as <code>delta.lim</code>, but it is about the 99.9th percentile of the lag distributions.</li> </ul>
quiet	<p>Logical value indicating whether prompt messages should be suppressed. Default is FALSE.</p>

## Details

The intercept captures the effect of time-invariant factors. When differencing is applied to the response variable, the effect of time-invariant factors is canceled out, thus the intercept represents the coefficient of a linear trend (drift). To avoid the presence of a drift, argument `add.intercept` should be set to FALSE.

All S3 methods for class `lm` are also available for class `gammadlm`. Notably:

- `plot` displays the estimated lag distributions;
- `summary` provides a summary of parameter estimation;
- `residuals` returns the residuals;
- `fitted.values` returns the fitted values;
- `predict` performs  $h$  step ahead forecast.

Furthermore, method `lagCoef` can be used to see the estimated dynamic coefficients.

## Value

An object of class `lm` and `gammadlm`, including all the components of an object of class `lm` plus the following components:

- `offset`: vector including the offset of the lag distributions;
- `add.intercept`: logical value indicating whether the model includes the intercept;
- `par`: matrix including the shape parameters for each variable in `x.names` (by column);
- `variables`: list including the names of the variables provided to arguments `y.name`, `x.names`, `z.names`, `unit` and `time`.
- `unit.id`: list including the row names of the observations for each unit. NULL if `unit` is NULL.
- `data`: `data.frame` including the data used for parameter estimation.
- `local.max`: list including all the models fitted at each restart.

## Note

Weak stationarity of all time series (expected value and autocorrelation function independent of time) is a basic assumption of the model. However, in order to get valid results, it is sufficient that no time series contains unit roots (Granger & Newbold, 1974). Before calling the function `gammadlm`, the user is strongly recommended to check the absence of unit roots in each time series through the function `unirootTest`, and to apply differencing through the function `preProcess` to the ones containing unit roots.

Function `gammadlm` automatically checks the absence of unit roots in the residuals and returns a warning in case of failure.

When the summary method is called on an object of class `gammadlm`, the order of auto-correlation of the residuals is estimated based on the Bayesian Information Criterion. If it is greater than 0, then the Heteroskedasticity and Autocorrelation Consistent (HAC, Newey & West, 1987) estimator of the covariance matrix of least squares estimates is applied to get robust standard errors. The same holds for the `confint` method.

## References

- C. W. J. Granger, and P. Newbold (1974). Spurious regressions in econometrics. *Journal of Econometrics*, 2(2): 111-120.
- A. Magrini (2022). A hill climbing algorithm for maximum likelihood estimation of the Gamma distributed-lag model with multiple explanatory variables. *Austrian Journal of Statistics*, 51(2): 40-46.
- W. K. Newey, K. D. West (1987). A simple, positive semi-definite, heteroskedasticity and autocorrelation consistent covariance matrix. *Econometrica*, 55(3): 703-708

## See Also

`unirootTest`; `lagCoef`; `plot.gammadlm`.

## Examples

```
data(USstock)
mydata <- USstock[which(USstock$Date>="2020-04-01"),]
mydataLR <- preProcess(time="Date", data=mydata, box.cox=0, ndiff=1)

# estimation with pre-specified shape parameters (simple OLS)
dval <- list(DJA=0.85,IXIC=0.75,GSPC=0.55)
lval <- list(DJA=0.05, IXIC=0.35,GSPC=0.45)
m0 <- gammadlm(y.name="BTC", x.names=c("DJA","IXIC","GSPC"), data=mydataLR,
  control=list(delta.lim=dval, lambda.lim=lval))
summary(m0) ## summary of estimation

# hill climbing algorithm without random restarts
# (initialization based on independent OLS-grid searches)
m1 <- gammadlm(y.name="BTC", x.names=c("DJA","IXIC","GSPC"), data=mydataLR)
summary(m1)

# add constraints: peak>=1 and 3<=length<=15
pklim <- list(DJA=c(1,Inf),IXIC=c(1,Inf),GSPC=c(1,Inf))
lenlim <- list(DJA=c(3,15),IXIC=c(3,15),GSPC=c(3,15))
m2 <- gammadlm(y.name="BTC", x.names=c("DJA","IXIC","GSPC"), data=mydataLR,
  control=list(peak.lim=pklim, length.lim=lenlim))
summary(m2)
```

```
## 50 random restarts without constraints: NOT RUN
#set.seed(100)
#m3 <- gammadlm(y.name="BTC", x.names=c("DJA","IXIC","GSPC"), data=mydataLR,
# control=list(nstart=50))
#summary(m3)
##
## 50 random restarts with constraints: NOT RUN
#set.seed(100)
#m4 <- gammadlm(y.name="BTC", x.names=c("DJA","IXIC","GSPC"), data=mydataLR,
# control=list(peak.lim=pklim, length.lim=lenlim, nstart=50))
#summary(m4) ## same as m2
```

---

gammaWeights

*Functionalities for the Gamma lag distribution*


---

## Description

Obtain weights, quantiles and kernel projection for the desired Gamma lag distribution.

## Usage

```
gammaWeights(k, par, offset=0, normalize=TRUE)
gammaQuantile(prob, par, offset=0)
gammaKernel(x, par, unit=NULL, offset=0, normalize=TRUE)
```

## Arguments

k	Numerical vector indicating the lags for which the weights should be computed.
prob	Numerical vector indicating the order of the quantiles to be computed.
x	Numerical vector representing temporally ordered data for which the kernel projection should be returned.
par	Numerical vector of length 2 representing the shape parameters of the Gamma lag distribution.
unit	Character containing the name of the units of observation in case of a panel structure. If a vector with length greater than 1 is provided, only the first element is considered. If NULL (the default), a single unit of observation is assumed.
offset	Numerical value representing the offset of the Gamma lag distribution. Default is 0.
normalize	Logical value indicating whether the weights should be normalized to have sum 1. Default is TRUE.

## Details

Function gammaWeights provides the weights, function gammaQuantile computes the quantiles, and function gammaKernel returns the kernel projection.

## Examples

```
## examples for a Gamma lag distribution with delta=0.6 and lambda=0.3

# weights
gammaWeights(0:12, par=c(0.6,0.3)) ## at lags from 0 to 12
gammaWeights(10, par=c(0.6,0.3))   ## at lag 10

# quantiles
gammaQuantile(0.5, par=c(0.6,0.3)) ## median
gammaQuantile(0.95, par=c(0.6,0.3)) ## 95th percentile
gammaQuantile(0.99, par=c(0.6,0.3)) ## 99th percentile

# kernel projection
set.seed(100); xval <- rnorm(10)
gammaKernel(xval, par=c(0.6,0.3))

# kernel projection under a panel structure
set.seed(100); xval <- rnorm(20)
gr <- c(rep(0,10),rep(1,10))
gammaKernel(xval, par=c(0.6,0.3), unit=gr)
```

---

lagCoef	<i>Estimated dynamic coefficients</i>
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## Description

See the estimated dynamic coefficients for each explanatory variable with lags.

## Usage

```
lagCoef(x, conf=0.95, cumulative=FALSE, max.lag=NULL, max.quantile=0.999)
```

## Arguments

x	An object of class <code>gammadlm</code> .
conf	Numerical value indicating the level of confidence intervals. If <code>NULL</code> , confidence intervals are not provided. Default is 0.95.
cumulative	Logical value indicating whether cumulative coefficients should be returned. Default is <code>FALSE</code> .
max.lag	Non-negative integer value indicating the lag up to which coefficients should be returned. If <code>NULL</code> (the default), it is set accordingly to argument <code>max.quantile</code> .
max.quantile	Numerical value indicating the order of the quantile lag up to which coefficients should be returned. Default is 0.999 (99.9th percentile).

## Value

A list with one component for each explanatory variable with lags. Each component is an object of class `data.frame` containing estimation, asymptotic standard error and confidence interval for each lag up to `max.lag`.

**See Also**

[gammadlm](#); [plot.gammadlm](#).

**Examples**

```
data(USstock)
mydata <- USstock[which(USstock$Date>="2020-04-01"),]
mydataLR <- preProcess(time="Date", data=mydata, box.cox=0, ndiff=1)

dval <- list(DJA=0.85,IXIC=0.75,GSPC=0.55)
lval <- list(DJA=0.05, IXIC=0.35,GSPC=0.45)
m0 <- gammadlm(y.name="BTC", x.names=c("DJA","IXIC","GSPC"), data=mydataLR,
  control=list(delta.lim=dval, lambda.lim=lval))

lagCoef(m0) ## coefficients shown up to the 99.9th percentile lag
lagCoef(m0, max.lag=11) ## coefficients shown up to lag 11
lagCoef(m0, cumulative=TRUE) ## cumulative coefficients
```

---

plot.gammadlm

*Graphics for the estimated lag distributions*

---

**Description**

Display the estimated lag distribution of each explanatory variable with lags.

**Usage**

```
## S3 method for class 'gammadlm'
plot(x, x.names=NULL, conf=0.95, max.lag=NULL, max.quantile=0.999,
  xlim=NULL, ylim=NULL, add.legend=TRUE, cex.legend=1, digits=4, grid.length=100,
  main=NULL, ylab=NULL, xlab=NULL, ...)
```

**Arguments**

x	Object of class gammadlm.
x.names	Character vector including the name of the variables for which the lag distribution should be displayed. If NULL (the default), the lag distribution of all the variables with lags will be displayed.
conf	Numerical value indicating the confidence level. Default is 0.95.
max.lag	Non-negative integer value indicating the lag up to which each lag distribution should be displayed. If NULL (the default), it is set accordingly to argument max.quantile.
max.quantile	Numerical value indicating the order of the quantile up to which each lag distribution should be displayed. Default is 0.999 (99.9th percentile).
xlim	Numerical vector of length 2 indicating the range of the x-axis, which is applied to all graphics (optional).
ylim	Numerical vector of length 2 indicating the range of the y-axis, which is applied to all graphics (optional).
add.legend	Logical value indicating whether a legend with numerical information should be added to the graphics. Default is TRUE.



cex.legend	Size of the legend. Default is 1.
digits	Integer non-negative value indicating the number of decimal places to be used in the legend. Default is 4. Ignored if add.legend=FALSE.
grid.length	Numerical value no less than 100 indicating the resolution of the interpolation. Default is 100.
main	Vector of characters including the title for each graphic. If NULL (the default), the name of the explanatory variables is used.
ylab	Text for y-axis, which is applied to all graphics (optional).
xlab	Text for x-axis, which is applied to all graphics (optional).
...	Further graphical parameters.

### See Also

[gammadlm](#).

### Examples

```
data(USstock)
mydata <- USstock[which(USstock$Date>="2020-04-01"),]
mydataLR <- preProcess(time="Date", data=mydata, box.cox=0, ndiff=1)

dval <- list(DJA=0.85,IXIC=0.75,GSPC=0.55)
lval <- list(DJA=0.05, IXIC=0.35,GSPC=0.45)
m0 <- gammadlm(y.name="BTC", x.names=c("DJA","IXIC","GSPC"), data=mydataLR,
  control=list(delta.lim=dval, lambda.lim=lval))

plot(m0) ## all the lag distributions
plot(m0, x.names=c("DJA","IXIC")) ## just the ones of 'DJA' and 'IXIC'
```

---

```
preProcess
```

---

```
Preprocessing of time series data
```

---

### Description

Application of the Box-Cox transformation and/or differencing to a multivariate time series, eventually structured as a panel.

### Usage

```
preProcess(var.names=NULL, unit=NULL, time=NULL, data, box.cox=1, ndiff=0,
  imputation=TRUE, em.control=list(nlags=NULL, tol=1e-4, maxit=1000, quiet=FALSE))
```

### Arguments

var.names	Character vector including the name of the variables to be differenced. If NULL (the default), all the quantitative variables in the dataset provided to argument data will be differenced, with the exception of the variable indicated in time.
unit	Character containing the name of the units of observation in case of a panel structure. If a vector with length greater than 1 is provided, only the first element is considered. If NULL (the default), a single unit of observation is assumed.

time	Character containing the name of the time variable, which must be in numeric or date format. If a vector with length greater than 1 is provided, only the first element is considered. If NULL (the default), data are assumed to be temporally ordered.
data	Object of class <code>data.frame</code> containing the variables in <code>var.names</code> , <code>unit</code> and <code>time</code> .
box.cox	Named vector indicating the parameter of the Box-Cox transformation (Box & Cox, 1964) for variables in <code>x.names</code> . If <code>box.cox</code> is of length one and has no names, the same parameter is used for all variables in <code>x.names</code> . If a variable in <code>x.names</code> has no name in <code>box.cox</code> , value 1 is assumed, meaning no transformation. Value 0 of the parameter equates to the logarithmic transformation. Default value of <code>box.cox</code> is 1, meaning no transformation for all variables in <code>x.names</code> .
ndiff	Named vector with non-negative integer values indicating the number of differences for variables in <code>x.names</code> . If <code>ndiff</code> is of length one and has no names, the same number of differences is used for all variables in <code>x.names</code> . If a variable in <code>x.names</code> has no name in <code>ndiff</code> , value 0 is assumed, meaning no differencing. Default value of <code>ndiff</code> is 0, meaning no differencing for all variables in <code>x.names</code> .
imputation	Logical value indicating whether imputation of missing values should be performed (see details). Default is TRUE.
em.control	List including control options for the EM algorithm (see details) with the following components: <ul style="list-style-type: none"> <li>• <code>nlags</code>: Non-negative integer indicating the number of lags to consider. If NULL (the default), it is taken as <math>\text{trunc}((n-1)^{(1/3)})</math>, where <math>n</math> is the sample size, otherwise it is right-truncated at <math>\text{trunc}(n*2/3)</math>.</li> <li>• <code>tol</code>: Positive number indicating the tolerance of the EM algorithm. Default is 0.001.</li> <li>• <code>maxit</code>: Positive integer indicating the number of iterations. Default is 1000.</li> <li>• <code>quiet</code>: Logical value indicating whether prompt messages should be suppressed. Default is FALSE.</li> </ul> Ignored if <code>imputation</code> is FALSE.

## Details

Imputation of missing values is performed after the Box-Cox transformation (Box & Cox, 1964) and differencing. A vector autoregressive model is assumed where the expectation of missing values is computed through the Expectation-Maximization (EM, Dempster et al., 1977) algorithm.

## Value

The object provided to argument `data` where variables in `var.names` have been transformed and/or differenced, and the following two attributes are added:

- `inits`: if `unit=NULL`, named vector indicating the first observed value for each variable in `x.names`, otherwise a named matrix indicating the first observed value for each unit of observation (by row) and each variable in `x.names` (by column);
- `box.cox`: named vector indicating the parameter of the Box-Cox transformation for each variable in `x.names`;
- `ndiff`: named vector indicating the number of differences for each variable in `x.names`.

**Note**

The first order difference of logarithmic values (`box.cox=0` and `ndiff=1`) provides the log returns, which approximate the proportional changes with respect to the previous time point.

If a variable contains negative values and the parameter of the Box-Cox transformation is different from 1, the transformation will not be applied and a warning is returned.

**References**

G. E. P. Box, and D. R. Cox (1964). An analysis of transformations. *Journal of the Royal Statistical Society, Series B (Methodological)*, 26(2): 211-252.

A. P. Dempster, N. M. Laird, and D. B. Rubin (1977). Maximum likelihood from incomplete data via the EM algorithm. *Journal of the Royal Statistical Society Series B (Methodological)*, 39(1): 1-38.

**See Also**

[unirootTest](#).

**Examples**

```
data(USstock)
mydata <- USstock[which(USstock$Date>="2020-04-01"),]

# first order differencing ('box.cox'=1 by default)
mydataD <- preProcess(var.names=c("BTC","DJA","IXIC","GSPC"), time="Date",
  data=mydata, ndiff=1)
summary(mydataD)
plot(ts(mydataD[,c("BTC","DJA","IXIC","GSPC")]), main="")

# setting box.cox=0 and ndiff=1 produces the log returns
mydataLR <- preProcess(var.names=c("BTC","DJA","IXIC","GSPC"), time="Date",
  data=mydata, box.cox=0, ndiff=1)
summary(mydataLR)
plot(ts(mydataLR[,c("BTC","DJA","IXIC","GSPC")]), main="")

# same result by omitting 'var.names'
mydataLR2 <- preProcess(time="Date", data=mydata, box.cox=0, ndiff=1)
summary(mydataLR2)
plot(ts(mydataLR2[,c("BTC","DJA","IXIC","GSPC")]), main="")
```

---

unirootTest

Unit root tests

---

**Description**

Augmented Dickey-Fuller (ADF) and Kwiatkowski-Phillips-Schmidt-Shin (KPSS) tests for a multivariate time series, eventually structured as a panel.

**Usage**

```
unirootTest(var.names=NULL, unit=NULL, time=NULL, data, box.cox=1, ndiff=0, max.lag=NULL)
```

## Arguments

<code>var.names</code>	Character vector including the name of the variables to be differenced. If NULL (the default), all the quantitative variables in the dataset provided to argument <code>data</code> will be differenced, with the exception of the variable indicated in <code>time</code> .
<code>unit</code>	Character containing the name of the units of observation in case of a panel structure. If a vector with length greater than 1 is provided, only the first element is considered. If NULL (the default), a single unit of observation is assumed.
<code>time</code>	Character containing the name of the time variable, which must be in numeric or date format. If a vector with length greater than 1 is provided, only the first element is considered. If NULL (the default), data are assumed to be temporally ordered.
<code>data</code>	Object of class <code>data.frame</code> containing the variables in <code>var.names</code> , <code>unit</code> and <code>time</code> .
<code>box.cox</code>	Named vector indicating the parameter of the Box-Cox transformation for variables in <code>x.names</code> . If <code>box.cox</code> is of length one and has no names, the same parameter is used for all variables in <code>x.names</code> . If a variable in <code>x.names</code> has no name in <code>box.cox</code> , value 1 is assumed, meaning no transformation. Value 0 of the parameter equates to the logarithmic transformation. Default value of <code>box.cox</code> is 1, meaning no transformation for all variables in <code>x.names</code> .
<code>ndiff</code>	Named vector with non-negative integer values indicating the number of differences for variables in <code>x.names</code> . If <code>ndiff</code> is of length one and has no names, the same number of differences is used for all variables in <code>x.names</code> . If a variable in <code>x.names</code> has no name in <code>ndiff</code> , value 0 is assumed, meaning no differencing. Default value of <code>ndiff</code> is 0, meaning no differencing for all variables in <code>x.names</code> .
<code>max.lag</code>	Non-negative integer representing the maximum lag length at which to perform the tests. If NULL (the default), it is taken as the squared root of the length of the time series.

## Details

The variable subjected to the test must be quantitative.

The null hypothesis of the ADF test is the presence of unit roots, while the null hypothesis of the KPSS test is the absence of unit roots. Therefore, p-value higher than 0.05 for the ADF test or p-value lower than 0.05 for the KPSS test suggest the presence of unit roots and the need of further differencing.

The lag length at which to perform the tests is selected through backward elimination starting from the lag length specified in argument `max.lag`.

In case of a panel structure, p-values are combined according to the method by Demetrescu *et al.* (2006).

Missing values internal to the time series are imputed through linear interpolation, otherwise they are deleted out.

## Value

One list for each variable in `var.names`, each with three components:

- `statistic`: test statistic for both tests;

- `lag.selected`: lag length selected for both tests;
- `p.value`: p-value of both tests, which is a single value if `unit` is `NULL`, otherwise one value for each unit of observation plus another one indicating the combined p-value;
- `box.cox`: parameter of the Box-Cox transformation;
- `ndiff`: order of differencing.

### Note

The first order difference of logarithmic values (`box.cox=0` and `ndiff=1`) provides the log returns, which approximate the proportional changes with respect to the previous time point.

If the parameter of the Box-Cox transformation is set to 0 for a non-negative variable, the transformation will not be applied. i.e., the parameter is set to 1, and a warning is returned.

### References

M. Demetrescu, U. Hassler, and A. Tarcolea (2006). Combining significance of correlated statistics with application to panel data. *Oxford Bulletin of Economics and Statistics*, 68(5), 647-663.

D. A. Dickey, and W. A. Fuller (1981). Likelihood ratio statistics for autoregressive time series with a unit root. *Econometrica*, 49(4): 1057-1072.

D. Kwiatkowski, P. C. B. Phillips, P. Schmidt, and Y. Shin (1992). Testing the null hypothesis of stationarity against the alternative of a unit root. *Journal of Econometrics*, 54(1-3): 159-178.

### See Also

[preProcess](#).

### Examples

```
data(USstock)
mydata <- USstock[which(USstock$Date>="2020-04-01"),]

# tests on time series in level ('box.cox'=1 by default):
# there are some unit roots
unirrootTest(var.names=c("BTC","DJA","IXIC","GSPC"), time="Date",
             data=mydata)

# tests on time series in log return: ok
unirrootTest(var.names=c("BTC","DJA","IXIC","GSPC"), time="Date",
             data=mydata, box.cox=0, ndiff=1)
```

---

USstock

*Bitcoin and US stock indices data*

---

### Description

Daily close exchange rate of Bitcoin and of three composite indices of the US stock market from 17 September 2014 to 30 September 2020.

### Usage

```
data(USstock)
```

**Format**

A data.frame with a total of 1521 observations on the following 5 variables:

Date Day of observation.

DJA Exchange rate of Dow Jones Average index.

IXIC Exchange rate of Nasdaq Composite index.

GSPC Exchange rate of Standard\&Poor 500 index.

BTC Exchange rate of Bitcoin.

# Index

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`gammadlm-package`, [1](#)  
`gammaKernel (gammaWeights)`, [6](#)  
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`preProcess`, [2](#), [5](#), [9](#), [13](#)  
  
`unirootTest`, [2](#), [5](#), [11](#), [11](#)  
`USstock`, [13](#)