

Package ‘ardl.nardl’

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Title Linear and Nonlinear Autoregressive Distributed Lag Models:
General-to-Specific Approach

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Description Estimate the linear and nonlinear autoregressive distributed lag (ARDL & NARDL) models and the corresponding error correction models, and test for longrun and short-run asymmetric. The general-to-specific approach is also available in estimating the ARDL and NARDL models. The Pesaran, Shin & Smith (2001) (<[doi:10.1002/jae.616](https://doi.org/10.1002/jae.616)>) bounds test for level relationships is also provided. The 'ardl.nardl' package also performs short-run and longrun symmetric restrictions available at Shin et al. (2014) <[doi:10.1007/978-1-4899-8008-3_9](https://doi.org/10.1007/978-1-4899-8008-3_9)> and their corresponding tests.

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ardl.nardl-package *Linear and Nonlinear Autoregressive Distributed Lag Models:
General-to-Specific Approach*

Description

Estimate the linear and nonlinear autoregressive distributed lag (ARDL & NARDL) models and the corresponding error correction models, and test for longrun and short-run asymmetric. The Pesaran, Shin & Smith (2001) Bounds test for level relationships is also provided with the aid of Jordan and Philips (2020) pssbounds function. In addition, the 'ardl.nardl' package also performs short-run and longrun symmetric restrictions available at Shin et al. (2014) and their corresponding tests.

References

- Jordan S, Philips A (2020). `_dynamac: Dynamic Simulation and Testing for Single-Equation ARDL Models_`. R package version 0.1.11
- Pesaran, M. H., Shin, Y., & Smith, R. J. (2001). Bounds testing approaches to the analysis of level relationships. *Journal of applied econometrics*, 16(3), 289-326. <https://doi.org/10.1002/jae.616>
- Shin, Y., Yu, B., & Greenwood-Nimmo, M. (2014). Modelling Asymmetric Cointegration and Dynamic Multipliers in a Nonlinear ARDL Framework. In: Sickles, R., Horrace, W. (eds) *Festschrift in Honor of Peter Schmidt*. Springer, New York, NY. https://doi.org/10.1007/978-1-4899-8008-3_9

Examples

```
## Not run:
data(fuel_price)
data(expectation)
out1 <- gets_ardl_uecm(x = expectation,
                      dep_var = c('nq_inf_exp'),
                      expl_var = c('food_inf', 'nethawkish'),
                      p_order = c(4),
```

```
      q_order = c(5,7),
      gets_pval = 0.1,
      case = 4,
      graph_save = FALSE,
      F_HC = FALSE,
      order_l = 7)

out1

out2 <- gets_nardl_uecm(x = expectation,
  decomp = 'food_inf',
  dep_var = 'nq_inf_exp',
  control = 'nethawkish',
  c_q_order = c(3),
  p_order = c(3),
  q_order = c(3),
  gets_pval = 0.1,
  graph_save = FALSE,
  case = 5,
  F_HC = FALSE)

out2

OUT3 <- auto_case_ardl(x = expectation,
  dep_var = 'n12m_inf_exp',
  expl_var = c('food_inf', "hawkish", "dovish"),
  p_order = 2,
  q_order = c(4,4,4),
  gets_pval = 0.05,
  graph_save = FALSE,
  order_l = 7)

OUT3

OUT4 <- nardl_auto_case(x = fuel_price,
  decomp = 'wti',
  dep_var = 'fpp',
  control = 'bdc',
  c_q_order = c(5),
  p_order = c(5),
  q_order = c(6),
  gets_pval = 0.1,
  order_l = 4,
  graph_save = FALSE)

OUT4

uecm_case3 <- ardl_uecm(x = fuel_price,
  dep_var = c('fpp'),
  expl_var = c('bdc', 'wti'),
  p_order = c(6),
  q_order = c(5,3),
  graph_save = FALSE,
  case = 3)

uecm_case3

output_n1_case5 <- nardl_uecm(x = fuel_price,
```

```

                                decomp = c('bdc'),
                                control =c('wti'),
                                c_q_order = c(2),
                                p_order = c(3),
                                q_order = c(5),
                                dep_var = c('fpp'),
                                graph_save = FALSE,
                                case = 5)

output_n1_case5

data(syg_data)
out_srsr <- nardl_uecm_sym(x = syg_data,
                          decomp = 'ca_u',
                          assumption = c('SRSR'),
                          control =NULL,
                          p_order =5,
                          q_order =3,
                          dep_var = 'ca_ip',
                          graph_save = FALSE,
                          case = 3)

out_srsr

out_lrsr <- nardl_uecm_sym(x = syg_data,
                           decomp = 'ca_u',
                           assumption = c('LRSR'),
                           control =NULL,
                           p_order =5,
                           q_order =3,
                           dep_var = 'ca_ip',
                           graph_save = FALSE,
                           case = 3)

out_lrsr

## End(Not run)

```

ArchTest

*Import from package **nardl***

Description

The ArchTest object is imported from package **nardl**. Help is available here: [nardl::ArchTest](#).

ardl_uecm

Estimate the ARDL and ARDL Error Correction Model.

Description

Estimate ARDL and ARDL error correction model, conduct the bounds test for cointegration

Usage

```
ardl_uecm(x, p_order = c(2), q_order, dep_var, order_l = 4,
          graph_save = FALSE, expl_var, case = 3)
```

Arguments

x	A dataframe object. S3 dataframe object not applicable.
p_order	is the lag for the dependent variable (dep_var). Default value is c(2)
q_order	A numeric vector containing the lags for each explanatory variable. A vector of two lags (i.e c(3,4)) which corresponds to the expl_var (say c('x1','x2') where x1 has lag of 3 and x2 has lag of 4).
dep_var	Character vector containing one dependent variable
order_l	Positive integer. lag for the order of serial correlation and ARCH test.
graph_save	Logical (default is FALSE). If TRUE, stability plot is displayed as well.
expl_var	Character vector containing the explanatory variable(s).
case	A numerical vector that can take a value between 1 to 5. The default value is 3 - restricted constant (no intercept), no trend

Value

ARDL_fit	This display the ARDL lm fit
ARDL_ECM_fit	This display the ARDL ECM lm fit
UECM	This display the summary results of the estimated unrestricted error correction models
UECM	This display the summary results of the estimated linear unrestricted error correction models
cointegration	This presents the Pesaran, Shin, and Smith (2001) cointegration test. The critical F and t statistics obtained if the sample is greater than 80 are asymptotic. If you adopt this segment, please consider citing Pesaran et al (2001) and Jordan and Philips (2020) dynamac package. If the sample is less than or equal to 80, for case 1, the F and t statistics adopted are large sample asymptotic critical value (CV) and no small-sample asymptotic critical value available. When case = 3 or 5, the critical value of F statistic is small sample asymptotic, and the adopted t statistics are large sample asymptotic CV. Consider including Narayan (2005) in your citation list. If the sample is greater than 80, for case 1, 3 and 5, large sample asymptotic CV for F and t statistics are adopted. In case 2 and 4, for sample <= 80, the F statistic are small sample asymptotic and no asymptotic CV for the t statistic. If sample > 80, for case 2 and 4, large sample asymptotic CV for F is adopted and no t-statistic CV.
Longrun_relation	The estimated longrun relations
diagnostics test	The diagnostic tests indicate the Breusch-Godfrey test for higher-order serial correlation (BG_SC_lm_test). The Engle (1982) test for conditional heteroscedasticity (LM_ARCH_test). The test for non-normality is that of Jarque and Bera

(1980). The RESET null hypothesis adopted implies - including the 2nd - degree terms improve the fit (over the model specified). This ARCH LM function is derived from Zaghoudi (2018)

References

- Jordan S, Philips A (2020). `_dynamac: Dynamic Simulation and Testing for Single-Equation /n ARDL Models_`. R package version 0.1.11
- Narayan, P. K. (2005). The saving and investment nexus for China: evidence from cointegration tests. *Applied economics*, 37(17), 1979-1990.
- Pesaran, M. H., Shin, Y., & Smith, R. J. (2001). Bounds testing approaches to the analysis of level relationships. *Journal of applied econometrics*, 16(3), 289-326.
- Zaghoudi, T. (2018). `nardl: Nonlinear Cointegrating Autoregressive Distributed Lag Model_`. R package version 0.1.5

See Also

[gets_ardl_uecm](#) [gets_nardl_uecm](#) [nardl_uecm](#) [nardl_uecm_sym](#)

Examples

```
data(fuel_price)
uecm_case3 <- ardl_uecm(x = fuel_price,
                       p_order =c(6),
                       q_order =c(5,3),
                       dep_var = c('fpp'),
                       expl_var = c('bdc', 'wti'),
                       graph_save = FALSE,
                       case = 3)

uecm_case3
uecm_case3$cointegration
uecm_case3$Longrun_relation
uecm_case3$`diagnostics test`

uecm_case2 <- ardl_uecm(x = fuel_price,
                       p_order =c(3),
                       q_order =c(4),
                       dep_var = c('fpp'),
                       expl_var = c('wti'),
                       graph_save = TRUE,
                       case = 2)

uecm_case2
uecm_case2$cointegration
uecm_case2$Longrun_relation
uecm_case2$`diagnostics test`
```

auto_case_ardl *Obtain the best ARDL model specification and bounds test.*

Description

This function finds the best ARDL model specification and conduct bounds test by relying on the general to specific approach.

Usage

```
auto_case_ardl(x, dep_var, expl_var, p_order, q_order,
  gets_pval = 0.05, order_l = 3, graph_save = FALSE)
```

Arguments

x	Dataframe.
dep_var	A Character vector that contain the response variable.
expl_var	Character vector containing the list of explanatory variable(s).
p_order	An integer. Lag differenced adopted for the differenced response variable
q_order	An integer. Lag differenced adopted for the differenced explanatory variable(s)
gets_pval	The p- value which served as the criteria for eliminating non-significant variable in the course of obtaining the best model based on the Schwarz information criteria.
order_l	Integer. Needed for the autocorrelation and heteroscedasticity test
graph_save	Logical. If TRUE, displays the stability plots

Details

The procedure of the general-to-specific approach in obtaining the parsimonious model involves conducting the multi-path backwards elimination; tests both single and multiple hypothesis tests, diagnostics tests and goodness-of-fit measures. See page 5 of Sucarrat, (2021) for more details.

The value for gets_pval is influential the final model based on the multipath backward elimination. For more details on the general-to-specific approach, see the vignette of the 'gets' package.

Value

Parsimonious_ARDL_fit	Return an estimated general-to-specific ARDL model
Parsimonious_ECM_fit	Return an estimated general-to-specific error correction model
Summary_ecm_fit	Return the summary of 'Parsimonious_ECM_fit'

Parsimonious_ECM_diagnostics_test
 Return the diagnostic test for 'Parsimonious_ECM_fit'. The diagnostic tests items are the Breusch-Godfrey test for higher-order serial correlation (BG_SC_lm_test). The Engle (1982) test for conditional heteroscedasticity (LM_ARCH_test). The test for non-normality is that of Jarque and Bera (1980). The RESET null hypothesis adopted implies - including the 2nd - degree terms improve the fit (over the model specified). Ljung and Box (1978) tests for autocorrelation in the residuals

cointegration Return the F statistic, the upper and lower critical values for PSS (2001) bounds test

Longrun_relation
 The estimated longrun relation from the error correction model

Note

Do not differenced the variables to be adopted in this function and all other functions for ARDL and NARDL estimation. The package inherently takes the difference and produced output with a prefix (D.) to the variable name and suffix the variable name with underscore (_) and the lag value.

References

Sucarrat, G. User-Specified General-to-Specific (GETS) and Indicator Saturation (ISAT) Methods. 28th September 2021. <https://mirror.epn.edu.ec/CRAN/web/packages/gets/vignettes/user-defined-gets-and-isat.pdf>

See Also

[gets](#) [gets_nardl_uecm](#) [ardl_uecm](#) [nardl_auto_case](#)

Examples

```
data("expectation")
out_aut <- auto_case_ardl(x = expectation,
  dep_var = 'n12m_inf_exp',
  expl_var = c('food_inf', "hawkish", "dovish"),
  p_order = 2,
  q_order = c(4,4,4),
  gets_pval = 0.05,
  graph_save = FALSE)

data("fuel_price")
out_aut <- auto_case_ardl(x = fuel_price,
  dep_var = 'fpp',
  expl_var = c('bdc', 'wti'),
  p_order = 2,
  q_order = c(4,4),
  gets_pval = 0.08,
  graph_save = TRUE)
```

cumsq	<i>Import from package nardl</i>
-------	---

Description

The cumsq object is imported from package **nardl**. Help is available here: [nardl::cumsq](#).

cusum	<i>Import from package nardl</i>
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Description

The cusum object is imported from package **nardl**. Help is available here: [nardl::cusum](#).

dynamac_pkg_bounds_test	<i>A function in 'dynamac' package to Perform Pesaran, Shin, and Smith (2001) cointegration test</i>
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Description

This is an Abridged version of Jordan and Philips (2020). Please, see the dynamac package for the full documentation and due reference be given to the authors of dynamac - Jordan and Philips (2020).

Usage

```
dynamac_pkg_bounds_test(obs, fstat, tstat = NULL, case, k)
```

Arguments

obs
fstat
tstat
case
k

Value

Fstat	The bounds F statistic, lower and upper bounds
tstat	The bounds t statistic, lower and upper bounds

References

Jordan S, Philips A (2020). `_dynamac`: Dynamic Simulation and Testing for Single-Equation ARDL Models. R package version 0.1.11 <<https://CRAN.R-project.org/package=dynamac>>

See Also

[pssbounds](#)

expectation

Inflation Expectation Dataset

Description

Dataset

Usage

```
data("expectation")
```

Format

A data frame with 47 observations on the following variables.

`date` date from 2009Q1 to 2020Q4

`nq_inf_exp` Natural log of Next quarter inflation expectation (index)

`n12m_inf_exp` Natural log of Next 12-month inflation expectation (index)

`hawkish` Hawkishness

`dovish` Dovishness

`nethawkish` nethawkishness

`headline_inf` actual headline inflation (percent)

`core_inf` actual core inflation (percent)

`food_inf` actual food inflation (percent)

Details

The sentiment indices were extracted (text mined) from CBN monetary policy committee ommuniques with the aid of the monetary policy dictionary adopted in Mate et al. (2021).

Source

....

References

Mate A, Sebok M, Barczikay T (2021) The effect of central bank communication on sovereign bond yields: The case of Hungary. PLoS ONE 16(2): e0245515. <https://doi.org/10.1371/journal.pone.0245515>

Examples

```
data(expectation)
```

fuel_price

Time series data from 2000M01 through 2021M03

Description

Time series data on fuel pump price, oil price and exchange rate

Usage

```
data("fuel_price")
```

Format

A data frame with 255 observations on the following variables.

fpp log of Fuel Pump Price Per Litre - Average (PMS)

wti log of West Texas Intermediate

bdc log of Bureau De-change exchange rate (N/\$)

date time series from 2000M01 through 2021M03

Source

fpp is obtained from Central Bank of Nigeria 2021Q1 Statistical Bulletin

wti obtained from World bank commodity price (pinksheet)

bdc sourced from <http://statistics.cbn.gov.ng/cbn-onlinestats/>

Examples

```
data(fuel_price)
```

gets_ardl_uecm	<i>General-to-specific approach for the autoregressive distributed lag model</i>
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Description

Adopt the general-to-specific approach to estimate the autoregressive distributed lag model

Usage

```
gets_ardl_uecm(x, dep_var, expl_var, p_order = c(2), q_order = c(3),
              gets_pval = 0.1, case = 3, F_HC = FALSE, order_l = 5,
              graph_save = FALSE)
```

Arguments

x	data frame
dep_var	A character vector. The dependent variable.
expl_var	Character vector. List of explanatory variable(s)
p_order	Integer. Maximum number of lags for 'dep_var'
q_order	Integer. Maximum number of lags for 'expl_var'
gets_pval	Integer value between 0 and 1 needed for the general-to-specific approach. The default is 0.1 (10 percent significance level). The chosen p-value is the criteria for determining non-significant repressors to be eliminated in a backward elimination path. The final parsimonious model is the best fit model based on the Schwarz information criteria
case	Positive integer 1 to 5. Default is 3
F_HC	Logical (default is FALSE). If TRUE, Heteroscedasticity-Consistent Covariance Matrix Estimation is applied to the model before when estimating F statistic
graph_save	Logical. If TRUE, display stability plot. Default is FALSE
order_l	Integer. order for the serial correlation, and heteroscedasticity test

Value

Parsimonious_ARDL_fit	Return an estimated general-to-specific ARDL model
Parsimonious_ECM_fit	Return an estimated general-to-specific error correction model
Summary_ecm_fit	Return the summary of 'Parsimonious_ECM_fit'
Parsimonious_ECM_diagnostics_test	Return the diagnostic test for 'Parsimonious_ECM_fit'. The diagnostic tests items are the Breusch-Godfrey test for higher-order serial correlation (BG_SC_lm_test). The Engle (1982) test for conditional heteroscedasticity (LM_ARCH_test). The

test for non-normality is that of Jarque and Bera (1980). The RESET null hypothesis adopted implies - including the 2nd - degree terms improve the fit (over the model specified). Ljung and Box (1978) tests for autocorrelation in the residuals

cointegration Return the F statistic, the upper and lower critical values for PSS (2001) bounds test

Longrun_relation
The estimated longrun relation from the error correction model

References

- Engle, R. F. (1982). Autoregressive conditional heteroscedasticity with estimates of the variance of United Kingdom inflations. *Econometrica* (50) 987 - 1007
- Ljung GM, Box GEP (1978). On a Measure of Lack of Fit in Time Series Models. *Biometrika*, 65(2), 297 - 303. <https://doi.org/10.2307/2335207>
- Jarque C, Bera A (1980). Efficient Tests for Normality, Homoskedasticity, and Serial Independence. *Economics Letters*, 6(3), 255 - 259. [https://doi.org/10.1016/0165-1765\(80\)90024-5](https://doi.org/10.1016/0165-1765(80)90024-5)
- Pesaran, M. H., Shin, Y., & Smith, R. J. (2001). Bounds testing approaches to the analysis of level relationships *Journal of applied econometrics*, 16(3), 289-326

See Also

[gets_nardl_uecm](#) [ardl_uecm](#)

Examples

```
data(expectation)
out <- gets_ardl_uecm(x = expectation,
                    dep_var = c('nq_inf_exp'),
                    expl_var = c('food_inf', 'nethawkish'),
                    p_order = c(4),
                    q_order = c(5,7),
                    gets_pval = 0.1,
                    case = 4,
                    graph_save = FALSE,
                    F_HC = FALSE,
                    order_l = 7)

out
```

gets_nardl_uecm

Parsimonious model for the autoregressive distributed lag model

Description

Adopt the general-to-specific approach to estimate the autoregressive distributed lag model

Usage

```
gets_nardl_uecm(x, decomp, dep_var, control = NULL, c_q_order = c(2),
               p_order = c(3), q_order = c(4), gets_pval = 0.1, order_1 = 4,
               F_HC = FALSE, graph_save = FALSE, case = 3)
```

Arguments

x	data frame
decomp	A character vector. The variable to be decomposed to positive (pos) and negative (neg) variable.
dep_var	A character vector. The dependent variable
control	A character vector. Default is NULL. The second dependent variable.
c_q_order	Integer. Maximum number of lags for 'control'
p_order	Integer. Maximum number of lags for 'dep_var'
q_order	Integer. Maximum number of lags for level and differenced 'decomp'
gets_pval	Integer value between 0 and 1 needed for the general-to-specific approach. The default is 0.1 (10 percent significance level). The chosen p-value is the criteria for determining non-significant repressors to be eliminated in a backward elimination path. The final parsimonious model is the best fit model based on the Schwarz information criteria.
order_1	Integer. order for the serial correlation, and heteroscedasticity test
F_HC	Logical (default is FALSE). If TRUE, heteroscedasticity-Consistent Covariance Matrix Estimation is applied to the model before when estimating F statistic
graph_save	Logical. If TRUE, display stability plot. Default is FALSE.
case	Positive integer 1 to 5. Default is 3

Value

Parsimonious_NARDL_fit	Return an estimated general-to-specific NARDL model.
Parsimonious_ECM_fit	Return an estimated general-to-specific error correction model.
Summary_uecm_fit	Return the summary of 'Parsimonious_ECM_fit'
ecm_diagnostics_test	Return the diagnostic test for the 'Parsimonious_ECM_fit'. The diagnostic tests indicate the Breusch-Godfrey test for higher-order serial correlation (BG_SC_lm_test). The Engle (1982) test for conditional heteroscedasticity (LM_ARCH_test). The test for non-normality is that of Jarque and Bera (1980). The RESET null hypothesis adopted implies - including the 2nd - degree terms improve the fit (over the model specified).
longrun_asym	Return the estimated longrun asymmetric test

lagm	<i>Lag a matrix</i>
------	---------------------

Description

Take the lag of a variable using the specified number of lag(s).

Usage

```
lagm(x, k)
```

Arguments

x	a matrix object
k	number of lags

Value

x	dimensional matrix with k
---	---------------------------

Examples

```
x <- cbind(x1 = c(23, 45, 56, 78, 98, 21, 17),  
           x2 = c(3, 4, 6, 7, 2, 9, 17))  
lagm(x, k = 3)
```

nardl_auto_case	<i>Obtain the best NARDL model specification and bounds test.</i>
-----------------	---

Description

This function finds the best NARDL model specification and conduct bounds test by relying on the general to specific approach.

Usage

```
nardl_auto_case(x, decomp, dep_var, control = NULL, c_q_order = c(2),  
               p_order = c(3), q_order, gets_pval = 0.1,  
               order_l = order_l, graph_save = FALSE)
```


Arguments

x	Dataframe
decomp	A character vector. The variable to be decomposed to positive (pos) and negative (neg) variable.
dep_var	A character vector. The dependent variable
control	A character vector. Default is NULL. The second dependent variable.
c_q_order	Integer. Maximum number of lags for 'control'
p_order	An integer. Lag differenced adopted for the differenced response variable
q_order	An integer. Lag differenced adopted for the differenced explanatory variable(s)
gets_pval	The p- value which served as criteria for eliminating non-significant variable in the course of obtaining the best model based on the Schwarz information criteria.
order_l	Integer. Needed for the autocorrelation and heteroscedasticity test
graph_save	Logical. If TRUE, displays the stability plots

Details

The procedure of the general-to-specific approach in obtaining the parsimonious model involves conducting the multi-path backwards elimination; tests both single and multiple hypothesis tests, diagnostics tests and goodness-of-fit measures. See page 5 - 6 of Sucarrat (2021) for more details.

The value for gets_pval is influential the final model based on the multipath backward elimination. For more details on the general-to-specific approach, see the vignette of the 'gets' package.

Value

Parsimonious_NARDL_fit	Return an estimated general-to-specific NARDL model.
Parsimonious_ECM_fit	Return an estimated general-to-specific error correction model.
Summary_uecm_fit	Return the summary of 'Parsimonious_ECM_fit'
ecm_diagnostics_test	Return the diagnostic test for the 'Parsimonious_ECM_fit'. The diagnostic tests indicate the Breusch-Godfrey test for higher-order serial correlation (BG_SC_lm_test). The Engle (1982) test for conditional heteroscedasticity (LM_ARCH_test). The test for non-normality is that of Jarque and Bera (1980). The RESET null hypothesis adopted implies - including the 2nd - degree terms improve the fit (over the model specified).
longrun_asym	Return the estimated longrun asymmetric test
Shortrun_asym	Return the estimated short-run asymmetric test. If one of the decomposed variable does not appear among the shortrun differenced variables of the parsimonious model, The value returned is a wald test of whether the sum of the coefficients of the remaining decomposed variable included does not have any significant effect on the best model

`cointegration` Return the F statistic, the upper and lower critical values for PSS (2001) bounds test. Please, disregard the `tstat` on the cointegration test.

`Longrun_relation`
The longrun relation

Note

Do not differenced the variables to be adopted in this function and all other functions for NARDL and ARDL estimation. The package inherently takes the difference and produced output with a prefix (D.) to the variable name and suffix the variable name with underscore (`_`) and the lag value.

References

Sucarrat, G. User-Specified General-to-Specific (GETS) and Indicator Saturation (ISAT) Methods. 28th September 2021. <https://mirror.epn.edu.ec/CRAN/web/packages/gets/vignettes/user-defined-gets-and-isat.pdf>

See Also

[gets](#) [gets_nardl_uecm](#) [ardl_uecm](#) [auto_case_ardl](#)

Examples

```
## Not run:
data("fuel_price")
out1 <- nardl_auto_case(x = fuel_price,
  decomp = 'wti',
  dep_var = 'fpp',
  control = 'bdc',
  c_q_order = c(5),
  p_order = c(5),
  q_order = c(6),
  gets_pval = 0.1,
  order_l = 4,
  graph_save = FALSE)

out1

out2 <- nardl_auto_case(x = fuel_price,
  decomp = 'wti',
  dep_var = 'fpp',
  control = NULL,
  c_q_order = c(4),
  p_order = c(5),
  q_order = c(6),
  gets_pval = 0.02,
  order_l = 4,
  graph_save = FALSE)

out2

## End(Not run)
```

nardl_uecm

*Estimate the nonlinear ARDL (NARDL) Error Correction Model***Description**

Function to estimate Shin, Yu, and Greenwood-Nimmo (2014) nonlinear ARDL (NARDL), the NARDL ECM, and conduct the bounds test for cointegration.

Usage

```
nardl_uecm(x, decomp, control = NULL, c_q_order = c(2),
p_order = c(3), q_order = c(4), dep_var, order_l = 4, graph_save = FALSE, case = 3)
```

Arguments

x	A dataframe object. S3 dataframe object not applicable.
decomp	A character vector. The right hand side variable to be decomposed into positive (pos) and negative (neg)
control	A character vector with one element serving as the second variable in the right hand side of the equation. The default is NULL.
c_q_order	Lag order for 'control'. the default value is c(2) and only applicable when control is specified.
p_order	Positive interger. Number of lags for dependent variable (dep_var).
q_order	Positive interger. Short-run lags for the decomposed variables. In case the LM test null hypothesis of no serial correlation is rejected, consider increasing the lags, in particular, the p_order or both p_order and q_order lags until the null hypothesis is accepted.
dep_var	Character vector. The dependent variable.
order_l	Positive integer. lag order for the serial correlation, and heteroscedasticity test
graph_save	Logical (default is FALSE). If TRUE, stability plot is displayed.
case	A numerical vector that can take a value between 1 to 5. The default value is 3 - restricted constant (no intercept), no trend. See details for more information.

Value

NARDL_fit	This display the NARDL lm fit
NARDL_ECM_fit	This display the NARDL ECM lm fit
UECM	This display the summary results of the estimated unrestricted error correction models
cointegration	This presents the Pesaran, Shin, and Smith (2001) cointegration test. The critical F and t statistics obtained if the sample is greater than 80 are asymptotic. If you adopt this segment, please consider citing Pesaran et al (2001) and Jordan and Philips (2020) dynamac package. If the sample is less than or equal to 80, for

case 1, the F and t statistics adopted are large sample asymptotic critical value (CV) and no small-sample asymptotic critical value available. When case = 3 or 5, the critical value of F statistic is small sample asymptotic, and the adopted t statistics are large sample asymptotic CV. Consider including Narayan (2005) in your citation list. If the sample is greater than 80, for case 1, 3 and 5, large sample asymptotic CV for F and t statistics are adopted. In case 2 and 4, for sample ≤ 80 , the F statistic are small sample asymptotic and no asymptotic CV for the t statistic. If sample > 80 , for case 2 and 4, large sample asymptotic CV for F is adopted and no t-statistic CV.

Longrun_relation

The estimated longrun relations

Longrun_asymmetric_test

This contains the asymmetric test, with the null hypothesis of longrun coefficients of the positive decomposed variable equals to the coefficient of negative decomposed variable.

Shortrun_asymmetric_test

This contains the asymmetric test, with the null hypothesis the sum of the coefficients of all the short run positive decomposed variable equals to the sum of the coefficients of the short-run negative decomposed variable.

diagnostics test

The diagnostic tests indicate the Breusch-Godfrey test for higher-order serial correlation (BG_SC_lm_test). The Engle (1982) test for conditional heteroscedasticity (LM_ARCH_test). The test for non-normality is that of Jarque and Bera (1980). The RESET null hypothesis adopted implies - including the 2nd - degree terms improve the fit (over the model specified). This ARCH LM function is derived from Zaghoudi (2018)

References

- Jordan S, Philips A (2020). `_dynamac: Dynamic Simulation and Testing for Single-Equation ARDL Models_`. R package version 0.1.11
- Narayan, P. K. (2005). The saving and investment nexus for China: evidence from cointegration tests. *Applied economics*, 37(17), 1979-1990.
- Pesaran, M. H., Shin, Y., & Smith, R. J. (2001). Bounds testing approaches to the analysis of level relationships. *Journal of applied econometrics*, 16(3), 289-326.
- Shin, Y., Yu, B., & Greenwood-Nimmo, M. (2014). Modelling Asymmetric Cointegration and Dynamic Multipliers in a Nonlinear ARDL Framework. In: Sickles, R., Horrace, W. (eds) *Festschrift in Honor of Peter Schmidt*. Springer, New York, NY. https://doi.org/10.1007/978-1-4899-8008-3_9
- Zaghoudi, T. (2018). `nardl: Nonlinear Cointegrating Autoregressive Distributed Lag Model_`. R package version 0.1.5

See Also

[ardl_uecm](#) [nardl_uecm_sym](#)

Examples

```

data(fuel_price)
output_n1_case5 <- nardl_uecm(x = fuel_price,
                             decomp = c('bdc'),
                             control = c('wti'),
                             c_q_order = c(2),
                             p_order = c(3),
                             q_order = c(5),
                             dep_var = c('fpp'),
                             graph_save = FALSE,
                             case = 5)

output_n1_case5
output_n1_case5$UECM

output_n2_case3 <- nardl_uecm(x = fuel_price,
                              decomp = c('bdc'),
                              control = NULL,
                              c_q_order = NULL,
                              p_order = c(3),
                              q_order = c(5),
                              dep_var = c('fpp'),
                              graph_save = FALSE,
                              case = 3)

output_n2_case3
output_n3_case3 <- nardl_uecm(x = fuel_price,
                              decomp = c('wti'),
                              control = NULL,
                              c_q_order = NULL,
                              p_order = c(3),
                              q_order = c(5),
                              dep_var = c('fpp'),
                              graph_save = FALSE,
                              case = 3)

output_n3_case3

```

nardl_uecm_sym	<i>Short-run symmetry restrictions (SRSR) and the longrun symmetry restrictions (LRSR).</i>
----------------	---

Description

~~ Performs one of Short-run symmetry restrictions or the longrun symmetry restrictions.

Usage

```

nardl_uecm_sym(x, assumption = c("SRSR"), decomp, control = c(2), c_q_order = c(2),
p_order = c(3), q_order = c(4), dep_var, graph_save = FALSE, case = 3)

```

Arguments

x	A dataframe object. S3 dataframe object not applicable.
assumption	The Short-run symmetry restrictions (SRSR), the default, assume the transitory asymmetric effect is insignificant and testing the longrun asymmetric effect. longrun symmetry restrictions (LRSR) imposes symmetric effect on the longrun coefficients of the positive and negative decomposed variable, and assuming the presence of transitory asymmetric effect.
decomp	A variable to be decomposed into positive (pos) and negative (neg)
control	A character vector with one element serving as the second variable single control variable. The default is c(2) and applicable when control variable is specified.
c_q_order	Lag order for the control. The default value is c(2) and only applicable when control is specified.
p_order	Number of lags for dependent variable (dep_var).
q_order	Short-run lags for each positive and negative decomposed variable. In case the LM test null hypothesis of no serial correlation is rejected, consider increasing the lags, in particular, the p_order or both p_order and q_order lags until the null hypothesis is accepted.
dep_var	Character vector which contain the dependent variable.
graph_save	Logical (default is FALSE). If TRUE, stability plot is displayed.
case	A numerical vector of either 1, 2 3, 4 or 5. The default value is 3 - restricted constant (no intercept), no trend. See details for more information.

Details

The Short-run symmetry restrictions (SRSR), the default may be useful in obtaining more accurate estimate when the sample is small (SYG, 2014). Besides, this may be necessary when the short-run asymmetric effect does not exist. longrun symmetry restrictions (LRSR) imposes symmetric effect on the positive and negative decomposed variable.

Value

NARDL_ECM_fit	This display the NARDL ECM Im fit
UECM	This display the summary results of the estimated unrestricted error correction models
cointegration	This presents the Pesaran, Shin, and Smith (2001) cointegration test. The critical F and t statistics obtained if the sample is greater than 80 are asymptotic. If you adopt this segment, please consider citing Pesaran et al (2001) and Jordan and Philips (2020) dynamac package. If the sample is less than or equal to 80, for case 1, the F and t statistics adopted are large sample asymptotic critical value (CV) and no small-sample asymptotic critical value available. When case = 3 or 5, the critical value of F statistic is small sample asymptotic, and the adopted t statistics are large sample asymptotic CV. Consider including Narayan (2005) in your citation list. If the sample is greater than 80, for case 1, 3 and 5, large sample asymptotic CV for F and t statistics are adopted. In case 2 and 4, for sample <= 80, the F statistic are small sample asymptotic and no asymptotic CV

for the t statistic. If sample > 80, for case 2 and 4, large sample asymptotic CV for F is adopted and no t-statistic CV.

Longrun_relation

The estimated longrun relations

asymmetric test

This contains the asymmetric test

diagnostics test

This presents the Breusch-Godfrey test for higher-order serial correlation with null hypothesis of no serial correlation, the Autoregressive Conditional Heteroscedasticity (ARCH) LM-test with the null hypothesis of no ARCH effect and the Jarque-Bera normality test with null hypothesis of normality. This ARCH LM function is derived from Zaghoudi (2018)

References

Jordan S, Philips A (2020). `_dynamac: Dynamic Simulation and Testing for Single-Equation ARDL Models_`. R package version 0.1.11

Narayan, P. K. (2005). The saving and investment nexus for China: evidence from cointegration tests. *Applied economics*, 37(17), 1979-1990.

Pesaran, M. H., Shin, Y., & Smith, R. J. (2001). Bounds testing approaches to the analysis of level relationships. *Journal of applied econometrics*, 16(3), 289-326.

Shin, Y., Yu, B., & Greenwood-Nimmo, M. (2014). Modelling Asymmetric Cointegration and Dynamic Multipliers in a Nonlinear ARDL Framework. In: Sickles, R., Horrace, W. (eds) *Festschrift in Honor of Peter Schmidt*. Springer, New York, NY. https://doi.org/10.1007/978-1-4899-8008-3_9

Zaghoudi, T. (2018). `nardl: Nonlinear Cointegrating Autoregressive Distributed Lag Model_`. R package version 0.1.5

See Also

[ardl_uecm](#) [nardl_uecm](#) [pssbounds](#)

Examples

```
data(syg_data)
out_srsr <- nardl_uecm_sym(x = syg_data,
  decomp = 'ca_ip',
  assumption = c('SRSR'),
  control = NULL,
  p_order = 5,
  q_order = 3,
  dep_var = 'ca_u',
  graph_save = FALSE,
  case = 3)
out_srsr

out_lrsr <- nardl_uecm_sym(x = syg_data,
  decomp = 'ca_ip',
  assumption = c('LRSR'),
  control = NULL,
```

```

                                p_order =5,
                                q_order =3,
                                dep_var = 'ca_u',
                                graph_save = FALSE,
                                case = 3)
out_lrslr

```

output_ren	<i>Rename the coefficient of the nardl_uecm summary output</i>
------------	--

Description

As the case might be, there could be need to rename the rownames in nardl_uecm summary output

Usage

```
output_ren(x, listn = listn, D.pattern, D.repl, l.pattern, l.repl)
```

Arguments

x	A summary model output, preferably, nardl_uecm object
listn	A character vector
D.pattern	The difference and lag difference variables contained the model
D.repl	Character vector containing the new names to replace the D.pattern
l.pattern	Character vector of variable appearing at level
l.repl	Character vector containing the names to replace l.pattern

Value

x	nardl_uecm or nardl_uecm_sym object
---	-------------------------------------

Examples

```

data(syg_data)
un <- c('ca_u', 'jp_u', 'us_u')
ip <- c('ca_ip', 'jp_ip', 'us_ip')
cnames <- c('Canada', 'Japan', 'United States')
case <- c(5,5,5)
p_lags <- c(3,4,5)
q_lags <- c(3,4,5)
#####
out_nardl <- lapply(1:length(un), function(i)
  nardl_uecm(x = syg_data,
            decomp = un[[i]],
            control =NULL,
            c_q_order = c(2),
            p_order =p_lags[[i]],
            q_order =q_lags[[i]],

```



```

        dep_var = ip[[i]],
        graph_save = FALSE,
        case = 3))
names(out_nardl) <- cnames
out_nardl$Canada$UECM

output <- output_ren(x=out_nardl,listn = un,
  D.patern="D.ca_ip|D.jp_ip|D.us_ip",D.repl="D.output",
  l.patern = "ca_ip|jp_ip|us_ip",l.repl="output")

output <- output_ren(x=output,listn = un,
  D.patern="D.ca_u|D.jp_u|D.us_u",D.repl="D.unemployment",
  l.patern = "ca_u|jp_u|us_u",l.repl="unemployment")

output$Canada$UECM

```

syg_data	<i>Time series data on output and unemployment in Canada, Japan and United States from 1981M01 to 2022M07</i>
----------	---

Description

Data on unemployment and industrial production for Canada, Japan and United States

Usage

```
data("syg_data")
```

Format

A data frame with 0 observations on the following variables.

ca_u Canada (CA) unemployment rate
 jp_u Japan (JP) unemployment rate
 us_u United States (US) unemployment rate
 us_ip Natural log of US industrial production (s.a)
 jp_ip Natural log of JP industrial production (s.a)
 ca_ip Natural log of CA industrial production (s.a)
 time 1981M01 - 2022M07

Details

_U is Unemployment rate (monthly), Total, All persons _IP is index of industrial production, s.a.

Source

<https://stats.oecd.org/index.aspx?lang=en>

Examples

```
data(syg_data)
```

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