Package 'erer'

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Depends R (>= 3.0.0), systemfit, lmtest, tseries, ggplot2, urca
Description This package contains functions and datasets for the book of 'Empirical Research in Economics: Growing up with R' by Dr. Changyou Sun. These functions can calculate marginal effects for a binary probit or logit model, estimate static and dynamic Almost Ideal Demand System (AIDS) models, and conduct event analysis.

License GPL (>= 2)

LazyLoad yes

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Description

This package contains functions and datasets for the book of 'Empirical Research in Economics: Growing up with R' by Dr. Changyou Sun. These functions can calculate marginal effects for a binary probit or logit model, estimate static and dynamic Almost Ideal Demand System (AIDS) models, and conduct event analysis.

Details

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License: GPL (>= 2)

LazyLoad: yes

Author(s)

Changyou Sun <csun@cfr.msstate.edu>

aiData

Transforming Raw Data for Static AIDS Model

Description

This function transforms import values and quantities into a data format that are needed for a static AIDS model.

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Usage

```
aiData(x, label, label.tot = "WD", prefix.value = "v",
   prefix.quant = "q", start = NULL, end = NULL, stone = TRUE,
   dummy = NULL, season = c("none", "m", "q"), ...)
```

Arguments

x raw time series data such as daBedRaw.

label names of supplying countries; this can be as long as needed.

label.tot names of the world total (default label is "WD").

prefix.value prefix for value variables.
prefix.quant prefix for quantity varibles.

start date for the transformed time series; this can be used to select a smaller

window; the default is the start date of the raw data x.

end date for the transformed time series.

stone whether the Stond Price Index is constructed (default TRUE); if FALSE, the

version of log-linear analog to the Paasche index with lagged budget shares in

Moschini (1995) is used. See references for detail.

dummy adding dummy variables if date ranges are provided as a list.

season adding seasonality variables or not; if yes, either monthly dummy or quarterly

dummy; this is prepared mainly for monthly data.

... additional arguments to be passed.

Details

This transforms raw import data into a format needed for a static AIDS model. This separation of data prepraration from model fitting allows greater flexibility in using aiStaFit in estimating a static AIDS model. In addition, when the raw data contain zero, a small number is substituted to avoid NA when the price variable (value/quantity) is calculated.

Value

Return a list object with two components:

out a time series object ready for static AIDS models.

share a time series object of the share data.

price a time series object of the price data.

quantity a time series object of the quantity data.

value a time series object of the value data.

m a vector of the total expenditure.

call a record of the system call; this allows update. default to be used.

Methods

One method is defined as follows:

print: print the first several observations of the final data needed for the AIDS model.

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Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

References

Moschini, G. 1995. Units of measurement and the Stone index in demand system estimation. American Journal of Agricultural Economics 77(1):63-68.

Singh, K., M.M. Dey, and G. Thapa. 2011. An error corrected almost ideal demand system for crustaceans in the United States. Journal of International Food & Agribusiness Marketing 23(3):271-284.

Wan, Y., C. Sun, and D.L. Grebner. 2010. Analysis of import demand for wooden beds in the United States. Journal of Agricultural and Applied Economics 42(4):643-658.

See Also

```
aiStaFit; daBedRaw; daBed.
```

Examples

```
data(daBedRaw)
imp8 <- aiData(x = daBedRaw,
 label = c("CN", "VN", "ID", "MY", "CA", "BR", "IT"),
  label.tot = "WD", prefix.value = "v", prefix.quant = "q",
  start = c(2001, 1), end = c(2008, 12), frequency = 12)
imp8; imp4; imp5
dat8 <- imp8$out
dum <- ts(0, start = start(dat8), end = end(dat8), frequency = 12)</pre>
dum1 < - replace(dum, time(dum) == 2003+(10-1)/12, 1)
dum2 < - replace(dum, time(dum) == 2004+(7 -1)/12, 1)
dum3 \leftarrow replace(dum, time(dum) == 2005+(1 -1)/12, 1)
daTest <- ts.union(dat8, dum1, dum2, dum3)</pre>
colnames(daTest) <- c(colnames(dat8), "dum1", "dum2", "dum3")</pre>
data(daBed)
identical(daBed, daTest)
```

aiDiag

Diagnostic Statitics for Static or Dynamic AIDS Model

Description

Report a set of diagnostic statistics for static or dynamic AIDS models

Usage

```
aiDiag(x, digits = 3, ...)
```

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Arguments

```
x an object of class aiFit from the function of aiStaFit or aiDynFit.digits number of digits used in rounding outputs.additional arguments to be passed.
```

Details

Compute several diagnostic statistics for each equation in a AIDS model. Tests includes are BG, BP, RESET, and JB. See the reference paper for detail.

Value

Return a data frame object with the statistics and p values for the four tests by equation.

Author(s)

```
Changyou Sun (<csun@cfr.msstate.edu>)
```

References

Wan, Y., C. Sun, and D.L. Grebner. 2010. Analysis of import demand for wooden beds in the United States. Journal of Agricultural and Applied Economics 42(4):643-658.

See Also

```
aiStaFit; aiDynFit.
```

Examples

```
# see the examples for 'aiDynFit'.
```

aiDynFit

Fitting a Dynamic AIDS Model

Description

Estimate a dynamic AIDS model for a system.

Usage

```
aiDynFit(w, dum.dif = FALSE, AR1 = FALSE,
  rho.sel = c("all", "mean"), ...)
```

Arguments

W	a object of class aiStaFit.
dum.dif	a logical value (default of FALSE) of whether to take a difference on the dummy variables passed from w.
AR1	whether first-degree autocorrelation should be corrected.
rho.sel	if AR1 $$ = TRUE, there are two ways of computing the autocorrelation coefficient.
	additional arguments to be passed.

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Details

This estimates a dynamic AIDS model. The residuals from the statis AIDS model are included. As it is programmed now, only one lag is allowed for the share variables on the right-hand side. Autocorrelation in the residuals can be corrected following the treatment in Berndt (1975).

Value

Return a list object of class "aiFit" and "aiDynFit" with the following components:

w a object of class aiStaFit.

y data for fitting the static AIDS model, passed down by w.

dum.dif a logical value (default of FALSE) of whether to take a difference on the dummy

variables passed from w.

daDyn data for fitting the dynamic AIDS model.

share names of shares by commodity, used as depedent variables.

price names of prices by commodity, used as independent variables.

expen names of expenditure variable.

shift names of the shifters.

omit names of the omitted share variable.

nOmit position of the omitted share variable in the name of share variable.

hom a logical value of homogeneity test.
sym a logical value of symmetry test.

nShare number of share variables.

nExoge number of exogenous variables (lagged share, residual, expenditure, and shifters).

nParam number of parameters in one equation.

nTotal number of parameters in the whole system estimated.

formula for estimating the system.

res.matrix restriction matrix for hom or sym, or both.

res.rhs right-hand values for tests of hom or sym, or both.

est the dynamic AIDS model estimated.

Methods

One method is defined as follows:

print: print the first several observations of the final data.

Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

References

Berndt, E.R., and N.E. Savin. 1975. Estimation and hypothesis testing in singular equation systems with autoregressive disturbances. Econometrica 43(5/6):937-957.

Wan, Y., C. Sun, and D.L. Grebner. 2010. Analysis of import demand for wooden beds in the United States. Journal of Agricultural and Applied Economics 42(4):643-658.

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See Also

```
systemfitAR; aiStaFit; aiDiag; aiElas; summary.aiFit.
```

```
# --- Step 1: Read data
data(daExp, daBedRaw, daBed)
# --- Step 2: Hausman Test
# 2.1 Getting started with a static AIDS model
sh <- c("sCN", "sVN", "sID", "sMY", "sCA", "sBR", "sIT", "sRW")
pr <- c("lnpCN", "lnpVN", "lnpID", "lnpMY",
         "lnpCA", "lnpBR", "lnpIT", "lnpRW")
du3 <- c("dum1","dum2","dum3")</pre>
rSta <- aiStaFit(y = daBed, share = sh, price = pr, shift = du3,
 expen = "rte", omit = "sRW", hom = TRUE, sym = TRUE)
summary(rSta)
# The following steps should work. It takes about 20 seconds.
## Not run:
# 2.2 The final Hausman test and new data
(dg <- daExp[, "dg"])
rHau <- aiStaHau(x = rSta, instr = dg, choice = FALSE)
names(rHau); colnames(rHau$daHau); colnames(rHau$daFit); rHau
two.exp <- rHau$daFit[, c("rte", "rte.fit")]</pre>
bsStat(two.exp, digits = 4)
plot(data.frame(two.exp)); abline(a = 0, b = 1)
daBedFit <- rHau$daFit</pre>
# --- Step 3: Static and dynamic AIDS models
# 3.1 Diagnostics and coefficients
hSta <- update(rSta, y = daBedFit, expen = "rte.fit")
hSta2 <- update(hSta, hom = FALSE, sym = FALSE)
hSta3 <- update(hSta, hom = FALSE, sym = TRUE)
hSta4 <- update(hSta, hom = TRUE, sym = FALSE)
lrtest(hSta2$est, hSta$est)
lrtest(hSta2$est, hSta3$est)
lrtest(hSta2$est, hSta4$est)
hDyn <- aiDynFit(hSta)
hDyn2 <- aiDynFit(hSta2); Irtest(hDyn2$est, hDyn$est)</pre>
hDyn3 <- aiDynFit(hSta3); lrtest(hDyn2$est, hDyn3$est)
hDyn4 <- aiDynFit(hSta4); lrtest(hDyn2$est, hDyn4$est)</pre>
(table.2 <- rbind(aiDiag(hSta), aiDiag(hDyn)))</pre>
(table.3 <- summary(hSta))</pre>
(table.4 <- summary(hDyn))</pre>
# 3.2 Elasticity calculation
es <- aiElas(hSta); esm <- es$marsh
ed <- aiElas(hDyn); edm <- ed$marsh
esm2 <- data.frame(c(esm[1:2, 2], esm[3:4, 3],</pre>
  esm[5:6, 4], esm[7:8, 5], esm[9:10, 6], esm[11:12, 7],
 esm[13:14, 8], esm[15:16, 9]))
edm2 <- data.frame(c(edm[1:2, 2], edm[3:4, 3],
```

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```
edm[5:6, 4], edm[7:8, 5], edm[9:10, 6], edm[11:12, 7],
edm[13:14, 8], edm[15:16, 9]))
eEM <- cbind(es$expen, esm2, ed$expen[2], edm2)
colnames(eEM) <- c("Country", "LR.expen", "LR.Marsh",
    "SR.expen", "SR.Marsh")
(table.5 <- eEM[-c(15:16),])
(table.6a <- es$hicks[-c(15:16), -9])
(table.6b <- ed$hicks[-c(15:16), -9])</pre>
```

aiElas

Computing Elasticity for Static or Dynamic AIDS Models

Description

Calculate expenditure elasticity, Marshalllian price elasticity, Hicksian price elasticity, and their variances for static or dynamic AIDS Models.

Usage

```
aiElas(z, digits = 3, ...)
```

Arguments

z an object of class aiFit from the function of aiStaFit or aiDynFit.
digits number of digits used in rounding outputs.
... additional arguments to be passed.

Details

Calculate expenditure elasticity, Marshalllian price elasticity, and Hicksian price elasticity for static or dynamic AIDS Models. The related variance, t-ratio, p-value, and significance are also reported.

Value

Return a list object with the following components:

name of the share variables; the omitted share name is the last one.

expen expenditure elasticity and related statistics.

marsh Marshalllian price elasticity and related statistics. hicks Hicksian price elasticity and related statistics.

Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

References

Wan, Y., C. Sun, and D.L. Grebner. 2010. Analysis of import demand for wooden beds in the United States. Journal of Agricultural and Applied Economics 42(4):643-658.

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See Also

```
aiStaFit; aiDynFit.
```

Examples

```
# see the examples for 'aiDynFit'.
```

aiStaFit

Fitting a Static AIDS Model

Description

Estimate a static AIDS model for a system.

Usage

```
aiStaFit(y, share, price, expen, shift = NULL, omit = NULL,
hom = TRUE, sym = TRUE, AR1 = FALSE, rho.sel = c("all", "mean"), ...)
```

Arguments

a multiple time series data.
names of the share variables.
names of the price variables.
name of the expenditure variables.
names of the shifter variables.
name of the share variable omitted; if not supplied, this is the last one of share.
a logical value of homogeneity test.
a logical value of symmetry test.
whether first-degree autocorrelation should be corrected.
if AR1 = TRUE, there are two ways of computing the autocorrelation coefficient.
additional arguments to be passed.

Details

This estimates a static AIDS model. The data supplied should be in the final format. Autocorrelation in the residuals can be corrected following the treatment in Berndt (1975).

Value

Return a list object of class "aiFit" and "aiStaFit" with the following components:

у	data for fitting the static AIDS model.
share	names of the share variables.
price	names of the price variables.
expen	name of the expenditure variables.
shift	names of the shifter variables.

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omit name of the share variable omitted; if not supplied, this is the last one of share.

nOmit position of the omitted share variable in the name of share variable.

hom a logical value of homogeneity test.

sym a logical value of symmetry test.

nShare number of share variables.

nExoge number of exogenous variables (lagged share, residual, expenditure, and shifters).

nParam number of parameters in one equation.

nTotal number of parameters in the whole system estimated.

formula for estimating the system.

res.matrix restriction matrix for hom or sym, or both.

res.rhs right-hand values for tests of hom or sym, or both.

est the static AIDS model estimated.

AR1 a logical value whether autocorrelation is corrected.

call a record of the system call; this allows update.default to be used.

Methods

One method is defined as follows. This is the print method related to three functions: aiStaFit, aiDynFit, and aiStaHau.

print: print the first several observations of selectec outputs.

Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

References

Berndt, E.R., and N.E. Savin. 1975. Estimation and hypothesis testing in singular equation systems with autoregressive disturbances. Econometrica 43(5/6):937-957.

Wan, Y., C. Sun, and D.L. Grebner. 2010. Analysis of import demand for wooden beds in the United States. Journal of Agricultural and Applied Economics 42(4):643-658.

See Also

```
aiDiag; aiElas; summary.aiFit; aiDynFit; aiStaHau; systemfitAR.
```

```
# see the examples for 'aiDynFit'.
```

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aiStaHau Conducting a Hausman Test on a Static AIDS Model

Description

Conduct a Hausman test on a static AIDS model and report the result of likelihood ratio test.

Usage

```
aiStaHau(x, instr, choice = FALSE, ...)
```

Arguments

x an object of class aiStaFit from a static AIDS model.

instr a single time series data as instrument for the expenditure variable in AIDS model.

choice a logical value of whether to take a difference on the right-hand price and instr

variables.

... additional arguments to be passed.

Details

Conduct a Hausman test on a static AIDS model and report the result of likelihood ratio test. Note that logarithm is taken on every variable in the auxiliary regression. These variables are the real total expenditure and its lagged value, instrumental variable, and the price variables.

Value

Return a data frame object with the statistics and p values for the four tests by equation.

daHau data used in estimating the Hausman test. formulau for estimating the Hausman test.

regHau regression for the Hausman test.

daFit revised data with the fitted value of expenditure included.

aiBase the base static AIDs model estimated.

aiHaus the reestimated static AIDS model using the fitted value of expenditure.

ratio result of the likelihood ration test for the Hausman test.

Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

References

Wan, Y., C. Sun, and D.L. Grebner. 2010. Analysis of import demand for wooden beds in the United States. Journal of Agricultural and Applied Economics 42(4):643-658.

See Also

```
aiStaFit; print.aiFit.
```

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Examples

```
\mbox{\tt\#} see the examples for 'aiDynFit'.
```

bsFormu

Generating Formula for Models

Description

Generate a single formula for models like 1m or a list of formula for models like systemfit.

Usage

```
bsFormu(name.y, name.x, intercept = TRUE, ...)
```

Arguments

name.y	a character vector of variables names for dependent variables; when the length is more than one, there will a list of formula generated for each variable in the name.
name.x	a character vector of indepedent variables.
intercept	a logical value (default of TRUE) of whether to include intercept or not.
	additional arguments to be passed.

Details

This function can generate a single formula for simple model like 1m or a list of formula for systems (systemfit. Note that the right-hand side variables are the same for each dependent variable. If different, a for loop can be added by users to address that, as demonstrated by the example below.

Value

a single formula object or a list of formula objects.

Author(s)

 $Changyou \ Sun \ (\verb|<csun@cfr.msstate.edu>|)$

```
# fake data
y <- c("y")
ym <- c("y1", "y2", "y3")
x <- c("x")
xAll <- c("x", "xx", "xxx", "xxxx")

bsFormu(name.y = y, name.x = x)
bsFormu(name.y = ym, name.x = xAll)
fm.ym <- bsFormu(name.y = ym, name.x = xAll, intercept = FALSE)
fm.ym

# If independent variables differ by equation,</pre>
```

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```
# add a loop to address the differentiation.
xInd <- c("x1", "x2", "x3")
fm.ym <- list()</pre>
for (i in 1:length(ym)) {
    ny \leftarrow ym[i]
    nx <- c(xInd[i], xAll)</pre>
    fm.ym[[i]] <- bsFormu(name.y = ny, name.x = nx, intercept = FALSE)</pre>
}
fm.ym
# real data
data(daIns)
(xx <- colnames(daIns)[-c(1, 14)])
fm.ins <- bsFormu(name.y = "Y", name.x = xx, intercept = TRUE)</pre>
fm.ins
(ra <- glm(formula = fm.ins,</pre>
           family = binomial(link="logit"),
           data = daIns, x = TRUE)
```

bsLag

Lagged Time Series

Description

Generate a set of lagged time series for time series data.

Usage

```
bsLag(h, lag, prefix = "", var.name, suffix = ".t_",
    include.orig = TRUE, by.lag = FALSE, ...)
```

Arguments

h	time series data
lag	number of lags
prefix	prefix for the name of lagged time series.
var.name	varible name of the lagged time series.
suffix	suffix of the name of lagged time series.
include.orig	logical value (default of TRUE) of whether to include the original series (i.e., lag zero) in the final output.
by.lag	logical value (default of FALSE) of whether to order the column by variable (FALSE) or by lag (TRUE).
	additional arguments to be passed.

Details

The input data can be a single time series or a set of multiple time series data. The output is a set of lagged time series with the specified lag dimension. All the series are aligned with the shortest window so the loss of observations is equal to lag. The original series (e.g., without lag but just loss of beginning observations) can be included or excluded by setting the logical value of include.org.

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The name of the output data is composed of four parts: prefix, var.name, suffix, and an index number of lag. Users can control the first three parts only because the lag number is added automatically. prefix and suffix can be fixed for all the output series. var.name provides some flexibility when bsLag is used within a function and the variable name is unknown *a priori*.

The column of the output can be ordered either by the variable name (e.g., diff.GA.t_0, diff.GA.t_1, diff.ND.t_0, diff.ND.t_1), or by the lag order ((e.g., diff.GA.t_0, diff.ND.t_0, diff.ND.t_1).

Value

Return a multiple time series object.

Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

```
# simple example
h1 <- ts(data=cbind(1:24), start=c(2001, 1), frequency=12)
h2 <- ts(data=cbind(1:24, 25:48), start=c(2001, 1), frequency=12)
h3 \leftarrow ts(data=cbind(1:4, 5:8, 9:12), start=c(2001, 1), frequency=4)
colnames(h2) <- c("aa", "bb")</pre>
colnames(h3) <- c("cc", "dd", "ee")</pre>
h1; h2; h3
bsLag(h=h1, lag=0, prefix="", suffix=".t_")
bsLag(h=h1, lag=2, prefix="price.", var.name="fl", suffix=".t_")
bsLag(h=h1, lag=2, prefix="price.", var.name="fl", suffix=".t_", by.lag=TRUE)
bsLag(h=h1,\ lag=23,\ prefix="price.",\ suffix=".t\_",\ include.orig=FALSE)
bsLag(h=h2, lag=4, prefix="", suffix=".t_", include.orig = TRUE)
bsLag(h=h2, lag=4, prefix="", suffix=".t_", include.orig = FALSE)
bsLag(h=h2, lag=4, prefix="", suffix=".t_", include.orig = FALSE, by.lag=TRUE)
bsLag(h=h2, lag=0, prefix="", var.name=c("nc", "sc"), suffix=".t_")
bsLag(h=h3, lag=2, prefix="", suffix=".t_", include.orig=FALSE)
bsLag(h=h3, lag=1, prefix="", var.name=c("nd", "sd", "mi"), suffix=".lag.")
bsLag(h=h3, lag=2, prefix="NY.", suffix=".t_", by.lag=TRUE)
bsLag(h=h3, lag=3, prefix="NY.", suffix=".t_", include.orig=FALSE)
# with real data
data(daBedRaw)
small <- daBedRaw[, c("vCN", "qCN")]</pre>
(lag.small <- bsLag(h=small, lag=4))</pre>
colnames(lag.small)
resid <- residuals(lm(qCN ~ vCN, data = small))</pre>
res <- ts(resid, start=start(small), end=end(small),</pre>
    frequency=tsp(small)[3])
lag.res <- bsLag(h=res, lag=2, prefix="resid.", var.name="china")</pre>
str(lag.res)
head(lag.res)
tail(lag.res)
```

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bsStat	Summary of Basic Statistics

Description

Calcluate basic statistics of data.

Usage

```
bsStat(y, two = NULL, digits = c(2, 2), use = 'complete.obs',
    na.rm = TRUE, ...)
```

Arguments

У	input data for summary statistics.
two	a logical value of whether to report the correlation and summary statistics separately; if NULL and the number of variables is less than 11, its value will be set to TRUE.
digits	digits for the output data, one for correlation coefficients and the other for mean and others; if a single scalar is supplied, it will be used for both.
use	an argument for correlation coefficient; see cor for detail.
na.rm	an argument for mean, sd, min, and max.
	additional arguments to be passed.

Details

Two set of summary statistics are generated. One is correlation coefficients and the other is mean, minimum, maximum, standard deviation, and number of observations. When two is unspecified and the number of variables is bigger than ten, the two sets are reported separately; otherwise, it is reported as a single data frame object.

Value

A dataframe or list of the summary statistics.

Author(s)

```
Changyou Sun (<csun@cfr.msstate.edu>)
```

```
data(daIns)
(sum.daIns <- bsStat(y=daIns, digits=c(3,2)))</pre>
```

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bsTab

Generating Pretty Statistical Tables

Description

Format statistics from regressions into pretty outputs

Usage

```
bsTab(w,
  need = c("1T", "1E", "2T", "2E", "3T", "3E", "4T", "4E", "5"),
  wrap.TE = c("(", "", "["),
  add.sig = c("coef", "TE"),
  percent = c(0.01, 0.05, 0.10),
  symbol = c("***", "**", "*", ""),
  digits = c(3, 3, 3, 2), ...)
```

Arguments

W	statistical results from regression models; an object of class glm, lm, and systemfit can be supplied directly, or a data frame with at least four columns with the sequence of estimates, errors, t-values, and p-values.
need	a choice of output formats; default of 1T is one column with t ratio and significance symbols; 1 to 5 is the number of columns; T is t ratios; E is standard errors. This argument must be a character string.
wrap.TE	parentheses, none, or brackets can be used to enclose t ratios or standard errors; default value is parenthese for one-column format and none for other formats.
add.sig	a chacter string to indicate where to add the significance symbol, either to the coefficents ("coef") or the t-value and error ("TE").
percent	percentage values used to categorize p values.
symbol	symbols used to represent p-value categories; the default values can be changed to symbols like a, b, c, or different combinations of *.
digits	digits for outputs; the default values are 3, 3, 3, and 2 for estimate, error, t value, and p value, correspondingly. A single value like 4 can be supplied and it will be recycled for all of them.
	additional arguments to be passed.

Details

Format statistics from regressions into tables that are often reported in economic journals. The column of 'Variable' in the outuput is the row names of the input data so the raw data should contain meaningful rownames. Besides the variable name column, the maximum number of output is five columns: estimate, error, t ratio, p value, and significance. wrap. TE and add. sig are only valid for column widths of 1 and 2.

Value

A dataframe of statistical results.

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Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

Examples

```
# a simulated data
tes <- data.frame(est = c(4, 56, 12), err = c(0.3, 0.56, 0.789),
  t.rat = c(2.56, 7.9, 1.2), p.val = c(0.002, 0.23, 0.061))
bsTab(tes)
bsTab(w = tes, need = "2E")
# real data
data(daIns)
ra <- glm(formula = Y ~ Injury + HuntYrs + Nonres +</pre>
                    Lspman + Lnong + Gender + Age +
                    Race + Marital + Edu + Inc + TownPop,
          family = binomial(link="logit"),
          data = daIns, x = TRUE)
(ca <- data.frame(summary(ra)$coefficients))</pre>
# an object of class 'glm' as input
bsTab(w = ra, add.sig = "TE")
bsTab(w = ra, wrap.TE = "[")
bsTab(w = ra, need = "5")
bsTab(w = ra, need = "4T", wrap.TE = "[")
final <- bsTab(w = ra, need = "3T",</pre>
  percent = c(0.01, 0.05, 0.10),
  symbol = c("a", "b", "c", ""), digits = 4)
print(final, right = FALSE)
# any matrix with at least four columns can be supplied
cbind(bsTab(ca), bsTab(ra))
```

daBed

Transformed Wooden Beds Import Data for Static AIDS Models

Description

This data set contains transformed values related to wooden beds imports by the United States from January 2001 to December 2008. There are 96 observations and 20 variables.

```
sCN
         monthly import share of wooden beds from China
sVN
         monthly import share of wooden beds from Vietnam
         monthly import share of wooden beds from Indonesia
sID
sMY
         monthly import share of wooden beds from Malaysia
         monthly import share of wooden beds from Canada
sCA
sBR
         monthly import share of wooden beds from Brazil
         monthly import share of wooden beds from Italy
sIT
sRW
         monthly import share of wooden beds from the rest of world
         real total expenditure in logarithm
rte
```

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lnpCN	monthly import price of wooden beds from China in logarithm
inpVN	monthly import price of wooden beds from Vietnam in logarithm
lnpID	monthly import price of wooden beds from Indonesia in logarithm
lnpMY	monthly import price of wooden beds from Malaysia in logarithm
lnpCA	monthly import price of wooden beds from Canada in logarithm
inpBR	monthly import price of wooden beds from Brazil in logarithm
lnpIT	monthly import price of wooden beds from Italy in logarithm
lnpRW	monthly import price of wooden beds from the rest of world in logarithm
dum1	a pulse dummy variable (1 for October 2003, 0 otherwise)
dum2	a pulse dummy variable (1 for July 2004, 0 otherwise)
dum3	a pulse dummy variable (1 for January 2005, 0 otherwise)

Usage

```
data(daBed)
```

Format

Monthly time series from January 2001 to December 2008 with 96 observations for each of the 20 variables.

Details

This is the transformated data set for static AIDS model. The transformation detail is described in Wan et al. (2010).

Source

U.S. ITC, 2010. Interactive tariff and trade data web. http://dataweb.usitc.gov (Assecced on March 1, 2010).

References

Wan, Y., C. Sun, and D.L. Grebner. 2010. Analysis of import demand for wooden beds in the United States. Journal of Agricultural and Applied Economics 42(4):643-658.

See Also

```
aiStaFit; daBedRaw.
```

```
data(daBed)
class(daBed); dim(daBed); colnames(daBed)
daBed
```

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daBedRaw	Wooden Beds Import Data	

Description

This data set contains a mulitiple time series related to wooden beds imports by the United States. The time covered is January 1996 to December 2008 with 156 observations. There are 34 variables in total: 17 import values (dollars) and 17 import quantities (dollars / piece). In total, 16 countries are covered and the world total is also reported.

```
cost-insurance-freight import values in dollar from Brazil
vBR
        cost-insurance-freight import values in dollar from Canada
vCA
vCN
        cost-insurance-freight import values in dollar from China
vDK
        cost-insurance-freight import values in dollar from Denmark
vFR
        cost-insurance-freight import values in dollar from France
νHK
        cost-insurance-freight import values in dollar from Hong Kong
νIA
        cost-insurance-freight import values in dollar from India
vID
        cost-insurance-freight import values in dollar from Indonesia
νIT
        cost-insurance-freight import values in dollar from Italy
vMY
        cost-insurance-freight import values in dollar from Malaysia
        cost-insurance-freight import values in dollar from Mexico
vMX
vPH
        cost-insurance-freight import values in dollar from Philippines
vTW
        cost-insurance-freight import values in dollar from Taiwan
νTΗ
        cost-insurance-freight import values in dollar from Thailand
vUK
        cost-insurance-freight import values in dollar from United Kingdom
vVN
        cost-insurance-freight import values in dollar from Vietnam
vWD
        cost-insurance-freight import values in dollar from World in total
qBR
        quantity in piece from Brazil
qCA
        quantity in piece from Canada
aCN
        quantity in piece from China
qDK
        quantity in piece from Denmark
qFR
        quantity in piece from France
qHK
        quantity in piece from Hong Kong
        quantity in piece from India
qIA
qID
        quantity in piece from Indonesia
        quantity in piece from Italy
qIT
qMY
        quantity in piece from Malaysia
qMX
        quantity in piece from Mexico
qPH
        quantity in piece from Philippines
qTW
        quantity in piece from Taiwan
qTH
        quantity in piece from Thailand
        quantity in piece from United Kingdom
aUK
aVN
        quantity in piece from Vietnam
qWD
        quantity in piece from World in total
```

Usage

data(daBedRaw)

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Format

Monthly time series from January 1996 to December 2008 with 156 observations for each of the 34 variables.

Details

Under the Harmonized Tariff Schedule (HTS) system, the commodity of wooden beds is classified as HTS 9403.50.9040. The monthly cost-insurance-freight values in dollar and quantities in piece are reported by country from U.S. ITC (2010).

Source

U.S. ITC, 2010. Interactive tariff and trade data web. http://dataweb.usitc.gov (Assecced on March 1, 2010).

References

Wan, Y., C. Sun, and D.L. Grebner. 2010. Analysis of import demand for wooden beds in the United States. Journal of Agricultural and Applied Economics 42(4):643-658.

See Also

```
aiStaFit; daBed.
```

Examples

```
data(daBedRaw)
class(daBedRaw); dim(daBedRaw); colnames(daBedRaw)
```

daEsa

Daily returns of 14 public firms from 1990 to 2004

Description

This data set contains daily returns of 14 public firms, three-month treasury bill, and SP 500 Index from 1990 to 2004.

```
Eight-digit numbers for date of 1990-2004; the format is YYMMDD, e.g., 19900102 for Jan 2, 1990
date
tb3m
         Daily returns for three-month treasury bills
sp500
         Daily returns for SP 500 Index
         Daily returns for Boise Cascade
bcc
bow
         Daily returns for Bowater
         Daily returns for Chesapeake Corp VA
csk
         Daily returns for Georgia-Pacific
gp
         Daily returns for International Paper
ip
kmb
         Daily returns for Kimberly Clark
         Daily returns for Louisiana Pacific
1px
         Daily returns for MeadWestvaco
mwv
pch
         Daily returns for Potlatch
pcl
         Daily returns for Plum Creek
         Daily returns for Pope and Talbot
pop
         Daily returns for Temple Inland
tin
```

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```
wpp Daily returns for Wausau Mosinee Paperwy Daily returns for Weyerhaeuser
```

Usage

```
data(daEsa)
```

Format

A data frame object with daily returns for firms or indexes from 1994 to 2004. There are 17 columns and 3747 rows. The date is not regular because there is no trading on weekends and holidays. Therefore, the date is represented by a number, not a date.

Details

This is the transformated data set used in the study of Sun and Liao (2011).

Source

The daily returns for SP 500 and individual firms are from the database of the Center for Research in Security Prices (CRSP). The risk-free rate of return is the secondary market rate for the 3-month US Treasury bills from the Federal Reserve Bank.

References

Sun, C., and X. Liao. 2011. Effects of litigation under the Endangered Species Act on forest firm values. Journal of Forest Economics 17(4):388-398.

See Also

```
evReturn; evRisk.
```

Examples

```
data(daEsa)
dim(daEsa); colnames(daEsa)
head(daEsa); tail(daEsa)
str(daEsa)

# if dates are stored as a date object in R, then it can be converted into
# numbers as the following example shows.

raw <- as.Date(c('1990-01-02', '1991-11-12')); raw; str(raw)
raw2 <- as.numeric(strftime(raw, format = "%Y%m%d")); raw2; str(raw2)</pre>
```

daExp

Expenditure Data for a Hausman Test in AIDS Model

Description

This data set contains seven monthly times series for expenditure from 2001 to 2008.

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pinc	Billions of dollars, personal income
dpi	Billions of dollars, disposable personal income
pce	Billions of dollars, personal consumption expenditures
dg	Billions of dollars, Personal consumption expenditures for durable goods
rdpi	Billions of dollars, real disposable personal income
rpce	Billions of dollars, real personal consumption expenditures
rdg	Billions of dollars, real personal consumption expenditures for durable goods

Usage

```
data(daExp)
```

Format

Monthly time series from January 2001 to December 2008 with 96 observations for each of the seven variables.

Details

This is the data set for conducting a Hausman test in a static AIDS model, as detailed in Wan et al. (2010). The test focuses on whether the expenditure variable in a AIDS model is exogenous or not. Each of the seven expenditure data can be used as an instrumental variable in an auxiliary regression.

Source

Federal Reserve Bank of St. Louis. Economic Data - Fred. Internet site: http://stlouisfed.org (Accessed February 25, 2010).

References

Wan, Y., C. Sun, and D.L. Grebner. 2010. Analysis of import demand for wooden beds in the United States. Journal of Agricultural and Applied Economics 42(4):643-658.

Examples

```
data(daExp)
class(daExp); dim(daExp); colnames(daExp)
daExp
```

daIns

Liability Insurance Coverage for Hunters and Anglers in Mississippi

Description

This data set contains a survey result about liability insurance purchase decision by hunters and anglers in Mississippi. There are 1653 observations for 14 variables.

Binary dependent variable = 1 if had liability insurance; 0 otherwise Injury
Times of bodily injuries or property damages in the past three years
Years of hunting

daIns 23

Nonres Dummy = 1 if nonresidents; 0 if Mississippi residents Dummy = 1 if purchased the license of resident sportsman; 0 otherwise Lspman Dummy = 1 if purchased the license of nonresident all game; 0 otherwise Lnong Gender Dummy = 1 if male; 0 otherwise Age of the hunter or angler Age Dummy = 1 if Caucasian; 0 otherwise Race Dummy = 1 if married; 0 otherwise Marital Years of education Edu Household income in 2004 (1,000 dollars) Population size of the residence town (1,000) TownPop FishYrs Years of fishing

Usage

```
data(daIns)
```

Format

A cross sectional data with 1653 observations and 14 variables.

Details

The data set is from a telephone survey conducted in 2005 in Mississippi.

Source

Sun, C., S. Pokharel, W.D. Jones, S.C. Grado, and D.L. Grebner. 2007. Extent of recreational incidents and determinants of liability insurance coverage for hunters and anglers in Mississippi. Southern Journal of Applied Forestry 31(3):151-158.

24 daLaw

daLaw	Data for statutory reforms and retention of prescribed fire liability laws on forest land by state

Description

This data set contains a cross-sectional data set for current adoption of statutory laws with regard to prescribed fire liability on forest landowners. It has 50 observations and 16 variables.

SHORT	Two-letter abbreviations of 50 state names
STATE	Full state names
Υ	Categorical dependent variable $(Y = 0, 1, 2, or 3)$
FYNFS	National Forests area in a state (million acres)
FYIND	Industrial forest land area in a state (million acres)
FYNIP	Nonindustrial private forest land area in a state (million acres)
AGEN	Permanent forestry program personnel in a state
POPRUR	Rural population in a state (million)
EDU	Population 25 years and older with advanced degrees in a state (million)
INC	Per capita income in a state (thousand dollars)
DAY	The maximum length of legislative sessions in calendar days in a state
BIANN	A dummy variable equal to one for states with annual legislative sessions, zero with biannual (or less)
SEAT	Total number of legislative seats (Senate plus House) in the legislative body in a state
BICAM	Level of bicameralism in a state, defined as the size of the Senate divided by the size of the House
COMIT	Total number of standing committees in a state
RATIO	Total number of standing committees in a state divided by the number of legislators

Usage

```
data(daLaw)
```

Format

A data frame object with 50 rows and 16 variables.

Details

This is the final data set used in the study of Sun (2006).

Source

See Table 2 in Sun (2006) for detail of data sources.

References

Sun, C. 2006. State statutory reforms and retention of prescribed fire liability laws on US forest land. Forest Policy and Economics 9(4):392-402.

```
data(daLaw)
str(daLaw); head(daLaw); names(daLaw)
```

daRoll 25

daPe	Program Effectiveness of a New Method of Teaching Economics
daPe	Program Effectiveness of a New Method of Teaching Economics

Description

This data set contains the evaluation results of a new program of teaching in economics. There are 32 observations for 4 variables.

grade a binary variable indicating grade increase (1) and decrease (0) after participation.
gpa a continous variable measuring studens' grade point average.
tuce a continous variable measuring students' scores on an economics test.
psi a binary variable indicating whether a student participates the program or not.

Usage

```
data(daPe)
```

Format

A data frame of cross sectional data with 32 observations and 4 variables.

Details

Evaluation results on 32 students of the impact of a new teaching methods.

Source

Spector, L.C., and M. Mazzeo. 1980. Probit analysis and economic education. Journal of Economic Education 11(2):37-44.

Examples

```
data(daPe)
dim(daPe)
summary(daPe)
daPe
```

daRoll

Voting records for the Healthy Forests Restoration Act in 2003 and the associated characteristics of congressmen

Description

This data set contains the voting records for the Healthy Forests Restoration Act in 2003, as used in Sun (2006). The characteristics of individual congressmen are also included. There are 537 observations and 22 variables.

```
state state name for a congressman district district for a congressman; 0 for senators
```

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name Family name of a congressman

voteMay voting record in May 2003 in the House; 1 if yes, 0 if no, and NA if not voted

voteNov voting record in Nov 2003 in both the House and Senate

RepParty Dummy equals one if Republican

East Regional dummy for 11 northeastern states
West Regional dummy for 11 western states
South Regional dummy for 13 southern states
PopDen Population density - 1000 persons per km2

PopRural Population density per km2

Edu Percentage of population over 25 with a Bachelor's degree

Income Median family income (\$1,000)

FYland Percentage of federal lands in total forestlands 2002

Size Value of shipments of forest industry 1997 (million dollars)

ContrFY Contribution from forest firms (1,000 dollars)

ContrEN Contribution from environmental groups (1,000 dollars)

Sex Dummy equals one if male Lawyer Dummy equals one if lawyer

Member Dummy equals one if a committee member for the HFRA

Year Number of years in the position

Chamber Dummy equals one if House and zero if Senate

Usage

```
data(daRoll)
```

Format

A data frame object with 537 rows and 22 variables. This is a cross-sectional dataset that are generating from merging several raw datasets.

Details

This is the combinded final data set used in the study of Sun (2006).

Source

See Table 1 in Sun (2006) for detail.

References

Sun, C. 2006. A roll call analysis of the Healthy Forests Restoration Act and constituent interests in fire policy. Forest Policy and Economics 9(2):126-138.

See Also

```
glm; maBina.
```

```
# generate four datasets used in Sun (2006)
data(daRoll)
xn <- c('RepParty', 'East', 'West', 'South', 'PopDen',</pre>
```

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```
'PopRural', 'Edu', 'Income', 'Fyland', 'Size',
    'ContrFY', 'ContrEN', 'Sex', 'Lawyer', 'Member', 'Year', 'Chamber')

f1 <- daRoll[!is.na(daRoll$voteMay), c('voteMay', xn)]

f2 <- daRoll[!is.na(daRoll$voteNov) & daRoll$Chamber == 1, c('voteNov', xn)]

f3 <- daRoll[!is.na(daRoll$voteNov), c('voteNov', xn)]

f4 <- daRoll[!is.na(daRoll$voteNov) & daRoll$RepParty == 0, c('voteNov', xn)]

rownames(f1) <- 1:nrow(f1); rownames(f2) <- 1:nrow(f2)

rownames(f3) <- 1:nrow(f3); rownames(f4) <- 1:nrow(f4)

colnames(f1)[1] <- colnames(f2)[1] <- 'Vote'

colnames(f3)[1] <- colnames(f4)[1] <- 'Vote'

dim(f1); dim(f2); dim(f3); dim(f4)

tail(f3)
```

download.lib

Downloading source and PDF version of R packages

Description

This function can download three documents for each package: zip version, source version (tar.gz), and PDF manual.

Usage

```
download.lib(pkgs, destdir = getwd(), mode = "wb",
   pdf.url = "http://cran.r-project.org/web/packages/",
   f.zip = TRUE, f.pdf = TRUE)
```

Arguments

pkgs	A character vector for the names of one or multiple packages
destdir	The directory where documents are saved (e.g., C:/myFile; the default directory is the current working directory.
mode	The argument for download.file; the default is 'wb' for PDF version.
pdf.url	The url for PDF documents online.
f.zip	A logical value indicating whether a zip version is downloaded; by default, a zip version is downloaded by download.packages
f.pdf	A logical value indicating whether a PDF mannual for a package should be downloaded from the internet.

Details

This function is mainly used to download the source version of one or multiple packages. A zip version and a PDF manual can also be downloaded at the same time.

Value

Return the packge names.

Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

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See Also

```
download.packages; download.file.
```

Examples

```
## Not run:
  download.lib(pkgs = 'erer', destdir = 'c:/aErer/Rcode',
    f.zip = FALSE, f.pdf = TRUE)
## End(Not run)
```

evReturn

Estimating Abnormal Return from Event Analysis

Description

Conduct an event analysis and estimate abnormal returns over time and across firms.

Usage

```
evReturn(y, firm, event.date, y.date = "date",
  index = "sp500", event.win = 3, est.win = 250, digits = 4, ...)
```

Arguments

У	a data frame object with one column for date, return series by firms, a return series for a stock market index, and a return series for a risk free asset.
firm	a character vector of firm names; this is the name of the return series in y.
event.date	event dates for each firm as specified in firm; this should be a numerical vector and can match the values in y\$y.date; if event dates are the same for all the firms, this can be specificed as a single number.
y.date	a character value for the column name of date in y.
index	a character value for the column name of index in y.
event.win	the one-side width of event window in days; the default value of 3 corresponds to a 7-day window (i.e., $3 + 1 + 3$).
est.win	the width of estimation window in days.
digits	number of digits used to format outputs.
	additional arguments to be passed.

Details

This is the core function for event analysis. It estimates a market model by firm and then calculate abnormal returns by firm and over time. The time series of stock returns have irregular time frequency because of varying trading days. Thus, the time dimension is explicitly specified as a y.date column in the data of y.

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Value

Return a list object of class "evReturn" with the following components:

y a data frame of raw return data.

y.date a character value for the column name of date in y..

firm a character vector of firm names.

N the number of firms.

index a character value for the column name of index in y.

event.date event dates for each firm as specified in firm.

event.win the one-side width of event window in days.

event.width total number of days in an event window.
est.win the width of estimation window in days...

data used to estimate the market model for the last firm as specified in codefirm.

daEve data over the event window for the last firm.

ra fitted market model for the last firm.

digits number of digits used to format outputs.

reg regression coefficients by firm.

abr abnormal returns by day over the event window and by firm.

abc average abnormal returns across firms.

call a record of the system call; this allows update.default to be used.

Methods

Two methods are defined as follows:

print: print three selected outputs.

plot: plot average cumulative abnormal returns from event analysis versus days in event window.

Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

References

Mei, B., and C. Sun. 2008. Event analysis of the impact of mergers and acquisitions on the financial performance of the U.S. forest products industry. Forest Policy and Economics 10(5):286-294.

Sun, C., and X. Liao. 2011. Effects of litigation under the Endangered Species Act on forest firm values. Journal of Forest Economics 17(4):388-398.

See Also

evRisk

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Examples

```
data(daEsa)

# event analysis for one firm and one event window
hh <- evReturn(y = daEsa, firm = "wpp",
    y.date = "date", index = "sp500", est.win = 250, digits = 3,
    event.date = 19990505, event.win = 5)
hh; plot(hh)

# event analysis for many firms and one event window
hh2 <- update(hh, firm = c("tin", "wy", "pcl", "pch")); hh2
# event analysis for many firms and many event windows: need a for loop</pre>
```

evRisk

Risk Evaluation for Event Analysis

Description

Conduct a risk analysis by firm and evaluate the change of risk before and after an event. The model used is the Captial Asset Pricing Model.

Usage

```
evRisk(x, m = 50, r.free = "tbill", ...)
```

Arguments

x a object from evReturn.
 m the number of days before and after the event date for estimating CAPM.
 r. free the column name of risk free asset in y.
 ... additional arguments to be passed.

Details

This fits CAPM for each firm and reports the statistics for alpha, beta, and gamma. The statistics of gamma reveal the change of risk before and after the event.

Value

Return a list object of class "evReturn" with the following components:

x a object from evReturn.

daEst data used to estimate CAPM for the last firm as specified in codefirm.

rb fitted CAPM for the last firm.

reg regression coefficients by firm.

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Methods

One method is defined as follows:

```
print: print selected outputs.
```

Author(s)

```
Changyou Sun (<csun@cfr.msstate.edu>)
```

References

Mei, B., and C. Sun. 2008. Event analysis of the impact of mergers and acquisitions on the financial performance of the U.S. forest products industry. Forest Policy and Economics 10(5):286-294.

See Also

evReturn

Examples

```
data(daEsa)

hh <- evReturn(y = daEsa, firm = "wpp",
    y.date = "date", index = "sp500", est.win = 250, digits = 3,
    event.date = 19990505, event.win = 5)
hh2 <- update(hh, firm = c("tin", "wy", "pcl", "pch"))

kk <- evRisk(x = hh2, m = 100, r.free="tb3m")
kk</pre>
```

head

Return the first or last part of time series data

Description

Return the first of last parts of an object of time series data.

Usage

```
## S3 method for class 'ts' head(x, n = 5, ...)
## S3 method for class 'ts' tail(x, n = 5, ...)
```

Arguments

```
x input time seires data.n a single integer for the length or row of returned data... additional arguments to be passed.
```

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Details

The data can be an univariate or multivariate time series data.

Value

An object like x but generally smaller.

Author(s)

```
Changyou Sun (<csun@cfr.msstate.edu>)
```

Examples

```
h1 <- ts(data=cbind(1:24), start=c(2001, 1), frequency=12)
h2 <- ts(data=cbind(1:24, 25:48), start=c(2001, 1), frequency=12)
h3 <- ts(data=cbind(1:4, 5:8, 9:12), start=c(2001, 1), frequency=4)
colnames(h2) <- c("aa", "bb")
colnames(h3) <- c("cc", "dd", "ee")
h1; h2; h3

h1; head(h1); tail(h1, 28)
h2; head(h2); tail(h2, 50)
h3; head(h3, 2); tail(h3); tail(h3, 8)

data(daBed); head(daBed); tail(daBed)
```

listn

Generate a list object with names

Description

Generate a list object with names.

Usage

```
listn(...)
```

Arguments

... individual objects to be included in a list.

Details

This generates a list object by addressing the naming problem. For list, if no names are given, the list generated will have no names. In some situations, the number of individual objects is large and the names of these individual objects can be used as the names. This function addresses this need. If names are given, they will be used. If not, the names of individual objects will be used.

Value

Return a list object with names.

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Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

Examples

```
y1 <- 1:10
y2 <- c("a", "b")
listn(y1, y2)
listn(y1 = y1, y2)
listn(y1 = y1, y2.rev = y2, y2, 5:8, c("d", "f"))

identical(listn(y1, y2), listn(y1 = y1, y2))  # TRUE
identical(listn(y1, y2), list(y1 = y1, y2))  # FALSE
identical(listn(y1, y2), list(y1 = y1, y2=y2))  # TRUE</pre>
```

lss

List Objects with Their Sizes

Description

Show the name and memory size of objects in an R session.

Usage

```
lss(n = 5, pos = 1, decreasing = TRUE, order.by=c("Size", "Type"))
```

Arguments

```
n number of objects to show.
pos specifying the environment; see 1s.
decreasing sorting order.
```

order.by sorting variable.

Details

This function shows the names and sizes of objects in an R session. This is useful for managing avaiable memory in an R Session.

Value

Return a dataframe object with the following columns: name, type, size, prettysize, rows, and columns.

See Also

1s.

```
lss()
ls()
```

34 maBina

maBina	Marginal Effect for Binary Probit and Logit Model	

Description

This function calculates marginal effects for a binary probit or logit model and their standard errors.

Usage

```
maBina(w, x.mean = TRUE, rev.dum = TRUE, digits = 3, ...)
```

Arguments

W	a binary probit or logit model object estimated from glm().
x.mean	a logical value (default of TRUE) of whether to calculate marginal effects at the means of independent variables. If FALSE, marginal effects are calculated for each observation and then averaged.
rev.dum	a logical value (default of TRUE) of whether to revise the estimates and standard erros for binary independant variables. If FALSE, derivatives are taken on binary independant variables as continuous variables.
digits	number of digits for output.
	additional arguments to be passed.

Details

Marginal effects from a binary probit or logit model is calculated. The two choices are the method of averaging effects and revising estimates for dummy variables. Marginal effects can be calculated at the mean of the independent variables (i.e., x.mean = TRUE), or as the average of individual marginal effects at each observation (i.e., x.mean = FALSE). rev.dum = TRUE allows marginal effects for dummy variables are calculated differently, instead of treating them as continuous variables.

Value

Return a list object of class "maBina" with the following components:

link	link function used in the binary model;
f.xb	scale factor of marginal effects, calculated as the density function evaluated at the means of the variables when $x.mean = TRUE$ is specified or the average density value for all individual observations when $x.mean = FALSE$ is specified;
W	a binary probit or logit model object estimated from glm();
out	a data frame object of marginal effects, t-value, and p-value.

Methods

One method is defined as follows:

```
print: print the key output of marginal effects.
```

maTrend 35

Author(s)

```
Changyou Sun (<csun@cfr.msstate.edu>)
```

References

Greene, W.H. 2003. Econometric Analysis (5th ed.). Prentice Hall, New York. 1026 P.

See Also

```
maTrend; plot.maTrend.
```

Examples

maTrend

Trend of Marginal Effects

Description

This function computes the change of probability for a continuous variable, and furthermore, stratifies the probability through a binary independant variable.

Usage

```
maTrend(q, n = 300, nam.c, nam.d, ...)
```

Arguments

q	a object of class of "maBina" estimated from maBina().
n	number of points for calculating probability; the large the number, the smoother the curve.
nam.c	a name of a continuous indepedent variable; this must be given for the function to work.
nam.d	an optional name of a binary independent variable; this is used to stratify the probability.
	additional arguments to be passed.

Details

Marginal effects are calcuated at each value of a continous variable. If specificied, the trend can be stratified by a binary independent variable.

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Value

Return a list object of class "maTrend" with the following components:

q a list object of class "maBina";
nam.c the name of a continous variable;
mm matrix of independant variables for all
trend a data frame of the continous variable and probability values; if nam.d is specificied, the data frame also contains the probability values stratified by the dummy variable;
nam.d if nam.d is specified, the name of a binary variable.
m1 if nam.d is specified, the matrix of mm with the column value for nam.d replaced by 1
m0 if nam.d is specified, the matrix of mm with the column value for nam.d replaced

Methods

Two methods are defined as follows:

by 0

```
print: print the probability output.
```

plot: Plot the probability values for a continuous variable. If a strata is specified through nam.d in maTrend(), then the stratified values also are shown.

Author(s)

```
Changyou Sun (<csun@cfr.msstate.edu>)
```

References

Greene, W.H. 2003. Econometric Analysis (5th ed.). Prentice Hall, New York. 1026 P.

See Also

```
maBina; print.maTrend; plot.maTrend.
```

summary.aiFit 37

summary.aiFit

Summary of Results from Static or Dynamic Models

Description

This summarizes the main results from AIDS models.

Usage

```
## S3 method for class 'aiFit'
summary(object, digits=3, ...)
```

Arguments

object an object of class aiFit from the function of aiStaFit or aiDynFit.

digits number of digits for rounding outputs

additional arguments to be passed.

Details

This wraps up the coefficents and statistics from aiFit by equation.

Value

A data frame object with coefficients and related statistics by equation.

Author(s)

```
Changyou Sun (<csun@cfr.msstate.edu>)
```

See Also

```
\verb"aiStaFit" and \verb"aiDynFit".
```

Examples

```
# see the examples for 'aiDynFit'.
```

systemfitAR

Linear Equation System Estimation with Correction on Autocorrelation

Description

Fits a set of linear structural equations using Ordinary Least Squares (OLS), Weighted Least Squares (WLS), Seemingly Unrelated Regression (SUR), with the option of autocorrelation correlation.

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Usage

```
systemfitAR(formula, method = "OLS", inst = NULL, data = list(),
  restrict.matrix = NULL, restrict.rhs = NULL, restrict.regMat = NULL,
  pooled = FALSE, control = systemfit.control( ... ),
  AR1 = FALSE, rho.sel = c("all", "mean"), model = c("static", "dynamic"), ...)
```

Arguments

formula an object of class formula (for single-equation models) or (typically) a list of

objects of class formula (for multiple-equation models).

method the estimation method, one of "OLS", "WLS", "SUR"; iterated estimation meth-

ods can be specified by setting control parameter maxiter larger than 1 (e.g.

500).

inst one-sided model formula specifying instrumental variables or a list of one-sided

model formulas if different instruments should be used for the different equa-

tions (only needed for 2SLS, W2SLS, and 3SLS estimations).

data an optional data framecontaining the variables in the model.

restrict.matrix

an optional j x k matrix to impose linear restrictions on the coefficients by restrict.matrix * b = restrict.rhs (j = number of restrictions, k = number

of all coefficients, b = vector of all coefficients)

restrict.rhs an optional vector with j elements to impose linear restrictions (see restrict.matrix);

default is a vector that contains j zeros.

restrict.regMat

an optional matrix to impose restrictions on the coefficients by post-multiplying

the regressor matrix with this matrix (see details).

control list of control parameters. The default is constructed by the function systemfit.control.

See the documentation of systemfit.control for details.

pooled logical, restrict coefficients to be equal in all equations (only for panel-like data).

AR1 whether first-order autocorrelation is corrected

rho.sel how rho is computed; rho.sel = "all" means that the system is estimated as

a single equation and the residuals are used to compute rho. If "mean", each equation in the system is estimated separately and the average of rhos from all

the equations are used.

model Static model has interceipt, while dynamic model has no intercept; see translog

cost function and the package for detail.

... arguments passed to systemfit.control.

Details

This is a wrapper of systemfit with an addition of autocorrelation correction. It is mainly used for SUR model with autocorrelation. The main reference sources are Greene (2003), LIMDEP 9.0 manual, Judge et al. (1985), and Berndt and Savin (1975).

Value

systemfit returns a list of the class systemfit. This list contains one special object: "eq". It is a list and contains one object for each estimated equation. These objects are of the class

ur.df2 39

systemfit.equation and contain the results that belong only to the regarding equation. In addition, there are four new items in the output:

The objects of the class systemfit and systemfit.equation have the following components (the elements of the latter are marked with an asterisk (*)):

rho autocorrelation coefficient

rho_ste standard error of rho; if rho.sel = "mean", then it is a vector of the standar

errors for individual equations.

data used for systemfit; this is data adjusted for autocorrelation if AR1 = TRUE;

otherwise, it is just the raw data.

formula forumula used for systemfit. This can be adjusted for autocorrelation; costant

is adjusted as (1 - rho).

Author(s)

Changyou Sun (<csun@cfr.msstate.edu>)

References

Greene, W. H. (2003) Econometric Analysis, Fifth Edition, Prentice Hall.

Judge, George G.; W. E. Griffiths; R. Carter Hill; Helmut Luetkepohl and Tsoung-Chao Lee (1985) *The Theory and Practice of Econometrics, Second Edition*, Wiley.

LIMDEP 9.0 software manual.

Berndt, E.R., and N.E. Savin. 1975. Estimation and hypothesis testing in singular equation systems with autoregressive disturbances. Econometrica 43(5/6):937-957.

See Also

```
lm; aiStaFit; systemfit
```

Examples

- # Check Berndt and Savin (1975) dataset
- # Check Kemenda data

ur.df2

Augmented-Dickey-Fuller Unit Root Test revised

Description

Augmented-Dickey-Fuller Unit Root Test revised

Usage

```
ur.df2(y, type = c("none", "drift", "trend"), lags = 1,
    selectlags = c("Fixed", "AIC", "BIC"))
```

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Arguments

y Vector to be tested for a unit root.

type Test type, either "none", "drift" or "trend"...

lags Number of lags for endogenous variable to be included.

selectlags Lag selection can be achieved according to the Akaike "AIC" or the Bayes "BIC"

information criteria. The maximum number of lags considered is set by lags.

The default is to use a "fixed" lag length set by lags.

Details

This is a modification of ur.df in the library of urca. The function was written in S4, and it is changed into S3. The lag selected by AIC or BIC is reported explicitly through lag.used in the output list. In addition, the values of AIC and BIC statistics are reported

Value

Return an object of class ur.df2: the new outputs are lag.used, aic, and bic.

y input of y
model input of model
cval critical values
lags input of lags
lag.used acutal lags used
teststat test statistics

res resiuals of the test regression

aic aic values
bic bic values
test.name test name

Methods

One method is defined as follows:

print: print test statistics and critical values.

Author(s)

 $Changyou \ Sun \ (\verb|<csun@cfr.msstate.edu>|)$

See Also

ur.df in urca library.

```
# see the code for TD.Link: Japan and China wood product imports
```

write.list 41

write.list	Output and Write a List Object
------------	--------------------------------

Description

Print or write its required argument z to a file

Usage

```
write.list(z, file, t.name = NULL, row.names = FALSE, ...)
```

Arguments

Z	A list object to be written. Each item in the list is perferably a data frame. If not, it is converted into a data frame. All the contents are coerced into characters to avoid loss of information (e.g., a loss of zero in 5.130.
file	a character string naming a file.
t.name	table names. This can be given explicitly, or given by the list name, or by default, named as "table 1" for the first item in z.
row.names	whether the row names in each table should be written (default is FALSE). If TRUE, a new column of the row names is added to each table.
	Other arguments that can be passed to write.table

Details

This function is a wrap-up of write.table. It is convenient to write a set of tables to C drive.

See Also

```
write.table.
```

```
h1 <- ts(data=cbind(1:24), start=c(2001, 1), frequency=12)
h2 <- ts(data=cbind(1:24, 25:48), start=c(2001, 1), frequency=12)
h3 <- ts(data=cbind(1:4, 5:8, 9:12), start=c(2001, 1), frequency=4)
colnames(h2) <- c("aa", "bb")
colnames(h3) <- c("cc", "dd", "ee")
h1; h2; h3

test <- list(t1 = h1, t2 = h2, t3 = h3)

## Not run:
    # test.csv can be saved at a specific working directory
    getwd(); setwd("c:/aERER"); getwd()
    write.list(z = test, file = "test.csv")

## End(Not run)</pre>
```

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