

Package ‘autostsm’

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Kim, Chang-Jin and Charles R. Nelson (1999) ``State-Space Models with Regime Switching: Classical and Gibbs-Sampling Approaches with Applications" <http://econ.korea.ac.kr/~cjkim/doi:10.7551/mitpress/6444.001.0001><http://econ.korea.ac.kr/~{ }cjkim/>.

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Imports maxLik (>= 1.4), forecast (>= 8.13), lubridate (>= 1.7),
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autostsm

AutoSTSM

Description

autostsm Automatic model selection for structural time series decomposition into trend, cycle, and seasonal components using the Kalman filter. See the package vignette using `vignette("autostsm")` or `browseVignettes("autostsm")` to view it in your browser.

Author(s)

Alex Hubbard

GDP

US GDP Seasonally Adjusted

Description

US GDP Seasonally Adjusted

Usage

`data(GDP)`

Format

data.table with columns DATE and GDP, quarterly frequency

Source

FRED

NA000334Q

US GDP Not Seasonally Adjusted

Description

US GDP Not Seasonally Adjusted

Usage

`data(NA000334Q)`

Format

data.table with columns DATE and NA000334Q, quarterly frequency

Source

FRED

SP500

S&P 500

Description

S&P 500

Usage`data(SP500)`**Format**

data.table with columns DATE and SP500, daily frequency

SourceFRED

stsm_bdiag

Build a block diagonal matrix from two matrices

Description

Build a block diagonal matrix from two matrices

Usage`stsm_bdiag(A, B)`**Arguments**

A The top left matrix

B The bottom right matrix

Value

A block diagonal matrix

stsm_build_dates	<i>Build the date sequence as a Date type</i>
------------------	---

Description

Build the date sequence as a Date type

Usage

```
stsm_build_dates(y)
```

Arguments

y a list object created from stsm_detect_frequency

Value

a list with the univariate time series and corrected dates

stsm_check_exo	<i>Data check for input exo</i>
----------------	---------------------------------

Description

Checks for proper input of the table exo

Usage

```
stsm_check_exo(exo, y)
```

Arguments

exo exo datagenous

y input data y

Value

none

stsm_check_exo_fc *Data check for input exo.fc*

Description

Checks for proper input of the table exo.fc

Usage

```
stsm_check_exo_fc(exo.fc, n.ahead)
```

Arguments

exo.fc	exogenous forecast data
n.ahead	forecast periods

Value

none

stsm_check_y *Data check for input y*

Description

Checks for proper input of the table y

Usage

```
stsm_check_y(y)
```

Arguments

y	input data y
---	--------------

Value

none

stsm_constraints *Set the inequality constraints for estimation*

Description

Inequality constraints: ineqA

Usage

```
stsm_constraints(
  prior,
  par,
  freq,
  unconstrained,
  det_trend,
  det_drift,
  det_cycle,
  det_seas,
  det_obs,
  saturating_growth
)
```

Arguments

prior	A data table created by stsm_prior
par	parameter values for the state space model
freq	Frequency of the data
unconstrained	Whether to remove inequality constraints on the trend during estimation
det_trend	Set the trend error variance to 0 (deterministic trend)
det_drift	Set the drift error variance to 0 (deterministic drift)
det_cycle	Set the cycle error variance to 0 (deterministic cycle)
det_seas	Set the seasonality error variances to 0 (deterministic seasonality)
det_obs	Set the observation equation error variance to 0 (deterministic observation equation)
saturating_growth	Force the growth rate to converge to 0 in the long term

Value

list containing the initial values for the Kalman filter

stsm_detect_anomalies *Detect Anomalies*

Description

Detect anomalies using the estimated structural time series model

Usage

```
stsm_detect_anomalies(
  model,
  y = NULL,
  freq = NULL,
  exo = NULL,
  sig_level = 0.01,
  smooth = TRUE,
  plot = FALSE
)
```

Arguments

model	Structural time series model estimated using stsm_estimate.
y	Univariate time series of data values. May also be a 2 column data frame containing a date column.
freq	Frequency of the data (1 (yearly), 4 (quarterly), 12 (monthly), 365.25/7 (weekly), 365.25 (daily)), default is NULL and will be automatically detected
exo	Matrix of exogenous variables used for the historical data. Can be used to specify regression effects or other seasonal effects like holidays, etc.
sig_level	Significance level to determine statistically significant anomalies
smooth	Whether or not to use the Kalman smoother
plot	Whether to plot everything

Value

data table (or list of data tables) containing the dates of detected anomalies from the filtered and/or smoothed series

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
```



```

NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
stsm = stsm_estimate(NA000334Q)
anomalies = stsm_detect_anomalies(model = stsm, y = NA000334Q, plot = TRUE)

## End(Not run)

```

stsm_detect_breaks *Detect Structural Breaks*

Description

Detect structural breaks using the estimated structural time series model

Usage

```

stsm_detect_breaks(
  model,
  y,
  components = c("trend", "cycle", "seasonal"),
  freq = NULL,
  exo = NULL,
  sig_level = 0.01,
  ci = 0.8,
  smooth = TRUE,
  plot = FALSE,
  cores = NULL,
  show_progress = FALSE
)

```

Arguments

model	Structural time series model estimated using stsm_estimate.
y	Univariate time series of data values. May also be a 2 column data frame containing a date column.
components	Vector of components to test for structural breaks
freq	Frequency of the data (1 (yearly), 4 (quarterly), 12 (monthly), 365.25/7 (weekly), 365.25 (daily)), default is NULL and will be automatically detected
exo	Matrix of exogenous variables used for the historical data. Can be used to specify regression effects or other seasonal effects like holidays, etc.
sig_level	Significance level to determine statistically significant anomalies
ci	Confidence interval, value between 0 and 1 exclusive.
smooth	Whether or not to use the Kalman smoother
plot	Whether to plot everything
cores	Number of cores to use for break detection
show_progress	Whether to show progress bar

Value

data table (or list of data tables) containing the dates of detected anomalies from the filtered and/or smoothed series

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
stsm = stsm_estimate(NA000334Q)
breaks = stsm_detect_breaks(model = stsm, y = NA000334Q, plot = TRUE, cores = 2)

## End(Not run)
```

stsm_detect_cycle *Detect cycle from the data*

Description

Detect cycle from the data

Usage

```
stsm_detect_cycle(
  y,
  freq,
  sig_level = 0.01,
  prior = NULL,
  cl = NULL,
  cores = NULL,
  show_progress = FALSE
)
```

Arguments

y	Univariate time series of data values.
freq	Frequency of the data (1 (yearly), 4 (quarterly), 12 (monthly), 365.25/7 (weekly), 365.25 (daily))
sig_level	Significance level to determine statistically significant seasonal frequencies
prior	A data table created by stsm_prior
cl	a parallel cluster object

cores Number of cores to use
 show_progress Whether to show progress bar

Value

Numeric value of cycle periodicity

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
cycle = stsm_detect_cycle(y = NA000334Q$y, freq = 4)

## End(Not run)
```

stsm_detect_frequency *Detect frequency and dates from the data*

Description

Detect frequency and dates from the data

Usage

```
stsm_detect_frequency(y, freq = NULL)
```

Arguments

y Univariate time series of data values. May also be a 2 column data frame containing a date column.

freq Initial setting for the frequency detection

Value

List giving the dates and frequency of the data

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
freq = stsm_detect_frequency(y = NA000334Q)

## End(Not run)
```

```
stsm_detect_multiplicative
```

Detect if log transformation is best

Description

Detect if log transformation is best

Usage

```
stsm_detect_multiplicative(y, freq, sig_level = 0.01, prior = NULL)
```

Arguments

y	an object created from stsm_detect_frequency
freq	Frequency of the data
sig_level	Significance level to determine statistically significant seasonal frequencies
prior	A data table created by stsm_prior

Value

a logical indicating if the model should be multiplicative or not

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
```

```

multiplicative = stsm_detect_multiplicative(y = NA000334Q$y, freq = 4)

## End(Not run)

```

```

stsm_detect_seasonality
Detect seasonality from the data

```

Description

Detect seasonality from the data

Usage

```

stsm_detect_seasonality(
  y,
  freq,
  sig_level = 0.01,
  prior = NULL,
  cl = NULL,
  cores = NULL,
  show_progress = FALSE
)

```

Arguments

y	Univariate time series of data values.
freq	Frequency of the data (1 (yearly), 4 (quarterly), 12 (monthly), 365.25/7 (weekly), 365.25 (daily))
sig_level	Significance level to determine statistically significant seasonal frequencies
prior	A data table created from stsm_prior
cl	a parallel cluster object
cores	Number of cores to use
show_progress	Whether to show progress bar

Value

Numeric vector of seasonal periodicities

Examples

```

## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)

```

```

colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
seasonality = stsm_detect_seasonality(y = NA000334Q$y, freq = 4)

## End(Not run)

```

stsm_detect_trend *Detect trend type*

Description

Detect trend type

Usage

```

stsm_detect_trend(
  y,
  freq,
  decomp = "",
  sig_level = 0.01,
  prior = NULL,
  seasons = NULL,
  cycle = NULL
)

```

Arguments

y	Univariate time series of data values. May also be a 2 column data frame containing a date column.
freq	Frequency of the data (1 (yearly), 4 (quarterly), 12 (monthly), 365.25/7 (weekly), 365.25 (daily))
decomp	Decomposition model ("tend-cycle-seasonal", "trend-seasonal", "trend-cycle", "trend-noise")
sig_level	Significance level to determine statistically significant seasonal frequencies
prior	A data table created by stsm_prior
seasons	The seasonal periods
cycle	The cycle period

Value

list with trend type and logical flag for deterministic trend if the trend is determined to have 0 differencing

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
trend = stsm_detect_trend(y = NA000334Q$y, freq = 4)

## End(Not run)
```

stsm_estimate

Trend cycle seasonal decomposition using the Kalman filter.

Description

Estimates a structural time series model using the Kalman filter and maximum likelihood. The seasonal and cycle components are assumed to be of a trigonometric form. The function checks three trend specifications to decompose a univariate time series into trend, cycle, and/or seasonal components plus noise. The function automatically detects the frequency and checks for a seasonal and cycle component if the user does not specify the frequency or decomposition model. This can be turned off by setting freq or specifying decomp. State space model for decomposition follows $Y_t = T_t + C_t + S_t + A * X_t + e_t$, $e_t \sim N(0, \sigma_e^2)$ Y is the data T is the trend component C is the cycle component S is the seasonal component X is the exogenous data with parameter vector B e is the observation error

Usage

```
stsm_estimate(
  y,
  exo = NULL,
  freq = NULL,
  decomp = NULL,
  trend = NULL,
  unconstrained = FALSE,
  saturating_growth = FALSE,
  multiplicative = NULL,
  par = NULL,
  seasons = NULL,
  cycle = NULL,
  cores = NULL,
  det_obs = FALSE,
  det_trend = NULL,
  det_seas = FALSE,
```

```

det_drift = FALSE,
det_cycle = FALSE,
sig_level = 0.01,
optim_methods = c("BFGS", "NM", "CG", "SANN"),
maxit = 10000,
verbose = FALSE
)

```

Arguments

y	Univariate time series of data values. May also be a 2 column data frame containing a date column.
exo	Matrix of exogenous variables. Can be used to specify regression effects or other seasonal effects like holidays, etc.
freq	Frequency of the data (1 (yearly), 4 (quarterly), 12 (monthly), 365.25/7 (weekly), 365.25 (daily)), default is NULL and will be automatically detected
decomp	Decomposition model ("tend-cycle-seasonal", "trend-seasonal", "trend-cycle", "trend-noise")
trend	Trend specification ("random-walk", "random-walk-drift", "double-random-walk", "random-walk2"). The default is NULL which will choose the best of all specifications based on the maximum likelihood. "random-walk" is the random walk trend. "random-walk-drift" is the random walk with constant drift trend. "double-random-walk" is the random walk with random walk drift trend. "random-walk2" is a 2nd order random walk trend as in the Hodrick-Prescott filter. If trend is "random-walk", the trend model is $T_t = T_{t-1} + e_t$, $e_t \sim N(0, \text{sig}_t^2)$ If trend is "random-walk-drift", the trend model is $T_t = T_{t-1} + D_{t-1} + e_t$, $e_t \sim N(0, \text{sig}_t^2)$ with $D_t = d + \text{phi}_d * D_{t-1} + n_t$, $n_t \sim N(0, \text{sig}_d^2)$ If trend is "double-random-walk", the trend model is $T_t = M_{t-1} + T_{t-1} + e_t$, $e_t \sim N(0, \text{sig}_t^2)$ with $M_t = M_{t-1} + n_t$, $n_t \sim N(0, \text{sig}_d^2)$ If trend is "random-walk2", the trend model is $T_t = 2T_{t-1} - T_{t-2} + e_t$, $e_t \sim N(0, \text{sig}_t^2)$
unconstrained	Logical whether to remove inequality constraints on the trend during estimation
saturating_growth	Force the growth rate to converge to 0 in the long term
multiplicative	If data should be logged to create a multiplicative model. If multiplicative = TRUE, then the data is logged and the original model becomes multiplicative ($Y_t = T_t * C_t * S_t * BX_t * e_t$)
par	Initial parameters, default is NULL
seasons	The seasonal periods: i.e. c(365.25, 7 if yearly and weekly seasonality). Default is NULL and will be estimated via wavelet analysis. Can set to FALSE if want no seasonality
cycle,	The period for the longer-term cycle. Default is NULL and will be estimated via wavelet analysis. Can set to FALSE if want no cycle.
cores	Number of cores to use for seasonality and cycle detection

det_obs	Set the observation equation error variance to 0 (deterministic observation equation) If det_obs = TRUE then the error variance of the observation equation (sig_e) is set to 0
det_trend	Set the trend error variance to 0 (deterministic trend) If det_trend = TRUE then the error variance of the trend equation (sig_t) is set to 0 and is referred to as a smooth trend
det_seas	Set the seasonality error variances to 0 (deterministic seasonality) If det_seas = TRUE then the error variance all seasonality frequency j equations (sig_s) are set to 0 and is referred to as deterministic seasonality
det_drift	Set the drift error variance to 0 (deterministic drift) If det_drift = TRUE then the error variance of the drift equation (sig_d) is set to 0 and is referred to as a deterministic drift
det_cycle	Set the cycle error variance to 0 (deterministic cycle) If det_cycle = TRUE then the error variance of the cycle equation (sig_c) is set to 0 and is referred to as a deterministic cycle
sig_level	Significance level to determine statistically significant seasonal frequencies
optim_methods	Vector of 1 to 3 optimization methods in order of preference ("NR", "BFGS", "CG", "BHHH", or "SANN")
maxit	Maximum number of iterations for the optimization
verbose	Logical whether to print messages or not

Value

List of estimation values including a data table with coefficients, convergence code, frequency, decomposition, seasonality, cyclicity, and trend specification as well as the a data table with the original data with dates. Any exogenous data given is also returned.

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
stsm = stsm_estimate(NA000334Q)

## End(Not run)
```

stsm_forecast

Kalman Filter and Forecast

Description

Kalman filter and forecast an estimated model from stsm_estimate output

Usage

```
stsm_forecast(
  model,
  y,
  n.ahead = 0,
  freq = NULL,
  exo = NULL,
  exo.fc = NULL,
  ci = 0.8,
  plot = FALSE,
  plot.decomp = FALSE,
  plot.fc = FALSE,
  n.hist = NULL,
  smooth = TRUE,
  dampen_cycle = FALSE
)
```

Arguments

model	Structural time series model estimated using stsm_estimate.
y	Univariate time series of data values. May also be a 2 column data frame containing a date column.
n.ahead	the number of periods to forecast
freq	Frequency of the data (1 (yearly), 4 (quarterly), 12 (monthly), 365.25/7 (weekly), 365.25 (daily)), default is NULL and will be automatically detected
exo	Matrix of exogenous variables used for the historical data. Can be used to specify regression effects or other seasonal effects like holidays, etc.
exo.fc	Matrix of exogenous variables used for the forecast
ci	Confidence interval, value between 0 and 1 exclusive.
plot,	Logical, whether to plot everything
plot.decomp	Logical, whether to plot the filtered historical data
plot.fc	Logical, whether to plot the forecast
n.hist	Number of historical periods to include in the forecast plot. If plot = TRUE and n.hist = NULL, defaults to 3 years.
smooth	Whether or not to use the Kalman smoother
dampen_cycle	Whether to remove oscillating cycle dynamics and smooth the cycle forecast into the trend using a sigmoid function that maintains the rate of convergence

Value

data table (or list of data tables) containing the filtered and/or smoothed series.

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
stsm = stsm_estimate(NA000334Q)
fc = stsm_forecast(stsm, y = NA000334Q, n.ahead = floor(stsm$freq)*3, plot = TRUE)

## End(Not run)
```

stsm_format_exo

Format exo

Description

Format the exo table

Usage

```
stsm_format_exo(exo, dates, range)
```

Arguments

exo	exogenous data
dates	dates vector
range	range of data to include

Value

a data table

stsm_init_pars *Get initial parameter estimates for estimation*

Description

Get initial parameter estimates for estimation

Usage

```
stsm_init_pars(
  y,
  freq,
  trend,
  cycle,
  decomp = "",
  seasons = NULL,
  prior = NULL
)
```

Arguments

y	an object created from stsm_detect_frequency
freq	Frequency of the data
trend	Trend specification ("random-walk", "random-walk-drift", "double-random-walk", "random-walk2").
cycle	The period for the longer-term cycle
decomp	Decomposition model ("tend-cycle-seasonal", "trend-seasonal", "trend-cycle", "trend-noise")
seasons	The seasonal lengths to split the seasonality into
prior	A data table created by stsm_prior

Value

named vector containing the initial parameter estimates for estimation

stsm_init_vals *Get initial values for the Kalman filter*

Description

Get initial values for the Kalman filter

Usage

```

stsm_init_vals(
  y,
  par,
  freq,
  trend,
  decomp = "",
  seasons = NULL,
  prior = NULL,
  cycle = NULL
)

```

Arguments

y	an object created from stsm_detect_frequency
par	parameter values for the state space model
freq	Frequency of the data
trend	Trend specification ("random-walk", "random-walk-drift", "double-random-walk", "random-walk2").
decomp	Decomposition model ("tend-cycle-seasonal", "trend-seasonal", "trend-cycle", "trend-noise")
seasons	The seasonal periods to split the seasonality into
prior	A data table created by stsm_prior
cycle	The cycle period

Value

list containing the initial values for the Kalman filter

stsm_prior	<i>Return a naive model prior decomposition</i>
------------	---

Description

Return a naive model prior decomposition

Usage

```

stsm_prior(y, freq, decomp = "", seasons = NULL, cycle = NULL)

```

Arguments

y	an object created from stsm_detect_frequency
freq	Frequency of the data
decomp	decomposition string
seasons	The seasonal periods to split the seasonality into
cycle	The cycle periods

Value

data table containing a naive decomposition using STL

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
prior = stsm_prior(y = NA000334Q$y, freq = 4)

## End(Not run)
```

stsm_ssm

State space model

Description

Creates a state space model in list form $yt = H*B + e_t$ $B = F*B_{t-1} + u_t$

Usage

```
stsm_ssm(
  par = NULL,
  yt = NULL,
  decomp = NULL,
  trend = NULL,
  init = NULL,
  model = NULL
)
```

Arguments

par	Vector of named parameter values, includes the harmonics
yt	Univariate time series of data values
decomp	Decomposition model ("tend-cycle-seasonal", "trend-seasonal", "trend-cycle", "trend-noise")
trend	Trend specification ("random-walk", "random-walk-drift", "double-random-walk", "random-walk2"). The default is NULL which will choose the best of all specifications based on the maximum likelihood. "random-walk" is the random walk trend. "random-walk-drift" is the random walk with constant drift trend. "double-random-walk" is the random walk with random walk drift trend. "random-walk2" is a 2nd order random walk trend as in the Hodrick-Prescott filter.

init Initial state values for the Kalman filter
 model a stsm_estimate model object

Value

List of space space matrices

Examples

```
## Not run:
#GDP Not seasonally adjusted
library(autostsm)
data("NA000334Q", package = "autostsm") #From FRED
NA000334Q = data.table(NA000334Q, keep.rownames = TRUE)
colnames(NA000334Q) = c("date", "y")
NA000334Q[, "date" := as.Date(date)]
NA000334Q[, "y" := as.numeric(y)]
NA000334Q = NA000334Q[date >= "1990-01-01", ]
stsm = stsm_estimate(NA000334Q)
ssm = stsm_ssm(model = stsm)

## End(Not run)
```

UNRATE *Unemployment Rate Seasonally Adjusted*

Description

Unemployment Rate Seasonally Adjusted

Usage

```
data(UNRATE)
```

Format

data.table with columns DATE and UNRATE, monthly frequency

Source

FRED

UNRATENSA

Unemployment Rate Not Seasonally Adjusted

Description

Unemployment Rate Not Seasonally Adjusted

Usage

```
data(UNRATENSA)
```

Format

data.table with columns DATE and UNRATENSA, monthly frequency

Source

FRED

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