

Package ‘autostsm’

April 7, 2021

Type Package

Title Automatic Structural Time Series Models

Version 1.2.2

Date 2021-04-01

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Description Automatic model selection for structural time series decomposition into trend, cycle, and seasonal components using the Kalman filter.
Koopman, Siem Jan and Marius Ooms (2012) ``Forecasting Economic Time Series Using Unobserved Components Time Series Models" <doi:10.1093/oxfordhb/9780195398649.013.0006>.
Kim, Chang-Jin and Charles R. Nelson (1999) ``State-Space Models with Regime Switching: Classical and Gibbs-Sampling Approaches with Applications" <doi:10.7551/mitpress/6444.001.0001><http://econ.korea.ac.kr/~cjkim/>.

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Imports Matrix (>= 1.2), maxLik (>= 1.4), forecast (>= 8.13),
lubridate (>= 1.7), tsutils (>= 0.9), ggplot2 (>= 3.3),
gridExtra (>= 2.3), strucchange (>= 1.5), imputeTS (>= 3.1),
foreach (>= 1.5), doSNOW (>= 1.0), parallel (>= 4.0), zoo (>= 1.8),
lmtest (>= 0.9), tseries (>= 0.1), ggrepel (>= 0.9),
progress (>= 1.2)

Depends R (>= 3.5.0), data.table (>= 1.13)

LinkingTo Rcpp, RcppArmadillo

RoxygenNote 7.1.1

Suggests knitr, rmarkdown

VignetteBuilder knitr

NeedsCompilation yes

Repository CRAN

Date/Publication 2021-04-07 08:40:12 UTC

R topics documented:

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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_build_dates

Build the date sequence as a Date type

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 | <i>Data check for input exo.fc</i> |
|-------------------|------------------------------------|

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 | <i>Data check for input y</i> |
|--------------|-------------------------------|

Description

Checks for proper input of the table y

Usage

```
stsm_check_y(y)
```

Arguments

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

Value

none

| | |
|------------------|--|
| stsm_constraints | <i>Set the inequality constraints for estimation</i> |
|------------------|--|

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 = 1e-04, prior = NULL)
```

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 |

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 = 1e-04,
  prior = NULL,
  full_spectrum = 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 |
| full_spectrum | check the full spectrum of seasonal frequencies for seasonality |

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 | <i>Detect trend type</i> |
|-------------------|--------------------------|

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 $Y_t = T_t + C_t + S_t + A * X_t + e_t$, $e_t \sim N(0, \text{sig}_e^2)$ 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,
```

```

full_spectrum = FALSE,
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 ("trend-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. |
| full_spectrum | whether to check the full spectrum of frequencies for seasonality rather than a limited number based on the frequency of the data |
| 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 | <i>Kalman Filter and Forecast</i> |
|---------------|-----------------------------------|

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 | <i>Get initial parameter estimates for estimation</i> |
|----------------|---|

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 | <i>Get initial values for the Kalman filter</i> |
|----------------|---|

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)
```

sw_dcf

Stock & Watson US Dynamic Common Factor (1991) Data

Description

Stock & Watson US Dynamic Common Factor (1991) Data

Usage

```
data(sw_dcf)
```

Format

data.table with columns date, ip (industrial production), gmyxpg (personal income less transfer payments), mtq (total manufacturing and trade sales), and lpnag (employeees on nonagricultural payrolls) taken from

Source

Kim and Nelson (1999) "State-Space Models with Regime Switching" (<http://econ.korea.ac.kr/~cjkim/>)

| | |
|--------|--|
| UNRATE | <i>Unemployment Rate Seasonally Adjusted</i> |
|--------|--|

Description

Unemployment Rate Seasonally Adjusted

Usage

```
data(UNRATE)
```

Format

data.table with columns DATE and UNRATE, monthly frequency

Source

FRED

| | |
|-----------|--|
| UNRATENSA | <i>Unemployment Rate Not Seasonally Adjusted</i> |
|-----------|--|

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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