# Package 'TSstudio'

August 9, 2023

```
Title Functions for Time Series Analysis and Forecasting
Version 0.1.7
Maintainer Rami Krispin < rami.krispin@gmail.com>
Description Provides a set of tools for descriptive and predictive analysis of time series data. That in-
     cludes functions for interactive visualization of time series objects and as well utility func-
     tions for automation time series forecasting.
License MIT + file LICENSE
Encoding UTF-8
LazyData true
Depends R (>= 3.0.2)
Imports data.table(>= 1.11.2), dplyr(>= 0.7.5), forecast (>= 8.2),
     forecastHybrid(>= 2.0.10), parallel(>= 4.1.2), lubridate (>=
     1.6.0), magrittr (>= 1.5), plotly (>= 4.7.1), purrr(>= 0.2.5),
     RColorBrewer(>= 1.1-2), reshape2 (>= 1.4.2), scales(>= 1.0.0),
     tidyr(>= 0.8.1), tsibble(>= 1.1.3), viridis (>= 0.5.1), xts (>=
     0.12-0), zoo (>= 1.8-0)
Suggests devtools, DT, knitr, quantmod, rmarkdown, UKgrid
VignetteBuilder knitr
RoxygenNote 6.1.1
URL https://github.com/RamiKrispin/TSstudio
BugReports https://github.com/RamiKrispin/TSstudio/issues
NeedsCompilation no
Author Rami Krispin [aut, cre]
Repository CRAN
Date/Publication 2023-08-09 04:40:07 UTC
```

Type Package

38

Index

## R topics documented:

arima_diag	3
ccf_plot	4
check_res	5
Coffee_Prices	5
create_model	6
EURO_Brent	8
forecast_sim	9
Michigan_CS	10
plot_error	11
plot_forecast	12
plot_grid	13
plot_model	13
res_hist	15
test_forecast	16
train_model	17
ts_cor	18
ts_decompose	19
ts_grid	20
ts_heatmap	22
ts_info	23
ts_lags	24
ts_ma	25
ts_plot	27
ts_polar	28
ts_quantile	28
ts_reshape	30
ts_seasonal	30
ts_split	31
ts_sum	32
ts_surface	33
ts_to_prophet	33
USgas	34
USUnRate	35
USVSales	35
US_indicators	36
xts_to_ts	36
zoo_to_ts	37

arima\_diag 3

arima_	_d1ag

Diagnostic Plots for ARIMA Models

#### **Description**

Diagnostic Plots for ARIMA Models

#### Usage

## **Arguments**

ts.obj A ts object

method A list, defines the transformation parameters of each plot. Each plot should

be defined by a list, where the name of the list defines the plot ID. The plot

parameters are:

diff - an integer, defines the degree of diffrence log - a boolean, optional, defines

if log transformation should be used title - optional, the plot title

cor A boolean, if TRUE (default), will plot the series ACF and PACF

#### **Details**

The arima\_diag function provides a set of diagnostic plots for identify the ARIMA model parameters. The ACF and PACF can assist in identifying the AR and MA process, and the diffrence plotting hel in identifying the degree of differencing that required to make the series stationary

## Value

A plot

4 ccf\_plot

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cct	nla	+
CCI.	_plo	··

Time Series Cross Correlation Lags Visualization

#### **Description**

Visualize the series y against the series x lags (according to the setting of the lags argument) and return the corresponding cross-correlation value for each lag

#### Usage

```
ccf_plot(x, y, lags = 0:12, margin = 0.02, n_plots = 3,
  Xshare = TRUE, Yshare = TRUE, title = NULL)
```

#### **Arguments**

x y lags	A univariate time series object of a class "ts"  A univariate time series object of a class "ts"  An integer, set the lags range, by default will plot the two series along with the
1463	first 12 lags
margin	Plotly parameter, either a single value or four values (all between 0 and 1). If four values provided, the first will be used as the left margin, the second will be used as the right margin, the third will be used as the top margin, and the fourth will be used as the bottom margin. If a single value provided, it will be used as all four margins.
n_plots	An integer, define the number of plots per row
Xshare	Plotly parameter, should the x-axis be shared amongst the subplots?

Plotly parameter, should the y-axis be shared amongst the subplots?

title A character, optional, set the plot title

#### Value

Plot

Yshare

```
data("USUnRate")
data("USVSales")

ccf_plot(x = USVSales, y = USUnRate)

#Plotting the first 6 lead and lags of the USVSales with the USUnRate
ccf_plot(x = USVSales, y = USUnRate, lags = -6:6)

# Setting the plot margin and number of plots in each raw
ccf_plot(x = USVSales, y = USUnRate, lags = c(0, 6, 12, 24),
margin = 0.01, n_plots = 2)
```

check\_res 5

С	heck_res	Visualization of the Residuals of a Time Series Model

## Description

Provides a visualization of the residuals of a time series model. That includes a time series plot of the residuals, and the plots of the autocorrelation function (acf) and histogram of the residuals

#### Usage

```
check_res(ts.model, lag.max = 36)
```

## Arguments

ts.model A time series model (or forecasted) object, support any model from the forecast

package with a residuals output

lag.max The maximum number of lags to display in the residuals' autocorrelation func-

tion plot

## **Examples**

```
library(forecast)
data(USgas)

# Create a model
fit <- auto.arima(USgas)

# Check the residuals of the model
check_res(fit)</pre>
```

Coffee\_Prices

Coffee Prices: Robusta and Arabica

#### **Description**

Coffee Prices: Robusta and Arabica: 1960 - 2018. Units: Dollars per Kg

## Usage

```
Coffee_Prices
```

#### **Format**

```
Time series data - 'mts' object
```

6 create\_model

#### **Source**

WIKI Commodity Prices - Quandle

#### **Examples**

```
ts_plot(Coffee_Prices)
```

create\_model

A Functional Approach for Building the train\_model Components

## **Description**

Add, edit, or remove the components of the train\_model function

## Usage

```
create_model()
add_input(model.obj, input)
add_methods(model.obj, methods)
remove_methods(model.obj, method_ids)
add_train_method(model.obj, train_method)
add_horizon(model.obj, horizon)
build_model(model.obj)
set_error(model.obj, error)
add_xreg(model.obj, xreg)
add_level(model.obj, level)
```

#### Arguments

model.obj The train\_model skeleton, created by the create\_model function or edited by

add\_input, add\_methods, remove\_methods, add\_train\_method or add\_horizon

input A univariate time series object (ts class)

methods A list, defines the models to use for training and forecasting the series. The list

must include a sub list with the model type, and the model's arguments (when applicable) and notes about the model. The sub-list name will be used as the

model ID. Possible models:

arima - model from the stats package

create\_model 7

auto.arima - model from the forecast package ets - model from the forecast package HoltWinters - model from the stats package nnetar - model from the forecast package tslm - model from the forecast package (note that the 'tslm' model must have the formula argument in the 'method\_arg' argument) A character, defines the IDs of the model methods to be remove with the remethod\_ids move methods function train\_method A list, defines the train approach, either using a single testing partition (sample out) or use multiple testing partitions (backtesting). The list should include the training method argument, (please see 'details' for the structure of the argument) An integer, defines the forecast horizon horizon A character, defines the error metrics to be used to sort the models leaderboard. error Possible metric - "MAPE" or "RMSE" Optional, a list with two vectors (e.g., data.frame or matrix) of external regresxreg sors, one vector corresponding to the input series and second to the forecast itself (e.g., must have the same length as the input and forecast horizon, respectively) An integer, set the confidence level of the prediction intervals level

```
## Not run:
### Building train_model function by adding its different components
# Create a skeleton model
md <- create_model()</pre>
class(md)
# Add input
data(USgas)
md <- add_input(model.obj = md, input = USgas)</pre>
# Add methods
methods <- list(ets1 = list(method = "ets",</pre>
                             method_arg = list(opt.crit = "lik"),
                             notes = "ETS model with opt.crit = lik"),
                 ets2 = list(method = "ets",
                             method_arg = list(opt.crit = "amse"),
                             notes = "ETS model with opt.crit = amse"),
                 arima1 = list(method = "arima",
                               method\_arg = list(order = c(1,1,1),
                                          seasonal = list(order = c(1,0,1)),
                               notes = "SARIMA(1,1,1)(1,0,1)"))
md <- add_methods(model.obj = md, methods = methods)</pre>
# Add additional methods
methods2 <- list(arima2 = list(method = "arima",</pre>
                               method_arg = list(order = c(2,1,2),
```

8 EURO\_Brent

```
seasonal = list(order = c(1,1,1)),
                               notes = "SARIMA(2,1,2)(1,1,1)"),
                hw = list(method = "HoltWinters",
                           method_arg = NULL,
                           notes = "HoltWinters Model"),
                tslm = list(method = "tslm",
                    method_arg = list(formula = input ~ trend + season),
                    notes = "tslm model with trend and seasonal components"))
md <- add_methods(model.obj = md, methods = methods2)</pre>
# Remove methods
md <- remove_methods(model.obj = md, method_ids = c("ets2"))</pre>
# Add train method
md <- add_train_method(model.obj = md, train_method = list(partitions = 6,</pre>
                                                         sample.out = 12,
                                                         space = 3))
# Set the forecast horizon
md <- add_horizon(model.obj = md, horizon = 12)</pre>
# Add the forecast prediction intervals confidence level
md \leftarrow add_level(model.obj = md, level = c(90, 95))
### Alternatively, pipe the function with the magrittr package
library(magrittr)
md <- create_model() %>%
      add_input(input = USgas) %>%
      add_methods(methods = methods) %>%
      add_methods(methods = methods2) %>%
      add_train_method(train_method = list(partitions = 4,
                                            sample.out = 12,
                                            space = 3)) %>%
       add_horizon(horizon = 12) %>%
       add_level(level = c(90, 95))
# Run the model
fc <- md %>% build_model()
## End(Not run)
```

EURO\_Brent

Crude Oil Prices: Brent - Europe

#### **Description**

Crude Oil Prices: Brent - Europe: 1987 - 2019. Units: Dollars per Barrel

forecast\_sim 9

#### Usage

EURO\_Brent

#### **Format**

Time series data - 'zoo' object

#### **Source**

U.S. Energy Information Administration, Crude Oil Prices: Brent - Europe [MCOILBRENTEU], retrieved from FRED, Federal Reserve Bank of St. Louis; https://fred.stlouisfed.org/series/MCOILBRENTEU, January 8, 2018.

## **Examples**

```
ts_plot(EURO_Brent)
ts_decompose(EURO_Brent, type = "both")
```

forecast\_sim

Forecasting simulation

## Description

Creating different forecast paths for forecast objects (when applicable), by utilizing the underline model distribution with the simulate function

## Usage

```
forecast_sim(model, h, n, sim_color = "blue", opacity = 0.05,
   plot = TRUE)
```

## **Arguments**

model	A forecasting model supporting Arima, auto.arima, ets, and nnetar models from the **forecast** package
h	An integer, defines the forecast horizon
n	An integer, set the number of iterations of the simulation
sim_color	Set the color of the simulation paths lines
opacity	Set the opacity level of the simulation path lines
plot	Logical, if TRUE will desplay the output plot

#### Value

The baseline series, the simulated values and a plot

10 Michigan\_CS

## **Examples**

```
## Not run:
library(forecast)
data(USgas)

# Create a model
fit <- auto.arima(USgas)

# Simulate 100 possible forecast path, with horizon of 60 months
forecast_sim(model = fit, h = 60, n = 100)

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

Michigan\_CS

University of Michigan Consumer Survey, Index of Consumer Sentiment

## Description

University of Michigan Consumer Survey, Index of Consumer Sentiment: 1980 - 2019. Units: Index 1966:Q1=100

## Usage

Michigan\_CS

#### **Format**

Time series data - 'xts' object

## **Source**

University of Michigan, University of Michigan: Consumer Sentiment

```
ts_plot(Michigan_CS)
ts_heatmap(Michigan_CS)
```

plot\_error 11

plot\_error

Plot the Models Error Rates on the Testing Partitions

#### **Description**

Plot the Models Error Rates on the Testing Partitions

## Usage

```
plot_error(model.obj, error = "MAPE", palette = "Set1")
```

#### **Arguments**

model.obj A train\_model object

error A character, defines the type of error metrics to plot, possible metric - "MAPE"

or "RMSE"

palette A character, defines the color type to used on the plot, use row.names(RColorBrewer::brewer.pal.info)

to view possible color palletes

#### **Details**

The plot\_model provides a visualization of the models performance on the testing paritions for the train\_model function output

#### Value

A plot with a summery of the models error rate by testing partition

```
## Not run:
# Defining the models and their arguments
methods <- list(ets1 = list(method = "ets",</pre>
                            method_arg = list(opt.crit = "lik"),
                            notes = "ETS model with opt.crit = lik"),
                ets2 = list(method = "ets",
                            method_arg = list(opt.crit = "amse"),
                            notes = "ETS model with opt.crit = amse"),
                arima1 = list(method = "arima",
                              method_arg = list(order = c(2,1,0)),
                              notes = "ARIMA(2,1,0)"),
                arima2 = list(method = "arima",
                              method_arg = list(order = c(2,1,2),
                                                 seasonal = list(order = c(1,1,1)),
                              notes = "SARIMA(2,1,2)(1,1,1)"),
                hw = list(method = "HoltWinters",
                          method_arg = NULL,
                          notes = "HoltWinters Model"),
```

plot\_forecast

plot\_forecast

Plotting Forecast Object

## Description

Visualization functions for forecast package forecasting objects

## Usage

```
plot_forecast(forecast_obj, title = NULL, Xtitle = NULL,
    Ytitle = NULL, color = NULL, width = 2)
```

#### **Arguments**

forecast_obj	A forecast object from the forecast, forecastHybrid, or bsts packages
title	A character, a plot title, optional
Xtitle	Set the X axis title, default set to NULL
Ytitle	Set the Y axis title, default set to NULL
color	A character, the plot, support both name and expression
width	An Integer, define the plot width, default is set to 2

```
data(USgas)
library(forecast)
fit <- ets(USgas)
fc<- forecast(fit, h = 60)
plot_forecast(fc)</pre>
```

plot\_grid 13

plot_grid	Visualizing Grid Search Results	
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## Description

Visualizing Grid Search Results

## Usage

```
plot_grid(grid.obj, top = NULL, highlight = 0.1, type = "parcoords",
  colors = list(showscale = TRUE, reversescale = FALSE, colorscale =
  "Jet"))
```

## Arguments

grid.obj	A ts_grid output object
top	An integer, set the number of hyper-parameters combinations to visualize (ordered by accuracy). If set to NULL (default), will plot the top 100 combinations
highlight	A proportion between 0 (excluding) and 1, set the number of hyper-parameters combinations to highlight (by accuracy), if the type argument is set to "parcoords"
type	The plot type, either "3D" for 3D plot or "parcoords" for parallel coordinates plot. Note: the 3D plot option is applicable whenever there are three tuning parameters, otherwise will use a 2D plot for two tuning parameters.
colors	A list of plotly arguments for the color scale setting: showscale - display the color scale if set to TRUE. reversescale - reverse the color scale if set to TRUE colorscale set the color scale of the plot, possible palettes are: Greys, YlGnBu, Greens , YlOrRd, Bluered, RdBu, Reds, Blues, Picnic, Rainbow, Portland, Jet, Hot, Blackbody, Earth, Electric, Viridis, Cividis

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Plot the Models Performance on the Testing Partitions

## Description

Plot the Models Performance on the Testing Partitions

## Usage

```
plot_model(model.obj, model_ids = NULL)
```

14 plot\_model

#### **Arguments**

model.obj A train\_model object

model\_ids A character, defines the trained models to plot, if set to NULL (default), will plot all the models

#### **Details**

The plot\_model provides a visualization of the models performance on the testing paritions for the train\_model function output

#### Value

Animation of models forecast on the testing partitions compared to the actuals

```
## Not run:
# Defining the models and their arguments
methods <- list(ets1 = list(method = "ets",</pre>
                            method_arg = list(opt.crit = "lik"),
                            notes = "ETS model with opt.crit = lik"),
                ets2 = list(method = "ets",
                            method_arg = list(opt.crit = "amse"),
                            notes = "ETS model with opt.crit = amse"),
                arima1 = list(method = "arima",
                              method_arg = list(order = c(2,1,0)),
                              notes = "ARIMA(2,1,0)"),
                arima2 = list(method = "arima",
                              method_arg = list(order = c(2,1,2),
                                                seasonal = list(order = c(1,1,1)),
                              notes = "SARIMA(2,1,2)(1,1,1)"),
                hw = list(method = "HoltWinters",
                          method_arg = NULL,
                          notes = "HoltWinters Model"),
                tslm = list(method = "tslm",
                            method_arg = list(formula = input ~ trend + season),
                            notes = "tslm model with trend and seasonal components"))
# Training the models with backtesting
md <- train_model(input = USgas,</pre>
                  methods = methods,
                  train_method = list(partitions = 6,
                                      sample.out = 12,
                                      space = 3),
                  horizon = 12,
                  error = "MAPE")
# Plot the models performance on the testing partitions
plot_model(model.obj = md)
# Plot only the ETS models
plot_model(model.obj = md , model_ids = c("ets1", "ets2"))
```

res\_hist

```
## End(Not run)
```

res\_hist

Histogram Plot of the Residuals Values

## Description

Histogram plot of the residuals values

## Usage

```
res_hist(forecast.obj)
```

## **Arguments**

forecast.obj A fitted or forecasted object (of the forecast package) with residuals output

```
## Not run:
library(forecast)
data(USgas)

# Set the horizon of the forecast
h <- 12

# split to training/testing partition
split_ts <- ts_split(USgas, sample.out = h)
train <- split_ts$train
test <- split_ts$trest

# Create forecast object
fc <- forecast(auto.arima(train, lambda = BoxCox.lambda(train)), h = h)

# Plot the fitted and forecasted vs the actual values
res_hist(forecast.obj = fc)

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

16 test\_forecast

test_forecast	Visualize of the Fitted and the Forecasted vs the Actual Values	
test_rorecast	visualize of the Fillea and the Forecastea vs the Actual values	

## Description

Visualize the fitted values of the training set and the forecast values of the testing set against the actual values of the series

## Usage

```
test_forecast(actual, forecast.obj, train = NULL, test, Ygrid = FALSE,
   Xgrid = FALSE, hover = TRUE)
```

## **Arguments**

actual	The full time series object (supports "ts", "zoo" and "xts" formats)
forecast.obj	The forecast output of the training set with horizon align to the length of the testing (support forecasted objects from the "forecast" package)
train	Training partition, a subset of the first n observation in the series (not required thed)
test	The testing (hold-out) partition
Ygrid	Logic, show the Y axis grid if set to TRUE
Xgrid	Logic, show the X axis grid if set to TRUE
hover	If TRUE add tooltip with information about the model accuracy

```
## Not run:
library(forecast)
data(USgas)

# Set the horizon of the forecast
h <- 12

# split to training/testing partition
split_ts <- ts_split(USgas, sample.out = h)
train <- split_ts$train
test <- split_ts$trest

# Create forecast object
fc <- forecast(auto.arima(train, lambda = BoxCox.lambda(train)), h = h)

# Plot the fitted and forecasted vs the actual values
test_forecast(actual = USgas, forecast.obj = fc, test = test)

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

17 train\_model

train_model Train, Test, Evaluate, and Forecast Multiple Time Series Forecasting  Models	train_model	Train, Test, Evaluate, and Forecast Multiple Time Series Forecasting Models
--	-------------	--

## **Description**

Method for train test and compare multiple time series models using either one partition (i.e., sample out) or multipe partitions (backtesting)

#### **Usage**

```
train_model(input, methods, train_method, horizon, error = "MAPE",
 xreg = NULL, level = c(80, 95))
```

#### **Arguments**

input

A univariate time series object (ts class)

methods

A list, defines the models to use for training and forecasting the series. The list must include a sub list with the model type, and the model's arguments (when applicable) and notes about the model. The sub-list name will be used as the model ID. Possible models:

arima - model from the stats package

auto.arima - model from the forecast package

ets - model from the forecast package

HoltWinters - model from the stats package

nnetar - model from the forecast package

tslm - model from the forecast package (note that the 'tslm' model must have

the formula argument in the 'method\_arg' argument)

train\_method

A list, defines the backtesting parameters:

partitions - an integer, set the number of training and testing partitions to be used in the backtesting process, where when partition is set to 1 it is a simple holdout training approach

space - an integer, defines the length of the backtesting window expansion

sample.in - an integer, optional, defines the length of the training partitions, and therefore the backtesting window structure. By default, it set to NULL and therefore, the backtesting using expending window. Otherwise, when the sample.in defined, the window structure is sliding

sample.in - an integer, optional, defines the length of the training partitions, and therefore the type of the backtesting window. By default, is set to NULL, which implay that the backtesting is using an expending window. Otherwise, when defining the size of the training partition, th defines the train approach, either using a single testing partition (sample out) or use multiple testing partitions (backtesting). The list should include the training method argument, (please see

'details' for the structure of the argument)

horizon

An integer, defines the forecast horizon

18 ts\_cor

A character, defines the error metrics to be used to sort the models leaderboard.

Possible metric - "MAPE" or "RMSE"

xreg

Optional, a list with two vectors (e.g., data.frame or matrix) of external regressors, one vector corresponding to the input series and second to the forecast itself (e.g., must have the same length as the input and forecast horizon, respectively)

level

An integer, set the confidence level of the prediction intervals

#### **Examples**

```
## Not run:
# Defining the models and their arguments
methods <- list(ets1 = list(method = "ets",</pre>
                            method_arg = list(opt.crit = "lik"),
                            notes = "ETS model with opt.crit = lik"),
                ets2 = list(method = "ets",
                            method_arg = list(opt.crit = "amse"),
                            notes = "ETS model with opt.crit = amse"),
                arima1 = list(method = "arima",
                              method_arg = list(order = c(2,1,0)),
                              notes = "ARIMA(2,1,0)"),
                arima2 = list(method = "arima",
                              method_arg = list(order = c(2,1,2),
                                                 seasonal = list(order = c(1,1,1)),
                              notes = "SARIMA(2,1,2)(1,1,1)"),
                hw = list(method = "HoltWinters",
                          method_arg = NULL,
                          notes = "HoltWinters Model"),
                tslm = list(method = "tslm",
                            method_arg = list(formula = input ~ trend + season),
                            notes = "tslm model with trend and seasonal components"))
# Training the models with backtesting
md <- train_model(input = USgas,</pre>
                  methods = methods,
                  train_method = list(partitions = 4,
                                       sample.out = 12,
                                       space = 3),
                  horizon = 12,
                  error = "MAPE")
# View the model performance on the backtesting partitions
md$leaderboard
## End(Not run)
```

ts\_cor

An Interactive Visualization of the ACF and PACF Functions

#### **Description**

An Interactive Visualization of the ACF and PACF Functions

ts\_decompose 19

## Usage

```
ts_cor(ts.obj, type = "both", seasonal = TRUE, ci = 0.95,
  lag.max = NULL, seasonal_lags = NULL)
```

## Arguments

ts.obj	A univariate time series object class 'ts'
type	A character, defines the plot type - 'acf' for ACF plot, 'pacf' for PACF plot, and 'both' (default) for both ACF and PACF plots
seasonal	A boolean, when set to TRUE (default) will color the seasonal lags
ci	The significant level of the estimation - a numeric value between 0 and 1, default is set for $0.95$
lag.max	maximum lag at which to calculate the acf. Default is $10*log10(N/m)$ where N is the number of observations and m the number of series. Will be automatically limited to one less than the number of observations in the series
seasonal_lags	A vector of integers, highlight specific cyclic lags (besides the main seasonal lags of the series). This is useful when working with multiseasonal time series data. For example, for a monthly series (e.g., frequency 12) setting the argument to 3 will highlight the quarterly lags

## **Examples**

```
data(USgas)

ts_cor(ts.obj = USgas)

# Setting the maximum number of lags to 72
ts_cor(ts.obj = USgas, lag.max = 72)

# Plotting only ACF
ts_cor(ts.obj = USgas, lag.max = 72, type = "acf")
```

ts\_decompose

Visualization of the Decompose of a Time Series Object

## **Description**

Interactive visualization the trend, seasonal and random components of a time series based on the decompose function from the stats package.

## Usage

```
ts_decompose(ts.obj, type = "additive", showline = TRUE)
```

20 ts\_grid

#### **Arguments**

ts.obj a univariate time series object of a class "ts", "zoo" or "xts"

type Set the type of the seasonal component, can be set to either "additive", "mul-

tiplicative" or "both" to compare between the first two options (default set to

"additive")

showline Logic, add a separation line between each of the plot components (default set to

TRUE)

#### **Examples**

```
# Defualt decompose plot
ts_decompose(AirPassengers)

# Remove the sepration lines between the plot components
ts_decompose(AirPassengers, showline = FALSE)

# Plot side by side a decompose of additive and multiplicative series
ts_decompose(AirPassengers, type = "both")
```

ts\_grid

Tuning Time Series Forecasting Models Parameters with Grid Search

## Description

Tuning time series models with grid search approach using backtesting method. If set to "auto" (default), will use all available cores in the system minus 1

#### **Usage**

```
ts_grid(ts.obj, model, optim = "MAPE", periods, window_length = NULL,
  window_space, window_test, hyper_params, parallel = TRUE,
  n.cores = "auto")
```

#### **Arguments**

ts.obj A univariate time series object of a class "ts"

model A string, defines the model c("HoltWinters"), currently support only Holt-Winters

model

optim A string, set the optimization method - c("MAPE", "RMSE")

periods A string, set the number backtesting periods

window\_length An integer, defines the length of the backtesting training window. If set to NULL

(default) will use an expending window starting the from the first observation,

otherwise will use a sliding window.

window\_space An integer, set the space length between each of the backtesting training parti-

tion

ts\_grid 21

window\_test An integer, set the length of the backtesting testing partition

hyper\_params A list, defines the tuning parameters and their range

parallel Logical, if TRUE use multiple cores in parallel

n.cores Set the number of cores to use if the parallel argument is set to TRUE. If set to

"auto" (default), will use n-1 of the available cores

#### Value

A list

```
## Not run:
data(USgas)
 # Starting with a shallow search (sequence between 0 and 1 with jumps of 0.1)
 # To speed up the process, will set the parallel option to TRUE
 # to run the search in parallel using 8 cores
 hw_grid_shallow <- ts_grid(ts.obj = USgas,</pre>
                             periods = 6,
                             model = "HoltWinters",
                             optim = "MAPE",
                             window_space = 6,
                             window_test = 12,
                             hyper_params = list(alpha = seq(0.01, 1, 0.1),
                                                   beta = seq(0.01, 1, 0.1),
                                                   gamma = seq(0.01, 1, 0.1)),
                             parallel = TRUE,
                             n.cores = 8)
 # Use the parameter range of the top 20 models
 # to set a narrow but more agressive search
 a_min <- min(hw_grid_shallow$grid_df$alpha[1:20])</pre>
 a_max <- max(hw_grid_shallow$grid_df$alpha[1:20])</pre>
 b_min <- min(hw_grid_shallow$grid_df$beta[1:20])</pre>
 b_max <- max(hw_grid_shallow$grid_df$beta[1:20])</pre>
 g_min <- min(hw_grid_shallow$grid_df$gamma[1:20])</pre>
 g_max <- max(hw_grid_shallow$grid_df$gamma[1:20])</pre>
 hw_grid_second <- ts_grid(ts.obj = USgas,</pre>
                            periods = 6,
                            model = "HoltWinters",
                            optim = "MAPE",
                            window_space = 6,
                            window_test = 12,
                            hyper_params = list(alpha = seq(a_min, a_max, 0.05),
```

22 ts\_heatmap

ts\_heatmap

Heatmap Plot for Time Series

## Description

Heatmap plot for time series object by it periodicity (currently support only daily, weekly, monthly and quarterly frequencies)

## Usage

```
ts_heatmap(ts.obj, last = NULL, wday = TRUE, color = "Blues",
   title = NULL, padding = TRUE)
```

## **Arguments**

ts.obj	A univariate time series object of a class "ts", "zoo", "xts", and the data frame family (data.frame, data.table, tbl, tibble, etc.) with a Date column and at least one numeric column. This function support time series objects with a daily, weekly, monthly and quarterly frequencies
last	An integer (optional), set a subset using only the last observations in the series
wday	An boolean, provides a weekday veiw for daily data (relevent only for objects with dates such as xts, zoo, data.frame, etc.)
color	A character, setting the color palette of the heatmap. Corresponding to any of the RColorBrewer palette or any other arguments of the col_numeric function. By default using the "Blues" palette
title	A character (optional), set the plot title
padding	A boolean, if TRUE will add to the heatmap spaces between the observations

ts\_info 23

## **Examples**

```
data(USgas)
ts_heatmap(USgas)

# Show only the last 4 years
ts_heatmap(USgas, last = 4 *12)
```

ts\_info

Get the Time Series Information

## Description

Returning the time series object main characteristics

## Usage

```
ts_info(ts.obj)
```

## **Arguments**

ts.obj

A time series object of a class "ts", "mts", "xts", or "zoo"

## Value

Text

```
# ts object
data("USgas")
ts_info(USgas)

# mts object
data("Coffee_Prices")
ts_info(Coffee_Prices)

# xts object
data("Michigan_CS")
ts_info(Michigan_CS)
```

24 ts\_lags

Time Series Lag Visualization

## **Description**

Visualization of series with its lags, can be used to identify a correlation between the series and it lags

## Usage

```
ts_lags(ts.obj, lags = 1:12, margin = 0.02, Xshare = TRUE,
    Yshare = TRUE, n_plots = 3)
```

## Arguments

ts.obj	A univariate time series object of a class "ts", "zoo" or "xts"
lags	An integer, set the lags range, by default will plot the first 12 lags
margin	Plotly parameter, either a single value or four values (all between 0 and 1). If four values provided, the first will be used as the left margin, the second will be used as the right margin, the third will be used as the top margin, and the fourth will be used as the bottom margin. If a single value provided, it will be used as all four margins.
Xshare	Plotly parameter, should the x-axis be shared amongst the subplots?
Yshare	Plotly parameter, should the y-axis be shared amongst the subplots?
n_plots	An integer, define the number of plots per row

```
data(USgas)
# Plot the first 12 lags (default)
ts_lags(USgas)
# Plot the seasonal lags for the first 4 years (hence, lag 12, 24, 36, 48)
ts_lags(USgas, lags = c(12, 24, 36, 48))
# Setting the margin between the plot
ts_lags(USgas, lags = c(12, 24, 36, 48), margin = 0.01)
```

ts\_ma 25

ts_ma	Moving Average Method for Time Series Data

## Description

Calculate the moving average (and double moving average) for time series data

## Usage

```
ts_ma(ts.obj, n = c(3, 6, 9), n_left = NULL, n_right = NULL,
double = NULL, plot = TRUE, show_legend = TRUE, multiple = FALSE,
separate = TRUE, margin = 0.03, title = NULL, Xtitle = NULL,
Ytitle = NULL)
```

## Arguments

ts.obj	a univariate time series object of a class "ts", "zoo" or "xts" (support only series with either monthly or quarterly frequency)
n	A single or multiple integers (by default using 3, 6, and 9 as inputs), define a two-sides moving averages by setting the number of past and future to use in each moving average window along with current observation.
n_left	A single integer (optional argument, default set to NULL), can be used, along with the n_right argument, an unbalanced moving average. The n_left defines the number of lags to includes in the moving average.
n_right	A single integer (optional argument, default set to NULL), can be used, along with the n_left argument, to set an unbalanced moving average. The n_right defines the number of negative lags to includes in the moving average.
double	A single integer, an optional argument. If not NULL (by default), will apply a second moving average process on the initial moving average output
1 .	A L L COMPANY III L L L
plot	A boolean, if TRUE will plot the results
plot show_legend	A boolean, if TRUE will plot the results  A boolean, if TRUE will show the plot legend
•	-
show_legend	A boolean, if TRUE will show the plot legend  A boolean, if TRUE (and n > 1) will create multiple plots, one for each moving
show_legend multiple	A boolean, if TRUE will show the plot legend  A boolean, if TRUE (and n > 1) will create multiple plots, one for each moving average degree. By default is set to FALSE  A boolean, if TRUE will separate the original series from the moving average
show_legend multiple separate	A boolean, if TRUE will show the plot legend  A boolean, if TRUE (and n > 1) will create multiple plots, one for each moving average degree. By default is set to FALSE  A boolean, if TRUE will separate the original series from the moving average output  A numeric, set the plot margin when using the multiple or/and separate option,
show_legend multiple separate margin	A boolean, if TRUE will show the plot legend  A boolean, if TRUE (and n > 1) will create multiple plots, one for each moving average degree. By default is set to FALSE  A boolean, if TRUE will separate the original series from the moving average output  A numeric, set the plot margin when using the multiple or/and separate option, default value is 0.03

26 ts\_ma

#### **Details**

A one-side moving averages (also known as simple moving averages) calculation for Y[t] (observation Y of the series at time t):

```
MA[t|n] = (Y[t-n] + Y[t-(n-1)] + ... + Y[t]) / (n + 1),
```

where n defines the number of consecutive observations to be used on each rolling window along with the current observation

Similarly, a two-sided moving averages with an order of (2\*n + 1) for Y[t]:

```
MA[t|n] = (Y[t-n] + Y[t-(n-1)] + ... + Y[t] + ... + Y[t+(n-1)] + Y[t+n]) / (2*n + 1)
```

Unbalanced moving averages with an order of (k1 + k2 + 1) for observation Y[t]:

$$MA[t|k1 & k2] = (Y[t-k1] + Y[t-(k1-1)] + ... + Y[t] + ... + Y[t+(k2-1)] + Y[t+k2]) / (k1 + k2 + 1)$$

The unbalanced moving averages is a special case of two-sides moving averages, where k1 and k2 represent the number of past and future periods, respectively to be used in each rolling window, and k1 = k2 (otherwise it is a normal two-sided moving averages function)

#### Value

A list with the original series, the moving averages outputs and the plot

```
## Not run:
# A one-side moving average order of 7
USgas_MA7 <- ts_ma(USgas, n_left = 6, n = NULL)</pre>
# A two-sided moving average order of 13
USgas_two_side_MA <- ts_ma(USgas, n = 6)</pre>
# Unbalanced moving average of order 12
USVSales_MA12 <- ts_ma(USVSales, n_left = 6, n_right = 5, n = NULL,
 title = "US Monthly Total Vehicle Sales - MA",
 Ytitle = "Thousand of Units")
# Adding double MA of order 2 to balanced the series:
USVSales_MA12 <- ts_ma(USVSales, n_left = 6, n_right = 5, n = NULL,
 double = 2,
 title = "US Monthly Total Vehicle Sales - MA",
 Ytitle = "Thousand of Units")
# Adding several types of two-sided moving averages along with the unblanced
# Plot each on a separate plot
USVSales_MA12 <- ts_ma(USVSales, n_left = 6, n_right = 5, n = c(3, 6, 9),
double = 2, multiple = TRUE,
title = "US Monthly Total Vehicle Sales - MA",
Ytitle = "Thousand of Units")
## End(Not run)
```

ts\_plot 27

ts_plot	Plotting Time Series Objects	

## Description

Visualization functions for time series object

## Usage

```
ts_plot(ts.obj, line.mode = "lines", width = 2, dash = NULL,
color = NULL, slider = FALSE, type = "single", Xtitle = NULL,
Ytitle = NULL, title = NULL, Xgrid = FALSE, Ygrid = FALSE)
```

## Arguments

ts.obj	A univariate or multivariate time series object of class "ts", "mts", "zoo", "xts", or any data frame object with a minimum of one numeric column and either a Date or POSIXt class column
line.mode	A plotly argument, define the plot type, c("lines", "lines+markers", "markers")
width	An Integer, define the plot width, default is set to 2
dash	A plotly argument, define the line style, c(NULL, "dot", "dash")
color	The color of the plot, support both name and expression
slider	Logic, add slider to modify the time axis (default set to FALSE)
type	A character, optional, if having multiple tims series object, will plot all series in one plot when set to "single" (default), or plot each series on a separate plot when set to "multiple"
Xtitle	A character, set the X axis title, default set to NULL
Ytitle	A character, set the Y axis title, default set to NULL
title	A character, set the plot title, default set to NULL
Xgrid	Logic, show the X axis grid if set to TRUE
Ygrid	Logic, show the Y axis grid if set to TRUE

```
data(USVSales)
ts_plot(USVSales)

# adding slider
ts_plot(USVSales, slider = TRUE)
```

28 ts\_quantile

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ts	pol	.ar

Polor Plot for Time Series Object

## **Description**

Polor plot for time series object (ts, zoo, xts), currently support only monthly and quarterly frequency

## Usage

```
ts_polar(ts.obj, title = NULL, width = 600, height = 600,
  left = 25, right = 25, top = 25, bottom = 25)
```

## Arguments

ts.obj	A univariate time series object of a class "ts", "zoo" or "xts" (support only series with either monthly or quarterly frequency)
title	Add a title for the plot, default set to NULL
width	The widht of the plot in pixels, default set to 600
height	The height of the plot pixels, default set to 600
left	Set the left margin of the plot in pixels, default set to 25
right	Set the right margin of the plot in pixels, default set to 25
top	Set the top margin of the plot in pixels, default set to 25
bottom	Set the bottom margin of the plot in pixels, default set to 25

## **Examples**

```
data(USgas)
ts_polar(USgas)
```

ts\_quantile

Quantile Plot for Time Series

## **Description**

A quantile plot of time series data, allows the user to display a quantile plot of a series by a subset period

## Usage

```
ts_quantile(ts.obj, upper = 0.75, lower = 0.25, period = NULL,
    n = 1, title = NULL, Xtitle = NULL, Ytitle = NULL)
```

ts\_quantile 29

## **Arguments**

ts.obj	A univariate time series object of a class "zoo", "xts", or data frame family ("data.frame", "data.table", "tbl")
upper	A numeric value between 0 and 1 (excluding 0, and greater than the "lower" argument) set the upper bound of the quantile plot (using the "probs" argument of the quantile function). By default set to 0.75
lower	A numeric value between 0 and 1 (excluding 1, and lower than the "upper" argument) set the upper bound of the quantile plot (using the "probs" argument of the quantile function). By default set to 0.25
period	A character, set the period level of the data for the quantile calculation and plot representation. Must be one level above the input frequency (e.g., an hourly data can represent by daily, weekdays, monthly, quarterly and yearly). Possible options c("daily", "weekdays", "monthly", "quarterly", "yearly")
n	An integer, set the number of plots rows to display (by setting the nrows argument in the subplot function), must be an integer between 1 and the frequency of the period argument.
title	A character, set the plot title, default set to NULL
Xtitle	A character, set the X axis title, default set to NULL
Ytitle	A character, set the Y axis title, default set to NULL

```
## Not run:
# Loading the UKgrid package to pull a multie seasonality data
require(UKgrid)
UKgrid_half_hour <- extract_grid(type = "xts", aggregate = NULL)</pre>
# Plotting the quantile of the UKgrid dataset
# No period subset
ts_quantile(UKgrid_half_hour,
period = NULL,
title = "The UK National Grid Net Demand for Electricity - Quantile Plot")
# Plotting the quantile of the UKgrid dataset
# Using a weekday subset
ts_quantile(UKgrid_half_hour,
period = "weekdays",
title = "The UK National Grid Net Demand for Electricity - by Weekdays")
# Spacing the plots by setting the
# number of rows of the plot to 2
ts_quantile(UKgrid_half_hour,
period = "weekdays",
title = "The UK National Grid Net Demand for Electricity - by Weekdays",
n = 2
```

30 ts\_seasonal

```
## End(Not run)
```

ts\_reshape

Transform Time Series Object to Data Frame Format

#### Description

Transform time series object into data frame format

## Usage

```
ts_reshape(ts.obj, type = "wide", frequency = NULL)
```

#### **Arguments**

ts.obj a univariate time series object of a class "ts", "zoo", "xts", and the data frame

family (data.frame, data.table, tbl, tibble, etc.) with a Date column and at least one numeric column. This function support time series objects with a daily,

weekly, monthly or quarterly frequencies

type The reshape type -

"wide" set the years as the columns and the cycle units (months or quarter) as

the rows, or

"long" split the time object to year, cycle unit and value

frequency An integer, define the series frequency when more than one option is available

and the input is one of the data frame family. If set to NULL will use the first

option by default when applicable - daily = c(7, 365)

#### **Examples**

```
data(USgas)
USgas_df <- ts_reshape(USgas)</pre>
```

ts\_seasonal

Seasonality Visualization of Time Series Object

## Description

Visualize time series object by it periodicity, currently support time series with daily, monthly and quarterly frequency

#### Usage

```
ts_seasonal(ts.obj, type = "normal", title = NULL, Ygrid = TRUE,
  Xgrid = TRUE, last = NULL, palette = "Set1",
  palette_normal = "viridis")
```

ts\_split 31

## **Arguments**

ts.obj	Input object, either a univariate time series object of a class "ts", "zoo", "xts", or a data frame object of a class "data.frame", "tbl", "data.table" as long as there is at least one "Date"/"POSIXt" and a "numeric" objects (if there are more then one, by defualt will use the first of each). Currently support only daily, weekly, monthly, and quarterly frequencies
type	The type of the seasonal plot - "normal" to split the series by full cycle units, or "cycle" to split by cycle units (applicable only for monthly and quarterly data), or "box" for box-plot by cycle units, or "all" for all the three plots together
title	Plot title - Character object
Ygrid	Logic, show the Y axis grid if set to TRUE (default)
Xgrid	Logic, show the X axis grid if set to TRUE (defualt)
last	Subset the data to the last number of observations
palette	A character, the color palette to be used when the "cycle" or "box" plot are being selected (by setting the type to "cycle", "box", or "all"). All the palettes in the RColorBrewer and viridis packages are available to be use, the default option is "Set1" from the RColorBrewer package
palette_normal	A character, the color palette to be used when the "normal" plot is being selected (by setting the type to "normal" or "all"). All the palettes in the RColorBrewer and viridis packages are available to be used, the default palette is "viridis" from the RColorBrewer package

## **Examples**

```
data(USgas)
ts_seasonal(USgas)

# Seasonal box plot
ts_seasonal(USgas, type = "box")

# Plot all the types
ts_seasonal(USgas, type = "all")
```

ts\_split

Split Time Series Object for Training and Testing Partitions

## Description

Split a time series object into training and testing partitions

## Usage

```
ts_split(ts.obj, sample.out = NULL)
```

32 ts\_sum

## **Arguments**

ts.obj A univariate time series object of a class "ts" or "tsibble"

sample.out An integer, set the number of periods of the testing or sample out partition,

defualt set for 30 percent of the lenght of the series

#### **Examples**

```
## Split the USgas dataset into training and testing partitions
## Set the last 12 months as a testing partition
## and the rest as a training partition
data(USgas, package = "TSstudio")
split_USgas <- ts_split(ts.obj = USgas, sample.out = 12)
training <- split_USgas$train
testing <- split_USgas$test
length(USgas)
length(training)
length(testing)</pre>
```

ts\_sum

Summation of Multiple Time Series Objects

## Description

A row sum function for multiple time series object ("mts"), return the summation of the "mts" object as a "ts" object

## Usage

```
ts_sum(mts.obj)
```

#### **Arguments**

mts.obj

A multivariate time series object of a class "mts"

```
x <- matrix(c(1:100, 1:100, 1:100), ncol = 3)
mts.obj <- ts(x, start = c(2000, 1), frequency = 12)
ts_total <- ts_sum(mts.obj)</pre>
```

ts\_surface 33

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3D Surface Plot for Time Series

## Description

3D surface plot for time series object by it periodicity (currently support only monthly and quarterly frequency)

## Usage

```
ts_surface(ts.obj)
```

#### **Arguments**

ts.obj

a univariate time series object of a class "ts", "zoo" or "xts" (support only series with either monthly or quarterly frequency)

## **Examples**

```
ts_surface(USgas)
```

ts_to_prophet	ts	prophe	t
---------------	----	--------	---

Transform Time Series Object to Prophet input

## **Description**

Transform a time series object to Prophet data frame input format

## Usage

```
ts_to_prophet(ts.obj, start = NULL)
```

## Arguments

	A univariate time series ob		
ts.obi			

monthly, quarterly or yearly frequency

start A date object (optional), if the starting date of the series is known. Otherwise,

the date would be derive from the series index

#### Value

A data frame object

34 USgas

## **Examples**

```
data(USgas)

ts_to_prophet(ts.obj = USgas)

# If known setting the start date of the input object

ts_to_prophet(ts.obj = USgas, start = as.Date("2000-01-01"))
```

**USgas** 

US monthly natural gas consumption

## Description

US monthly natural gas consumption: 2000 - 2019. Units: Billion Cubic Feet

## Usage

USgas

#### **Format**

Time series data - 'ts' object

## **Source**

U.S. Bureau of Transportation Statistics, Natural Gas Consumption [NATURALGAS], retrieved from FRED, Federal Reserve Bank of St. Louis; https://fred.stlouisfed.org/series/NATURALGAS, January 7, 2018.

```
ts_plot(USgas)
ts_seasonal(USgas, type = "all")
```

USUnRate 35

USUnRate

US Monthly Civilian Unemployment Rate

## **Description**

US monthly civilian unemployment rate: 1948 - 2019. Units: Percent

## Usage

USUnRate

#### **Format**

Time series data - 'ts' object

#### **Source**

U.S. Bureau of Labor Statistics, Civilian Unemployment Rate [UNRATENSA], retrieved from FRED, Federal Reserve Bank of St. Louis; https://fred.stlouisfed.org/series/UNRATENSA, January 6, 2018.

## **Examples**

```
ts_plot(USUnRate)
ts_seasonal(USUnRate)
```

USVSales

US Monthly Total Vehicle Sales

## **Description**

US monthly total vehicle sales: 1976 - 2019. Units: Thousands of units

## Usage

USVSales

## **Format**

Time series data - 'ts' object

#### **Source**

U.S. Bureau of Economic Analysis, Total Vehicle Sales [TOTALNSA], retrieved from FRED, Federal Reserve Bank of St. Louis; https://fred.stlouisfed.org/series/TOTALNSA, January 7, 2018.

36 xts\_to\_ts

#### **Examples**

```
ts_plot(USVSales)
ts_seasonal(USVSales)
```

US\_indicators

US Key Indicators - data frame format

## **Description**

Monthly total vehicle sales and unemployment rate: 1976 - 2019. Units: Dollars per Kg

## Usage

US\_indicators

#### **Format**

Time series data - 'data.frame' object

#### **Source**

U.S. Bureau of Economic Analysis, Total Vehicle Sales [TOTALNSA], retrieved from FRED, Federal Reserve Bank of St. Louis; https://fred.stlouisfed.org/series/TOTALNSA, January 7, 2018. U.S. Bureau of Labor Statistics, Civilian Unemployment Rate [UNRATENSA], retrieved from FRED, Federal Reserve Bank of St. Louis; https://fred.stlouisfed.org/series/UNRATENSA, January 6, 2018.

#### **Examples**

```
ts_plot(US_indicators)
```

xts\_to\_ts

Converting 'xts' object to 'ts' object

## Description

Converting 'xts' object to 'ts' object

## Usage

```
xts_to_ts(xts.obj, frequency = NULL, start = NULL)
```

zoo\_to\_ts 37

## **Arguments**

xts.obj A univariate 'xts' object

frequency A character, optional, if not NULL (default) set the frequency of the series

start A Date or POSIXct/lt object, optional, can be used to set the starting date or

time of the series

## **Examples**

```
data(Michigan_CS)
class(Michigan_CS)
ts_plot(Michigan_CS)
Michigan_CS_ts <- xts_to_ts(Michigan_CS)
ts_plot(Michigan_CS_ts)

# Defining the frequency and starting date of the series
Michigan_CS_ts1 <- xts_to_ts(Michigan_CS, start = as.Date("1980-01-01"), frequency = 12 )
ts_plot(Michigan_CS_ts1)</pre>
```

zoo\_to\_ts

Converting 'zoo' object to 'ts' object

## **Description**

Converting 'zoo' object to 'ts' object

## Usage

```
zoo_to_ts(zoo.obj)
```

## **Arguments**

zoo.obj a univariate 'zoo' object

```
data("EURO_Brent", package = "TSstudio")
class(EURO_Brent)
ts_plot(EURO_Brent)
EURO_Brent_ts <- zoo_to_ts(EURO_Brent)
class(EURO_Brent_ts)
ts_plot(EURO_Brent_ts)</pre>
```

# **Index**

* datasets Coffee_Prices, 5 EURO_Brent, 8	<pre>plot_grid, 13 plot_model, 13</pre>
Michigan_CS, 10	quantile, 29
US_indicators, 36 USgas, 34 USUnRate, 35	<pre>remove_methods (create_model), 6 res_hist, 15</pre>
USVSales, 35	<pre>set_error(create_model), 6</pre>
<pre>add_horizon(create_model), 6 add_input(create_model), 6</pre>	simulate, 9 subplot, 29
add_level (create_model), 6	test_forecast, 16
add_methods(create_model), 6	train_model, 6, 17
add_train_method(create_model), 6	ts_cor, 18
<pre>add_xreg(create_model), 6 Arima, 9</pre>	ts_decompose, 19
arima, 6, 17	ts_grid, 20
arima_diag, 3	ts_heatmap, 22
auto.arima, $7$ , $9$ , $17$	ts_info, 23
du to. di 1111d, 7, 9, 17	ts_lags, 24
<pre>build_model (create_model), 6</pre>	ts_ma, 25 ts_plot, 27
ccf_plot, 4	ts_polar,28
check_res, 5	ts_quantile, 28
Coffee_Prices, 5	ts_reshape, 30
col_numeric, 22	ts_seasonal, 30
create_model, 6	ts_split,31
	ts_sum, 32
ets, 7, 9, 17	ts_surface, 33
EURO_Brent, 8	ts_to_prophet, 33 tslm, 7, 17
forecast_sim, 9	US_indicators, 36
HoltWinters, 7, 17	USgas, 34
1101 CH 111 CC 3, 7, 17	USUnRate, 35
Michigan_CS, 10	USVSales, 35
nnetar, 7, 9, 17	xts_to_ts, 36
plot_error, 11	zoo_to_ts, 37
plot_forecast, 12	