

Package ‘forecastML’

July 22, 2025

Type Package

Title Time Series Forecasting with Machine Learning Methods

Version 0.9.0

Author Nickalus Redell

Maintainer Nickalus Redell <nickalusredell@gmail.com>

Description The purpose of 'forecastML' is to simplify the process of multi-step-ahead forecasting with standard machine learning algorithms. 'forecastML' supports lagged, dynamic, static, and grouping features for modeling single and grouped numeric or factor/sequence time series. In addition, simple wrapper functions are used to support model-building with most R packages. This approach to forecasting is inspired by Bergmeir, Hyndman, and Koo's (2018) paper ``A note on the validity of cross-validation for evaluating autoregressive time series prediction" <doi:10.1016/j.csda.2017.11.003>.

License MIT + file LICENSE

URL <https://github.com/nredell/forecastML/>

Encoding UTF-8

LazyData true

Imports tidyrr (>= 0.8.1), rlang (>= 0.4.0), magrittr (>= 1.5),
lubridate (>= 1.7.4), ggplot2 (>= 3.1.0), future.apply (>= 1.3.0), methods, purrr (>= 0.3.2), data.table (>= 1.12.6),
dplyr (>= 1.0.0), tibble (>= 2.1.3)

RoxxygenNote 7.1.0

Collate 'fill_gaps.R' 'create_windows.R' 'create_skeleton.R'
'combine_forecasts.R' 'lagged_df.R' 'return_error.R'
'return_hyper.R' 'train_model.R' 'data_seatbelts.R'
'data_buoy.R' 'data_buoy_gaps.R' 'zzz.R'

Depends R (>= 3.5.0), dplyr (>= 0.8.3)

Suggests glmnet (>= 2.0.16), DT (>= 0.5), knitr (>= 1.22), rmarkdown
(>= 1.12.6), xgboost (>= 0.82.1), randomForest (>= 4.6.14),
testthat (>= 2.2.1), covr (>= 3.3.1)

VignetteBuilder knitr

NeedsCompilation no

Repository CRAN
Date/Publication 2020-05-07 15:10:17 UTC

Contents

combine_forecasts	2
create_lagged_df	5
create_skeleton	9
create_windows	9
data_buoy	11
data_buoy_gaps	12
data_seatbelts	12
fill_gaps	13
plot.forecastML	14
plot.forecast_error	15
plot.forecast_model_hyper	16
plot.forecast_results	19
plot.lagged_df	20
plot.training_results	21
plot.validation_error	22
plot.windows	23
predict.forecast_model	24
return_error	27
return_hyper	29
summary.lagged_df	31
train_model	32
Index	35

combine_forecasts	<i>Combine multiple horizon-specific forecast models to produce one forecast</i>
-------------------	--

Description

The horizon-specific models can either be combined to (a) produce final forecasts for only those horizons at which they were trained (i.e., shorter-horizon models override longer-horizon models when producing final short-horizon h-step-ahead forecasts) or (b) produce final forecasts using any combination of horizon-specific models that minimized error over the validation/training dataset.

Usage

```
combine_forecasts(  
  ...,  
  type = c("horizon", "error"),  
  aggregate = stats::median,  
  data_error = list(NULL),
```

```
metric = NULL
)
```

Arguments

...	One or more objects of class 'forecast_results' from running <code>predict.forecast_model()</code> on an input forward-looking forecast dataset. These are the forecasts from the horizon-specific direct forecasting models trained over the entire training dataset by setting <code>create_windows(..., window_length = 0)</code> . If multiple models are passed in ... with the same direct forecast horizon, for <code>type = 'horizon'</code> , forecasts for the same direct forecast horizon are combined with <code>aggregate</code> ; for <code>type = 'error'</code> , the model that minimizes the error metric at the given direct forecast horizon produces the forecast.
type	Default: 'horizon'. A character vector of length 1 that identifies the forecast combination method.
aggregate	Default median for <code>type = 'horizon'</code> . A function—without parentheses—that aggregates forecasts if more than one model passed in ... has the same direct forecast horizon and <code>type = 'horizon'</code>].
data_error	Optional. A list of objects of class 'validation_error' from running <code>return_error()</code> on a training dataset. The length and order of <code>data_error</code> should match the models passed in
metric	Required if <code>data_error</code> is given. A length 1 character vector naming the forecast error metric used to select the optimal model at each forecast horizon from the models passed in '...' e.g., 'mae'.

Value

An S3 object of class 'forecastML' with final h-step-ahead forecasts.

Forecast combination type:

- `type = 'horizon'`: 1 final h-step-ahead forecast is returned for each model object passed in
- `type = 'error'`: 1 final h-step-ahead forecast is returned by selecting, for each forecast horizon, the model that minimized the chosen error metric at that horizon on the outer-loop validation data sets.

Columns in returned 'forecastML' dataframe:

- `model`: User-supplied model name in `train_model()`.
- `model_forecast_horizon`: The direct-forecasting time horizon that the model was trained on.
- `horizon`: Forecast horizons, 1:h, measured in dataset rows.
- `forecast_period`: The forecast period in row indices or dates. The forecast period starts at either `attributes(create_lagged_df())$data_stop + 1` for row indices or `attributes(create_lagged_df())$date_start + 1 * frequency` for date indices.
- `"groups"`: If given, the user-supplied groups in `create_lagged_df()`.
- `"outcome_name"_pred`: The final forecasts.

- "outcome_name"_pred_lower: If given, the lower forecast bounds returned by the user-supplied prediction function.
- "outcome_name"_pred_upper: If given, the upper forecast bounds returned by the user-supplied prediction function.

Methods and related functions

The output of `combine_forecasts()` has the following generic S3 methods

- `plot`

Examples

```
# Example with "type = 'horizon'".
data("data_seatbelts", package = "forecastML")

horizons <- c(1, 3, 12)
lookback <- 1:15

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback = lookback, horizon = horizons)

windows <- create_windows(data_train, window_length = 0)

model_function <- function(data, my_outcome_col) {
  model <- lm(DriversKilled ~ ., data = data)
  return(model)
}

model_results <- train_model(data_train, windows, model_name = "LM", model_function)

data_forecast <- create_lagged_df(data_seatbelts, type = "forecast", outcome_col = 1,
                              lookback = lookback, horizon = horizons)

prediction_function <- function(model, data_features) {
  x <- data_features
  data_pred <- data.frame("y_pred" = predict(model, newdata = x))
  return(data_pred)
}

data_forecasts <- predict(model_results, prediction_function = list(prediction_function),
                        data = data_forecast)

data_combined <- combine_forecasts(data_forecasts)

plot(data_combined)
```

create_lagged_df	<i>Create model training and forecasting datasets with lagged, grouped, dynamic, and static features</i>
------------------	--

Description

Create a list of datasets with lagged, grouped, dynamic, and static features to (a) train forecasting models for specified forecast horizons and (b) forecast into the future with a trained ML model.

Usage

```
create_lagged_df(
  data,
  type = c("train", "forecast"),
  method = c("direct", "multi_output"),
  outcome_col = 1,
  horizons,
  lookback = NULL,
  lookback_control = NULL,
  dates = NULL,
  frequency = NULL,
  dynamic_features = NULL,
  groups = NULL,
  static_features = NULL,
  predict_future = NULL,
  use_future = FALSE,
  keep_rows = FALSE
)
```

Arguments

data	A data.frame with the (a) target to be forecasted and (b) features/predictors. An optional date column can be given in the dates argument (required for grouped time series). Note that 'orecastML only works with regularly spaced date/time intervals and that missing rows—usually due to periods when no data was collected—will result in incorrect feature lags. Use fill_gaps to fill in any missing rows/data prior to running this function.
type	The type of dataset to return—(a) model training or (b) forecast prediction. The default is train.
method	The type of modeling dataset to create. direct returns 1 data.frame for each forecast horizon and multi_output returns 1 data.frame for simultaneously modeling all forecast horizons. The default is direct.
outcome_col	The column index—an integer—of the target to be forecasted. If outcome_col != 1, the outcome column will be moved to position 1 and outcome_col will be set to 1 internally.

horizons	A numeric vector of one or more forecast horizons, <i>h</i> , measured in dataset rows. If dates are given, a horizon of 1, for example, would equal 1 * frequency in calendar time.
lookback	A numeric vector giving the lags—in dataset rows—for creating the lagged features. All non-grouping, non-static, and non-dynamic features in the input dataset, data, are lagged by the same values. The outcome is also lagged by default. Either lookback or lookback_control need to be specified—but not both.
lookback_control	A list of numeric vectors, specifying potentially unique lags for each feature. The length of the list should equal ncol(data) and be ordered the same as the columns in data. Lag values for any grouping, static, or dynamic feature columns are automatically coerced to 0 and not lagged. list(NULL) lookback_control values drop columns from the input dataset. Either lookback or lookback_control need to be specified—but not both.
dates	A vector or 1-column data.frame of dates/times with class 'Date' or 'POSIXt'. The length of dates should equal nrow(data). Required if groups are given.
frequency	Date/time frequency. Required if dates are given. A string taking the same input as base::seq.Date(..., by = "frequency") or base::seq.POSIXt(..., by = "frequency") e.g., '1 hour', '1 month', '7 days', '10 years' etc. The highest frequency supported at present is '1 sec'.
dynamic_features	A character vector of column names that identify features that change through time but which are not lagged (e.g., weekday or year). If type = "forecast" and method = "direct", these features will receive NA values; though, they can be filled in by the user after running this function.
groups	A character vector of column names that identify the groups/hierarchies when multiple time series are present. These columns are used as model features but are not lagged. Note that combining feature lags with grouped time series will result in NA values throughout the data.
static_features	For grouped time series only. A character vector of column names that identify features that do not change through time. These columns are not lagged. If type = "forecast", these features will be filled forward using the most recent value for the group.
predict_future	When type = "forecast", a function for predicting the future values of any dynamic features. This function takes data and dates as positional arguments and returns a data.frame with (a) one or more rows, (b) an "index" column of future dates, (c) group columns if needed, and (d) 1 or more columns with name(s) in dynamic_features.
use_future	Boolean. If TRUE, the future.apply package is used for creating lagged data.frames. multisession or multicore futures are especially useful for (a) grouped time series with many groups and (b) high-dimensional datasets with many lags per feature. Run future::plan(future::multiprocess) prior to this function to set up multisession or multicore parallel dataset creation.
keep_rows	Boolean. For non-grouped time series, keep the 1:max(lookback) rows at the beginning of the time series. These rows will contain missing values for lagged features that "look back" before the start of the dataset.

Value

An S3 object of class 'lagged_df' or 'grouped_lagged_df': A list of data.frames with new columns for the lagged/non-lagged features. For method = "direct", the length of the returned list is equal to the number of forecast horizons and is in the order of horizons supplied to the horizons argument. Horizon-specific datasets can be accessed with my_lagged_df\$horizon_h where 'h' gives the forecast horizon. For method = "multi_output", the length of the returned list is 1. Horizon-specific datasets can be accessed with my_lagged_df\$horizon_1_3_5 where "1_3_5" represents the forecast horizons passed in horizons.

The contents of the returned data.frames are as follows:

type = 'train', non-grouped: A data.frame with the outcome and lagged/dynamic features.

type = 'train', grouped: A data.frame with the outcome and unlagged grouping columns followed by lagged, dynamic, and static features.

type = 'forecast', non-grouped: (1) An 'index' column giving the row index or date of the forecast periods (e.g., a 100 row non-date-based training dataset would start with an index of 101). (2) A 'horizon' column that indicates the forecast period from 1:max(horizons). (3) Lagged features identical to the 'train', non-grouped dataset.

type = 'forecast', grouped: (1) An 'index' column giving the date of the forecast periods. The first forecast date for each group is the maximum date from the dates argument + 1 * frequency which is the user-supplied date/time frequency. (2) A 'horizon' column that indicates the forecast period from 1:max(horizons). (3) Lagged, static, and dynamic features identical to the 'train', grouped dataset.

Attributes

- names: The horizon-specific datasets that can be accessed with my_lagged_df\$horizon_h.
- type: Training, train, or forecasting, forecast, dataset(s).
- method: direct or multi_output.
- horizons: Forecast horizons measured in dataset rows.
- outcome_col: The column index of the target being forecasted.
- outcome_cols: If method = multi_output, the column indices of the multiple outputs in the transformed dataset.
- outcome_name: The name of the target being forecasted.
- outcome_names: If method = multi_output, the column names of the multiple outputs in the transformed dataset. The names take the form "outcome_name_h" where 'h' is a horizon passed in horizons.
- predictor_names: The predictor or feature names from the input dataset.
- row_indices: The row.names() of the output dataset. For non-grouped datasets, the first lookback + 1 rows are removed from the beginning of the dataset to remove NA values in the lagged features.
- date_indices: If dates are given, the vector of dates.
- frequency: If dates are given, the date/time frequency.
- data_start: min(row_indices) or min(date_indices).

- data_stop: max(row_indices) or max(date_indices).
- groups: If groups are given, a vector of group names.
- class: grouped_lagged_df, lagged_df, list

Methods and related functions

The output of create_lagged_df() is passed into

- [create_windows](#)

and has the following generic S3 methods

- [summary](#)
- [plot](#)

Examples

```
# Sampled Seatbelts data from the R package datasets.
data("data_seatbelts", package = "forecastML")
#-----
# Example 1 - Training data for 2 horizon-specific models w/ common lags per predictor.
horizons <- c(1, 12)
lookback <- 1:15

data <- data_seatbelts

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              horizons = horizons, lookback = lookback)
head(data_train[[length(horizons)]])

# Example 1 - Forecasting dataset
# The last 'nrow(data_seatbelts) - horizon' rows are automatically used from data_seatbelts.
data_forecast <- create_lagged_df(data_seatbelts, type = "forecast", outcome_col = 1,
                                  horizons = horizons, lookback = lookback)
head(data_forecast[[length(horizons)]])

#-----
# Example 2 - Training data for one 3-month horizon model w/ unique lags per predictor.
horizons <- 3
lookback <- list(c(3, 6, 9, 12), c(4:12), c(6:15), c(8))

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              horizons = horizons, lookback_control = lookback)
head(data_train[[length(horizons)]])
```

create_skeleton	<i>Remove the features from a lagged training dataset to reduce memory consumption</i>
-----------------	--

Description

create_skeleton() strips the feature data from a create_lagged_df() object but keeps the outcome column(s), any grouping columns, and meta-data which allows the resulting lagged_df to be used downstream in the forecastML pipeline. The main benefit is that the custom modeling function passed in train_model() can read data directly from the disk or a database when the dataset is too large to fit into memory.

Usage

```
create_skeleton(lagged_df)
```

Arguments

lagged_df An object of class 'lagged_df' from create_lagged_df(..., type = 'train').

Value

An S3 object of class 'lagged_df' or 'grouped_lagged_df': A list of data.frames with the outcome column(s) and any grouping columns but with all other features removed. A special attribute skeleton = TRUE is added.

Methods and related functions

The output of create_skeleton can be passed into

- [create_windows](#)

create_windows	<i>Create time-contiguous validation datasets for model evaluation</i>
----------------	--

Description

Flexibly create blocks of time-contiguous validation datasets to assess the forecast accuracy of trained models at various times in the past. These validation datasets are similar to the outer loop of a nested cross-validation model training setup.

Usage

```
create_windows(
  lagged_df,
  window_length = 12L,
  window_start = NULL,
  window_stop = NULL,
  skip = 0,
  include_partial_window = TRUE
)
```

Arguments

lagged_df	An object of class 'lagged_df' or 'grouped_lagged_df' from create_lagged_df .
window_length	An integer that defines the length of the contiguous validation dataset in dataset rows/dates. If dates were given in <code>create_lagged_df()</code> , the validation window is 'window_length' * 'date frequency' in calendar time. Setting window_length = 0 trains the model on (a) the entire dataset or (b) between a single window_start and window_stop value. Specifying multiple window_start and window_stop values with vectors of length > 1 overrides window_length.
window_start	Optional. A row index or date identifying the row/date to start creating contiguous validation datasets. A vector of start rows/dates can be supplied for greater control. The length and order of window_start should match window_stop. If length(window_start) > 1, window_length, skip, and include_partial_window are ignored.
window_stop	Optional. An index or date identifying the row/date to stop creating contiguous validation datasets. A vector of start rows/dates can be supplied for greater control. The length and order of window_stop should match window_start. If length(window_stop) > 1, window_length, skip, and include_partial_window are ignored.
skip	An integer giving a fixed number of dataset rows/dates to skip between validation datasets. If dates were given in <code>create_lagged_df()</code> , the time between validation windows is skip * 'date frequency'.
include_partial_window	Boolean. If TRUE, keep validation datasets that are shorter than window_length.

Value

An S3 object of class 'windows': A data.frame giving the indices for the validation datasets.

Methods and related functions

The output of `create_windows()` is passed into

- [train_model](#)

and has the following generic S3 methods

- [plot](#)

Examples

```
# Sampled Seatbelts data from the R package datasets.
data("data_seatbelts", package = "forecastML")

# Example - Training data for 2 horizon-specific models w/ common lags per feature.
horizons <- c(1, 12)
lookback <- 1:15

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback = lookback, horizon = horizons)

# All historical window lengths of 12 plus any partial windows at the end of the dataset.
windows <- create_windows(data_train, window_length = 12)
windows

# Two custom validation windows with different lengths.
windows <- create_windows(data_train, window_start = c(20, 80), window_stop = c(30, 100))
windows
```

data_buoy

*NOAA buoy weather data***Description**

A dataset containing daily average sensor measurements of several environmental conditions collected by 14 buoys in Lake Michigan from 2012 through 2018.

Usage

data_buoy

Format

A data.frame with 30,821 rows and 9 columns:

date date

wind_spd average daily wind speed in kts

buoy_id the station ID for each buoy

lat latitude

lon longitude

day day of year

year calendar year

air_temperature air temperature in degrees Fahrenheit

sea_surface_temperature water temperature in degrees Fahrenheit

Source

<http://www.ndbc.noaa.gov/>

data_buoy_gaps	<i>NOAA buoy weather data</i>
----------------	-------------------------------

Description

A dataset containing daily average sensor measurements of several environmental conditions collected by 14 buoys in Lake Michigan from 2012 through 2018. This dataset is identical to the [data_buoy](#) dataset except that there are gaps in the daily sensor data. Running `fill_gaps()` on `data_buoy_gaps` will produce `data_buoy`.

Usage

```
data_buoy_gaps
```

Format

A `data.frame` with 23,646 rows and 9 columns:

date date

wind_spd average daily wind speed in kts

buoy_id the station ID for each buoy

lat latitude

lon longitude

day day of year

year calendar year

air_temperature air temperature in degrees Fahrenheit

sea_surface_temperature water temperature in degrees Fahrenheit

Source

<http://www.ndbc.noaa.gov/>

data_seatbelts	<i>Road Casualties in Great Britain 1969-84</i>
----------------	---

Description

This is the Seatbelts dataset from the `datasets` package.

Usage

```
data_seatbelts
```

Format

A `data.frame` with 192 rows and 8 columns

Source

Harvey, A.C. (1989). Forecasting, Structural Time Series Models and the Kalman Filter. Cambridge University Press, pp. 519–523.

Durbin, J. and Koopman, S. J. (2001). Time Series Analysis by State Space Methods. Oxford University Press.

<https://stat.ethz.ch/R-manual/R-devel/library/datasets/html/UKDriverDeaths.html>

fill_gaps	<i>Prepare a dataset for modeling by filling in temporal gaps in data collection</i>
-----------	--

Description

In order to create a modeling dataset with feature lags that are temporally correct, the entry function in `forecastML`, `create_lagged_df`, needs evenly-spaced time series with no gaps in data collection. `fill_gaps()` can help here. This function takes a `data.frame` with (a) dates, (b) the outcome being forecasted, and, optionally, (c) dynamic features that change through time, (d) group columns for multiple time series modeling, and (e) static or non-dynamic features for multiple time series modeling and returns a `data.frame` with rows evenly spaced in time. Specifically, this function adds rows to the input dataset while filling in (a) dates, (b) grouping information, and (c) static features. The (a) outcome and (b) dynamic features will be NA for any missing time periods; these NA values can be left as-is, user-imputed, or removed from modeling in the user-supplied modeling wrapper function for `train_model`.

Usage

```
fill_gaps(data, date_col = 1, frequency, groups = NULL, static_features = NULL)
```

Arguments

data	A <code>data.frame</code> or object coercible to a <code>data.frame</code> with, minimally, dates and the outcome being forecasted.
date_col	The column index—an integer—of the date index. This column should have class 'Date' or 'POSIXt'.
frequency	Date/time frequency. A string taking the same input as <code>base::seq.Date(..., by = "frequency")</code> or <code>base::seq.POSIXt(..., by = "frequency")</code> e.g., '1 hour', '1 month', '7 days', '10 years' etc. The highest frequency supported at present is '1 sec'.
groups	Optional. A character vector of column names that identify the unique time series (i.e., groups/hierarchies) when multiple time series are present.

static_features

Optional. For grouped time series only. A character vector of column names that identify features that do not change through time. These columns are expected to be used as model features but are not lagged (e.g., a ZIP code column). The most recent values for each static feature for each group are used to fill in the resulting missing data in static features when new rows are added to the dataset.

Value

An object of class 'data.frame': The returned data.frame has the same number of columns and column order but with additional rows to account for gaps in data collection. For grouped data, any new rows added to the returned data.frame will appear between the minimum—or oldest—date for that group and the maximum—or most recent—date across all groups. If the user-supplied forecasting algorithm(s) cannot handle missing outcome values or missing dynamic features, these should either be imputed prior to `create_lagged_df()` or filtered out in the user-supplied modeling function for `train_model`.

Methods and related functions

The output of `fill_gaps()` is passed into

- `create_lagged_df`

Examples

```
# NOAA buoy dataset with gaps in data collection
data("data_buoy_gaps", package = "forecastML")

data_buoy_no_gaps <- fill_gaps(data_buoy_gaps, date_col = 1, frequency = '1 day',
                              groups = 'buoy_id', static_features = c('lat', 'lon'))

# The returned data.frame has the same number of columns but the time-series
# are now evenly spaced at 1 day apart. Additionally, the unchanging grouping
# columns and static features columns have been filled in for the newly created dataset rows.
dim(data_buoy_gaps)
dim(data_buoy_no_gaps)

# Running create_lagged_df() is the next step in the forecastML forecasting
# process. If there are long gaps in data collection, like in this buoy dataset,
# and the user-supplied modeling algorithm cannot handle missing outcomes data,
# the best option is to filter these rows out in the user-supplied modeling function
# for train_model()
```

plot.forecastML

Plot an object of class 'forecastML'

Description

A forecast plot of h-step-ahead forecasts produced from multiple horizon-specific forecast models using `combine_forecasts()`.

Usage

```
## S3 method for class 'forecastML'
plot(
  x,
  data_actual = NULL,
  actual_indices = NULL,
  facet = ~model,
  models = NULL,
  group_filter = NULL,
  drop_facet = FALSE,
  ...
)
```

Arguments

<code>x</code>	An object of class 'forecastML' from <code>combine_forecasts()</code> .
<code>data_actual</code>	A data.frame containing the target/outcome name and any grouping columns. The data can be historical actuals and/or holdout/test data.
<code>actual_indices</code>	Required if <code>data_actual</code> is given. A vector or 1-column data.frame of numeric row indices or dates (class 'Date' or 'POSIXt') with length <code>nrow(data_actual)</code> . The data can be historical actuals and/or holdout/test data.
<code>facet</code>	Optional. A formula with any combination of <code>model</code> , or <code>group</code> (for grouped time series) passed to <code>ggplot2::facet_grid()</code> internally (e.g., <code>~ model</code> , <code>model ~ .</code> , <code>~ model + group</code>).
<code>models</code>	Optional. Filter results by user-defined model name from <code>train_model()</code> .
<code>group_filter</code>	Optional. A string for filtering plot results for grouped time-series (e.g., <code>"group_col_1 == 'A'"</code>); passed to <code>dplyr::filter()</code> internally.
<code>drop_facet</code>	Optional. Boolean. If actuals are given when forecasting factors, the plot facet with 'actual' data can be dropped.
<code>...</code>	Not used.

Value

Forecast plot of class 'ggplot'.

<code>plot.forecast_error</code>	<i>Plot forecast error</i>
----------------------------------	----------------------------

Description

Plot forecast error at various levels of aggregation.

Usage

```
## S3 method for class 'forecast_error'
plot(
  x,
  type = c("global"),
  metric = NULL,
  facet = NULL,
  models = NULL,
  horizons = NULL,
  windows = NULL,
  group_filter = NULL,
  ...
)
```

Arguments

x	An object of class 'forecast_error' from return_error().
type	Select plot type; type = "global" is the default plot.
metric	Select error metric to plot (e.g., "mae"); attributes(x)\$error_metrics[1] is the default metric.
facet	Optional. A formula with any combination of horizon, model, or group (for grouped time series). passed to ggplot2::facet_grid() internally (e.g., horizon ~ model, horizon + model ~ ., ~ horizon + group). Can be NULL. The default faceting is set internally depending on the plot type.
models	Optional. A vector of user-defined model names from train_model() to filter results.
horizons	Optional. A numeric vector to filter results by horizon.
windows	Optional. A numeric vector to filter results by validation window number.
group_filter	A string for filtering plot results for grouped time series (e.g., "group_col_1 == 'A'").
...	Not used.

Value

Forecast error plots of class 'ggplot'.

plot.forecast_model_hyper

Plot hyperparameters

Description

Plot hyperparameter stability and relationship with error metrics across validation datasets and horizons.

Usage

```
## S3 method for class 'forecast_model_hyper'
plot(
  x,
  data_results,
  data_error,
  type = c("stability", "error"),
  horizons = NULL,
  windows = NULL,
  ...
)
```

Arguments

x	An object of class 'forecast_model_hyper' from return_hyper().
data_results	An object of class 'training_results' from predict.forecast_model().
data_error	An object of class 'validation_error' from return_error().
type	Select plot type; 'stability' is the default.
horizons	Optional. A numeric vector to filter results by horizon.
windows	Optional. A numeric vector to filter results by validation window number.
...	Not used.

Value

Hyper-parameter plots of class 'ggplot'.

Examples

```
# Sampled Seatbelts data from the R package datasets.
data("data_seatbelts", package = "forecastML")

# Example - Training data for 2 horizon-specific models w/ common lags per predictor.
horizons <- c(1, 12)
lookback <- 1:15

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback = lookback, horizon = horizons)

# One custom validation window at the end of the dataset.
windows <- create_windows(data_train, window_start = 181, window_stop = 192)

# User-define model - LASSO
# A user-defined wrapper function for model training that takes the following
# arguments: (1) a horizon-specific data.frame made with create_lagged_df(..., type = "train")
# (e.g., my_lagged_df$horizon_h) and, optionally, (2) any number of additional named arguments
# which are passed as '...' in train_model().
library(glmnet)
model_function <- function(data, my_outcome_col) {
```

```

x <- data[, -(my_outcome_col), drop = FALSE]
y <- data[, my_outcome_col, drop = FALSE]
x <- as.matrix(x, ncol = ncol(x))
y <- as.matrix(y, ncol = ncol(y))

model <- glmnet::cv.glmnet(x, y, nfolds = 3)
return(model)
}

# my_outcome_col = 1 is passed in ... but could have been defined in model_function().
model_results <- train_model(data_train, windows, model_name = "LASSO", model_function,
                             my_outcome_col = 1)

# User-defined prediction function - LASSO
# The predict() wrapper takes two positional arguments. First,
# the returned model from the user-defined modeling function (model_function() above).
# Second, a data.frame of predictors--identical to the datasets returned from
# create_lagged_df(..., type = "train"). The function can return a 1- or 3-column data.frame
# with either (a) point forecasts or (b) point forecasts plus lower and upper forecast
# bounds (column order and column names do not matter).
prediction_function <- function(model, data_features) {

  x <- as.matrix(data_features, ncol = ncol(data_features))

  data_pred <- data.frame("y_pred" = predict(model, x, s = "lambda.min"))
  return(data_pred)
}

# Predict on the validation datasets.
data_valid <- predict(model_results, prediction_function = list(prediction_function),
                      data = data_train)

# User-defined hyperparameter function - LASSO
# The hyperparameter function should take one positional argument--the returned model
# from the user-defined modeling function (model_function() above). It should
# return a 1-row data.frame of the optimal hyperparameters.
hyper_function <- function(model) {

  lambda_min <- model$lambda.min
  lambda_1se <- model$lambda.1se

  data_hyper <- data.frame("lambda_min" = lambda_min, "lambda_1se" = lambda_1se)
  return(data_hyper)
}

data_error <- return_error(data_valid)

data_hyper <- return_hyper(model_results, hyper_function)

plot(data_hyper, data_valid, data_error, type = "stability", horizons = c(1, 12))

```

plot.forecast_results *Plot an object of class forecast_results*

Description

A forecast plot for each horizon for each model in `predict.forecast_model()`.

Usage

```
## S3 method for class 'forecast_results'
plot(
  x,
  data_actual = NULL,
  actual_indices = NULL,
  facet = horizon ~ model,
  models = NULL,
  horizons = NULL,
  windows = NULL,
  group_filter = NULL,
  ...
)
```

Arguments

<code>x</code>	An object of class 'forecast_results' from <code>predict.forecast_model()</code> .
<code>data_actual</code>	A data.frame containing the target/outcome name and any grouping columns. The data can be historical actuals and/or holdout/test data.
<code>actual_indices</code>	Required if <code>data_actual</code> is given. A vector or 1-column data.frame of numeric row indices or dates (class 'Date' or 'POSIXt') with length <code>nrow(data_actual)</code> . The data can be historical actuals and/or holdout/test data.
<code>facet</code>	Optional. For numeric outcomes, a formula with any combination of <code>horizon</code> , <code>model</code> , or <code>group</code> (for grouped time series) passed to <code>ggplot2::facet_grid()</code> internally (e.g., <code>horizon ~ model</code> , <code>horizon + model ~ .</code> , <code>~ horizon + group</code>). Can be <code>NULL</code> .
<code>models</code>	Optional. Filter results by user-defined model name from <code>train_model()</code> .
<code>horizons</code>	Optional. Filter results by horizon.
<code>windows</code>	Optional. Filter results by validation window number.
<code>group_filter</code>	Optional. A string for filtering plot results for grouped time-series (e.g., <code>"group_col_1 == 'A'"</code>); passed to <code>dplyr::filter()</code> internally.
<code>...</code>	Not used.

Value

Forecast plot of class 'ggplot'.

plot.lagged_df	<i>Plot datasets with lagged features</i>
----------------	---

Description

Plot datasets with lagged features to view their direct forecasting setup across horizons.

Usage

```
## S3 method for class 'lagged_df'
plot(x, ...)
```

Arguments

x	An object of class 'lagged_df' from create_lagged_df().
...	Not used.

Value

A single plot of class 'ggplot' if lookback was specified in create_lagged_df(); a list of plots, one per feature, of class 'ggplot' if lookback_control was specified.

Examples

```
# Sampled Seatbelts data from the R package datasets.
data("data_seatbelts", package = "forecastML")
#-----
# Example 1 - Training data for 3 horizon-specific models w/ common lags per predictor.
horizons <- c(1, 6, 12)
lookback <- 1:15

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback = lookback, horizon = horizons)
plot(data_train)
#-----
# Example 2 - Training data for one 3-month horizon model w/ unique lags per predictor.
horizons <- 3
lookback <- list(c(3, 6, 9, 12), c(4:12), c(6:15), c(8))

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback_control = lookback, horizon = horizons)
plot(data_train)
```

plot.training_results *Plot an object of class training_results*

Description

Several diagnostic plots can be returned to assess the quality of the forecasts based on predictions on the validation datasets.

Usage

```
## S3 method for class 'training_results'
plot(
  x,
  type = c("prediction", "residual", "forecast_stability"),
  facet = horizon ~ model,
  models = NULL,
  horizons = NULL,
  windows = NULL,
  valid_indices = NULL,
  group_filter = NULL,
  keep_missing = FALSE,
  ...
)
```

Arguments

x	An object of class 'training_results' from predict.forecast_model().
type	Plot type. The default plot is "prediction" for validation dataset predictions.
facet	Optional. For numeric outcomes, a formula with any combination of horizon, model, or group (for grouped time series) passed to ggplot2::facet_grid() internally (e.g., horizon ~ model, horizon + model ~ ., ~ horizon + group).
models	Optional. Filter results by user-defined model name from train_model().
horizons	Optional. A numeric vector of model forecast horizons to filter results by horizon-specific model.
windows	Optional. A numeric vector of window numbers to filter results.
valid_indices	Optional. A numeric or date vector to filter results by validation row indices or dates.
group_filter	Optional. A string for filtering plot results for grouped time series (e.g., "group_col_1 == 'A'"). The results are passed to dplyr::filter() internally.
keep_missing	Boolean. If TRUE, predictions are plotted for indices/dates where the outcome is missing.
...	Not used.

Value

Diagnostic plots of class 'ggplot'.

plot.validation_error *Plot validation dataset forecast error*

Description

Plot forecast error at various levels of aggregation across validation datasets.

Usage

```
## S3 method for class 'validation_error'
plot(
  x,
  type = c("window", "horizon", "global"),
  metric = NULL,
  facet = NULL,
  models = NULL,
  horizons = NULL,
  windows = NULL,
  group_filter = NULL,
  ...
)
```

Arguments

x	An object of class 'validation_error' from return_error().
type	Select plot type; type = "window" is the default plot.
metric	Select error metric to plot (e.g., "mae"); attributes(x)\$error_metrics[1] is the default metric.
facet	Optional. A formula with any combination of horizon, model, or group (for grouped time series). passed to ggplot2::facet_grid() internally (e.g., horizon ~ model, horizon + model ~ ., ~ horizon + group). Can be NULL. The default faceting is set internally depending on the plot type.
models	Optional. A vector of user-defined model names from train_model() to filter results.
horizons	Optional. A numeric vector to filter results by horizon.
windows	Optional. A numeric vector to filter results by validation window number.
group_filter	A string for filtering plot results for grouped time series (e.g., "group_col_1 == 'A'").
...	Not used.

Value

Forecast error plots of class 'ggplot'.

plot.windows	<i>Plot validation datasets</i>
--------------	---------------------------------

Description

Plot validation datasets across time.

Usage

```
## S3 method for class 'windows'
plot(x, lagged_df, show_labels = TRUE, group_filter = NULL, ...)
```

Arguments

x	An object of class 'windows' from create_windows().
lagged_df	An object of class 'lagged_df' from create_lagged_df().
show_labels	Boolean. If TRUE, show validation dataset IDs on the plot.
group_filter	Optional. A string for filtering plot results for grouped time series (e.g., "group_col_1 == 'A' "). This string is passed to dplyr::filter() internally.
...	Not used.

Value

A plot of the outer-loop nested cross-validation windows of class 'ggplot'.

Examples

```
# Sampled Seatbelts data from the R package datasets.
data("data_seatbelts", package = "forecastML")

# Example - Training data for 3 horizon-specific models w/ common lags per predictor.
horizons <- c(1, 6, 12)
lookback <- 1:15

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback = lookback, horizon = horizons)

# All historical window lengths of 12 plus any partial windows at the end of the dataset.
windows <- create_windows(data_train, window_length = 12)
plot(windows, data_train)

# Two custom validation windows with different lengths.
windows <- create_windows(data_train, window_start = c(20, 80), window_stop = c(30, 100))
plot(windows, data_train)
```

predict.forecast_model

Predict on validation datasets or forecast

Description

Predict with a 'forecast_model' object from train_model(). If data = create_lagged_df(..., type = "train"), predictions are returned for the outer-loop nested cross-validation datasets. If data is an object of class 'lagged_df' from create_lagged_df(..., type = "forecast"), predictions are returned for the horizons specified in create_lagged_df(horizons = ...).

Usage

```
## S3 method for class 'forecast_model'
predict(..., prediction_function = list(NULL), data)
```

Arguments

...	One or more trained models from train_model().
prediction_function	A list of user-defined prediction functions with length equal to the number of models supplied in The prediction functions take 2 required positional arguments—(1) a 'forecast_model' object from train_model() and (2) a data.frame of model features from create_lagged_df(). For numeric outcomes and method = "direct", the function should return() 1- or 3-column data.frame of model predictions. If the prediction function returns a 1-column data.frame, point forecasts are assumed. If the prediction function returns a 3-column data.frame, lower and upper forecast bounds are assumed (the order and names of the 3 columns does not matter). For factor outcomes and method = "direct", the function should return() (1) 1-column data.frame of the model-predicted factor level or (2) an L-column data.frame of class probabilities where 'L' equals the number of levels in the outcome; columns should be ordered, from left to right, the same as levels(data\$outcome) which is the default behavior for most predict(..., type = "prob") functions. Column names do not matter. For numeric outcomes and method = "multi_output", the function should return() and h-column data.frame of model predictions—1 column for each horizon. Forecast intervals and factor outcomes are not currently supported with method = "multi_output".
data	If data is a training dataset from create_lagged_df(..., type = "train"), validation dataset predictions are returned; else, if data is a forecasting dataset from create_lagged_df(..., type = "forecast"), forecasts from horizons 1:h are returned.

Value

If data = create_lagged_df(..., type = "forecast"), an S3 object of class 'training_results'.
 If data = create_lagged_df(..., type = "forecast"), an S3 object of class 'forecast_results'.

Columns in returned 'training_results' data.frame:

- `model`: User-supplied model name in `train_model()`.
- `model_forecast_horizon`: The direct-forecasting time horizon that the model was trained on.
- `window_length`: Validation window length measured in dataset rows.
- `window_number`: Validation dataset number.
- `valid_indices`: Validation dataset row names from `attributes(create_lagged_df())$row_indices`.
- `date_indices`: If given and `method = "direct"`, validation dataset date indices from `attributes(create_lagged_df())`. If given and `method = "multi_output"`, `date_indices` represents the date of the forecast.
- `"groups"`: If given, the user-supplied groups in `create_lagged_df()`.
- `"outcome_name"`: The target being forecasted.
- `"outcome_name"_pred`: The model predictions.
- `"outcome_name"_pred_lower`: If given, the lower prediction bounds returned by the user-supplied prediction function.
- `"outcome_name"_pred_upper`: If given, the upper prediction bounds returned by the user-supplied prediction function.
- `forecast_indices`: If `method = "multi_output"`, the validation index of the h-step-ahead forecast.
- `forecast_date_indices`: If `method = "multi_output"`, the validation date index of the h-step-ahead forecast.

Columns in returned 'forecast_results' data.frame:

- `model`: User-supplied model name in `train_model()`.
- `model_forecast_horizon`: If `method = "direct"`, the direct-forecasting time horizon that the model was trained on.
- `horizon`: Forecast horizons, 1:h, measured in dataset rows.
- `window_length`: Validation window length measured in dataset rows.
- `forecast_period`: The forecast period in row indices or dates. The forecast period starts at either `attributes(create_lagged_df())$data_stop + 1` for row indices or `attributes(create_lagged_df())$data_stop + 1 * frequency` for date indices.
- `"groups"`: If given, the user-supplied groups in `create_lagged_df()`.
- `"outcome_name"`: The target being forecasted.
- `"outcome_name"_pred`: The model forecasts.
- `"outcome_name"_pred_lower`: If given, the lower forecast bounds returned by the user-supplied prediction function.
- `"outcome_name"_pred_upper`: If given, the upper forecast bounds returned by the user-supplied prediction function.

Examples

```

# Sampled Seatbelts data from the R package datasets.
data("data_seatbelts", package = "forecastML")

# Example - Training data for 2 horizon-specific models w/ common lags per predictor.
horizons <- c(1, 12)
lookback <- 1:15

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback = lookback, horizon = horizons)

# One custom validation window at the end of the dataset.
windows <- create_windows(data_train, window_start = 181, window_stop = 192)

# User-define model - LASSO
# A user-defined wrapper function for model training that takes the following
# arguments: (1) a horizon-specific data.frame made with create_lagged_df(..., type = "train")
# (e.g., my_lagged_df$horizon_h) and, optionally, (2) any number of additional named arguments
# which are passed as '...' in train_model().
library(glmnet)
model_function <- function(data, my_outcome_col) {

  x <- data[, -(my_outcome_col), drop = FALSE]
  y <- data[, my_outcome_col, drop = FALSE]
  x <- as.matrix(x, ncol = ncol(x))
  y <- as.matrix(y, ncol = ncol(y))

  model <- glmnet::cv.glmnet(x, y, nfolds = 3)
  return(model)
}

# my_outcome_col = 1 is passed in ... but could have been defined in model_function().
model_results <- train_model(data_train, windows, model_name = "LASSO", model_function,
                             my_outcome_col = 1)

# User-defined prediction function - LASSO
# The predict() wrapper takes two positional arguments. First,
# the returned model from the user-defined modeling function (model_function() above).
# Second, a data.frame of predictors--identical to the datasets returned from
# create_lagged_df(..., type = "train"). The function can return a 1- or 3-column data.frame
# with either (a) point forecasts or (b) point forecasts plus lower and upper forecast
# bounds (column order and column names do not matter).
prediction_function <- function(model, data_features) {

  x <- as.matrix(data_features, ncol = ncol(data_features))

  data_pred <- data.frame("y_pred" = predict(model, x, s = "lambda.min"))
  return(data_pred)
}

# Predict on the validation datasets.
data_valid <- predict(model_results, prediction_function = list(prediction_function),

```

```

        data = data_train)

# Forecast.
data_forecast <- create_lagged_df(data_seatbelts, type = "forecast", outcome_col = 1,
                                lookback = lookback, horizon = horizons)

data_forecasts <- predict(model_results, prediction_function = list(prediction_function),
                          data = data_forecast)

```

return_error

*Compute forecast error***Description**

Compute forecast error metrics on the validation datasets or a new test dataset.

Usage

```

return_error(
  data_results,
  data_test = NULL,
  test_indices = NULL,
  aggregate = stats::median,
  metrics = c("mae", "mape", "mdape", "smape", "rmse", "rmsse"),
  models = NULL,
  horizons = NULL,
  windows = NULL,
  group_filter = NULL
)

```

Arguments

data_results	An object of class 'training_results' or 'forecast_results' from running (a) predict on a trained model or (b) combine_forecasts() .
data_test	Required for forecast results only. If data_results is an object of class 'forecast_results', a data.frame used to assess the accuracy of a 'forecast_results' object. data_test should have the outcome/target columns and any grouping columns.
test_indices	Required if data_test is given or 'rmsse' row indices or dates (class 'Date' or 'POSIXt') with length nrow(data_test).
aggregate	Default median. A function—without parentheses—that aggregates historical prediction or forecast error across time series. All error metrics are first calculated at the level of the individual time series. aggregate is then used to combine error metrics across validation windows and horizons. Aggregations are returned at the group level if data_results contains groups.
metrics	A character vector of common forecast error metrics. The default behavior is to return all metrics.

models	Optional. A character vector of user-defined model names supplied to <code>train_model()</code> to filter results.
horizons	Optional. A numeric vector to filter results by horizon.
windows	Optional. A numeric vector to filter results by validation window number.
group_filter	Optional. A string for filtering plot results for grouped time series (e.g., "group_col_1 == 'A'"). <code>group_filter</code> is passed to <code>dplyr::filter()</code> internally.

Value

An S3 object of class 'validation_error', 'forecast_error', or 'forecastML_error': A list of data.frames of error metrics for the validation or forecast dataset depending on the class of `data_results`: 'training_results', 'forecast_results', or 'forecastML' from `combine_forecasts()`.

A list containing:

- Error metrics by model, horizon, and validation window
- Error metrics by model and horizon, collapsed across validation windows
- Global error metrics by model collapsed across horizons and validation windows

Error Metrics

- `mae`: Mean absolute error (works with factor outcomes)
- `mape`: Mean absolute percentage error
- `mdape`: Median absolute percentage error
- `smape`: Symmetrical mean absolute percentage error
- `rmse`: Root mean squared error
- `rmsse`: Root mean squared scaled error from the M5 competition

Methods and related functions

The output of `return_error()` has the following generic S3 methods

- `plot` from `return_error()`
- `plot` from `return_error()`

Examples

```
# Sampled Seatbelts data from the R package datasets.
data("data_seatbelts", package = "forecastML")

# Example - Training data for 2 horizon-specific models w/ common lags per predictor.
horizons <- c(1, 12)
lookback <- 1:15

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback = lookback, horizon = horizons)
```

```

# One custom validation window at the end of the dataset.
windows <- create_windows(data_train, window_start = 181, window_stop = 192)

# User-define model - LASSO
# A user-defined wrapper function for model training that takes the following
# arguments: (1) a horizon-specific data.frame made with create_lagged_df(..., type = "train")
# (e.g., my_lagged_df$horizon_h) and, optionally, (2) any number of additional named arguments
# which are passed as '...' in train_model().
library(glmnet)
model_function <- function(data, my_outcome_col) {

  x <- data[, -(my_outcome_col), drop = FALSE]
  y <- data[, my_outcome_col, drop = FALSE]
  x <- as.matrix(x, ncol = ncol(x))
  y <- as.matrix(y, ncol = ncol(y))

  model <- glmnet::cv.glmnet(x, y, nfolds = 3)
  return(model)
}

# my_outcome_col = 1 is passed in ... but could have been defined in model_function().
model_results <- train_model(data_train, windows, model_name = "LASSO", model_function,
                             my_outcome_col = 1)

# User-defined prediction function - LASSO
# The predict() wrapper takes two positional arguments. First,
# the returned model from the user-defined modeling function (model_function() above).
# Second, a data.frame of predictors--identical to the datasets returned from
# create_lagged_df(..., type = "train"). The function can return a 1- or 3-column data.frame
# with either (a) point forecasts or (b) point forecasts plus lower and upper forecast
# bounds (column order and column names do not matter).
prediction_function <- function(model, data_features) {

  x <- as.matrix(data_features, ncol = ncol(data_features))

  data_pred <- data.frame("y_pred" = predict(model, x, s = "lambda.min"))
  return(data_pred)
}

# Predict on the validation datasets.
data_valid <- predict(model_results, prediction_function = list(prediction_function),
                      data = data_train)

# Forecast error metrics for validation datasets.
data_error <- return_error(data_valid)

```

Description

The purpose of this function is to support investigation into the stability of hyperparameters in the nested cross-validation and across forecast horizons.

Usage

```
return_hyper(forecast_model, hyper_function)
```

Arguments

forecast_model An object of class 'forecast_model' from [train_model](#).
hyper_function A user-defined function for retrieving model hyperparameters. See the example below for details.

Value

An S3 object of class 'forecast_model_hyper': A data.frame of model-specific hyperparameters.

Methods and related functions

The output of `return_hyper()` has the following generic S3 methods

- [plot](#)

Examples

```
# Sampled Seatbelts data from the R package datasets.
data("data_seatbelts", package = "forecastML")

# Example - Training data for 2 horizon-specific models w/ common lags per predictor.
horizons <- c(1, 12)
lookback <- 1:15

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback = lookback, horizon = horizons)

# One custom validation window at the end of the dataset.
windows <- create_windows(data_train, window_start = 181, window_stop = 192)

# User-define model - LASSO
# A user-defined wrapper function for model training that takes the following
# arguments: (1) a horizon-specific data.frame made with create_lagged_df(..., type = "train")
# (e.g., my_lagged_df$horizon_h) and, optionally, (2) any number of additional named arguments
# which are passed as '...' in train_model().
library(glmnet)
model_function <- function(data, my_outcome_col) {

  x <- data[, -(my_outcome_col), drop = FALSE]
  y <- data[, my_outcome_col, drop = FALSE]
  x <- as.matrix(x, ncol = ncol(x))
  y <- as.matrix(y, ncol = ncol(y))
```

```

    model <- glmnet::cv.glmnet(x, y, nfolds = 3)
    return(model)
}

# my_outcome_col = 1 is passed in ... but could have been defined in model_function().
model_results <- train_model(data_train, windows, model_name = "LASSO", model_function,
                             my_outcome_col = 1)

# User-defined prediction function - LASSO
# The predict() wrapper takes two positional arguments. First,
# the returned model from the user-defined modeling function (model_function() above).
# Second, a data.frame of predictors--identical to the datasets returned from
# create_lagged_df(..., type = "train"). The function can return a 1- or 3-column data.frame
# with either (a) point forecasts or (b) point forecasts plus lower and upper forecast
# bounds (column order and column names do not matter).
prediction_function <- function(model, data_features) {

  x <- as.matrix(data_features, ncol = ncol(data_features))

  data_pred <- data.frame("y_pred" = predict(model, x, s = "lambda.min"))
  return(data_pred)
}

# Predict on the validation datasets.
data_valid <- predict(model_results, prediction_function = list(prediction_function),
                      data = data_train)

# User-defined hyperparameter function - LASSO
# The hyperparameter function should take one positional argument--the returned model
# from the user-defined modeling function (model_function() above). It should
# return a 1-row data.frame of the optimal hyperparameters.
hyper_function <- function(model) {

  lambda_min <- model$lambda.min
  lambda_1se <- model$lambda.1se

  data_hyper <- data.frame("lambda_min" = lambda_min, "lambda_1se" = lambda_1se)
  return(data_hyper)
}

data_error <- return_error(data_valid)

data_hyper <- return_hyper(model_results, hyper_function)

plot(data_hyper, data_valid, data_error, type = "stability", horizons = c(1, 12))

```

Description

Return a summary of a lagged_df object

Usage

```
## S3 method for class 'lagged_df'
summary(object, ...)
```

Arguments

object	An object of class 'lagged_df' from create_lagged_df().
...	Not used.

Value

A printed summary of the contents of the lagged_df object.

train_model	<i>Train a model across horizons and validation datasets</i>
-------------	--

Description

Train a user-defined forecast model for each horizon, 'h', and across the validation datasets, 'd'. If method = "direct", a total of 'h' * 'd' models are trained. If method = "multi_output", a total of 1 * 'd' models are trained. These models can be trained in parallel with the future package.

Usage

```
train_model(
  lagged_df,
  windows,
  model_name,
  model_function,
  ...,
  use_future = FALSE
)
```

Arguments

lagged_df	An object of class 'lagged_df' from create_lagged_df .
windows	An object of class 'windows' from create_windows .
model_name	A name for the model.
model_function	A user-defined wrapper function for model training that takes the following arguments: (1) a horizon-specific data.frame made with create_lagged_df(..., type = "train") (i.e., the dataset(s) stored in lagged_df) and, optionally, (2) any number of additional named arguments which can be passed in ... in this function.

... Optional. Named arguments passed into the user-defined model_function.

use_future Boolean. If TRUE, the future package is used for training models in parallel. The models will train in parallel across either (1) model forecast horizons or (b) validation windows, whichever is longer (i.e., length(create_lagged_df()) or nrow(create_windows())). The user should run future::plan(future::multiprocess) or similar prior to this function to train these models in parallel.

Value

An S3 object of class 'forecast_model': A nested list of trained models. Models can be accessed with my_trained_model\$horizon_h\$window_w\$model where 'h' gives the forecast horizon and 'w' gives the validation dataset window number from create_windows().

Methods and related functions

The output of train_model can be passed into

- `return_error`
- `return_hyper`

and has the following generic S3 methods

- `predict`
- `plot` (from predict.forecast_model(data = create_lagged_df(..., type = "train")))
- `plot` (from predict.forecast_model(data = create_lagged_df(..., type = "forecast")))

Examples

```
# Sampled Seatbelts data from the R package datasets.
data("data_seatbelts", package = "forecastML")

# Example - Training data for 2 horizon-specific models w/ common lags per predictor.
horizons <- c(1, 12)
lookback <- 1:15

data_train <- create_lagged_df(data_seatbelts, type = "train", outcome_col = 1,
                              lookback = lookback, horizon = horizons)

# One custom validation window at the end of the dataset.
windows <- create_windows(data_train, window_start = 181, window_stop = 192)

# User-define model - LASSO
# A user-defined wrapper function for model training that takes the following
# arguments: (1) a horizon-specific data.frame made with create_lagged_df(..., type = "train")
# (e.g., my_lagged_df$horizon_h) and, optionally, (2) any number of additional named arguments
# which are passed as '...' in train_model().
library(glmnet)
model_function <- function(data, my_outcome_col) {

  x <- data[, -(my_outcome_col), drop = FALSE]
```

```
y <- data[, my_outcome_col, drop = FALSE]
x <- as.matrix(x, ncol = ncol(x))
y <- as.matrix(y, ncol = ncol(y))

model <- glmnet::cv.glmnet(x, y, nfolds = 3)
return(model)
}

# my_outcome_col = 1 is passed in ... but could have been defined in model_function().
model_results <- train_model(data_train, windows, model_name = "LASSO", model_function,
                             my_outcome_col = 1)

# View the results for the model (a) trained on the first horizon
# and (b) to be assessed on the first outer-loop validation window.
model_results$horizon_1$window_1$model
```

Index

* datasets

- data_buoy, [11](#)
- data_buoy_gaps, [12](#)
- data_seatbelts, [12](#)

- combine_forecasts, [2](#)
- create_lagged_df, [5](#), [10](#), [13](#), [14](#), [32](#)
- create_skeleton, [9](#)
- create_windows, [8](#), [9](#), [9](#), [32](#)

- data_buoy, [11](#), [12](#)
- data_buoy_gaps, [12](#)
- data_seatbelts, [12](#)

- fill_gaps, [5](#), [13](#)

- plot, [4](#), [8](#), [10](#), [28](#), [30](#), [33](#)
- plot.forecast_error, [15](#)
- plot.forecast_model_hyper, [16](#)
- plot.forecast_results, [19](#)
- plot.forecastML, [14](#)
- plot.lagged_df, [20](#)
- plot.training_results, [21](#)
- plot.validation_error, [22](#)
- plot.windows, [23](#)
- predict, [27](#), [33](#)
- predict.forecast_model, [24](#)

- return_error, [27](#), [33](#)
- return_hyper, [29](#), [33](#)

- summary, [8](#)
- summary.lagged_df, [31](#)

- train_model, [10](#), [13](#), [14](#), [30](#), [32](#)