

# Package ‘HighFreq’

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**Type** Package

**Title** High Frequency Data Management

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**Author** Jerzy Pawlowski (algoquant)

**Maintainer** Jerzy Pawlowski <jp3900@nyu.edu>

**Description** Functions for chaining and joining time series, scrubbing bad data, managing time zones and alligning time indices, converting TAQ data to OHLC format, aggregating data to lower frequency, estimating volatility, skew, and higher moments.

**License** GPL (>= 2)

**Depends** rutils,  
quantmod,

**Imports** TTR,  
caTools,  
lubridate,

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**LazyData** true

**Repository** GitHub

**URL** <https://github.com/algoquant/HighFreq>

**RoxygenNote** 5.0.1

## R topics documented:

agg_regate . . . . .	2
extreme_values . . . . .	3
hf_data . . . . .	4
price_jumps . . . . .	5
roll_apply . . . . .	5
roll_hurst . . . . .	7
roll_moment . . . . .	8
roll_sharpe . . . . .	9
roll_vwap . . . . .	10
run_returns . . . . .	10
run_sharpe . . . . .	11

run_skew . . . . .	12
run_variance . . . . .	13
save_rets . . . . .	14
save_rets_ohlc . . . . .	15
save_scrub_agg . . . . .	15
save_TAQ . . . . .	16
scrub_agg . . . . .	17
scrub_TAQ . . . . .	18
season_ality . . . . .	19
to_period . . . . .	20

<b>Index</b>	<b>22</b>
--------------	-----------

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agg_regate	<i>Calculate the aggregation (weighted sum) of a statistical estimator over a OHLC time series.</i>
------------	---

---

## Description

Calculate the aggregation (weighted sum) of a statistical estimator over a OHLC time series.

## Usage

```
agg_regate(oh_lc, mo_moment = "run_variance", weight_ed = TRUE, ...)
```

## Arguments

oh_lc	OHLC time series of prices and trading volumes, in xts format.
mo_moment	character string representing function for estimating the moment.
weight_ed	Boolean should estimate be weighted by the trading volume? (default is TRUE)
...	additional parameters to the mo_moment function.

## Details

Calculates a single number representing the volume weighted sum of an estimator over the OHLC time series of prices. By default the sum is trade volume weighted.

## Value

Single numeric value equal to the volume weighted sum of an estimator over the time series.

## Examples

```
# create time index of one minute intervals over several days
in_dex <- seq(from=as.POSIXct("2015-02-09 09:30:00"),
              to=as.POSIXct("2015-02-28 16:00:00"), by="1 min")
# create synthetic xts of random prices
x_ts <- xts(exp(cumsum(0.001*rnorm(length(in_dex))))), order.by=in_dex)
# add trade Volume data
x_ts <- merge(x_ts,
              volume=sample(x=10*(2:18),
                            size=length(in_dex), replace=TRUE))
# aggregate to hours OHLC data
```

```
oh_lc <- to.period(x=x_ts, period="hours")
# calculate time series of daily skew estimates
skew_daily <- apply.daily(x=oh_lc, FUN=agg_regate, mo_moment="run_skew")
```

---

extreme\_values

*Identify extreme values in a single-column xts time series.*


---

## Description

Identifies extreme values as those that exceed a multiple of the rolling volatility.

## Usage

```
extreme_values(x_ts, win_dow = 51, vol_mult = 2)
```

## Arguments

<code>x_ts</code>	single-column xts time series.
<code>win_dow</code>	number of data points for estimating rolling volatility.
<code>vol_mult</code>	volatility multiplier.

## Details

Calculates the rolling volatility as a quantile of values over a rolling window. Extreme values are those that exceed the product of the volatility multiplier times the rolling volatility. Extreme values are the very tips of the tails when the distribution of values becomes very fat-tailed. The volatility multiplier `vol_mult` controls the threshold at which values are identified as extreme. Smaller volatility multiplier values will cause more values to be identified as extreme.

## Value

Boolean vector with the same number of rows as input time series.

## Examples

```
# create xts time series
x_ts <- xts(x=rnorm(1000), order.by=(Sys.time()-3600*(1:1000)))
# scrub extreme values
x_ts <- x_ts[!extreme_values(x_ts, vol_mult=1)]
```

---

hf_data	<i>High frequency data sets</i>
---------	---------------------------------

---

## Description

hf\_data.RData is a file containing the datasets:

**SPY** an xts time series containing 1-minute OHLC bar data for the SPY etf, from 2008-01-02 to 2014-05-19. SPY contains 625,425 rows of data, each row contains a single minute bar.

**TLT** an xts time series containing 1-minute OHLC bar data for the TLT etf, up to 2014-05-19.

**VXX** an xts time series containing 1-minute OHLC bar data for the VXX etf, up to 2014-05-19.

## Usage

```
data(hf_data) # not required - data is lazy load
```

## Format

Each xts time series contains OHLC data, with each row containing a single minute bar:

**Open** Open price in the bar

**High** High price in the bar

**Low** Low price in the bar

**Close** Close price in the bar

**Volume** trading volume in the bar

## Source

<https://wrds-web.wharton.upenn.edu/wrds/>

## References

Wharton Research Data Service (**WRDS**)

## Examples

```
# data(hf_data) # not required - data is lazy load
head(SPY)
chart_Series(x=SPY["2009"])
```

---

price_jumps	<i>Identify isolated price jumps in a single-column xts time series of prices, based on pairs of large neighboring returns of opposite sign.</i>
-------------	--

---

**Description**

Identify isolated price jumps in a single-column xts time series of prices, based on pairs of large neighboring returns of opposite sign.

**Usage**

```
price_jumps(x_ts, win_dow = 51, vol_mult = 2)
```

**Arguments**

x_ts	single-column xts time series of prices.
win_dow	number of data points for estimating rolling volatility.
vol_mult	volatility multiplier.

**Details**

Isolated price jumps are single prices that are very different from neighboring values. Price jumps create pairs of large neighboring returns of opposite sign. The function `price_jumps` first calculates simple returns from prices. Then it calculates the rolling volatility of returns as a quantile of returns over a rolling window. Jump prices are identified as those where neighboring returns both exceed a multiple of the rolling volatility, but the sum of those returns doesn't exceed it.

**Value**

Boolean vector with the same number of rows as input time series.

**Examples**

```
# create xts time series
x_ts <- xts(x=rnorm(1000), order.by=(Sys.time()-3600*(1:1000)))
# scrub jump prices
x_ts <- x_ts[!price_jumps(x_ts, vol_mult=1.0)]
```

---

roll_apply	<i>Apply an aggregation function over a rolling lookback window and the end points of an OHLC time series.</i>
------------	--

---

**Description**

Apply an aggregation function over a rolling lookback window and the end points of an OHLC time series.

**Usage**

```
roll_apply(oh_lc, agg_fun = "run_variance", win_dow = 11,
  end_points = (0:NROW(oh_lc)), by_columns = FALSE, ...)
```

## Arguments

<code>oh_ltc</code>	OHLC time series of prices and trading volumes, in xts format.
<code>agg_fun</code>	character string representing an aggregation function to be applied over a rolling lookback window.
<code>win_dow</code>	the size of the lookback window, equal to the number of bars of data used for applying the aggregation function.
<code>end_points</code>	an integer vector of end points.
<code>by_columns</code>	Boolean should the function <code>agg_fun</code> be applied column-wise, or should it be applied to all the columns combined? (default is FALSE)
<code>...</code>	additional parameters to the <code>agg_fun</code> function.

## Details

Applies an aggregation function over a rolling lookback window and the end points of an OHLC time series.

Performs similar operations to the functions `rollapply()` and `period.apply()` from package **xts**, and also the function `apply.rolling()` from package **PerformanceAnalytics**. (The function `rollapply()` isn't exported from the package **xts**.)

But the function `roll_apply()` is faster because it performs less type-checking and other overhead. Unlike the other functions, `roll_apply()` doesn't produce any leading NA values.

The function `roll_apply()` can be called in two different ways, depending on the argument `end_points`. If the argument `end_points` isn't explicitly passed to `roll_apply()`, then the default value is used, and `roll_apply()` performs aggregations over overlapping windows at each point in time. If the argument `end_points` is explicitly passed to `roll_apply()`, then `roll_apply()` performs aggregations over windows spanned by the `end_points`.

The aggregation function `agg_fun` can return either a single value or a vector of values. If the aggregation function `agg_fun` returns a single value, then `roll_apply()` returns an xts time series with a single column. If the aggregation function `agg_fun` returns a vector of values, then `roll_apply()` returns an xts time series with multiple columns equal to the length of the vector returned by the aggregation function `agg_fun`.

## Value

An xts time series with the same number of rows as the argument `oh_ltc`.

## Examples

```
# extract a single day of SPY data
price_s <- SPY["2012-02-13"]
win_dow <- 11
# Calculate the rolling sums of the columns of price_s
agg_regations <- roll_apply(price_s, agg_fun=sum, win_dow=win_dow, by_columns=TRUE)
# Apply a vector-valued aggregation function over a rolling window
agg_function <- function(x_ts) c(max(x_ts[, 1]), min(x_ts[, 4]))
agg_regations <- roll_apply(price_s, agg_fun=agg_function, win_dow=win_dow)
# define end points at 11-minute intervals (SPY is minutely bars)
end_points <- rutils::end_points(price_s, inter_val=win_dow)
# Calculate the rolling sums of the columns of price_s over end_points
agg_regations <- roll_apply(price_s, agg_fun=sum, win_dow=2, end_points=end_points, by_columns=TRUE)
# Apply a vector-valued aggregation function over the end_points of price_s
agg_regations <- roll_apply(price_s, agg_fun=agg_function, win_dow=2, end_points=end_points)
```

---

roll_hurst	<i>Calculate the rolling Hurst exponent over a rolling lookback window or the end points of an OHLC time series.</i>
------------	--

---

## Description

Calculate the rolling Hurst exponent over a rolling lookback window or the end points of an OHLC time series.

## Usage

```
roll_hurst(oh_lc, win_dow = 11, off_set = 0, roll_end_points = FALSE)
```

## Arguments

oh_lc	an OHLC time series of prices in xts format.
win_dow	the size of the lookback window, equal to the number of bars of data used for aggregating the OHLC prices.
off_set	the number of bars of data in the first, stub window.
roll_end_points	Boolean should the Hurst exponent be calculated using aggregations over the end points, or by rolling over a lookback window? (default is FALSE)

## Details

Calculates the rolling Hurst exponent in two different ways, depending on the argument `roll_end_points`.

If `roll_end_points` is `TRUE`, then the rolling Hurst exponent is calculated as the logarithm of the ratios of two rolling variance estimates. The Hurst exponent is defined as the logarithm of the ratio of the variance of aggregated returns, divided by the variance of simple returns. The aggregated returns are calculated over non-overlapping windows spanned by the end points, using the function `to_period()`. The Hurst exponent values are calculated only at the end points. The non-overlapping aggregation windows can be shifted by using the argument `off_set`, which produces a slightly different series of rolling hurst exponent values.

If `roll_end_points` is `FALSE`, then the rolling Hurst exponent is calculated as the logarithm of the ratios of two rolling price range estimates. The Hurst exponent is defined as the logarithm of the ratio of the range of aggregated prices, divided by the average range of prices in each bar. The aggregated prices are calculated over overlapping windows, and the Hurst exponent values are calculated at each point in time.

## Value

An xts time series with a single column and the same number of rows as the argument `oh_lc`.

## Examples

```
# calculate rolling Hurst over SPY
hurst_rolling <- roll_hurst(oh_lc=SPY, win_dow=10)
# calculate Hurst over end points of SPY
hurst_rolling <- roll_hurst(oh_lc=SPY, win_dow=10, off_set=0, roll_end_points=TRUE)
# calculate a series of rolling hurst values using argument off_set
hurst_rolling <- lapply(0:9, roll_hurst, oh_lc=SPY, win_dow=10, roll_end_points=TRUE)
```

```

hurst_rolling <- rutils::do_call_rbind(hurst_rolling)
# remove daily warmup periods
hurst_rolling <- hurst_rolling["T09:41:00/T16:00:00"]
chart_Series(x=hurst_rolling["2012-02-13"],
  name=paste(colnames(hurst_rolling), "10-minute aggregations"))

```

---

roll_moment	<i>Calculate a vector of statistics over an OHLC time series, and calculate a rolling mean over the statistics.</i>
-------------	---

---

## Description

Calculate a vector of statistics over an OHLC time series, and calculate a rolling mean over the statistics.

## Usage

```

roll_moment(oh_lc, mo_moment = "run_variance", win_dow = 11,
  weight_ed = TRUE, ...)

```

## Arguments

oh_lc	OHLC time series of prices and trading volumes, in xts format.
mo_moment	character string representing a function for estimating statistics of a single bar of OHLC data, such as volatility, skew, and higher moments.
win_dow	the size of the lookback window, equal to the number of bars of data used for calculating the rolling mean.
weight_ed	Boolean should statistic be weighted by trade volume? (default TRUE)
...	additional parameters to the mo_moment function.

## Details

Calculates a vector of statistics over an OHLC time series, such as volatility, skew, and higher moments. The statistics could also be any other aggregation of a single bar of OHLC data, for example the High price minus the Low price squared. The length of the vector of statistics is equal to the number of rows of the argument oh\_lc. Then it calculates a trade volume weighted rolling mean over the vector of statistics over and calculate statistics.

## Value

An xts time series with a single column and the same number of rows as the argument oh\_lc.

## Examples

```

# create time index of one minute intervals over several days
in_dex <- seq(from=as.POSIXct("2015-02-09 09:30:00"),
  to=as.POSIXct("2015-02-28 16:00:00"), by="1 min")
# create synthetic xts of random prices
x_ts <- xts(exp(cumsum(0.001*rnorm(length(in_dex))))), order.by=in_dex)
# add trade Volume data
x_ts <- merge(x_ts,
  volume=sample(x=10*(2:18),

```



```

        size=length(in_dex), replace=TRUE))
# aggregate to hours OHLC data
oh_lc <- to.period(x=x_ts, period="hours")
# calculate time series of rolling variance and skew estimates
var_rolling <- roll_moment(oh_lc=oh_lc)
skew_rolling <- roll_moment(oh_lc=oh_lc, mo_moment="run_skew")
skew_rolling <- skew_rolling/(var_rolling)^(1.5)
skew_rolling[1, ] <- 0
skew_rolling <- na.locf(skew_rolling)

```

---

roll_sharpe	<i>Calculate the rolling Sharpe ratio over a rolling lookback window for an OHLC time series.</i>
-------------	---

---

## Description

Calculate the rolling Sharpe ratio over a rolling lookback window for an OHLC time series.

## Usage

```
roll_sharpe(oh_lc, win_dow = 11)
```

## Arguments

oh_lc	an OHLC time series of prices in xts format.
win_dow	the size of the lookback window, equal to the number of bars of data used for aggregating the OHLC prices.

## Details

Calculates the rolling Sharpe ratio as the ratio of absolute returns over the lookback window (not percentage returns), divided by the average volatility of returns.

## Value

An xts time series with a single column and the same number of rows as the argument oh\_lc.

## Examples

```

# calculate rolling Sharpe ratio over SPY
sharpe_rolling <- roll_sharpe(oh_lc=SPY, win_dow=10)

```

---

roll_vwap	<i>Calculate the volume-weighted average price of an OHLC time series over a rolling window (lookback period).</i>
-----------	--

---

### Description

Performs the same operation as function `VWAP()` from package **VWAP**, but using vectorized functions, so it's a little faster.

### Usage

```
roll_vwap(oh_lc, x_ts = Cl(oh_lc), win_dow)
```

### Arguments

oh_lc	an OHLC time series of prices in xts format.
x_ts	single-column xts time series.
win_dow	the size of the lookback window, equal to the number of bars of data used for calculating the average price.

### Details

Calculates the volume-weighted average closing price, defined as the sum of the prices multiplied by trading volumes in the lookback window, divided by the sum of trading volumes in the window. If the argument `x_ts` is passed in explicitly, then its volume-weighted average value over time is calculated.

### Value

An xts time series with a single column and the same number of rows as the argument `oh_lc`.

### Examples

```
# calculate the volume-weighted average closing price (VWAP)
roll_vwap(oh_lc=get("SPY"), win_dow=11)
# calculate the volume-weighted average returns
roll_vwap(oh_lc=get("SPY"), x_ts=re_returns, win_dow=11)
```

---

run_returns	<i>Calculate single period returns from either TAQ or OHLC prices.</i>
-------------	--

---

### Description

Calculate single period returns from either TAQ or OHLC prices.

### Usage

```
run_returns(x_ts)
```

**Arguments**

`x_ts` xts time series of either TAQ or OHLC data.

**Details**

Calculates single period returns as the ratio of differenced mid prices divided by the time index differences.

**Value**

A single-column xts time series of returns.

**Examples**

```
# create time index of one second intervals for a single day
in_dex <- seq(from=as.POSIXct("2015-02-09 09:30:00"), to=as.POSIXct("2015-02-09 16:00:00"), by="1 sec")
# create xts of random TAQ prices
taq_data <- xts(cumsum(rnorm(length(in_dex))), order.by=in_dex)
# create vector of random bid-offer prices
bid_offer <- abs(rnorm(length(in_dex)))/10
# create TAQ data using cbind
taq_data <- cbind(taq_data-bid_offer, taq_data+bid_offer)
# add Trade.Price
taq_data <- cbind(taq_data, taq_data+rnorm(length(in_dex))/10)
# add Volume
taq_data <- cbind(taq_data, sample(x=10*(2:18), size=length(in_dex), replace=TRUE))
colnames(taq_data) <- c("Bid.Price", "Ask.Price", "Trade.Price", "Volume")
re_returns <- run_returns(taq_data)
```

---

run_sharpe	<i>Calculate time series of Sharpe-like statistics for each bar of a OHLC time series.</i>
------------	--

---

**Description**

Calculate time series of Sharpe-like statistics for each bar of a OHLC time series.

**Usage**

```
run_sharpe(oh_lc, calc_method = "close")
```

**Arguments**

`oh_lc` an OHLC time series of prices in xts format.  
`calc_method` character string representing method for estimating the Sharpe-like exponent.

**Details**

Calculates Sharpe-like statistics for each bar of a OHLC time series. The Sharpe-like statistic is defined as the ratio of the difference between Close minus Open prices divided by the difference between High minus Low prices. This statistic may also be interpreted as something like a Hurst exponent for a single bar of data. The motivation for the Sharpe-like statistic is the notion that if prices are trending in the same direction inside a given time bar of data, then this statistic is close to either 1 or -1.

**Value**

An xts time series with the same number of rows as the argument oh\_lc.

**Examples**

```
# create time index of one second intervals for a single day
in_dex <- seq(from=as.POSIXct("2015-02-09 09:30:00"),
              to=as.POSIXct("2015-02-09 16:00:00"), by="1 sec")
# create synthetic xts of random prices
x_ts <- xts(cumsum(rnorm(length(in_dex))), order.by=in_dex)
# aggregate to minutes OHLC data
oh_lc <- to.period(x=x_ts, period="minutes")
# calculate running Sharpe ratio
sharpe_running <- run_sharpe(oh_lc)
```

---

run\_skew

---

*Calculate time series of skew estimates from a OHLC time series.*


---

**Description**

Calculate time series of skew estimates from a OHLC time series.

**Usage**

```
run_skew(oh_lc, calc_method = "rogers.satchell")
```

**Arguments**

oh\_lc                    an OHLC time series of prices in xts format.  
calc\_method            character string representing method for estimating skew.

**Details**

Calculates skew estimates from OHLC prices at each point in time (row). The methods include Garman-Klass and Rogers-Satchell.

**Value**

Time series of skew estimates.

**Examples**

```
# create time index of one second intervals for a single day
in_dex <- seq(from=as.POSIXct("2015-02-09 09:30:00"),
              to=as.POSIXct("2015-02-09 16:00:00"), by="1 sec")
# create synthetic xts of random prices
x_ts <- xts(cumsum(rnorm(length(in_dex))), order.by=in_dex)
# aggregate to minutes OHLC data
oh_lc <- to.period(x=x_ts, period="minutes")
# calculate skew estimates
sk_ew <- run_skew(oh_lc)
```

---

run_variance	<i>Calculate a time series of variance estimates for an OHLC time series.</i>
--------------	---

---

## Description

Calculates the variance estimates for each bar of OHLC prices at each point in time (row), using the squared differences of OHLC prices at each point in time.

## Usage

```
run_variance(oh_lc, calc_method = "garman.klass_yz")
```

## Arguments

oh_lc	an OHLC time series of prices in xts format.
calc_method	character string representing method for estimating variance. The methods include: <ul style="list-style-type: none"> <li>• "close" close to close,</li> <li>• "garman.klass" Garman-Klass,</li> <li>• "garman.klass_yz" Garman-Klass with account for close-to-open price jumps,</li> <li>• "rogers.satchell" Rogers-Satchell,</li> <li>• "yang.zhang" Yang-Zhang,</li> </ul>

## Details

Performs a similar operation as function `volatility()` from package **TTR**, but without calculating a running sum using `runSum()`. It's also a little faster because it performs less data validation. The variance estimation methods "close", "garman.klass\_yz", and "yang.zhang" do account for close-to-open price jumps, while the methods "garman.klass" and "rogers.satchell" do not account for close-to-open price jumps.

## Value

An xts time series with a single column and the same number of rows as the argument `oh_lc`.

## Examples

```
# create time index of one second intervals for a single day
in_dex <- seq(from=as.POSIXct("2015-02-09 09:30:00"),
              to=as.POSIXct("2015-02-09 16:00:00"), by="1 sec")
# create synthetic xts of random prices
x_ts <- xts(cumsum(rnorm(length(in_dex))), order.by=in_dex)
# aggregate to minutes OHLC data
oh_lc <- to.period(x=x_ts, period="minutes")
# calculate variance estimates for oh_lc
var_running <- run_variance(oh_lc)
# calculate variance estimates for SPY
var_running <- run_variance(SPY, calc_method="yang.zhang")
# calculate SPY variance without overnight jumps
var_running <- run_variance(SPY, calc_method="rogers.satchell")
```

---

save_rets	<i>Load, scrub, aggregate, and rbind multiple days of TAQ data for a single symbol. Calculate returns and save them to a single '*.RData' file.</i>
-----------	---

---

## Description

Load, scrub, aggregate, and rbind multiple days of TAQ data for a single symbol. Calculate returns and save them to a single '\*.RData' file.

## Usage

```
save_rets(sym_bol, data_dir = "E:/mktdata/sec/",
          output_dir = "E:/output/data/", win_dow = 51, vol_mult = 2,
          period = "minutes", tzzone = "America/New_York")
```

## Arguments

sym_bol	character string representing symbol or ticker.
data_dir	character string representing directory containing input '*.RData' files.
output_dir	character string representing directory containing output '*.RData' files.
win_dow	number of data points for estimating rolling volatility.
vol_mult	volatility multiplier.
period	aggregation period.
tzzone	timezone to convert.

## Details

The function `save_rets` loads multiple days of TAQ data, then scrubs, aggregates, and rbinds them into a OHLC time series. It then calculates returns using function `run_returns`, and stores them in a variable named `'symbol.rets'`, and saves them to a file called `'symbol.rets.RData'`. The TAQ data files are assumed to be stored in separate directories for each `'symbol'`. Each `'symbol'` has its own directory (named `'symbol'`) in the `'data_dir'` directory. Each `'symbol'` directory contains multiple daily '\*.RData' files, each file containing one day of TAQ data.

## Value

Time series of returns and volume in xts format.

## Examples

```
## Not run:
save_rets("SPY")

## End(Not run)
```

---

save_rets_ohlc	<i>Load OHLC time series data for a single symbol, calculate its returns, and save them to a single '*.RData' file, without aggregation.</i>
----------------	--

---

### Description

Load OHLC time series data for a single symbol, calculate its returns, and save them to a single '\*.RData' file, without aggregation.

### Usage

```
save_rets_ohlc(sym_bol, data_dir = "E:/output/data/",
               output_dir = "E:/output/data/")
```

### Arguments

sym_bol	character string representing symbol or ticker.
data_dir	character string representing directory containing input '*.RData' files.
output_dir	character string representing directory containing output '*.RData' files.

### Details

The function `save_rets_ohlc` loads OHLC time series data from a single file. It then calculates returns using function `run_returns`, and stores them in a variable named `'symbol.rets'`, and saves them to a file called `'symbol.rets.RData'`.

### Value

Time series of returns and volume in xts format.

### Examples

```
## Not run:
save_rets_ohlc("SPY")

## End(Not run)
```

---

save_scrub_agg	<i>Load, scrub, aggregate, and rbind multiple days of TAQ data for a single symbol, and save the OHLC time series to a single '*.RData' file.</i>
----------------	---

---

### Description

Load, scrub, aggregate, and rbind multiple days of TAQ data for a single symbol, and save the OHLC time series to a single '\*.RData' file.

**Usage**

```
save_scrub_agg(sym_bol, data_dir = "E:/mktdata/sec/",
               output_dir = "E:/output/data/", win_dow = 51, vol_mult = 2,
               period = "minutes", tzzone = "America/New_York")
```

**Arguments**

sym_bol	character string representing symbol or ticker.
data_dir	character string representing directory containing input '*.RData' files.
output_dir	character string representing directory containing output '*.RData' files.
win_dow	number of data points for estimating rolling volatility.
vol_mult	volatility multiplier.
period	aggregation period.
tzzone	timezone to convert.

**Details**

The function `save_scrub_agg` loads multiple days of TAQ data, then scrubs, aggregates, and rbinds them into a OHLC time series, and finally saves it to a single '\*.RData' file. The OHLC time series is stored in a variable named 'symbol', and then it's saved to a file named 'symbol.RData' in the 'output\_dir' directory. The TAQ data files are assumed to be stored in separate directories for each 'symbol'. Each 'symbol' has its own directory (named 'symbol') in the 'data\_dir' directory. Each 'symbol' directory contains multiple daily '\*.RData' files, each file containing one day of TAQ data.

**Value**

OHLC time series in xts format.

**Examples**

```
## Not run:
save_scrub_agg("SPY")

## End(Not run)
```

---

save_TAQ	<i>Load and scrub multiple days of TAQ data for a single symbol, and save it to multiple '*.RData' files.</i>
----------	---

---

**Description**

Load and scrub multiple days of TAQ data for a single symbol, and save it to multiple '\*.RData' files.

**Usage**

```
save_TAQ(sym_bol, data_dir = "E:/mktdata/sec/",
         output_dir = "E:/output/data/", win_dow = 51, vol_mult = 2,
         tzzone = "America/New_York")
```



**Arguments**

sym_bol	character string representing symbol or ticker.
data_dir	character string representing directory containing input '*.RData' files.
output_dir	character string representing directory containing output '*.RData' files.
win_dow	number of data points for estimating rolling volatility.
vol_mult	volatility multiplier.
tzone	timezone to convert.

**Details**

The function `save_TAQ` loads multiple days of TAQ data, scrubs it, and saves it to '\*.RData' files. It uses the same file names for output as the input file names. The TAQ data files are assumed to be stored in separate directories for each 'symbol'. Each 'symbol' has its own directory (named 'symbol') in the 'data\_dir' directory. Each 'symbol' directory contains multiple daily '\*.RData' files, each file containing one day of TAQ data.

**Value**

TAQ time series in xts format.

**Examples**

```
## Not run:
save_TAQ("SPY")

## End(Not run)
```

---

scrub_agg	<i>Scrub a single day of TAQ data, aggregate it, and convert to OHLC format.</i>
-----------	--

---

**Description**

Scrub a single day of TAQ data, aggregate it, and convert to OHLC format.

**Usage**

```
scrub_agg(taq_data, win_dow = 51, vol_mult = 2, period = "minutes",
          tzone = "America/New_York")
```

**Arguments**

taq_data	TAQ xts time series.
win_dow	number of data points for estimating rolling volatility.
vol_mult	volatility multiplier.
period	aggregation period.
tzone	timezone to convert.

## Details

The function `scrub_agg` performs:

- index timezone conversion,
- data subset to trading hours,
- removal of duplicate time stamps,
- scrubbing of quotes with suspect bid-offer spreads,
- scrubbing of quotes with suspect price jumps,
- cbinding of mid prices with volume data,
- aggregation to OHLC using function `to.period` from package `xts`,

Valid 'period' character strings include: "minutes", "3 min", "5 min", "10 min", "15 min", "30 min", and "hours". The time index of the output time series is rounded up to the next integer multiple of 'period'.

## Value

OHLC time series in `xts` format.

## Examples

```
# create time index of one second intervals for a single day
in_dex <- seq(from=as.POSIXct("2015-02-09 09:30:00"), to=as.POSIXct("2015-02-09 16:00:00"), by="1 sec")
# create xts of random TAQ prices
x_ts <- xts(cumsum(rnorm(length(in_dex))), order.by=in_dex)
# create vector of random bid-offer prices
bid_offer <- abs(rnorm(length(in_dex)))/10
# create TAQ data using cbind
taq_data <- cbind(x_ts-bid_offer, x_ts+bid_offer)
# add Trade.Price
taq_data <- cbind(taq_data, x_ts+rnorm(length(in_dex))/10)
# add Volume
taq_data <- cbind(taq_data, sample(x=10*(2:18), size=length(in_dex), replace=TRUE))
colnames(taq_data) <- c("Bid.Price", "Ask.Price", "Trade.Price", "Volume")
# aggregate to ten minutes OHLC data
ohlc_data <- scrub_agg(taq_data, period="10 min")
chartSeries(ohlc_data, name=SPY, theme=chartTheme("white"))
```

---

`scrub_TAQ`

*Scrub a single day of TAQ data in xts format, without aggregation.*

---

## Description

Scrub a single day of TAQ data in `xts` format, without aggregation.

## Usage

```
scrub_TAQ(taq_data, win_dow = 51, vol_mult = 2,
          tzzone = "America/New_York")
```

**Arguments**

taq_data	TAQ xts time series.
win_dow	number of data points for estimating rolling volatility.
vol_mult	volatility multiplier.
tzone	timezone to convert.

**Details**

The function `scrub_TAQ` performs the same scrubbing operations as `scrub_agg`, except it doesn't aggregate, and returns the TAQ data in xts format.

**Value**

TAQ xts time series.

**Examples**

```
# create time index of one second intervals for a single day
in_dex <- seq(from=as.POSIXct("2015-02-09 09:30:00"),
              to=as.POSIXct("2015-02-09 16:00:00"), by="1 sec")
# create xts of random TAQ prices
x_ts <- xts(cumsum(rnorm(length(in_dex))), order.by=in_dex)
# create vector of random bid-offer prices
bid_offer <- abs(rnorm(length(in_dex)))/10
# create TAQ data using cbind
taq_data <- cbind(x_ts-bid_offer, x_ts+bid_offer)
# add Trade.Price
taq_data <- cbind(taq_data, x_ts+rnorm(length(in_dex))/10)
# add Volume
taq_data <- cbind(taq_data, sample(x=10*(2:18), size=length(in_dex), replace=TRUE))
colnames(taq_data) <- c("Bid.Price", "Ask.Price", "Trade.Price", "Volume")
taq_data <- scrub_TAQ(taq_data)
taq_data <- scrub_TAQ(taq_data, win_dow=11, vol_mult=1)
```

---

season_ality	<i>Perform seasonality aggregations over a single-column xts time series.</i>
--------------	---

---

**Description**

Perform seasonality aggregations over a single-column xts time series.

**Usage**

```
season_ality(x_ts, in_dex = format(index(x_ts), "%H:%M"))
```

**Arguments**

x_ts	single-column xts time series.
in_dex	vector of character strings representing points in time, of the same length as the argument x_ts.

### Details

Calculates the mean of values observed at the same points in time specified by the argument `in_dex`. An example of a daily seasonality aggregation is the average price of a stock between 9:30AM and 10:00AM every day, over many days. The argument `in_dex` is passed into function `tapply()`, and must be the same length as the argument `x_ts`.

### Value

An xts time series with mean aggregations over the seasonality interval.

### Examples

```
# calculate variance of each minutely OHLC bar of data
x_ts <- run_variance(get("SPY"))
# remove overnight variance spikes at "09:31"
in_dex <- format(index(x_ts), "%H:%M")
x_ts <- x_ts[!in_dex=="09:31", ]
# calculate daily seasonality of variance
seasonality(x_ts=x_ts)
```

---

to\_period

---

Aggregate an OHLC time series to a lower periodicity.

---

### Description

Given an OHLC time series at high periodicity (say seconds), calculates the OHLC prices at lower periodicity (say minutes).

### Usage

```
to_period(oh_lc, period = "minutes", k = 1,
  end_points = xts::endpoints(oh_lc, period, k))
```

### Arguments

<code>oh_lc</code>	an OHLC time series of prices in xts format.
<code>period</code>	aggregation interval ("seconds", "minutes", "hours", "days", "weeks", "months", "quarters", and "years").
<code>k</code>	number of periods to aggregate over (for example if <code>period="minutes"</code> and <code>k=2</code> , then aggregate over two minute intervals.)
<code>end_points</code>	an integer vector of end points.

### Details

#' Performs a similar aggregation as function `to.period()` from package `xts`, but has the flexibility to aggregate to a user-specified vector of end points. The function `to_period()` simply calls the compiled function `toPeriod()` (from package `xts`), to perform the actual aggregations. If `end_points` are passed in explicitly, then the `period` argument is ignored.

### Value

OHLC time series of prices in xts format, with a lower periodicity defined by the `end_points`.

**Examples**

```
# define end points at 10-minute intervals (SPY is minutely bars)
end_points <- rutils::end_points(SPY["2009"], inter_val=10)
# aggregate over 10-minute end_points:
to_period(x_ts=SPY["2009"], end_points=end_points)
# aggregate over days:
to_period(x_ts=SPY["2009"], period="days")
# equivalent to:
to.period(x=SPY["2009"], period="days", name=rutils::na_me(SPY))
```

# Index

## \*Topic **datasets**

hf\_data, [4](#)

agg\_regate, [2](#)

extreme\_values, [3](#)

hf\_data, [4](#)

price\_jumps, [5](#)

roll\_apply, [5](#)

roll\_hurst, [7](#)

roll\_moment, [8](#)

roll\_sharpe, [9](#)

roll\_vwap, [10](#)

run\_returns, [10](#)

run\_sharpe, [11](#)

run\_skew, [12](#)

run\_variance, [13](#)

save\_rets, [14](#)

save\_rets\_ohlc, [15](#)

save\_scrub\_agg, [15](#)

save\_TAQ, [16](#)

scrub\_agg, [17](#)

scrub\_TAQ, [18](#)

seasonality, [19](#)

SPY (hf\_data), [4](#)

to\_period, [20](#)