Package ‘lazytrade’

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Type    Package
Title   Learn Computer and Data Science using Algorithmic Trading
Version 0.3.10
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Description Provide sets of functions and methods to learn and practice data science using idea of algorithmic trading.
Main goal is to process information within ”Decision Support System" to come up with analysis or predictions.
There are several utilities such as dynamic and adaptive risk management using reinforcement learning
and even functions to generate predictions of price changes using pattern recognition deep regression learning.
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URL https://vladdsm.github.io/myblog_attempt/topics/lazy20trading/,
https://github.com/vzhomeexperiments/lazytrade

BugReports https://github.com/vzhomeexperiments/lazytrade/issues
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aml_collect_data .................................................. 3
aml_make_model .................................................. 5
aml_score_data .................................................. 6
aml_test_model .................................................. 8
check_if_optimize .................................................. 10
create_labelled_data ................................................. 12
create_transposed_data ............................................... 13
data_trades .................................................. 14
decrypt_mykeys .................................................. 15
DFR .................................................. 16
encrypt_api_key .................................................. 16
EURUSDM15X75 .................................................. 18
evaluate_macroeconomic_event ....................................... 18
evaluate_market_type .................................................. 20
generate_RL_policy .................................................. 21
generate_RL_policy_mt ............................................... 22
get_profit_factorDF .................................................. 23
import_data .................................................. 24
import_data_mt .................................................. 25
indicator_dataset .................................................. 26
indicator_dataset_big ............................................... 26
load_asset_data .................................................. 27
log_RL_progress .................................................. 28
log_RL_progress_mt .................................................. 29
macd_100 .................................................. 30
macd_df .................................................. 31
macd_ML2_small .................................................. 31
mt_make_model .................................................. 32
opt_aggregate_results ............................................... 33
opt_create_graphs .................................................. 34
policy_tr_systDF .................................................. 35
price_dataset .................................................. 35
price_dataset_big .................................................. 36
profit_factor .................................................. 36
profit_factorDF .................................................. 37
profit_factor_data .................................................. 38
record_policy .................................................. 38
record_policy_mt .................................................. 39
result_prev .................................................. 40
result_R .................................................. 41
self_learn_ai_R .................................................. 41
test_data_pattern .................................................. 43
test_model .................................................. 43
to_m .................................................. 44
TradeStatePolicy .................................................. 45
trading_systemDF .................................................. 46
**aml_collect_data**

Function to read new data, transform data, save data for further re-training of regression model for a single currency pair

**Description**

Function is collecting data from the files using dedicated function load_asset_data.R. One file with a prices of the asset and another file with the corresponding indicator pattern. Both data objects are transformed to be suitable for Regression Modelling. Indicator values will be placed into the column X1-X75 and price change is in the column 'LABEL'. Result would be written to new or aggregated to the existing file.

Function is also checking that generated dataset is not too big. Should the dataset is too big (e.g. > 1000000 rows), then only latest 950000 rows will be used. Note: the amount 1000000 rows is not verified in practice, further testing is required.

**Usage**

```r
aml_collect_data(
  price_dataset,      # Dataset containing assets prices. It will be used as a label
  indicator_dataset,  # Dataset containing assets indicator which pattern will be used as predictor
  symbol,             # Character symbol of the asset for which to train the model
  num_bars,           # Number of bars used to detect pattern
  timeframe,          # Data timeframe e.g. 1 min
  path_data           # Path where the aggregated historical data is stored, if exists in rds format
)
```

**Arguments**

- `price_dataset`  Dataset containing assets prices. It will be used as a label
- `indicator_dataset`  Dataset containing assets indicator which pattern will be used as predictor
- `symbol`  Character symbol of the asset for which to train the model
- `num_bars`  Number of bars used to detect pattern
- `timeframe`  Data timeframe e.g. 1 min
- `path_data`  Path where the aggregated historical data is stored, if exists in rds format
Details

Function is handling shift of the price and indicator datasets. New data will be always on the 'bottom' of the dataset

The amount of rows is customizable however it must be selected once for the function to start working. Other 'aml_*' functions will rely on this selections, use the same number accordingly!

Value

Function is writing files into Decision Support System folder, mainly file object with the model

Author(s)

(C) 2019 Vladimir Zhbanko

Examples

# write examples for the function
library(dplyr)
library(readr)
library(lubridate)
library(lazytrade)

path_terminal <- system.file("extdata", package = "lazytrade")
macd <- load_asset_data(path_terminal = path_terminal, trade_log_file = "AI_Macd",
  time_period = 15, data_depth = "300")
prices <- load_asset_data(path_terminal = path_terminal, trade_log_file = "AI_CP",
  time_period = 15, data_depth = "300")

path_data <- normalizePath(tempdir(), winslash = "/")

# data transformation using the custom function for one symbol
aml_collect_data(price_dataset = prices,
  indicator_dataset = macd,
  symbol = 'EURUSD',
  num_bars = 75,
  timeframe = 15,
  path_data = path_data)
aml_make_model

Function to train Deep Learning regression model for a single currency pair.

Description

Function is training h2o deep learning model to match future prices of the asset to the indicator pattern. Main idea is to be able to predict future prices by solely relying on the most recent indicator pattern. This is to mimic traditional algorithmic systems based on the indicator rule attempting to automate optimization process with AI.

Deep learning model structure is obtained from the 6 random combinations of neurons within 4 layers of the network, the most accurate model configuration will be automatically selected.

In addition the function will check if there is a need to update the model. To do that function will check results of the function aml_test_model.R.

Usage

```r
aml_make_model(
  symbol,
  num_bars,
  timeframe,
  path_model,
  path_data,
  force_update = FALSE
)
```

Arguments

- **symbol**: Character symbol of the asset for which to train the model
- **num_bars**: Number of bars used to detect pattern
- **timeframe**: Data timeframe e.g. 1 min
- **path_model**: Path where the models are be stored
- **path_data**: Path where the aggregated historical data is stored, if exists in rds format
- **force_update**: Boolean, by setting this to TRUE function will generate new model (useful after h2o engine update)

Details

Function is using the dataset prepared by the function aml_collect_data.R. Function will start to train the model as soon as there are more than 100 rows in the dataset.

Value

Function is writing file object with the model
aml_score_data

**Author(s)**

(C) 2019 Vladimir Zhbanko

**Examples**

```r
library(dplyr)
library(readr)
library(h2o)
library(lazytrade)

path_model <- normalizePath(tempdir(), winslash = "/")
path_data <- normalizePath(tempdir(), winslash = "/")

data(EURUSDM15X75)
write_rds(EURUSDM15X75, file.path(path_data, 'EURUSDM15X75.rds'))

# start h2o engine (using all CPU's by default)
h2o.init()

# performing Deep Learning Regression using the custom function
aml_make_model(symbol = 'EURUSD',
    num_bars = 75,
    timeframe = 15,
    path_model = path_model,
    path_data = path_data)

# stop h2o engine
h2o.shutdown(prompt = FALSE)

#set delay to insure h2o unit closes properly before the next test
Sys.sleep(5)
```

---

**aml_score_data**

*Function to score new data and predict change for each single currency pair*

**Description**

Function is using the latest data from the financial assets indicator pattern and deep learning model. Prediction is a price change in the future for that asset will be used by the trading system.
aml_score_data

Usage

aml_score_data(
    symbol,
    num_bars,
    timeframe,
    path_model,
    path_data,
    path sbxm,
    path sbxs
)

Arguments

symbol Character symbol of the asset for which the model shall predict
num_bars Number of bars used to detect pattern
timeframe Data timeframe e.g. 1 min
path_model Path where the models are be stored
path_data Path where the aggregated historical data is stored, if exists in rds format
path sbxm Path to the sandbox where file with predicted price should be written (master terminal)
path sbxs Path to the sandbox where file with predicted price should be written (slave terminal)

Details

Performs fresh data reading from the rds file

Value

Function is writing file into Decision Support System folder, mainly file with price change prediction in pips

Author(s)

(C) 2019 Vladimir Zhbanko

Examples

# test of function aml_make_model is duplicated here
library(dplyr)
library(readr)
library(h2o)
library(magrittr)
library(lazytrade)
```r
path_model <- normalizePath(tempdir(), winslash = "/")
path_data <- normalizePath(tempdir(), winslash = "/")

data(EURUSD15X75)
write_rds(EURUSD15X75, file.path(path_data, 'EURUSD15X75.rds'))

# start h2o engine (using all CPU's by default)
h2o.init()

# performing Deep Learning Regression using the custom function
aml_make_model(symbol = 'EURUSD',
    num_bars = 75,
    timeframe = 15,
    path_model = path_model,
    path_data = path_data)

path_sbxm <- normalizePath(tempdir(), winslash = "/")
path_sbxs <- normalizePath(tempdir(), winslash = "/")

# score the latest data to generate predictions for one currency pair
aml_score_data(symbol = 'EURUSD',
    num_bars = 75,
    timeframe = 15,
    path_model = path_model,
    path_data = path_data,
    path_sbxm = path_sbxm,
    path_sbxs = path_sbxs)

# stop h2o engine
h2o.shutdown(prompt = FALSE)

# set delay to insure h2o unit closes properly before the next test
Sys.sleep(5)
```

---

**aml_test_model**

*Function to test the model and conditionally decide to update existing model for a single currency pair*

**Description**

Function is designed to test the trading decision generated by the Deep learning regression model. The outcome of this function will be used to perform update of existing model with a fresh data.
Usage

aml_test_model(symbol, num_bars, timeframe, path_model, path_data)

Arguments

symbol Character symbol of the asset for which to train the model
num_bars Number of bars used to detect pattern
timeframe Data timeframe e.g. 1 min
path_model Path where the models are be stored
path_data Path where the aggregated historical data is stored, if exists in rds format

Details

Function is reading shifted price data and corresponding indicator. Starting from the trained model function will test the trading strategy using simplified trading approach. Trading approach will entail using the last available indicator data, predict the price change for every row, shift predicted value by 75 bars as we will hold the asset for 75 bars. Account for the real price change after 75 bars by creating a cumulative sum column. Verify obtained summary results on the model and obtain virtual real/simulated result is consolidated to calculate model quality. Whenever this value is less than 0 function is writing dedicated decision using simple *.csv file. Such file will be used by the function aml_make_model.R to decide whether model must be updated...

Value

Function is writing file into Decision Support System folder

Author(s)

(C) 2019 Vladimir Zhbanko

Examples

```r
library(dplyr)
library(magrittr)
library(readr)
library(h2o)
library(lazytrade)

path_model <- normalizePath(tempdir(), winslash = "/")
path_data <- normalizePath(tempdir(), winslash = "/")

data(EURUSDM15X75)
write_rds(EURUSDM15X75, file.path(path_data, "EURUSDM15X75.rds"))

# start h2o engine (using all CPU's by default)
h2o.init()
```
check_if_optimize

Function check_if_optimize.

Description

Purpose of this function is to verify trading system functionality by analysing profit factor on the last trades. Whenever trading robot has profit factor value below certain limit function will write a file log indicating which trading systems need to be maintained.

Learn by example how to manipulate data
check_if_optimize

Usage

check_if_optimize(
  x,
  path_trading_robot = "",
  num_trades_to_consider = 10,
  profit_factor_limit = 0.7,
  demo_mode = FALSE,
  write_mode = FALSE
)

Arguments

x - dataframe containing trading results
path_trading_robot - path of trading robot repository. must contain folder TEST with file Setup.csv. File Setup.csv contains a table with magic numbers under test
num_trades_to_consider - Number of trades to calculate profit factor
profit_factor_limit - Limit below which trading robot considered not working properly
demo_mode - When true function uses package test dataset
write_mode - When true function will write result to the file located in the temporary directory

Details

Whenever there will be not enough trades then empty file will be written to the destination

Value

function returns a dataframe with systems that should be optimized

Author(s)

(C) 2019 Vladimir Zhbanko

Examples

library(lazytrade)
library(dplyr)
library(readr)
library(lubridate)
DFT1 <- import_data(trade_log_file = system.file("extdata",
  "OrdersResultsT1.csv",
  package = "lazytrade"),
  demo_mode = TRUE)
create_labelled_data

# without writing to the file
DFT1 %>% check_if_optimize(num_trades_to_consider = 10,
  profit_factor_limit = 1.2,
  demo_mode = TRUE,
  write_mode = FALSE)

# function will write to the temporary file
DFT1 %>% check_if_optimize(num_trades_to_consider = 10,
  profit_factor_limit = 1.2,
  demo_mode = TRUE,
  write_mode = TRUE)

create_labelled_data  Create labelled data

Description

FUNCTION create_labelled_data. PURPOSE: function gets price data of every currency in each column. It is splitting this data by periods and transposes the data. Additionally function is capable to label the data based on the simple logic. Each row will be assigned into 2 categories based on the difference between beginning and end of the row elements Finally all data will be stacked on top and joined into the table

Learn by example how to manipulate data

Usage

create_labelled_data(x, n = 50, type = "regression")

Arguments

x - data set containing a table where 1st column is a Time index and other columns containing financial asset price values
n - number of rows we intend to split and transpose the data to
type - type of the label required. Can be either "classification" or "regression". "classification" will return either "BU" or "BE", "regression" will return the difference between first value and the last value in each row (in pips)

Details

see more info in the udemy course self-learning-trading-robot

Value

function returns transposed data. One column called 'LABEL' indicate achieved value of the label. Transposed values from every column are stacked one to each other
create_transposed_data

Examples

```r
library(dplyr)
library(readr)
library(lazytrade)

# using a sample data
data(price_dataset)

# price change as a label
create_labelled_data(x = price_dataset, n = 75, type = "regression")

# factors 'BU'/'BE' as a label
create_labelled_data(x = price_dataset, n = 75, type = "classification")
```

create_transposed_data

Create Transposed Data

Description

PURPOSE: function gets indicator data in each column. Goal is to splitting this data into periods and transpose the data.

Learn by example how to manipulate data.

Usage

```r
create_transposed_data(x, n = 50)
```

Arguments

- **x** - data set containing a table where 1st column is a Time index and other columns containing financial asset indicator values
- **n** - number of rows we intend to split and transpose the data

Details

Each column contains records of the indicator value of the assets every column will be split into chunks of n observations and transposed into rows. This repeated for all the columns coming up with a matrix. Function works in combination with a function create_labelled_data.

Value

Function returns transposed data. Transposed values from every column are stacked one to each other.
Examples

```r
library(dplyr)
library(readr)

# using a sample data
data(indicator_dataset)

create_transposed_data(indicator_dataset, n = 75)
```

<table>
<thead>
<tr>
<th>data_trades</th>
<th>Table with Trade results samples</th>
</tr>
</thead>
</table>

Description

Table with Trade results samples

Usage

data_trades

Format

A dataframe with several columns

- **MagicNumber**  Unique identifiers of the Trading Robots
- **TicketNumber**  Ticket Number of closed position
- **OrderStartTime**  Date and Time when order started
- **OrderCloseTime**  Date and Time when order closed
- **Profit**  Monetary result of the trade
- **Symbol**  Symbol of the Asset e.g. EURUSD
- **OrderType**  Order Type 0 - buy, 1 - sell
*decrypt_mykeys*  
*Function that decrypt encrypted content*

**Description**
Function that decrypt encrypted content

**Usage**
```r
decrypt_mykeys(path_encrypted_content, path_private_key)
```

**Arguments**
- `path_encrypted_content` - path to the encrypted content of the API key
- `path_private_key` - path to the private RSA key, should be without password

**Details**
It is possible to generate private/public key pair using R-Studio Project Options Menu. Alternatively possible to use `openssl` R package

**Value**
-a string with decrypted key

**Examples**
```r
library(dplyr)
library(magrittr)
library(openssl)
library(readr)

path_ssh <- normalizePath(tempdir(), winslash = "/")
rsa_keygen() %>>% write_pem(path = file.path(path_ssh, "id_api"))
# extract and write your public key
read_key(file = file.path(path_ssh, "id_api"), password = "") %>>%
  `[["pubkey"] %>>% write_pem(path = file.path(path_ssh, "id_api.pub"))

path_private_key <- file.path(path_ssh, "id_api")
path_public_key <- file.path(path_ssh, "id_api.pub")

#encrypting string 'my_key'...
encrypt_api_key(api_key = 'my_key', enc_name = 'api_key.enc.rds', path_ssh = path_ssh)

#encrypted content
out <- read_rds(file.path(path_ssh, "api_key.enc.rds"))
```
# Consumer API keys
ConsumerAPIkeys <- decrypt_mykeys(path_encrypted_content = file.path(path_ssh, 'api_key.enc.rds'), path_private_key = path_private_key)

---

DFR

*Table with aggregated trade results*

**Description**

Table with aggregated trade results

**Usage**

DFR

**Format**

A dataframe with one column

- **MagicNumber**: Unique identifiers of the Trading Robots from Trade Log
- **TicketNumber**: Ticket Number of closed position
- **OrderStartTime**: Date and Time when order started
- **OrderCloseTime**: Date and Time when order closed
- **Profit**: Monetary result of the trade
- **Symbol**: Symbol of the Asset e.g. EURUSD
- **OrderType**: Order Type 0 - buy, 1 - sell
- **CUMSUM_PNL**: Cumulative sum of the ordered data

---

encrypt_api_key

*Encrypt api keys*

**Description**

Provide easy interface to encrypt the api key. In order to use function simply provide a string with an API key. In addition provide the path to the .ssh folder and names of the private and public keys.
encrypt_api_key

Usage

encrypt_api_key(
   api_key,
   enc_name = "api_key.enc.rds",
   path_ssh = "path_ssh",
   file_rsa = "id_api",
   file_rsa_pub = "id_api.pub"
)

Arguments

api_key String with API key
enc_name String with a name of the file with encrypted key. Default name is 'api_key.enc.rds'
path_ssh String with path to the file with rsa keys. Same place will be used to store encrypted data
file_rsa String with a name of the file with a private key. Default name is 'id_api'
file_rsa_pub String with a name of the file with a public key. Default name is 'id_api.pub'

Details

Make sure to clean the history of the R session

Value

Writes a file with encrypted key

References

for more info on how to use RSA cryptography in R check my course https://www.udemy.com/course/keep-your-secrets-under-control/?referralCode=5B78D58E7C06AFFD80AE

Examples

library(openssl)
library(magrittr)
library(readr)
path_ssh <- normalizePath(tempdir(), winslash = "/")
rsa_keygen() %>% write_pem(path = file.path(path_ssh, "id_api"))
# extract and write your public key
read_key(file = file.path(path_ssh, "id_api"), password = "") %>%
  ["("pubkey") %>% write_pem(path = file.path(path_ssh, "id_api.pub"))

path_private_key <- file.path(path_ssh, "id_api")
path_public_key <- file.path(path_ssh, "id_api.pub")

#encrypting string 'my_key'...
encrypt_api_key(api_key = 'my_key', enc_name = 'api_key.enc.rds', path_ssh = path_ssh)
```
out <- read_rds(file.path(path_ssh, "api_key.enc.rds"))

# decrypting the password using public data list and private key
api_key <- decrypt_envelope(out$data,
    out$iv,
    out$session,
    path_private_key, password = "") %>%
    unserialize()

# outcome of the encryption will be a string 'my_key'
```

EURUSDM15X75  Table with indicator and price change dataset

Description

Table with indicator and price change dataset

Usage

EURUSDM15X75

Format

A dataframe with several columns

X1  future price change
X2-X76  Values of the macd indicator

evaluate_macroeconomic_event

Function used to evaluate market type situation by reading the file with Macroeconomic Events and writing a trigger to the trading robot

Description

Function is reading the content of the file 01_MacroeconomicEvent.csv. Content of the file can be either 1 or 0. 1 - when Macro Economic event is present, 0 - when it's not. Function will also read magic number of the trading robots. This is indicated in the file 'Setup.csv'. Final outcome of the function is the series of files written to the destination directories. These files will either enable or disable opening of new positions in the trading robots #
evaluate_macroeconomic_event

Usage

```r
evaluate_macroeconomic_event(
    setup_file_path,
    setup_file_name = "Setup.csv",
    macro_event_path,
    macro_file_name = "01_MacroeconomicEvent.csv",
    path_T1,
    path_T3
)
```

Arguments

- `setup_file_path` string, path to the folder with Setup.csv file
- `setup_file_name` string, name of the file 'Setup.csv'
- `macro_event_path` string, path to the folder with a file '01_MacroeconomicEvent.csv'
- `macro_file_name` string, name of the file '01_MacroeconomicEvent.csv'
- `path_T1` Path of the Terminal 1
- `path_T3` Path of the Terminal 3

Details

This function is used exclusively with Market Type recognition system.
Final evaluation will consist in writing a dedicated file with a simple information:
When Macro economic even is not present:
"Magic","IsEnabled" 8139125,1
or, when Macro economic event is present:
"Magic","IsEnabled" 8139125,0

Value

Function will write files indicating to enable or disable trading systems to open new orders

Examples

```r
# evaluate data on macroeconomic event (required to start trading)
library(dplyr)
library(readr)

dir <- normalizePath(tempdir(), winslash = "/")

evaluate_macroeconomic_event(setup_file_path = system.file('extdata', package = "lazytrade"),
    setup_file_name = "Setup.csv",
```
evaluate_market_type

Function to score data and predict current market type using pre-trained classification model

Description

PURPOSE: Function that uses Deep Learning model and Time Series Column of the dataframe to find out specific market type of the financial asset it will also discard bad result outputting -1 if it is the case

Usage

evaluate_market_type(x, model_path, num_cols)

Arguments

x - dataframe with one column containing asset indicator in the time descending order, typically 64 or more values
model_path - path to the model
num_cols - number of columns (features) in the final vector input to the model

Details

it is mandatory to switch on the virtual h2o machine with h2o.init() also to shut it down with h2o.shutdown(prompt = F)

Value

dataframe with predicted value of the market type

Examples

library(h2o)
library(magrittr)
library(dplyr)
library(readr)
library(lazytrade)
generate_RL_policy

Function performs RL and generates model policy

Description

This function will perform Reinforcement Learning using Trading Data. It will suggest whether or not it is better to keep using trading systems or not. Function is just using results of the past performance to generate the recommendation (not a holy grail).

Usage

generate_RL_policy(x, states, actions, control)
generate_RL_policy_mt

**Arguments**

- **x** - Dataframe containing trading data
- **states** - Selected states of the System
- **actions** - Selected actions executed under environment
- **control** - Control parameters as defined in the Reinforcement Learning Package

**Details**

Initial policy is generated using a dummy zero values. This way function starts working directly from the first observation. However policy ‘ON’ value will only be generated once the Q value is greater than zero.

**Value**

Function returns data frame with reinforcement learning model policy

**Author(s)**

(C) 2019 Vladimir Zhbanko

**Examples**

```r
library(dplyr)
library(ReinforcementLearning)
library(magrittr)

data(data_trades)
states <- c("tradewin", "tradeloss")
actions <- c("ON", "OFF")
control <- list(alpha = 0.7, gamma = 0.3, epsilon = 0.1)
generate_RL_policy(data_trades, states, actions, control)
```

---

**generate_RL_policy_mt**  
*Function performs RL and generates model policy for each Market Type*

---

**Description**

This function will perform Reinforcement Learning using Trading Data. It will suggest whether or not it is better to keep using trading systems or not. Function is just using results of the past performance to generate the recommendation (not a holy grail).

**Usage**

`generate_RL_policy_mt(x, states, actions, control)`
**get_profit_factorDF**

**Arguments**

- x - Dataframe containing trading data
- states - possible states for Reinforcement Learning
- actions - possible actions
- control - control parameters

**Details**

Initial policy is generated using a dummy zero values. This way function starts working directly from the first observation. However policy 'ON' value will only be generated once the Q value is greater than zero.

**Value**

Function returns data frame with reinforcement learning model policy

**Examples**

```r
library(dplyr)
library(ReinforcementLearning)
data(trading_systemDF)
states <- c("BUN", "BUY", "BEN", "BEV", "RAN", "RAV")
actions <- c("ON", "OFF")
control <- list(alpha = 0.7, gamma = 0.3, epsilon = 0.1)
generate_RL_policy_mt(trading_systemDF, states, actions, control)
```

---

**get_profit_factorDF**

Function that returns the profit factors of the systems in a form of a DataFrame

**Description**

Function that returns the profit factors of the systems in a form of a DataFrame

**Usage**

`get_profit_factorDF(x, num_orders)`

**Arguments**

- x - data frame with orders. Note x must contain MagicNumber and Profit columns!
- num_orders - desired number of orders to base profit factor calculation
Value

- Function returns dataframe with column PrFact with calculated profit factor value for each trading robot

Examples

```r
library(lazytrade)
library(dplyr)
library(magrittr)
data(profit_factorDF)
get_profit_factorDF(profit_factorDF, 10)
```

import_data

Import Data file with Trade Logs to R.

Description

Function is capable to import file with executed trades log. Files do not have column headers hence function will take care to name columns as well as to perform relevant cleansing

Usage

```r
import_data(path_terminal, trade_log_file, demo_mode = FALSE)
```

Arguments

- `path_terminal` - path of the Trading Terminal where the file with data is written
- `trade_log_file` - File name where the order results are written
- `demo_mode` - When true function uses data stored in the package data folder

Value

Function will return the dataframe with trade data and automatically set proper column types

Author(s)

(C) 2019 Vladimir Zhbanko
**Examples**

```r
library(lazytrade)
library(dplyr)
library(readr)
library(lubridate)
DFT1 <- import_data(trade_log_file = system.file("extdata", "OrdersResultsT1.csv", package = "lazytrade"),
                    demo_mode = TRUE)
```

---

**import_data_mt**

*Import Market Type related Data to R from the Sandbox*

**Description**

Function imports file from the MetaTrader sandbox. Function performs necessary cleansing of the data column types

**Usage**

```r
import_data_mt(path_terminal, trade_log_file, system_number, demo_mode = FALSE)
```

**Arguments**

- `path_terminal` - path to the sandbox
- `trade_log_file` - direct path to the log file (used for demo purposes)
- `system_number` - magic number id of the trading system
- `demo_mode` - when true, uses sample datafile stored in the package

**Value**

function returns the data frame with 3 columns including market type code

**Author(s)**

(C) 2019 Vladimir Zhbanko
Examples

```r
library(dplyr)
library(readr)
import_data_mt(trade_log_file = system.file("extdata", "MarketTypeLog8132101.csv", 
       package = "lazytrade"),
       demo_mode = TRUE)
```

---

**indicator_dataset**  
*Table with indicator dataset*

**Description**  
Table with indicator dataset

**Usage**  
`indicator_dataset`

**Format**  
A dataframe with several columns

- **X1**  Date and time of the indicator sample
- **X2-X29**  Values of the assets

---

**indicator_dataset_big**  
*Table with indicator dataset, 30000 rows*

**Description**  
Table with indicator dataset, 30000 rows

**Usage**  
`indicator_dataset_big`

**Format**  
A dataframe with several columns

- **X1**  Date and time of the indicator sample
- **X2-X29**  Values of the assets
load_asset_data

Load and Prepare Asset Data

Description

Function imports file with financial asset data. Each column represent one asset, rows represent observations. Values in specific columns will be normalized by dividing them by 100. This is specifically done for pairs with JPY. In addition, X1 column will be converted to the ymd_hms format.

Usage

```r
load_asset_data(
  path_terminal,
  trade_log_file,
  time_period = 1,
  data_depth = 50000
)
```

Arguments

- `path_terminal` - path to the MT4 terminal, string
- `trade_log_file` - csv file name where the data is stored, without ".csv"
- `time_period` - data periodicity in minutes, can be 1, 15, 60
- `data_depth` - collected data deepth in rows. describe how many rows in original file to read

Details

Works for both price and indicator values, function parameters allowing to import different files. File names are selected to account different time periodicity and amount of the data.

Value

- dataframe with asset data in columns where X1 column is in a POSIXct format

Examples

```r
library(readr)
library(dplyr)
library(lubridate)
library(magrittr)
path_terminal <- system.file("extdata", package = "lazytrade")

# load and prepare prices data
prices <- load_asset_data(path_terminal = path_terminal, 
                          trade_log_file = "AI_CP",
```
```r
# load and prepare indicator data
macd <- load_asset_data(path_terminal = path_terminal,
                         trade_log_file = "AI_Macd",
                         time_period = 60,
                         data_depth = "300")
```

---

### log_RL_progress

*Function to log RL progress.*

#### Description

Function will record Q values during updating of the model. These values will be used by another function.

#### Usage

```r
log_RL_progress(x, states, actions, control)
```

#### Arguments

- `x` - dataframe containing trading results
- `states` - Selected states of the System
- `actions` - Selected actions executed under environment
- `control` - control parameters as defined in the Reinforcement Learning Package

#### Value

dataframe with log of RL model

#### Examples

```r
# retrieve RL model Q values progress
library(ReinforcementLearning)
library(dplyr)
library(magrittr)
data(data_trades)
x <- data_trades
states <- c("tradewin", "tradeloss")
actions <- c("ON", "OFF")
control <- list(alpha = 0.7, gamma = 0.3, epsilon = 0.1)
log_RL_progress(x = x, states = states, actions = actions, control = control)
```
log_RL_progress_mt

Function to log RL progress, dedicated to Market Types

Description

Function will record Q values during updating of the model. These values will be used by another function.

Usage

log_RL_progress_mt(x, states, actions, control)

Arguments

- x - dataframe containing trading results
- states - Selected states of the System
- actions - Selected actions executed under environment
- control - control parameters as defined in the Reinforcement Learning Package

Value

dataframe with log of RL model

Examples

# retrieve RL model Q values progress
library(ReinforcementLearning)
library(dplyr)
library(magrittr)
data(trading_systemDF)
x <- trading_systemDF
states <- c("BUN", "BUY", "BEN", "BEV", "RAN", "RAV")
actions <- c("ON", "OFF") # 'ON' and 'OFF' are referring to decision to trade with Slave system
control <- list(alpha = 0.7, gamma = 0.3, epsilon = 0.1)

log_RL_progress_mt(x = x, states = states, actions = actions, control = control)
### Description

Table with indicator only used to train model, 128 rows

### Usage

macd_100

### Format

A dataframe with several columns

- **EURUSD** Values of the macd indicator
- **GBPUSD** Values of the macd indicator
- **AUDUSD** Values of the macd indicator
- **NZDUSD** Values of the macd indicator
- **USDCAD** Values of the macd indicator
- **USDCHF** Values of the macd indicator
- **USDJPY** Values of the macd indicator
- **EURGBP** Values of the macd indicator
- **EURJPY** Values of the macd indicator
- **EURCHF** Values of the macd indicator
- **EURNZD** Values of the macd indicator
- **EURCAD** Values of the macd indicator
- **EURAUD** Values of the macd indicator
- **GBPUSD** Values of the macd indicator
- **GBPCAD** Values of the macd indicator
- **GBPCHF** Values of the macd indicator
- **GBPJPY** Values of the macd indicator
- **GBPUSD** Values of the macd indicator
- **AUDCAD** Values of the macd indicator
- **AUDCHF** Values of the macd indicator
- **AUDJPY** Values of the macd indicator
- **AUDNZD** Values of the macd indicator
- **CADJPY** Values of the macd indicator
- **CHFJPY** Values of the macd indicator
- **NZDJPY** Values of the macd indicator
- **NZDCAD** Values of the macd indicator
- **NZDCHF** Values of the macd indicator
- **CADCHF** Values of the macd indicator
macd_df

Table with one column indicator dataset

Description

Table with one column indicator dataset

Usage

macd_df

Format

A dataframe with one column

CADCHF  Indicator values of the asset

macd_ML2_small

Table with indicator and market type category used to train model

Description

Table with indicator and market type category used to train model

Usage

macd_ML2_small

Format

A dataframe with several columns

X1-X128  Values of the macd indicator

X129  Category of Market Type
Function to train Deep Learning Classification model for Market Type recognition

Description

Function is training h2o deep learning model to match manually classified patterns of the financial indicator. Main idea is to be able to detect Market Type by solely relying on the current indicator pattern. This is in the attempt to evaluate current market type and to use proper trading strategy.


Usage

```r
mt_make_model(num_bars, path_model, path_data, f_name_data)
```

Arguments

- `num_bars`: Number of bars used to detect pattern
- `path_model`: Path where the models are be stored
- `path_data`: Path where the aggregated historical data is stored, if exists in rds format
- `f_name_data`: Name of the file with the data

Details

Function is using manually prepared dataset

Value

Function is writing file object with the model

Author(s)

(C) 2020 Vladimir Zhbanko

Examples

```r
library(dplyr)
library(readr)
library(h2o)
library(lazytrade)

path_model <- normalizePath(tempdir(), winslash = "/")
```
opt_aggregate_results

Function to aggregate trading results from multiple folders and files

Description

PURPOSE: Read multiple files stored in different folders Store results to the intermediate dataframe.

Usage

opt_aggregate_results(fold_path)

Arguments

fold_path - path to the folder containing subfolders

Details

user must provide the path to the files in the folders all files in subfolders are read and aggregated into one data object. Data object is sorted in descending order by order close time

Value

Dataframe with trading results
Examples

```r
library(lazytrade)
library(readr)
library(dplyr)
library(magrittr)
library(lubridate)
DFOLDER <- system.file("extdata/RES", package = "lazytrade")
#dir <- normalizePath(tempdir(), winslash = "/")
opt_aggregate_results(fold_path = DFOLDER)
```

opt_create_graphs  Function to create summary graphs of the trading results

Description

Create graphs and store them into pdf file

Usage

```r
opt_create_graphs(x, outp_path, graph_type = "pdf")
```

Arguments

- `x` - dataframe with aggregated trading results
- `outp_path` - path to the folder where to write file
- `graph_type` - character, one of the options c('ts', 'bars', 'pdf')

Details

Bar graph and time series optionally written to the pdf file. File is named with a date of analysis to the location specified by the user

Value

graphic output

Examples

```r
library(lazytrade)
library(readr)
library(dplyr)
library(magrittr)
library(lubridate)
library(ggplot2)
```
data(DFR)
dir <- normalizePath(tempdir(), winslash = "/")
# create pdf file with two graphs
opt_create_graphs(x = DFR, outp_path = dir)

# only show time series plot
opt_create_graphs(x = DFR, graph_type = 'ts')

---

**policy_tr_systDF**  
*Table with Market Types and sample of actual policy for those states*

**Description**

Table with Market Types and sample of actual policy for those states

**Usage**

`policy_tr_systDF`

**Format**

A dataframe with 2 columns:

- **MarketType**  Current Market Type status
- **Policy**  Policy choice

---

**price_dataset**  
*Table with price dataset*

**Description**

Table with price dataset

**Usage**

`price_dataset`

**Format**

A dataframe with several columns

- **X1**  Date and time of the price sample
- **X2-X29**  Values of the assets
### price_dataset_big

**Table with price dataset, 30000 rows**

**Description**

Table with price dataset, 30000 rows

**Usage**

```r
price_dataset_big
```

**Format**

A dataframe with several columns

- **X1** Date and time of the price sample
- **X2-X29** Values of the assets

### profit_factor

**Calculate Profit Factor**

**Description**

Calculate profit factor using a data vector with the trading results

**Usage**

```r
profit_factor(x)
```

**Arguments**

- **x** column vector with profit or loss of the orders for one system

**Value**

function should calculate profit factor for this vector and return one value also as vector

**Author(s)**

(C) 2019 Vladimir Zhbanko
Examples

```r
library(magrittr)
library(dplyr)
data(profit_factor_data)
profit_factor_data %>%
group_by(X1) %>%
summarise(PnL = sum(X5),
    NumTrades = n(),
    PrFact = profit_factor(X5)) %>%
select(PrFact) %>% head(1) %>% as.vector() %>% round(3)
```

---

**profit_factorDF**  
*Table with Trade results samples*

**Description**

Table with Trade results samples

**Usage**

profit_factorDF

**Format**

A dataframe with several columns

- **MagicNumber**  Unique identifiers of the Trading Robots
- **TicketNumber**  Ticket Number of closed position
- **OrderStartTIme**  Date and Time when order started
- **OrderCloseTime**  Date and Time when order closed
- **Profit**  Monetary result of the trade
- **Symbol**  Symbol of the Asset e.g. EURUSD
- **OrderType**  Order Type 0 - buy, 1 - sell
**profit_factor_data**

*Table with Trade results samples*

**Description**

Table with Trade results samples

**Usage**

```
profit_factor_data
```

**Format**

A dataframe with several columns

- **X1** Unique identifiers of the Trading Robots
- **X2** Ticket Number of closed position
- **X3** Date and Time when order started
- **X4** Date and Time when order closed
- **X5** Monetary result of the trade
- **X6** Symbol of the Asset e.g. EURUSD
- **X7** Order Type 0 - buy, 1 - sell

---

**record_policy**

*Record Reinforcement Learning Policy.*

**Description**

Function will write a policy 'decision' to the csv file specific for each Expert Advisor

**Usage**

```
record_policy(
    x,  
    last_result, 
    trading_system, 
    path_terminal, 
    fileName = "SystemControl"
)
```
record_policy_mt

Record Reinforcement Learning Policy for Market Types

Description

Function will write a policy 'decision' to the csv file specific for each Expert Advisor

Usage

```r
record_policy_mt(
  x,
  trading_system,  # character vector of length 1 with Trading System Magic Number information
  path_terminal,    # path to the sandbox where this Policy/Decision must be written
  fileName          # string, desired control file prefix e.g. 'SystemControl'
)
```
Arguments

- **x** - Dataframe containing columns MarketType and Policy
- **trading_system** - numeric vector of length 1 with Trading System Magic Number information
- **path_terminal** - string, path to the terminal where this Policy/Decision must be written
- **fileName** - string, desired control file prefix e.g. 'SystemControlMT'

Value

nothing is returned but function will write csv file to the supplied directory

Examples

```r
library(stringr)
data(policy_tr_systDF)
dir <- normalizePath(tempdir(), winslash = "/")

record_policy_mt(x = policy_tr_systDF,
                 trading_system = 8118101,
                 path_terminal = dir,
                 fileName = "SystemControlMT")
```

---

**result_prev**

| Table with one column as result from the model prediction |

Description

Table with one column as result from the model prediction

Usage

result_prev

Format

A dataframe with one column

- **predict** Predicted values from the model
### Description

Table with predicted price change.

### Usage

```r
result_R
```

### Format

A dataframe with one column

- **predict**: predicted future price change

### self_learn_ai_R

*Function to train Deep Learning regression model*

### Description

Function is training h2o deep learning model to match future prices of the asset to the indicator pattern. Main idea is to be able to predict future prices by solely relying on the most recent indicator pattern.

### Usage

```r
self_learn_ai_R(
  price_dataset,
  indicator_dataset,
  num_bars,
  timeframe,
  path_model,
  setup_mode = FALSE,
  research_mode = FALSE,
  write_log = TRUE
)
```

### Arguments

- **price_dataset**: Dataset containing assets prices. It will be used as a label
- **indicator_dataset**: Dataset containing assets indicator which pattern will be used as predictor
- **num_bars**: Number of bars used to detect pattern
self_learn_ai_R

timeframe Data timeframe e.g. 1 min
path_model Path where the models are be stored
setup_mode When TRUE function will attempt to write model to the disk without checking it
research_mode When TRUE model will be saved and model result will be stored as well. To be used at the first run.
write_log Writes results of the newly trained model and previously used model to the file

Details
Performs data manipulation and training of the model. Function is handling shift of the price and indicator datasets. Function will also check how the model predict by using trading objective. NOTE: Always run parameter research_mode = TRUE for the first time
Because of the function is intended to periodically re-train the model it would always check how the previous model was working. In case new model is better, the better model will be used.
Function can also write a log files with a results of the strategy test

Value
Function is writing files into Decision Support System folder

Author(s)
(C) 2019 Vladimir Zhbanko

Examples

library(dplyr)
library(readr)
library(magrittr)
library(h2o)
library(lazytrade)
# start h2o engine (using all CPU's by default)
h2o.init()

path_model <- normalizePath(tempdir(), winslash = "/")
path_data <- normalizePath(tempdir(), winslash = "/")
data(indicator_dataset_big)
data(price_dataset_big)

prices <- price_dataset_big
macd <- indicator_dataset_big

# performing Deep Learning Regression using the custom function
self_learn_ai_R(price_dataset = prices,
test_data_pattern

indicator_dataset = macd, num_bars = 75, timeframe = 60, path_model = path_model, setup_mode = FALSE, research_mode = FALSE, write_log = FALSE)

# stop h2o engine
h2o.shutdown(prompt = FALSE)

#set delay to insure h2o unit closes properly before the next test
Sys.sleep(5)

test_data_pattern

<table>
<thead>
<tr>
<th>Description</th>
<th>Table with several columns containing indicator values and Label values</th>
</tr>
</thead>
</table>

Usage
test_data_pattern

Format
A dataframe with several columns

LABEL Asset values as were recorded in the future
V1-V49 Transposed values of the indicator

test_model

Test model using independent price data.

Description
Goal of the function is to verify how good predicted results are.

Usage
test_model(test_dataset, predictor_dataset, test_type)
Arguments

- **test_dataset** - Dataset containing the column 'LABEL' which will correspond to the real outcome of Asset price change. This column will be used to verify the trading strategy.

- **predictor_dataset** - Dataset containing the column 'predict'. This column is corresponding to the predicted outcome of Asset change. This column will be used to verify strategy outcomes.

- **test_type** - can be either "regression" or "classification" used to distinguish which type of model is being used.

Details

This function should work to backtest any possible dataset length. It could be that we will need to use it for testing 1 week or 1 month. It should also work for both Regression and Classification models. Note: strategy outcomes assumes trading on all 28 major forex pairs.

Value

Function will return a data frame with several quality score metrics for the best model. In case quality score is positive or more than 1 the model would likely be working good. In case the score will be negative then the model is not predicting good. Internal logic will test several predictor thresholds and will indicate the best one.

Examples

```r
library(dplyr)
data(result_prev)
data(test_data_pattern)

## evaluate hypothetical results of trading using the model
test_model(test_dataset = test_data_pattern,
            predictor_dataset = result_prev,
            test_type = "regression")
```

Description

Converts Time Series Column of the dataframe to the matrix with specified number of columns. Number of rows will be automatically found. Eventually not complete last row will be discarded.
Usage

to_m(x, n_cols)

Arguments

x - dataframe with one column
n_cols - number of columns in the matrix

Value

- matrix with specified amount of rows

Examples

library(magrittr)
macd_m <- seq(1:1000) %>% as.data.frame() %>% to_m(64)

<table>
<thead>
<tr>
<th>TradeStatePolicy</th>
<th>Table with Trade States and sample of actual policy for those states</th>
</tr>
</thead>
</table>

Description

Table with Trade States and sample of actual policy for those states

Usage

TradeStatePolicy

Format

A dataframe with 2 columns:

TradeState Current trade state status
Policy Policy choice
trading_systemDF | Table with trade data and joined market type info

Description
Table with trade data and joined market type info

Usage
```
trading_systemDF
```

Format
A dataframe with several columns

"MagicNumber.x" Unique identifiers of the Trading Robots from Trade Log
TicketNumber Ticket Number of closed position
OrderStartTime Date and Time when order started
OrderCloseTime Date and Time when order closed
Profit Monetary result of the trade
Symbol Symbol of the Asset e.g. EURUSD
OrderType Order Type 0 - buy, 1 - sell
"MagicNumber.y" Unique identifiers of the Trading Robots from Ticket Opening Log
"MarketType Logged Market Type of the asset at the moment of Ticket Opening

util_generate_password | R function to generate random passwords for MT4 platform or other needs

Description
Utility function to generate random passwords. Wrapper of cryptographic functions from `openssl` library in R. Password length can be customized. By default function just output randomly generated 8 symbol password suitable for MT4 logins. It is also possible to create other passwords and include special symbols. When required, it’s possible to write resulting password to the txt file. Once generated, password is written to the destination supplied by the user.
**util_generate_password**

**Usage**

```r
def util_generate_password(
    salt = "something random",
    pass_len = 8,
    write_file = FALSE,
    file_name = "",
    special_symbols = FALSE
)
```

**Arguments**

- **salt**: string, random text supplied by the user
- **pass_len**: integer, number specifying how long should the password be
- **write_file**: bool, if true writes result to the txt file
- **file_name**: string, indicate path of the file where to write text result
- **special_symbols**: bool, if true adds special symbols

**Details**

Passwords are generated using sha512 cryptographic function from openssl package. System date and user 'salt' is used to supply initial text for cryptographic function. Hashing function is using additional 'salt' which will be based on the current System time. Additionally, only a part of generated string is selected and used for password. Some letters of generated string are converted from lower to upper case.

**Value**

- string or text file with password

**Author(s)**

(C) 2019 Vladimir Zhbanko

**Examples**

```r
library(stringr)
library(magrittr)
library(openssl)
library(readr)

dir <- normalizePath(tempdir(), winslash = "/")
file_path <- file.path(dir, "p.txt")

#write to file
util_generate_password(salt = "random text", file_name = file_path)

#generate 8digit
```
util_generate_password(salt = 'random text')

# generate password with special symbols
util_generate_password(salt = 'random text', special_symbols = TRUE)

# generate longer password with special symbols
util_generate_password(salt = 'random text', pass_len = 10, special_symbols = TRUE)

writeCommandViaCSV

Write csv files with indicated commands to the external system

Description
Function is capable to read the data and writing multiple files e.g. 'SystemControl8139124.csv'

Usage
writeCommandViaCSV(x, path_terminal = "", fileName = "SystemControl")

Arguments
x - dataframe object with resulting command e.g. 1 - enable; 0 - disable
path_terminal - path to the terminal
fileName - desired control file prefix e.g. 'SystemControl'

Value
Function is writing multiple files e.g. 'SystemControl8139124.csv' to the Sandbox
typical content of the file: "Magic","IsEnabled" 8139124,1

Author(s)
(C) 2019 Vladimir Zhbanko

Examples
library(dplyr)
library(readr)
library(lubridate)
DFT1 <- import_data(trade_log_file = system.file("extdata",
"OrdersResultsT1.csv", package = "lazytrade"),
demo_mode = TRUE)
write_command_via_csv

```
dir <- normalizePath(tempdir(), winslash = "/")
DFT1 %>%
group_by(MagicNumber) %>% select(MagicNumber) %>% mutate(IsEnabled = 0) %>%
# write commands to disable systems
writeCommandViaCSV(path_terminal = file.path(dir))
```

---

**write_command_via_csv**  
*Write csv files with indicated commands to the external system*

**Description**

Function is capable to read the data and writing multiple files e.g. 'SystemControl8139124.csv'

**Usage**

```
write_command_via_csv(x, path_terminal = "", fileName = "SystemControl")
```

**Arguments**

- `x` - dataframe object with resulting command e.g. 1 - enable; 0 - disable
- `path_terminal` - path to the terminal
- `fileName` - desired control file prefix e.g. 'SystemControl'

**Value**

Function is writing multiple files e.g. 'SystemControl8139124.csv' to the Sandbox 
typical content of the file: "Magic","IsEnabled" 8139124,1

**Author(s)**

(C) 2019 Vladimir Zhbanko

**Examples**

```
library(dplyr)
library(readr)
library(lubridate)
library(lazytrade)
DFT1 <- import_data(trade_log_file = system.file("extdata",  
  "OrdersResultsT1.csv",  
  package = "lazytrade"),  
  demo_mode = TRUE)

dir <- normalizePath(tempdir(), winslash = "/")
```
```r
DFT1 %>%
group_by(MagicNumber) %>% select(MagicNumber) %>% mutate(IsEnabled = 0) %>%
# write commands to disable/enable systems
write_command_via_csv(path_terminal = file.path(dir))
```

---

**write_control_parameters**

*Function to find and write the best control parameters.*

**Description**

This function is supposed to run on a weekly basis. Purpose of this function is to perform RL and trading simulation and find out the best possible control parameters for the RL function.

**Usage**

```r
write_control_parameters(x, path_control_files)
```

**Arguments**

- `x` - dataset containing the trading results for one trading robot
- `path_control_files` - path where control parameters will be saved

**Details**

Function is used by the R script Adapt_RL_control.R

**Value**

Function writes best control parameters to be used by the Reinforcement Learning Function

**Author(s)**

(C) 2019 Vladimir Zhabenko

**Examples**

```r
#test lasts 15 sec:
library(dplyr)
library(readr)
library(ReinforcementLearning)
library(magrittr)
```
write_control_parameters_mt

Function to find and write the best control parameters.

Description

This function is supposed to run on a weekly basis. Purpose of this function is to perform RL and trading simulation and find out the best possible control parameters for the RL function.

Usage

write_control_parameters_mt(x, path_control_files)

Arguments

x - dataset containing the trading results for one trading robot
path_control_files - path where control parameters will be saved

Details

Function is used by the R script Adapt_RL_MT_control.R

Value

Function writes best control parameters to be used by the Reinforcement Learning Function

Author(s)

(C) 2019 Vladimir Zhbanko

Examples

# test lasts 15 sec:
library(dplyr)
library(readr)
library(ReinforcementLearning)
library(magrittr)
data(trading_systemDF)

data(data_trades)
write_control_parameters(data_trades, path_control_files = tempfile())

# use optimal control parameters found by auxiliary function
write_ini_file

Create initialization files to launch MT4 platform with specific configuration

Description

Function generate initialization files suitable for launching MT4 terminal with specific parameters. Several options available for generating files specific for each purpose. Option 'prod' will just use existing profile and connect to the broker server Option 'backtest' will generate file for the robot backtest Option 'opt' will generate file needed for the robot optimization Option 'full' allows to specify any desired parameter

Usage

```r
write_ini_file(
  mt4_Profile = "Default",
  mt4_MarketWatch = "Forex.set",
  mt4_Login = "1234567",
  mt4_Password = "xxxxxXX",
  mt4_Server = "BrokerServerName",
  mt4_AutoConfiguration = "false",
  mt4_EnableNews = "false",
  mt4_ExpertsEnable = "true",
  mt4_ExpertsDllImport = "true",
  mt4_ExpertsExpImport = "true",
  mt4_ExpertsTrades = "true",
  mt4_Symbol = "EURUSD",
  mt4_Period = "H1",
  mt4_Template = "Default",
  mt4_Expert = "",
  mt4_ExpertParameters = "",
  mt4_Script = "",
  mt4_ScriptParameters = "",
  mt4_TestExpert = "",
  mt4_TestExpertParameters = "",
  mt4_TestSymbol = "EURUSD",
  mt4_TestPeriod = "H1",
  mt4_TestModel = "",
  mt4_TestSpread = "",
  mt4_TestOptimization = "false",
  mt4_TestDateEnable = "true",
  mt4_TestFromDate = "",
)```
write_init_file

mt4_TestToDate = "", mt4_TestReport = "test report",
mt4_TestReplaceReport = "false",
mt4_TestShutdownTerminal = "", mt4_TestVisualEnable = "false",
dss_inifilepath = "", dss_inifilename = "test.ini",
dss_mode = "prod"
)

Arguments

mt4_Profile string, the subdirectory name in the /profiles directory. The charts will be opened
in the client terminal according to the given profile. If this parameter is not
specified, the current profile will be opened

mt4_MarketWatch string, file name (the symbolsets directory) that contains the symbol list to be
shown in the Market Watch window.

mt4_Login string, the number of the account to connect to at startup. If this parameter is not
specified, the current login will be used.

mt4_Password string, the password that allows entering the system. This parameter will be
ignored if the client terminal stores personal data on the disk and the account to
be connected is in the list

mt4_Server string, the name of the trade server to be connected to. The server name is the
same as the name of the corresponding .srv file stored in the /config directory

mt4_AutoConfiguration string, "true" or "false" depending on whether the autoconfiguration of Data
Center setting should be enabled or not. If this parameter is not specified, the
value from the current server settings will be used.

mt4_EnableNews string, either 'false' or 'true'

mt4_ExpertsEnable string, enable/disable experts.

mt4_ExpertsDllImport string, enable/disable DLL imports

mt4_ExpertsExpImport string, enable/disable import of functions from external experts or MQL4 li-

braries.

mt4_ExpertsTrades string, enable/disable the experts trading

mt4_Symbol string, the symbol of the security the chart of which should be opened immedi-
ately after the terminal startup

mt4_Period string, the chart timeframe (M1, M5, M15, M30, H1, H4, D1, W1, MN). If this
parameter is not specified, H1 is used

mt4_Template string, the name of the template file (the templates directory), which should be
applied to the chart.
mt4_Expert string, the name of the expert that should be launched after the client terminal has started
mt4_ExpertParameters string, the name of the file containing the expert parameters (the MQL4 Presets directory).
mt4_Script string, the name of the script, which must be launched after the client terminal startup
mt4_ScriptParameters string, the name of the file containing the script parameters (the MQL5 Presets directory).
mt4_TestExpert string, the name of the expert to be launched for testing. If this parameter has not been specified, no testing is launched.
mt4_TestExpertParameters string, the name of the file containing parameters (the tester directory).
mt4_TestSymbol string, the name of the symbol used for the expert testing. If this parameter has not been specified, the latest value used in the tester is used.
mt4_TestPeriod string, the chart period (M1, M5, M15, M30, H1, H4, D1, W1, MN). If this parameter has not been specified, H1 is used.
mt4_TestModel string, 0, 1, or 2, depending on the testing model (Every tick, Control points, Open prices only). If this parameter has not been specified, 0 is used (Every tick)
mt4_TestSpread string, spread value that will be used for modeling Ask prices during testing. If 0 value is specified, the strategy tester will use the current spread of a symbol at the beginning of testing
mt4_TestOptimization string, enable/disable optimization. The values that can be taken are "true" or "false". If this parameter had not been specified, the "false" value is used.
mt4_TestDateEnable string, enable/disable the "Use date" flag. The values that can be taken are "true" or "false". If this parameter had not been specified, the "false" value is used.
mt4_TestFromDate string, the date, from which to start testing, appeared as YYYY.MM.DD. If this parameter has not been specified, this date is 1970.01.01.
mt4_TestToDate string, the date, on which to finish testing, appeared as YYYY.MM.DD. If this parameter has not been specified, this date is 1970.01.01.
mt4_TestReport string, the name of the test report file. The file will be created in the client terminal directory. A relative path can be specified, for example: tester\MovingAverageReport". If the extension has not been specified in the file name, the ".htm" will be set automatically. If this parameter has not been specified, the test report will not be formed
mt4_TestReplaceReport string, enable/disable the repeated report file record. The values that can be taken are "true" or "false"
mt4_TestShutdownTerminal string, enable/disable shutdown of the terminal after the testing has been finished.
```
write_ini_file

mt4_TestVisualEnable
    string, enable (true) or disable (false) the visual test mode. If the parameter is not specified, the current setting is used.
dss_inifilepath
    string, path on the computer where file will be stored
dss_inifilename
    string, file name that should be written
dss_mode
    string,

Details

added value of this function is the ability to generate multiple files to backtest several robots for several timeframes. For example it is solves the problem of doing repetitive tasks to 'backtest' robots for several currencies and repeat this procedure over time.

Most of the variables present in the function are starting with a prefix mt4_, the remainder of the name comes from the platform documentation, see references

Remaining variables are named with a prefix 'dss_' stands for 'Decision Support System', as these are the variables used for further automation purposes

Note that for simplicity reasons not all parameters are present in this function. e.g. FTP Settings and Proxy Server settings are not present

Value

output is a file with desired parameters

Author(s)

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References

All parameters used are taken from the reference documentation https://www.metatrader4.com/en/trading-platform/help/service/start_conf_file

Examples

library(lazytrade)

dir <- normalizePath(tempdir(),winslash = "/")

# test file to launch MT4 terminal with parameters
write_ini_file(mt4_Profile = "Default",
              mt4_Login = "12345678",
              mt4_Password = "password",
              mt4_Server = "BrokerServerName",
              dss_inifilepath = dir,
              dss_inifilename = "prod_T1.ini",
              dss_mode = "prod")
```
# test file to launch robot backtest
TO <- format(as.Date(Sys.Date()), "%Y.%m.%d")
FROM <- format(as.Date(Sys.Date()-60), "%Y.%m.%d")

# test file for MT4 use for backtesting
write_ini_file(mt4_Profile = "Default",
              mt4_Login = "12345678",
              mt4_Password = "password",
              mt4_Server = "BrokerServerName",
              mt4_TestExpert="FALCON_D\Falcon_D",
              mt4_TestExpertParameters="Falcon_D.set",
              mt4_TestSymbol="EURUSD",
              mt4_TestPeriod="H1",
              mt4_TestModel="2",
              mt4_TestSpread="20",
              mt4_TestOptimization="false",
              mt4_TestDateEnable="true",
              mt4_TestFromDate=FROM,
              mt4_TestToDate=TO,
              mt4_TestReport="EURUSD_Report",
              mt4_TestReplaceReport="false",
              mt4_TestShutdownTerminal="true",
              mt4_TestVisualEnable="false",
              dss_inifilepath = dir,
              dss_inifilename = "backtest.ini",
              dss_mode = "backtest")

---

x_test_model

Table with a dataset to test the Model

<table>
<thead>
<tr>
<th>Description</th>
<th>Table with a dataset to test the Model</th>
</tr>
</thead>
<tbody>
<tr>
<td>Usage</td>
<td>x_test_model</td>
</tr>
<tr>
<td>Format</td>
<td>A dataframe with several columns</td>
</tr>
<tr>
<td>X1</td>
<td>future price change</td>
</tr>
<tr>
<td>X2-X76</td>
<td>Values of the macd indicator</td>
</tr>
</tbody>
</table>
Index

*Topic datasets
  data_trades, 14
  DFR, 16
  EURUSDM15X75, 18
  indicator_dataset, 26
  indicator_dataset_big, 26
  macd_100, 30
  macd_df, 31
  macd_ML2_small, 31
  policy_tr_systDF, 35
  price_dataset, 35
  price_dataset_big, 36
  profit_factor_data, 38
  profit_factorDF, 37
  result_prev, 40
  result_R, 41
  test_data_pattern, 43
  TradeStatePolicy, 45
  trading_systemDF, 46
  x_test_model, 56

  aml_collect_data, 3
  aml_make_model, 5
  aml_score_data, 6
  aml_test_model, 8

  check_if_optimize, 10
  create_labelled_data, 12
  create_transposed_data, 13

  data_trades, 14
  decrypt_mykeys, 15
  DFR, 16

  encrypt_api_key, 16
  EURUSDM15X75, 18
  evaluate_macroeconomic_event, 18
  evaluate_market_type, 20

  generate_RL_policy, 21
  generate_RL_policy_mt, 22

  get_profit_factorDF, 23
  import_data, 24
  import_data_mt, 25
  indicator_dataset, 26
  indicator_dataset_big, 26

  load_asset_data, 27
  log_RL_progress, 28
  log_RL_progress_mt, 29

  macd_100, 30
  macd_df, 31
  macd_ML2_small, 31
  mt_make_model, 32

  opt_aggregate_results, 33
  opt_create_graphs, 34

  policy_tr_systDF, 35
  price_dataset, 35
  price_dataset_big, 36
  profit_factor, 36
  profit_factor_data, 38
  profit_factorDF, 37

  record_policy, 38
  record_policy_mt, 39
  result_prev, 40
  result_R, 41

  self_learn_ai_R, 41

  test_data_pattern, 43
  test_model, 43
  to_m, 44
  TradeStatePolicy, 45
  trading_systemDF, 46

  util_generate_password, 46

  write_command_via_csv, 49
write_control_parameters, 50
write_control_parameters_mt, 51
write_ini_file, 52
writeCommandViaCSV, 48

x_test_model, 56