Welcome to the course!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Joshua Ulrich
Instructor

datacamp

About me

- Author and/or maintainer of several packages
  - `TTR`, `xts`, `quantmod`, `blotter`, `quantstrat`
- R/Finance Conference Organizing Committee
- St. Louis R User Group

What is getSymbols()?

- Provides access to multiple data sources
- Returns `xts` object by default
- Can import data two ways:
  - Return data like an ordinary function
  - Create an object like `load()` does in base R

```
getSymbols(Symbols = "AAPL", src = "yahoo")
```

```
"AAPL"
```

```
getSymbols("AAPL")
```

```
"AAPL"
```

```
head(AAPL)
```

```
           AAPL.Open AAPL.High AAPL.Low AAPL.Close AAPL.Volume AAPL.Adjusted
2007-01-03     86.29     86.58    81.90      83.80   309579900      10.85709
2007-01-04     84.05     85.95    83.82      85.66   211815100      11.09807
2007-01-05     85.77     86.20    84.40      85.05   208685400      11.01904
2007-01-08     85.96     86.53    85.28      85.47   199276700      11.07345
2007-01-09     86.45     92.98    85.15      92.57   837324600      11.99333
2007-01-10     94.75     97.80    93.45      97.00   738220000      12.56728
```

getSymbols() data sources

Yahoo! Finance	YAHOO! FINANCE
Google Finance	Google Finance
FRED	FRED ECONOMIC DATA | ST. LOUIS FED
Oanda	OANDA
CSV	.csv

Other getSymbols() data sources

- Yahoo! Finance Japan
- MySQL
- SQLite
- RData
- rds (created by `saveRDS()`)

getSymbols() example

```
# Load data like load()
getSymbols("AAPL", auto.assign = TRUE)
```

```
"AAPL"
```

```
head(AAPL, n = 3)
```

```
           AAPL.Open AAPL.High AAPL.Low AAPL.Close AAPL.Volume AAPL.Adjusted
2007-01-03     86.29     86.58    81.90       83.80   309579900      10.85709
2007-01-04     84.05     85.95    83.82       85.66   211815100      11.09807
2007-01-05     85.77     86.20    84.40       85.05   208685400      11.01904
```

getSymbols() example

```
# Return data like a normal function
aapl <- getSymbols("AAPL", auto.assign = FALSE)
head(aapl, n = 3)
```

```
           AAPL.Open AAPL.High AAPL.Low AAPL.Close AAPL.Volume AAPL.Adjusted
2007-01-03     86.29     86.58    81.90      83.80   309579900      10.85709
2007-01-04     84.05     85.95    83.82      85.66   211815100      11.09807
2007-01-05     85.77     86.20    84.40      85.05   208685400      11.01904
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Introduction to Quandl

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What is Quandl?

- Data service:
  - **https://www.quandl.com/**
- R package:
  - **https://CRAN.R-project.org/package=Quandl**
- Function:
  - `Quandl::Quandl()`

Quandl() versus getSymbols()

- Both provide access to multiple data sources
- `Quandl()` always returns data (i.e. does not behave like `load()`)

```
# Instrument and source specified by Symbols and src arguments
quantmod::getSymbols(Symbols = "DGS10", src = "FRED")
```

```
# Instrument and source specified by code argument
dgs10 <- Quandl::Quandl(code = "FRED/DGS10")
```

Quandl() versus getSymbols()

- `type` argument controls class of return object:
  - "raw" (`data.frame` object), `xts`, `zoo`, `ts`, `timeSeries`

```
# Return xts object instead of data.frame
dgs10 <- Quandl::Quandl(code = "FRED/DGS10", type = "xts")
```

- Defaults
  - `getSymbols()` returns `xts`
  - `Quandl()` returns `data.frame`

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IMPORTING AND MANAGING FINANCIAL DATA IN R

Finding data from internet sources

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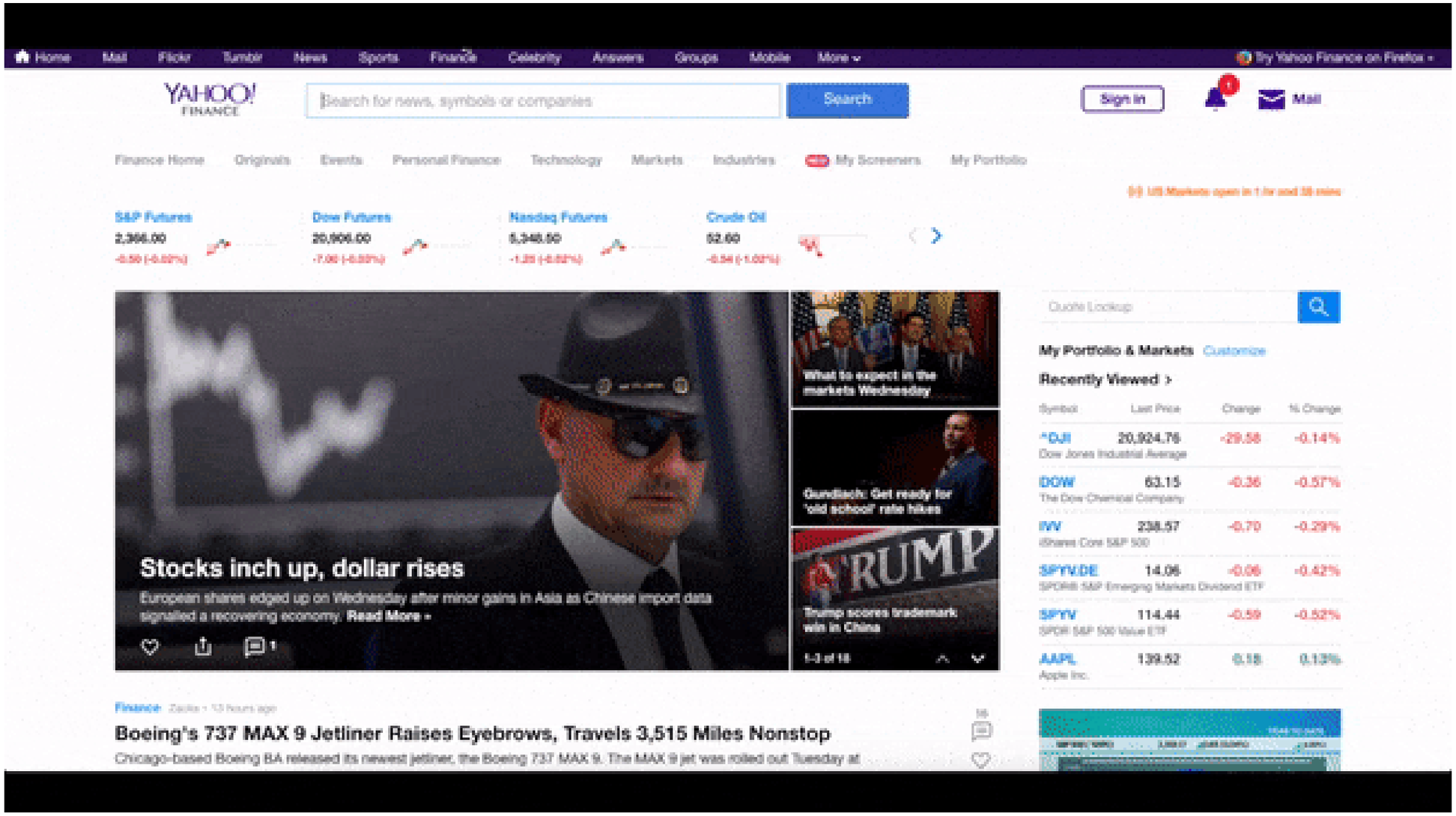
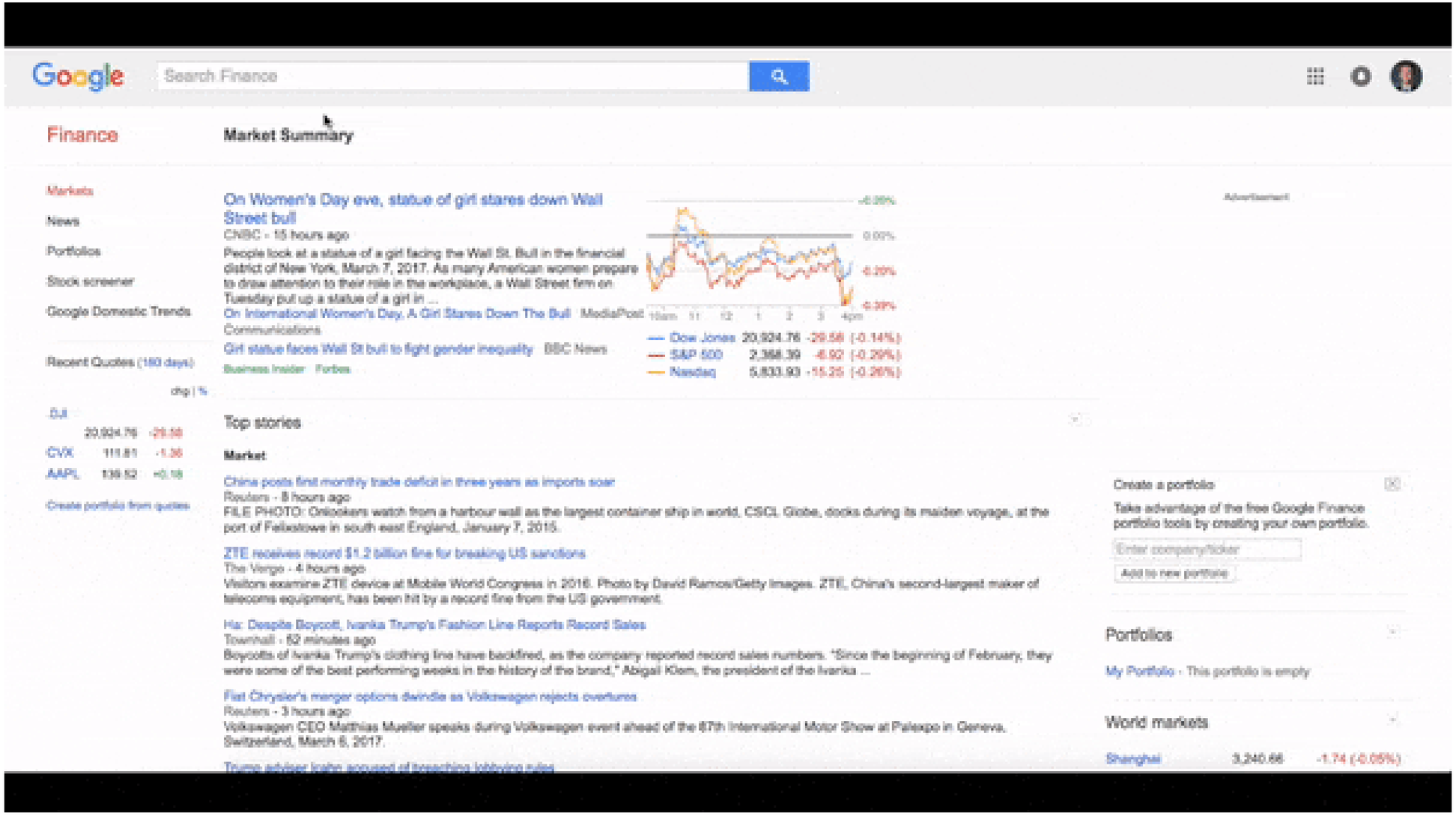
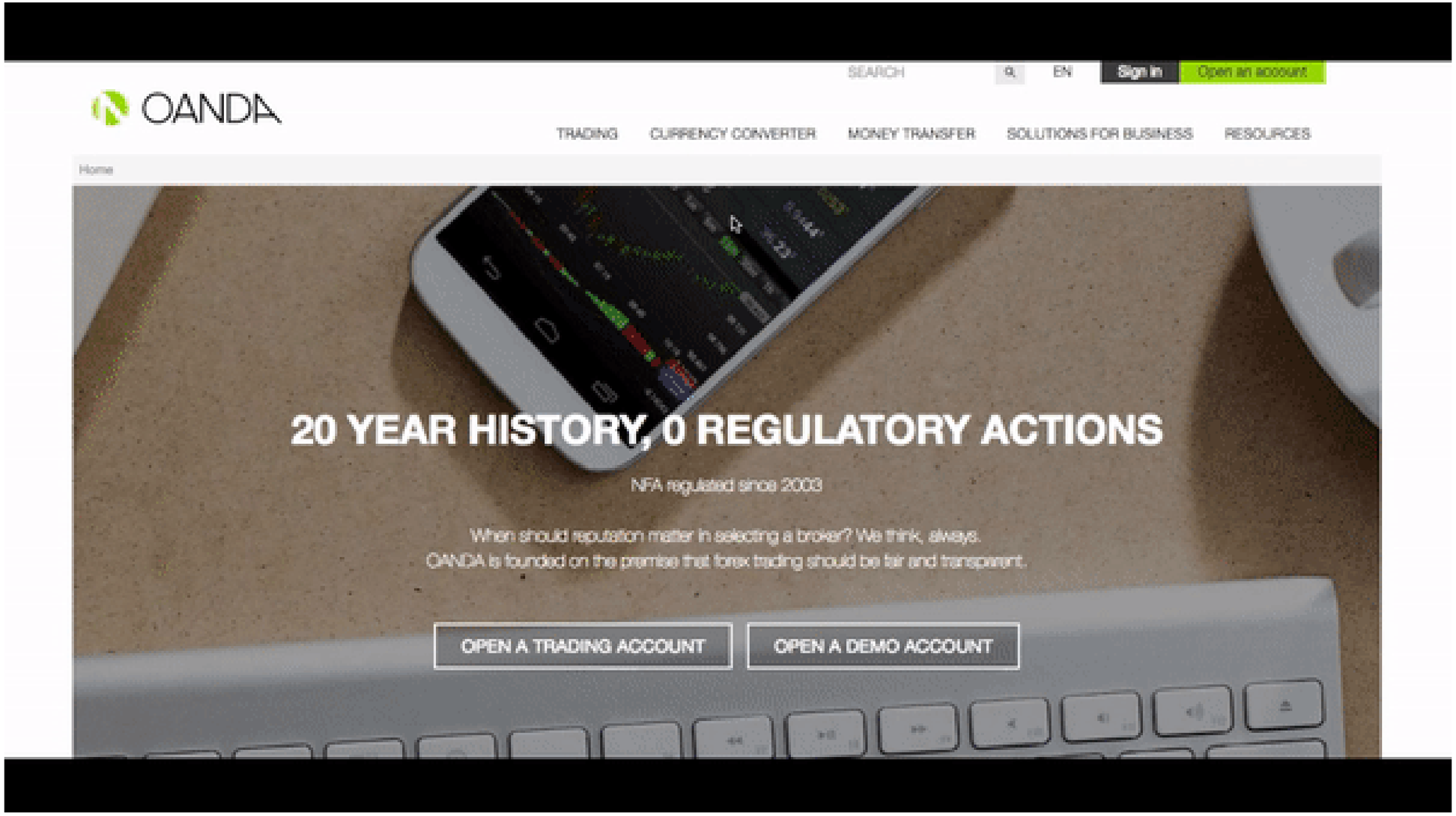
Yahoo Finance



Google Finance



Oanda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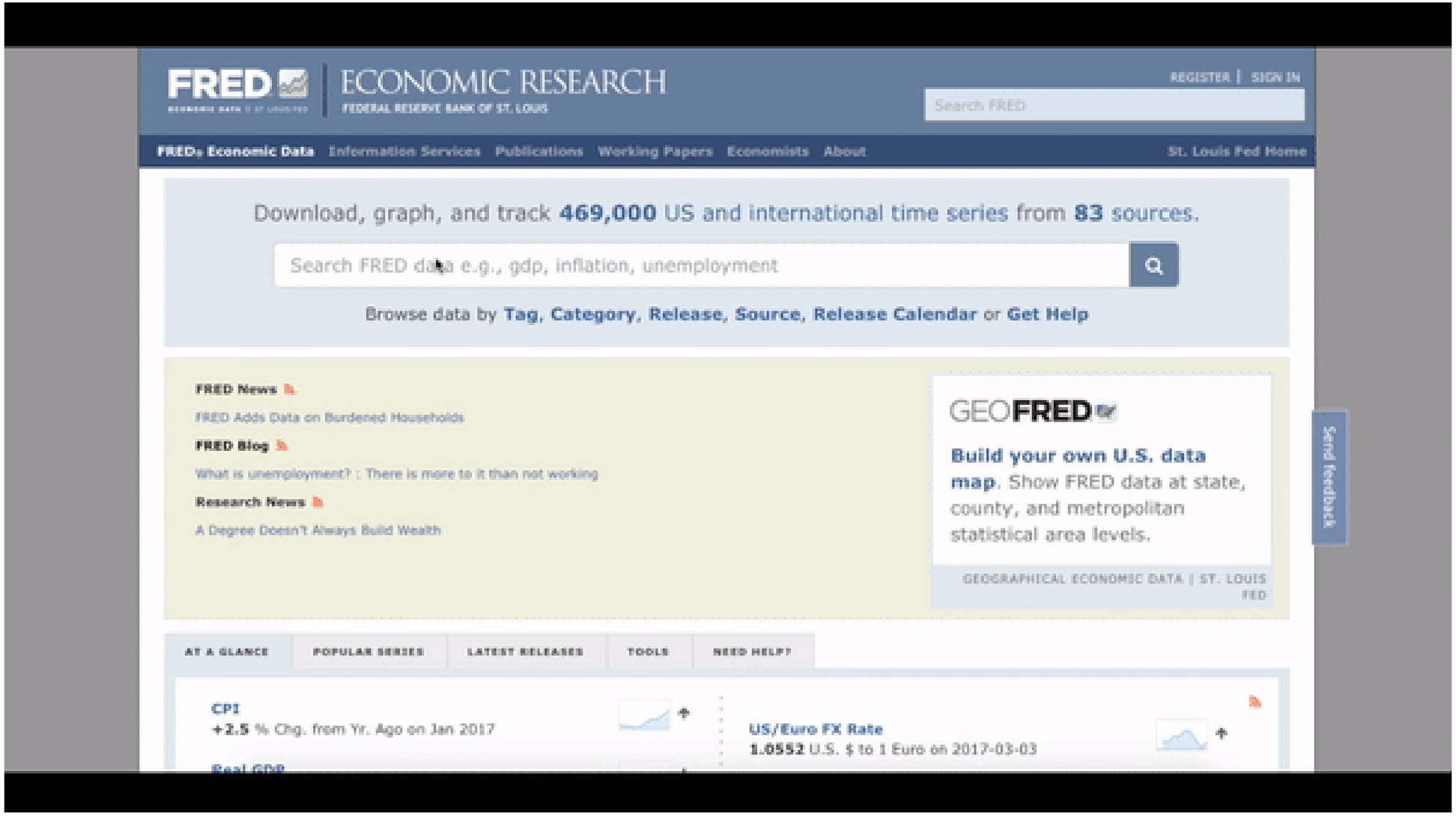
FRED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TOOLS
NEED HELP?
CPI
+2.5 % Chg. from Yr. Ago on Jan 2017
US/Euro FX Rate
1.0552 U.S. $ to 1 Euro on 2017-03-03

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Extracting columns from financial time series

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OHLC

- Stands for "Open High Low Close"
- *Open and Close*: first and last observed prices
- *High and Low*: largest and smallest observed prices
- Often *Volume*: sum of all contracts traded

OHLC data

```
head(DC)
```

```
                    DC.Open DC.High DC.Low DC.Close DC.Volume
2016-01-16 01:00:00  20.845  20.850 20.835   20.845       157
2016-01-16 02:00:00  20.845  20.850 20.835   20.845       214
2016-01-16 03:00:00  20.845  20.850 20.835   20.845       103
2016-01-16 04:00:00  20.845  20.855 20.835   20.845       180
2016-01-16 05:00:00  20.845  20.845 20.845   20.845       211
2016-01-16 06:00:00  20.845  20.845 20.840   20.845        35
```

Single-column extractor functions

- `Op()` - opening price
- `Hi()` - high price
- `Lo()` - low price
- `Cl()` - close price
- `Vo()` - traded volume
- `Ad()` - adjusted close price

Single-column extractor functions

```
# Open price
dc_open <- Op(DC)
head(dc_open, 4)
```

```
                    DC.Open
2016-01-16 01:00:00   20.84
2016-01-16 02:00:00   20.85
2016-01-16 03:00:00   20.85
2016-01-16 04:00:00   20.85
```

```
# High price
dc_high <- Hi(DC)
head(dc_high, 4)
```

```
                    DC.High
2016-01-16 01:00:00   20.85
2016-01-16 02:00:00   20.85
2016-01-16 03:00:00   20.85
2016-01-16 04:00:00   20.85
```

Multi-column extractor functions

```
# Extract multiple columns
dc_ohlc <- OHLC(DC)
head(dc_ohlc)
```

```
                    DC.Open DC.High DC.Low DC.Close
2016-01-16 01:00:00   20.84   20.85  20.83    20.84
2016-01-16 02:00:00   20.85   20.85  20.83    20.85
2016-01-16 03:00:00   20.85   20.85  20.84    20.85
2016-01-16 04:00:00   20.85   20.85  20.84    20.85
2016-01-16 05:00:00   20.85   20.85  20.84    20.85
2016-01-16 06:00:00   20.84   20.85  20.84    20.85
```

getPrice()

- 3 arguments
  - `x` : object that contains data
  - `symbol` : optional symbol if `x` contains multiple symbols
  - `prefer` : optional preferred price
- If `prefer` not specified:
  - `"price"` , then `"trade"` , then `"close"`

```
head(DC)
```

```
                    Price Volume Bid.Price Bid.Size Ask.Price Ask.Size
2016-01-16 00:00:07    NA     NA     20.84      198     20.85      684
2016-01-16 00:00:08    NA     NA     20.84      198     20.85      683
2016-01-16 00:00:08    NA     NA     20.84      198     20.85      682
2016-01-16 00:00:11    NA     NA     20.84      198     20.85      683
2016-01-16 00:00:25    NA     NA     20.84      198     20.85      684
2016-01-16 00:00:44 20.84      1     20.84      198     20.85      684
```

```
dc_bid <- getPrice(DC, prefer = "bid")
head(dc_bid)
```

```
                    Bid.Price
2016-01-16 00:00:07     20.84
2016-01-16 00:00:08     20.84
2016-01-16 00:00:08     20.84
2016-01-16 00:00:11     20.84
2016-01-16 00:00:25     20.84
2016-01-16 00:00:44     20.84
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Importing and transforming multiple instruments

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Aggregating with Quandl()

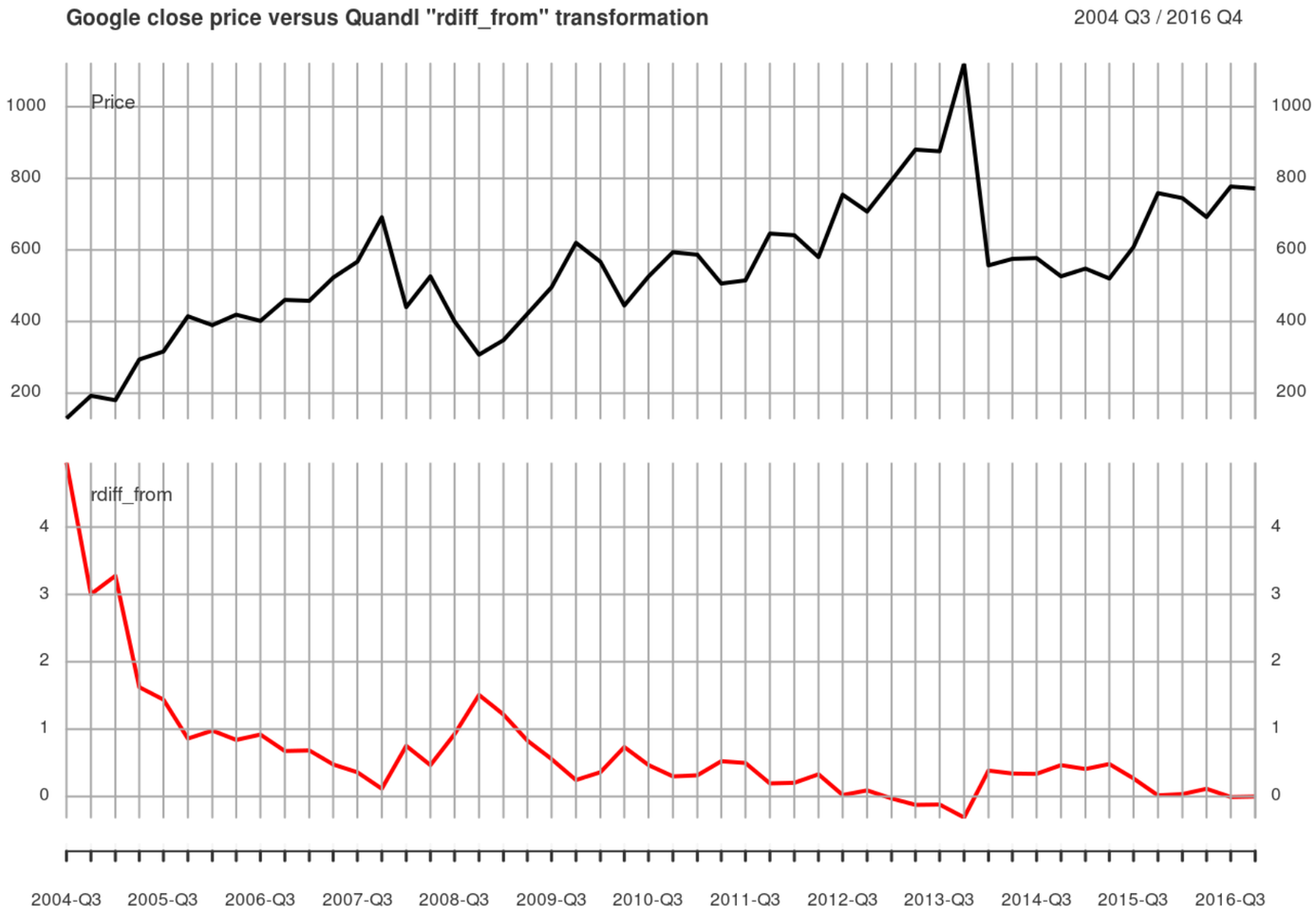
- Use `collapse` argument to aggregate
  - daily
  - weekly
  - monthly
  - quarterly
  - annual
- Always returns last observation for given time period
  - Can cause issues for some columns (e.g., opening price)

Transforming with Quandl()

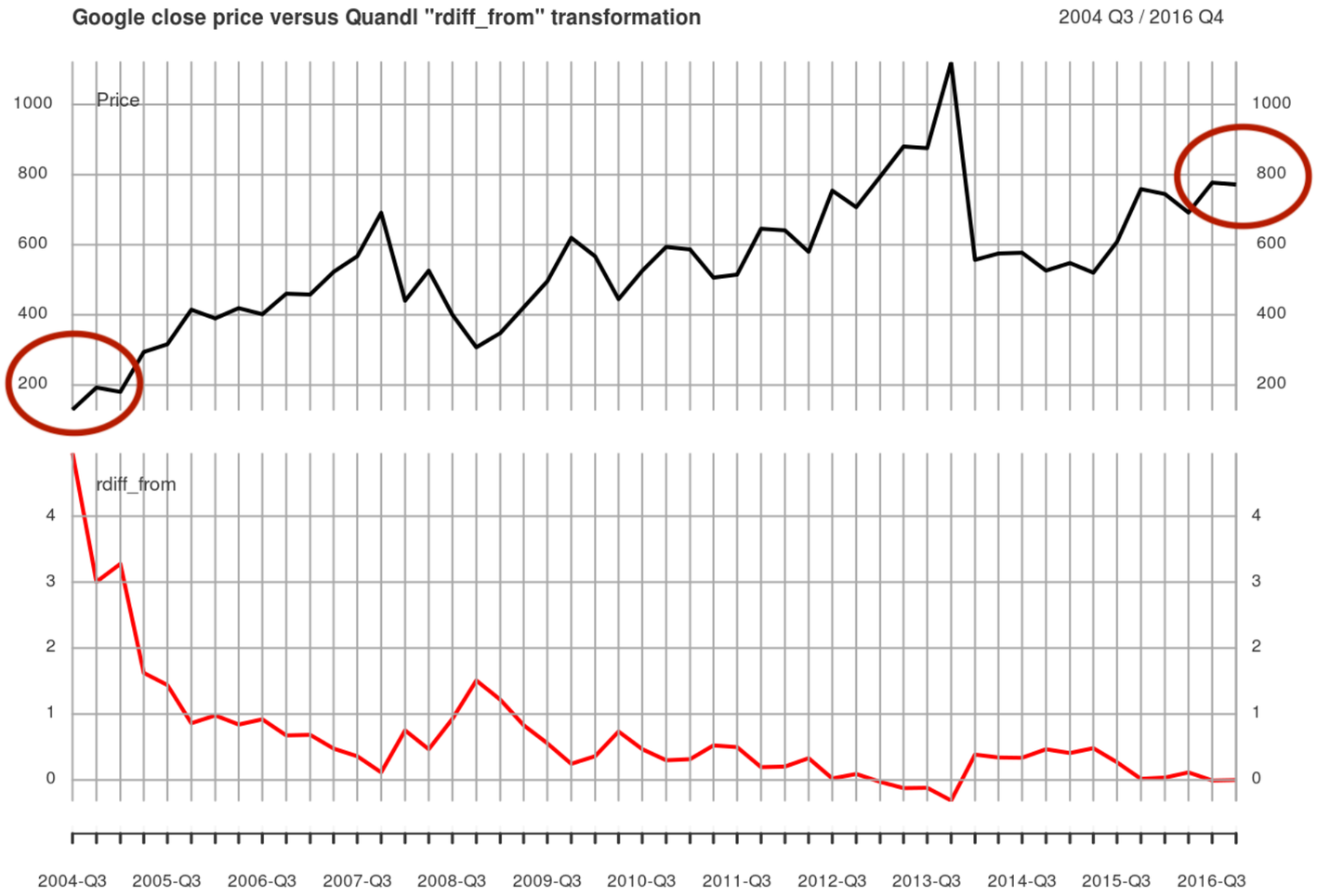
- Use `transform` argument to perform simple calculations before downloading

Name	Effect	Formula
none	no effect	$y'_t = y_t$
diff	row-on-row change	$y'_t = y_t - y_{t-1}$
rdiff	row-on-row % change	$y'_t = (y_t - y_{t-1})/y_{t-1}$
rdiff-from	latest value as % increment	$y'_t = (y_{latest} - y_t)/y_t$
cumul	cumulative sum	$y'_t = y_0 + y_1 + \cdots + y_t$
normalize	scale series to start at 100	$y'_t = y_t \div y_0 * 100$

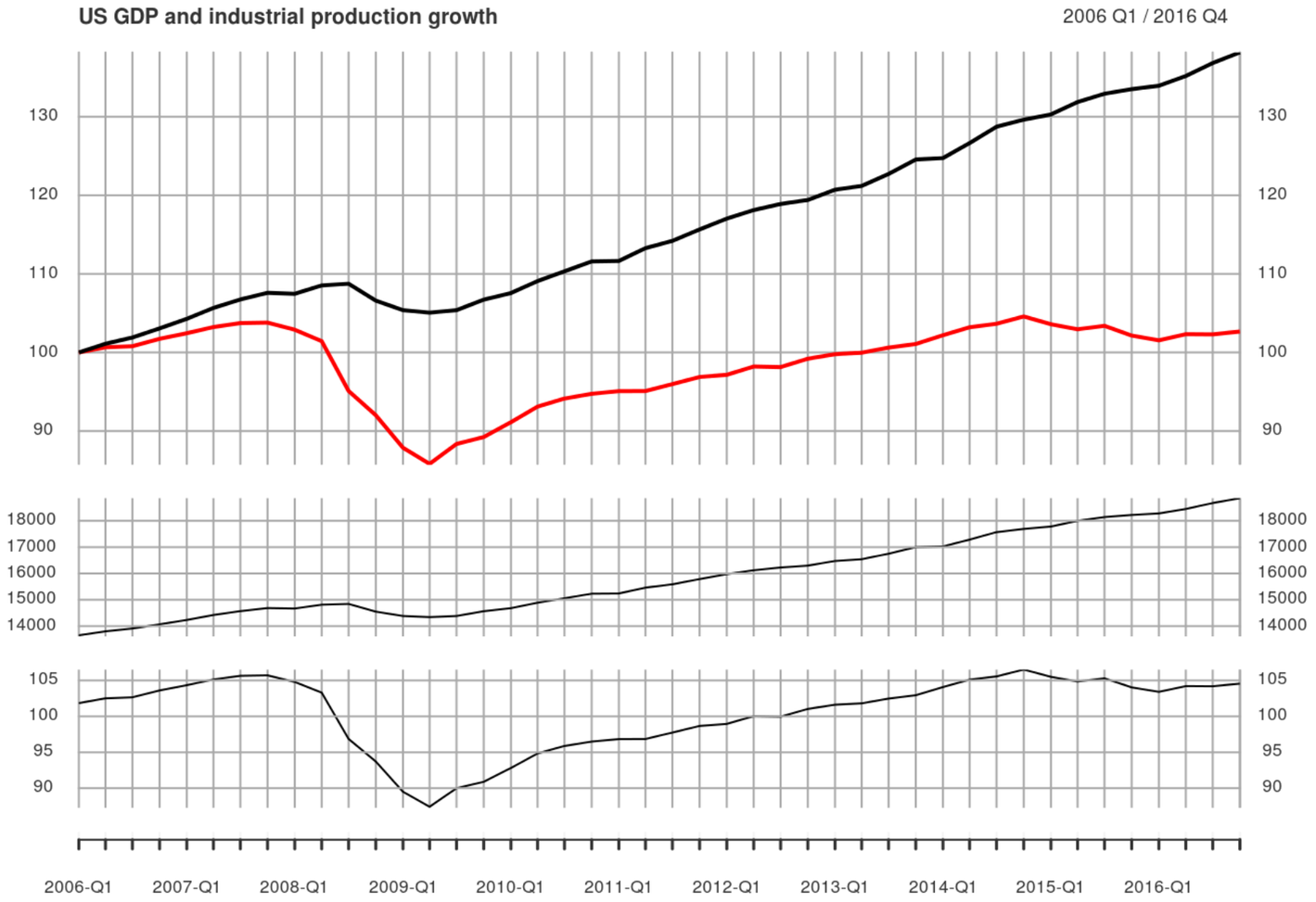
Quandl() rdiff_from transformation



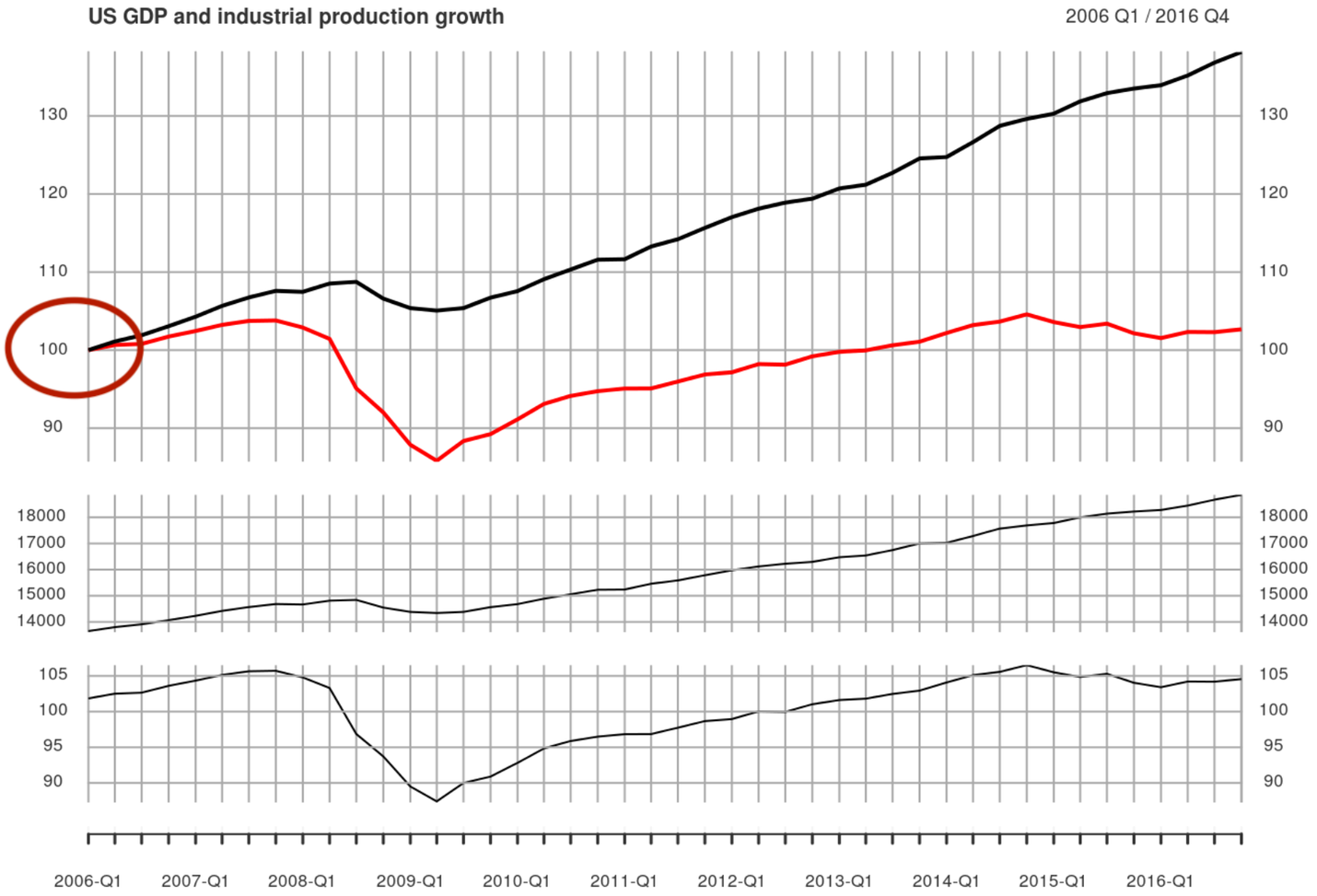
Quandl() rdiff_from transformation



Quandl() normalize transformation



Quandl() normalize transformation



Download instruments into a custom environment

```
# Create new environment
data_env <- new.env()
# Use getSymbols to load data into the environment
getSymbols(c("SPY", "QQQ"), env = data_env, auto.assign = TRUE)
```

```
"SPY" "QQQ"
```

```
# Look at a few rows of the SPY data
head(data_env$SPY, 3)
```

```
           SPY.Open SPY.High SPY.Low SPY.Close SPY.Volume SPY.Adjusted
2007-01-03   142.25   142.86  140.57    141.37   94807600     114.8094
2007-01-04   141.23   142.05  140.61    141.67   69620600     115.0530
2007-01-05   141.33   141.40  140.38    140.54   76645300     114.1353
```

Using lapply()

- Loops over all objects in environment
- Combine list of function results into one object using `do.call()`
  - First argument (`what`) is the function to be called
  - Second argument (`args`) is a list of arguments to pass

Extract volume and merge into one object

```
# Extract volume column from each object
adjusted_list <- lapply(data_env, Ad)
# Merge each list element into one object
adjusted <- do.call(merge, adjusted_list)
```

```
head(adjusted)
```

```
           QQQ.Adjusted SPY.Adjusted
2007-01-03     39.47694     114.8094
2007-01-04     40.22558     115.0530
2007-01-05     40.03385     114.1353
2007-01-08     40.06124     114.6632
2007-01-09     40.26210     114.5658
2007-01-10     40.73684     114.9475
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Setting default arguments for getSymbols()

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getSymbols() “methods”

- `getSymbols()` doesn't contain code to import data
- Code for each data source are in a `getSymbols.[source]` "method"
- For example:

```
# You call getSymbols()
getSymbols("GDP", src = "FRED")

# getSymbols() calls source "method"
getSymbols.FRED("GDP")
```

- Users should not call `getSymbols()` "methods" directly

Use setDefaults() to change default data source

```
setDefaults(getSymbols, src = "FRED")
```

```
gdp <- getSymbols("GDP", auto.assign = FALSE)
# Note the 'src' attribute
str(gdp)
```

```
An 'xts' object on 1947-01-01/2016-10-01 containing:
  Data: num [1:280, 1] 243 246 250 260 266 ...
 - attr(*, "dimnames")=List of 2
  ..$ : NULL
  ..$ : chr "GDP"
  Indexed by objects of class: [Date] TZ: UTC
  xts Attributes:
List of 2
 $ src    : chr "FRED"
 $ updated: POSIXct[1:1], format: "2017-02-13 08:46:50"
```

setDefaults()

- Sets new default arguments using `name = value` pairs
- Only alters behavior for `getSymbols()`
- Stores values in global `options()`

Other arguments

- Find formal arguments for a `getSymbols()` source method
  - Use `args()` : `args(getSymbols.yahoo)`
  - Use `help()` : `help("getSymbols.yahoo")`

Default from and to values

```
args(getSymbols.yahoo)
 function (Symbols, env, return.class = "xts", index.class = "Date",
     from = "2007-01-01", to = Sys.Date(), ...)
```

```
setDefaults(getSymbols.yahoo, from = "2016-01-01", to = "2016-12-31")?
aapl <- getSymbols("AAPL", auto.assign = FALSE)
str(aapl)
```

```
 An 'xts' object on 2016-01-04/2016-12-30 containing:
   Data: num [1:252, 1:6] 102.6 105.8 100.6 98.7 98.6 ...
  - attr(*, "dimnames")=List of 2
   ..$ : NULL
   ..$ : chr [1:6] "AAPL.Open" "AAPL.High" "AAPL.Low" "AAPL.Close" ...
   Indexed by objects of class: [Date] TZ: UTC
   xts Attributes:
 List of 2
  $ src    : chr "yahoo"
  $ updated: POSIXct[1:1], format: "2017-02-13 08:46:50"
```

getDefaults()

```
getDefaults()
```

```
"getSymbols.yahoo"
```

```
getDefaults(getSymbols.yahoo)
```

```
$from
"'2016-01-01'"

$to
"'2016-12-31'"
```

- Values returned do not imply those functions to accept user-specified defaults

```
setDefaults(load,
            file = "my_file.RData")

# Will not alter behavior
getDefaults(load)
```

```
$file
"'my_file.RData'"
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Setting per-instrument default arguments

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Use setSymbolLookup() to set data source

```
setSymbolLookup(AAPL = "google")
```

```
aapl <- getSymbols("AAPL", auto.assign = FALSE)
str(aapl)  # note the 'src' attribute
```

```
An ‘xts’ object on 2007-01-03/2017-02-22 containing:
  Data: num [1:2552, 1:5] 12.3 12 12.2 12.3 12.3 ...
 - attr(*, "dimnames")=List of 2
  ..$ : NULL
  ..$ : chr [1:5] "AAPL.Open" "AAPL.High" "AAPL.Low" "AAPL.Close" ...
  Indexed by objects of class: [Date] TZ: UTC
  xts Attributes:
List of 2
 $ src    : chr "google"
 $ updated: POSIXct[1:1], format: "2017-02-23 14:16:55"
```

Use setSymbolLookup() to set other arguments

```
setSymbolLookup(MSFT = list(src = "google", from = "2016-01-01"))
```

```
msft <- getSymbols("MSFT", auto.assign = FALSE)
str(msft)  # note the 'src' attribute and first date
```

```
An ‘xts’ object on 2016-01-04/2017-02-27 containing:
  Data: num [1:290, 1:5] 54.3 54.9 54.3 52.7 52.4 ...
 - attr(*, "dimnames")=List of 2
  ..$ : NULL
  ..$ : chr [1:5] "MSFT.Open" "MSFT.High" "MSFT.Low" "MSFT.Close" ...
  Indexed by objects of class: [Date] TZ: UTC
  xts Attributes:
List of 2
 $ src    : chr "google"
 $ updated: POSIXct[1:1], format: "2017-02-23 14:20:21"
```

Save and restore defaults (1)

```
# Set default
setSymbolLookup(AAPL = "google")
```

```
# Verify the default changed
getSymbolLookup()
```

```
$AAPL
$AAPL$src
"google"
```

```
# Save lookup
saveSymbolLookup("symbol_lookup.rda")
```

```
# Remove lookup
setSymbolLookup(AAPL = NULL)
```

Save and restore defaults (2)

```
# Verify the default is removed
getSymbolLookup()
```

```
named list()
```

```
# Load lookup
loadSymbolLookup("symbol_lookup.rda")
```

```
# Verify the default is restored
getSymbolLookup()
```

```
$AAPL
$AAPL$src
"google"
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Handling instrument symbols that clash or are not valid R names

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Syntactically valid names

- Valid names contain letters, numbers, `.` and `_`
- Must start with a letter, or a `.` followed by a non-number
- May not be one of the **reserved** words
- Not valid:
  - `.4times`, `_one`, `for`

Accessing objects with non-syntactic names (1)

- `getSymbols()` makes some names valid
  - S&P 500 Index: `"^GSPC"`

```
getSymbols("^GSPC")
```

```
"GSPC"
```

```
head(GSPC, 3)
```

```
           GSPC.Open GSPC.High GSPC.Low GSPC.Close GSPC.Volume GSPC.Adjusted
2007-01-03   1418.03   1429.42  1407.86    1416.60  3429160000       1416.60
2007-01-04   1416.60   1421.84  1408.43    1418.34  3004460000       1418.34
2007-01-05   1418.34   1418.34  1405.75    1409.71  2919400000       1409.71
```

Accessing objects with non-syntactic names (2)

- Some ticker symbols are not valid names
  - Shanghai Stock Exchange Composite Index: `"000001.SS"`

```
getSymbols("000001.SS", auto.assign = TRUE)
```

```
"000001.SS"
```

```
str(000001.SS)
```

```
Error: unexpected symbol in "str(000001.SS)"
```

```
head(`000001.SS`, n = 3)
```

```
           000001.SS.Open   000001.SS.High     000001.SS.Low
2007-01-04        2715.72          2715.72           2715.72
2007-01-05        2641.33          2641.33           2641.33
2007-01-08        2707.20          2707.20           2707.20
          000001.SS.Close 000001.SS.Volume 000001.SS.Adjusted
2007-01-04        2715.72                0            2715.72
2007-01-05        2641.33                0            2641.33
2007-01-08        2707.20                0            2707.20
```

```
head(get("000001.SS"), n = 3)
```

```
           000001.SS.Open   000001.SS.High     000001.SS.Low
2007-01-04        2715.72          2715.72           2715.72
2007-01-05        2641.33          2641.33           2641.33
2007-01-08        2707.20          2707.20           2707.20
          000001.SS.Close 000001.SS.Volume 000001.SS.Adjusted
2007-01-04        2715.72                0            2715.72
2007-01-05        2641.33                0            2641.33
2007-01-08        2707.20                0            2707.20
```

Valid name for one instrument

- Assign `getSymbols()` output to valid name
- Convert column names to valid names

```
sse <- getSymbols("000001.SS", auto.assign = FALSE)
```

```
colnames(sse) <- paste("SSE",
                       c("Open", "High", "Low", "Close",
                         "Volume", "Adjusted"), sep = ".")
head(sse, n = 2)
```

```
           SSE.Open    SSE.High     SSE.Low
2007-01-04   2715.72     2715.72     2715.72
2007-01-05   2641.33     2641.33     2641.33
           SSE.Close  SSE.Volume  SSE.Adjusted
2007-01-04   2715.72           0       2715.72
2007-01-05   2641.33           0       2641.33
```

- Create symbol-to-R object mapping with `setSymbolLookup()`

```
setSymbolLookup(SSE = list(name = "000001.SS"),
                FORD = list(name = "F"))
getSymbols(c("SSE", "FORD"))
```

```
"SSE"  "FORD"
```

```
head(SSE, n = 2)
```

```
            SSE.Open  SSE.High  SSE.Low  SSE.Close  SSE.Volume  SSE.Adjusted
2007-01-04   2715.72   2715.72  2715.72    2715.72           0       2715.72
2007-01-05   2641.33   2641.33  2641.33    2641.33           0       2641.33
```

```
head(FORD, n = 2)
```

```
           FORD.Open  FORD.High FORD.Low  FORD.Close  FORD.Volume  FORD.Adjusted
2007-01-03      7.56       7.67      7.44        7.51     78652200       6.150263
2007-01-04      7.56       7.72      7.43        7.70     63454900       6.305862
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Making irregular data regular

IMPORTING AND MANAGING FINANCIAL DATA IN R

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Instructor

datacamp

Regular date-time sequences

- Time observations are same distance apart
- Create regular date-time sequences using `seq()` methods:
  - `seq.Date()`
  - `seq.POSIXt()` (`POSIXct` and `POSIXlt`)

```
from_date <- as.Date("2017-01-01")
to_date <- as.Date("2017-01-03")
date_seq <- seq(from = from_date, to = to_date, by = "day")
```

- `start()` first index value
- `end()` last index value

```
regular_xts <- xts(seq_along(date_seq), order.by = date_seq)
start(regular_xts)
```

```
"2017-01-01"
```

```
end(regular_xts)
```

```
"2017-01-03"
```

```
seq(from = start(regular_xts), to = end(regular_xts), by = "day")
```

```
"2017-01-01" "2017-01-02" "2017-01-03"
```

Zero-width xts objects

- `xts` object with an index, no data

```
zero_width_xts <- xts(, order.by = date_seq)
zero_width_xts
```

```
Data:
numeric(0)
Index:
 Date[1:3], format: "2017-01-01" "2017-01-02" "2017-01-03"
```

```
str(zero_width_xts)
```

```
An 'xts' object of zero-width
```

Creating regular from irregular data

- Add observation at each date-time in regular sequence
- `NA` in the result

Merge irregular xts with regular zero-width xts

```
irregular
```

```
           Price
2017-01-02 20.01
2017-01-04 20.02
2017-01-10 20.05
```

```
date_seq <- seq(from = start(irregular),
                to = end(irregular),
                by = "day")
```

```
regular_xts <- xts(, date_seq)
```

Merge irregular xts with regular zero-width xts

```
merge(irregular, regular_xts)
```

```
           Price
2017-01-02 20.01
2017-01-03    NA
2017-01-04 20.02
2017-01-05    NA
2017-01-06    NA
2017-01-07    NA
2017-01-08    NA
2017-01-09    NA
2017-01-10 20.05
```

Filling missing values

```
merged_xts <- merge(irregular, regular_xts)
na.locf(merged_xts)
```

```
           Price
2017-01-02 20.01
2017-01-03 20.01
2017-01-04 20.02
2017-01-05 20.02
2017-01-06 20.02
2017-01-07 20.02
2017-01-08 20.02
2017-01-09 20.02
2017-01-10 20.05
```

Filling missing values

```
merge(irregular, regular_xts, fill = na.locf)
```

```
           Price
2017-01-02 20.01
2017-01-03 20.01
2017-01-04 20.02
2017-01-05 20.02
2017-01-06 20.02
2017-01-07 20.02
2017-01-08 20.02
2017-01-09 20.02
2017-01-10 20.05
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Aggregating to lower frequency

IMPORTING AND MANAGING FINANCIAL DATA IN R

Joshua Ulrich
Instructor

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Low frequency data

- Timestamps have too much resolution
- Represent the first quarter of 2017
  - `"2017-01-01"` (first)
  - `"2017-03-31"` (last)
  - `"2017-02-01"` (middle)

Example

- Compare the daily 10-year Treasury constant maturity rate with USA Gross Domestic Product (quarterly)
- FRED symbols:
  - `DGS10`
  - `GDP`

Merge aggregated data with low-frequency data

```
# Aggregate to quarterly
QGS10 <- apply.quarterly(DGS10, median, na.rm = TRUE)
# Merge quarterly aggregate with quarterly GDP
QGS10_GDP <- merge(QGS10, GDP)
QGS10_GDP
```

```
           DGS10     GDP
2015-01-01    NA 17783.6
2015-03-31  1.97      NA
2015-04-01    NA 17998.3
2015-06-30  2.19      NA
2015-07-01    NA 18141.9
2015-09-30  2.20      NA
2015-10-01    NA 18222.8
2015-12-31  2.23      NA
```

Low frequency date-time classes

- `yearmon()` for monthly data
- `yearqtr()` for quarterly data

```
as.Date("2017-01-01")
```

```
"2017-01-01"
```

```
as.yearmon("2017-01-01")
```

```
"Jan 2017"
```

```
as.yearqtr("2017-01-01")
```

```
"2017 Q1"
```

Convert index to lowest frequency

```
# Convert both indexes to yearqtr
index(QGS10) <- as.yearqtr(index(QGS10))
index(GDP) <- as.yearqtr(index(GDP))
```

```
# Merging 'just works'
merge(QGS10, GDP)
```

```
        DGS10     GDP
2015 Q1  1.97 17783.6
2015 Q2  2.19 17998.3
2015 Q3  2.20 18141.9
2015 Q4  2.23 18222.8
```

Align with beginning-of-period timestamp

```
# Last observation carried backward
QGS10_GDP_locb <- na.locf(QGS10_GDP, fromLast = TRUE)
```

```
# Subset by beginning-of-period index
QGS10_GDP_first_period <- QGS10_GDP_locb[index(GDP)]
QGS10_GDP_first_period
```

```
           DGS10     GDP
2015-01-01  1.97 17783.6
2015-04-01  2.19 17998.3
2015-07-01  2.20 18141.9
2015-10-01  2.23 18222.8
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Aggregating and combining intraday data

IMPORTING AND MANAGING FINANCIAL DATA IN R

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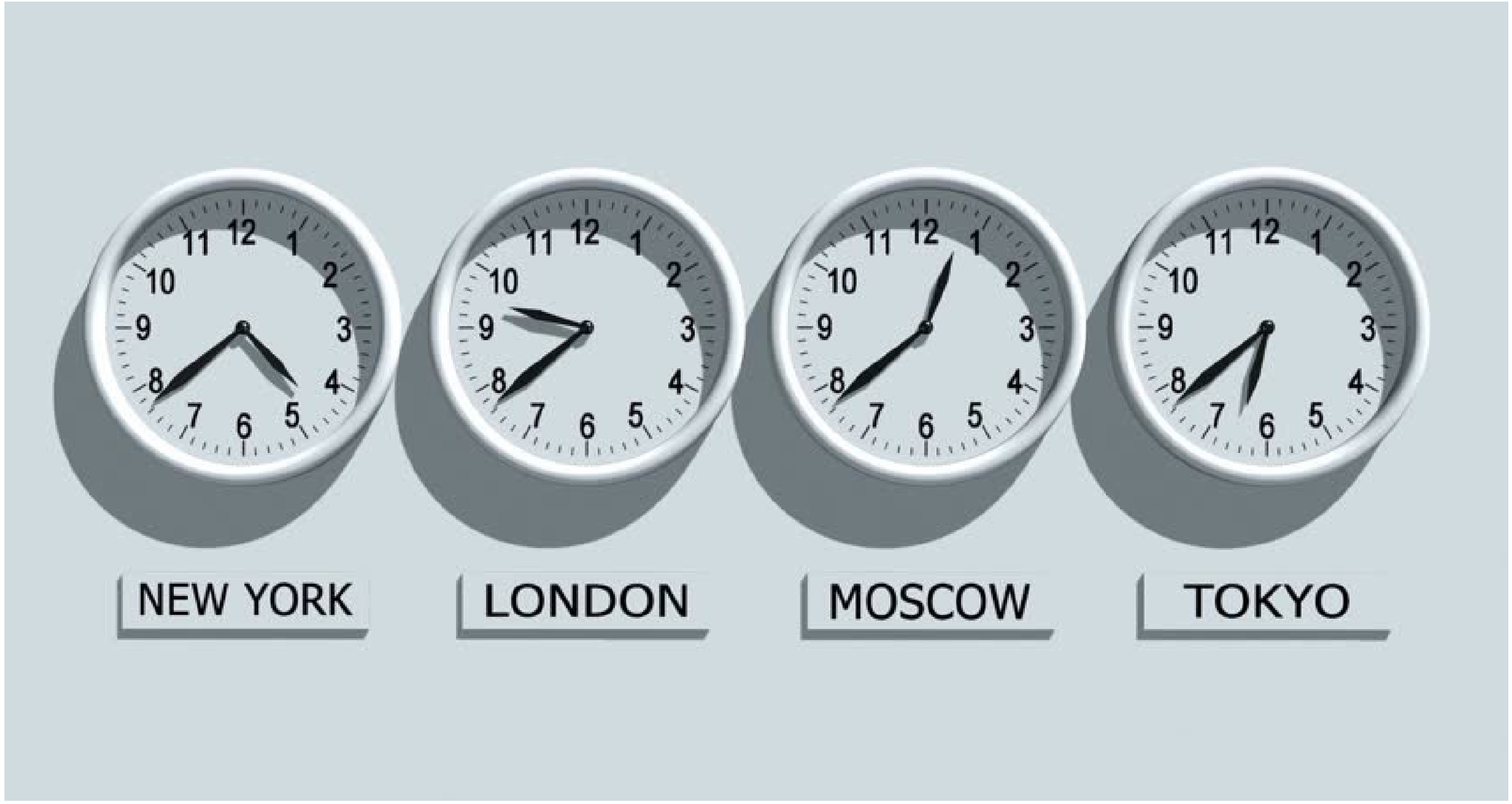
Timezones!



[1] Source: https://www.shutterstock.com/video/search/time-zone-clocks

Timezones!

- Internally, `xts` index is seconds since midnight `1970-01-01` in UTC
- `merge()` uses internal index
- `merge()` result will have timezone of the first object

Timezones!

```
datetime <- as.POSIXct("2017-01-18 10:00:00", tz = "UTC")
london <- xts(1, datetime, tzone = "Europe/London")
tokyo <- xts(1, datetime, tzone = "Asia/Tokyo")
```

```
merge(london, tokyo)
```

```
                    london tokyo
2017-01-18 10:00:00      1     1
```

```
merge(tokyo, london)
```

```
                    tokyo london
2017-01-18 19:00:00     1      1
```

Creating regular intraday data

```
head(dc_trades)
```

```
                     Price
2016-01-16 08:00:58 20.85
2016-01-16 08:01:56 20.85
2016-01-16 08:03:35 20.85
2016-01-16 08:07:44 20.84
2016-01-16 08:45:58 20.85
2016-01-16 08:46:49 20.85
```

Creating regular intraday data

```
datetimes <- seq(from = as.POSIXct("2016-01-16 08:00"),
                 to = as.POSIXct("2016-01-17 18:00"),
                 by = "1 min")
```

```
regular_xts <- xts(, order.by = datetimes)
```

```
merged_xts <- merge(dc_trades, regular_xts)
head(merged_xts)
```

```
                     Price
2016-01-16 08:00:00    NA
2016-01-16 08:00:58 20.85
2016-01-16 08:01:00    NA
2016-01-16 08:01:56 20.85
2016-01-16 08:02:00    NA
2016-01-16 08:03:00    NA
```

Subset to trading hours

```
# All observations should be NA
all(is.na(merged_xts["2016-01-16 19:00/2016-01-17 07:00"]))
```

```
TRUE
```

```
# xts time-of-day subsetting
merged_trade_day <- merged_xts["T08:00/T18:00"]
```

```
# Now there are no observations
nrow(merged_trade_day["2016-01-16 19:00/2016-01-17 07:00"])
```

```
0
```

Fill missing values by trading day

- `split()` - `lapply()` - `rbind()` paradigm from **this DataCamp course about manipulating time series**

```
# split() data into list of non-overlapping chunks
trade_day_list <- split(merged_trade_day, "days")
```

```
# lapply() a function to each chunk (list element)
filled_trade_day_list <- lapply(trade_day_list, na.locf)
```

```
# Combine list of chunks using do.call() and rbind()
filled_trade_day <- do.call(rbind, filled_trade_day_list)
```

Aggregate irregular intraday data

- Aggregate dense intraday data with `to.period()`
  - `period` : new periodicity (e.g. seconds, hours, days, etc)
  - `k` : number of periods per new observation

Aggregate irregular intraday data (1)

```
head(dc_price)
```

```
                    DC.Price
2016-01-16 00:00:07 20.84224
2016-01-16 00:00:08 20.84225
2016-01-16 00:00:08 20.84225
2016-01-16 00:00:11 20.84225
2016-01-16 00:00:25 20.84224
2016-01-16 00:00:44 20.84224
```

```
xts_5min <- to.period(dc_price, period = "minutes", k = 5)
head(xts_5min, n = 4)
```

```
                    dc_price.Open  dc_price.High  dc_price.Low  dc_price.Close
2016-01-16 00:03:49       20.84224       20.84227      20.84140        20.84160
2016-01-16 00:09:50       20.84160       20.84160      20.84156        20.84156
2016-01-16 00:14:57       20.84156       20.84156      20.84154        20.84154
2016-01-16 00:19:23       20.84154       20.84154      20.83206        20.83211
```

```
xts_aligned <- align.time(xts_5min, n = 60 * 5)
head(xts_aligned, n = 4)
```

```
                    dc_price.Open  dc_price.High  dc_price.Low  dc_price.Close
2016-01-16 00:05:00       20.84224       20.84227      20.84140        20.84160
2016-01-16 00:05:00       20.84160       20.84160      20.84156        20.84156
2016-01-16 00:15:00       20.84156       20.84156      20.84154        20.84154
2016-01-16 00:20:00       20.84154       20.84154      20.83206        20.83211
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Importing text files

IMPORTING AND MANAGING FINANCIAL DATA IN R

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Instructor

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getSymbols() with CSV files

- Well-formatted
  - One instrument per file
  - Columns: date, open, high, low, close, volume, adjusted close
- Files named `"[symbol].csv"`
- Can use `dir` argument to specify directory

getSymbols() with CSV files

- AMZN.csv

```
"Date","AMZN.Open","AMZN.High","AMZN.Low","AMZN.Close","AMZN.Volume","AMZN.Adjusted"
2002-01-02,11.13,11.01,10.46,10.87,6674703,10.87
2002-01-03,11.26,12.25,10.76,11.99,11441553,11.99
2002-01-04,12.46,12.62,11.71,12.1,12619402,12.1
```

```
getSymbols("AMZN", src = "csv")
```

```
"AMZN"
```

```
head(AMZN, 3)
```

```
           AMZN.Open AMZN.High AMZN.Low AMZN.Close AMZN.Volume AMZN.Adjusted
2002-01-02     11.13     11.01    10.46      10.87     6674703         10.87
2002-01-03     11.26     12.25    10.76      11.99    11441553         11.99
2002-01-04     12.46     12.62    11.71      12.10    12619402         12.10
```

read.zoo()

- AMZN.csv

```
"Date","AMZN.Open","AMZN.High","AMZN.Low","AMZN.Close","AMZN.Volume","AMZN.Adjusted"
2002-01-02,11.13,11.01,10.46,10.87,6674703,10.87
2002-01-03,11.26,12.25,10.76,11.99,11441553,11.99
2002-01-04,12.46,12.62,11.71,12.1,12619402,12.1
```

```
amzn_zoo <- read.zoo("AMZN.csv", sep = ",", header = TRUE)
amzn_xts <- as.xts(amzn_zoo)
head(amzn_xts, n = 3)
```

```
           AMZN.Open AMZN.High AMZN.Low AMZN.Close AMZN.Volume AMZN.Adjusted
2002-01-02     11.13     11.01    10.46      10.87     6674703         10.87
2002-01-03     11.26     12.25    10.76      11.99    11441553         11.99
2002-01-04     12.46     12.62    11.71      12.10    12619402         12.10
```

Date and time in separate columns

- FOO.csv

```
"Date","Time","Open","High","Low","Close"
2016-11-08,09:05:00,80.9,81,80.87,81
2016-11-08,09:10:00,80.92,80.93,80.89,80.89
2016-11-08,09:15:00,80.93,80.94,80.92,80.93
```

```
foo_zoo <- read.zoo("FOO.csv", sep = ",", header = TRUE,
                    index.column = c("Date", "Time"))
head(foo_zoo,  n = 3)
```

```
                     Open  High   Low Close
2016-11-08 09:05:00 80.90 81.00 80.87 81.00
2016-11-08 09:10:00 80.92 80.93 80.89 80.89
2016-11-08 09:15:00 80.93 80.94 80.92 80.93
```

File contains multiple instruments

- BAR.csv

```
Date,Symbol,Type,Price
2016-01-01 10:43:01,A,Bid,58.23
2016-01-01 10:43:01,A,Ask,58.24
2016-01-01 10:43:01,B,Bid,28.96
2016-01-01 10:43:01,B,Ask,28.98
```

```
bar_zoo <- read.zoo("BAR.csv",
                    split = c("Symbol", "Type"),
                    sep = ",", header = TRUE)
bar_zoo
```

```
                    A.Ask B.Ask A.Bid B.Bid
2016-01-01 10:43:01 58.24 28.98 58.23 28.96
2016-01-01 10:43:02 58.25 28.99 58.24 28.97
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Checking for weirdness

IMPORTING AND MANAGING FINANCIAL DATA IN R

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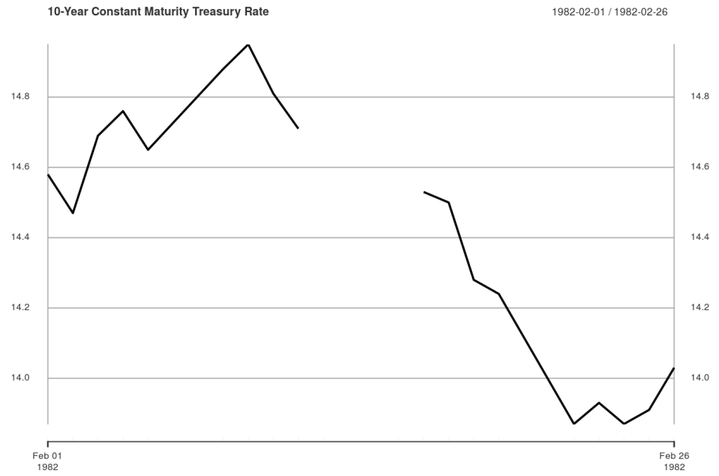
Visualize Data

```
getSymbols("DGS10", src = "FRED")
```

```
"DGS10"
```

```
treasury_10 <- DGS10["1982-02"]
plot(treasury_10, main = "10-Year Constant Maturity Treasury Rate")
```



Handle missing values

```
# Fill NA using last observation carried forward
locf <- na.locf(treasury_10)
```

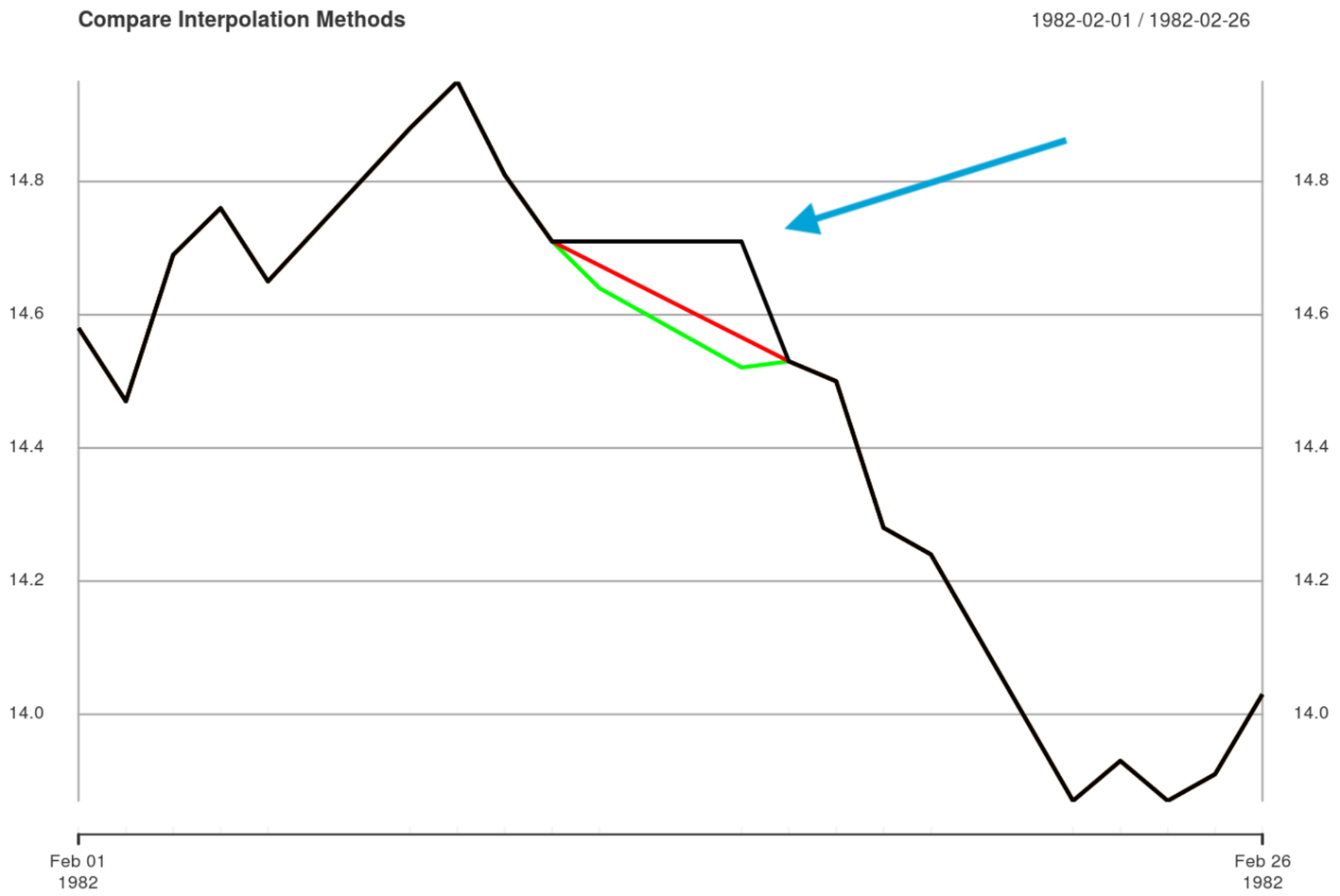
```
# Fill NA using linear interpolation
approx <- na.approx(treasury_10)
```

```
# Fill NA using spline interpolation
spline <- na.spline(treasury_10)
```

```
# Merge into one object
na_filled <- merge(locf, approx, spline)
# Plot combined object
plot(na_filled, col = c("black", "red", "green"),
     main = "Compare Interpolation Methods")
```

Handle missing values

Handle missing values

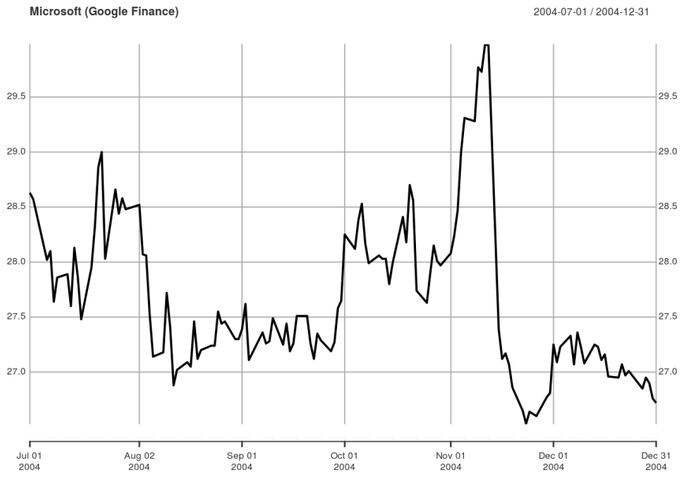


Visualize data

```
getSymbols("MSFT", from = "2004-07-01", to = "2004-12-31", src = "google")
```

```
"MSFT"
```

```
plot(Cl(MSFT), main = "Microsoft (Google Finance)")
```

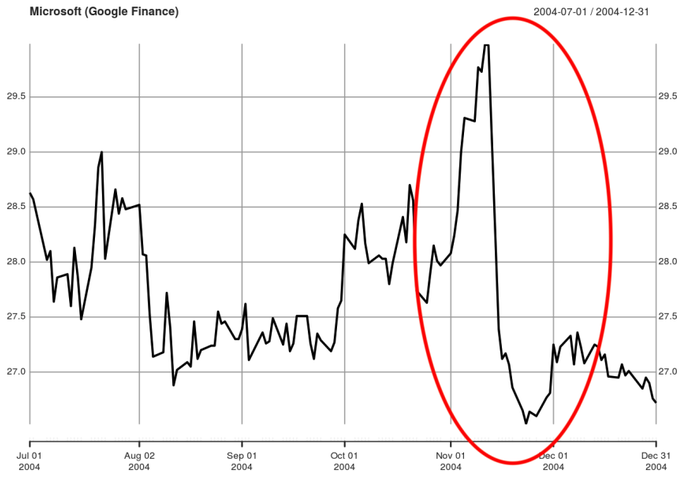


Visualize data

```
getSymbols("MSFT", from = "2004-07-01", to = "2004-12-31", src = "google")
```

```
"MSFT"
```

```
plot(Cl(MSFT), main = "Microsoft (Google Finance)")
```

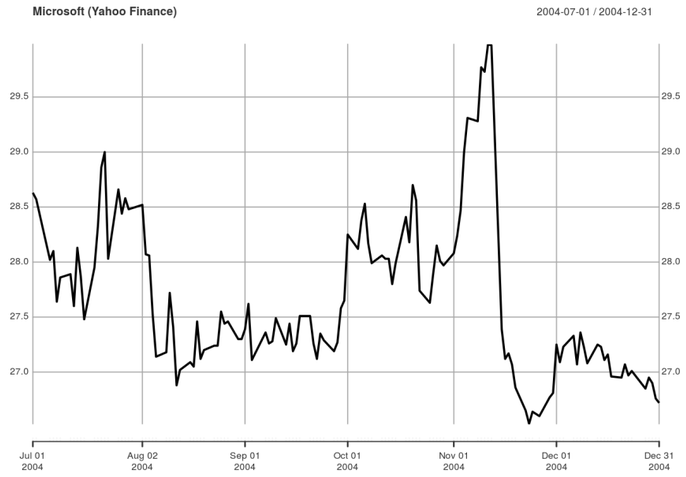


Cross-reference sources

```
getSymbols("MSFT", from = "2004-07-01", to = "2004-12-31")
```

```
"MSFT"
```

```
plot(Cl(MSFT), main = "Microsoft (Yahoo Finance)")
```

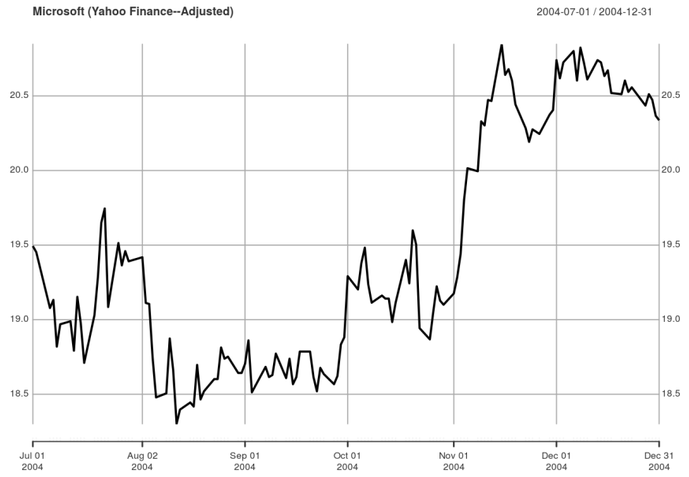


Cross-reference sources

```
getSymbols("MSFT", from = "2004-07-01", to = "2004-12-31")
```

```
"MSFT"
```

```
plot(Ad(MSFT), main = "Microsoft (Yahoo Finance-Adjusted)")
```

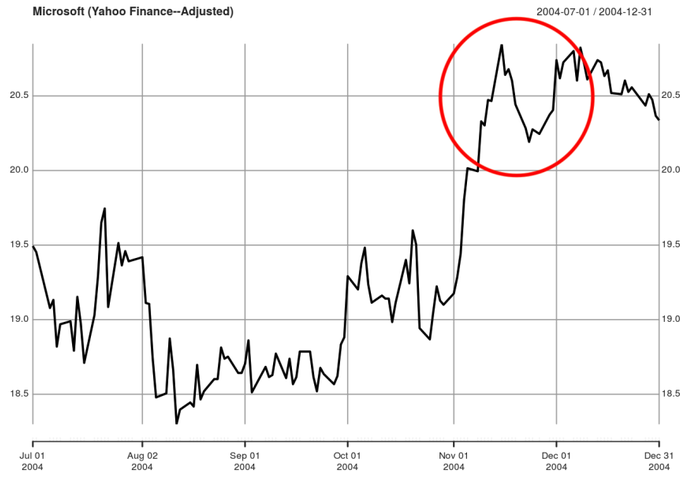


Cross-reference sources

```
getSymbols("MSFT", from = "2004-07-01", to = "2004-12-31")
```

```
"MSFT"
```

```
plot(Ad(MSFT), main = "Microsoft (Yahoo Finance—Adjusted)")
```



Stock split example

- MSFT stock splits 2-for-1

	Pre-split	Post-split
Shares	100	200
Price	$50	$25
Value	$5,000	$5,000

Stock dividend example

- MSFT issues a $3 per share dividend

	Pre-dividend	Post-dividend
Cash	$0	$300
Shares	100	100
Price	$50	$47
Value	$5,000	$5,000

Data source differences

- Yahoo Finance:
  - Raw OHLC prices
  - Split- and dividend-adjusted close
- Google Finance:
  - Split-adjusted OHLC prices

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Adjusting for corporate actions

IMPORTING AND MANAGING FINANCIAL DATA IN R

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Instructor

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Adjust for stock splits and dividends (1)

```
getSymbols("MSFT", from = "2004-07-01", to = "2004-12-31")
```

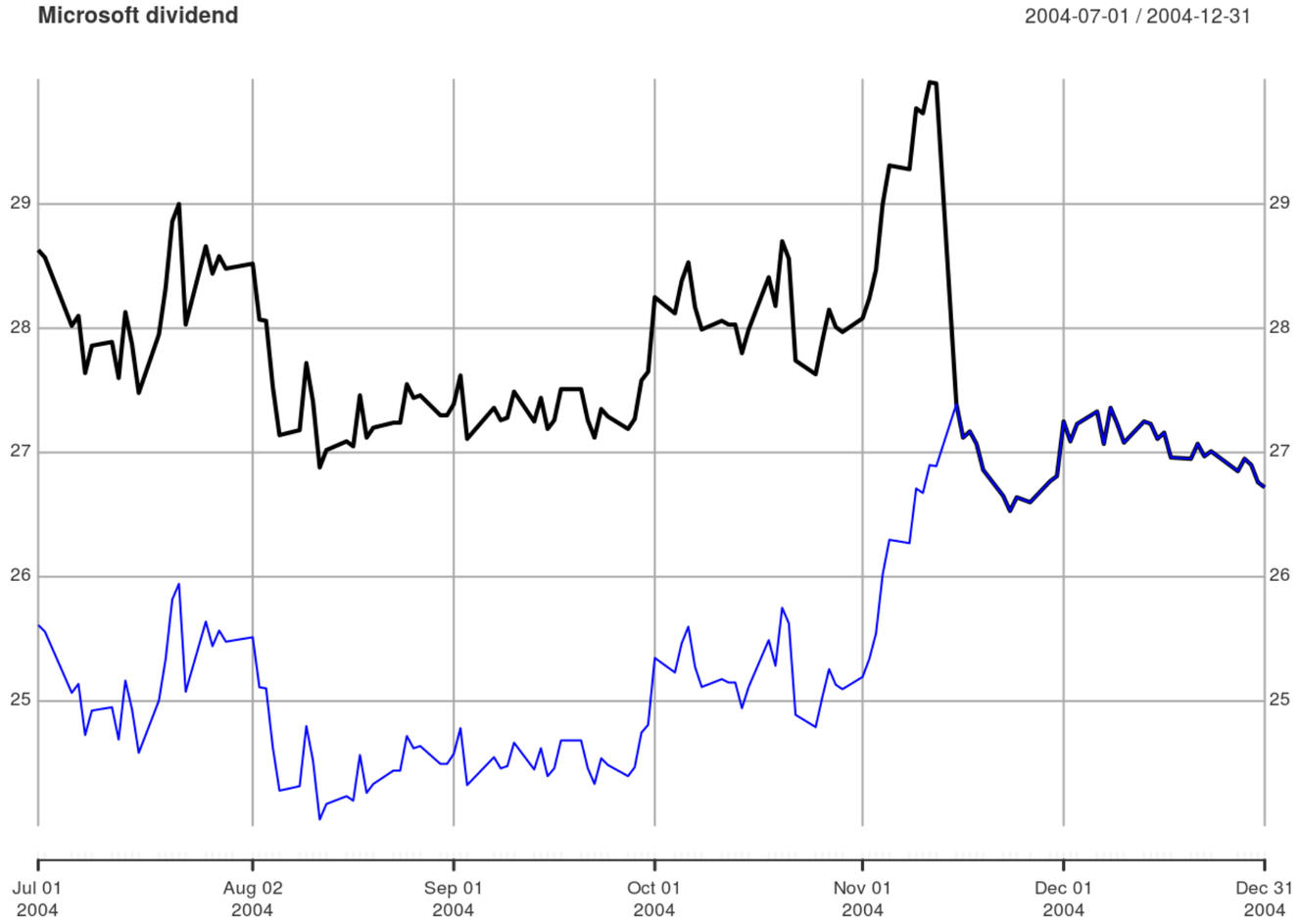
```
"MSFT"
```

```
# Adjust data for splits and dividends
msft_adjusted <- adjustOHLC(MSFT)
```

```
# Object name is not ticker symbol
my_data <- MSFT

# Use symbol.name argument
my_data_adjusted <- adjustOHLC(my_data, symbol.name = "MSFT")
```

Adjust for stock splits and dividends (2)



```r
# Download split data from Yahoo Finance
splits <- getSplits("GE")
head(splits, n = 4)
```

```
           GE.spl
1971-06-08    0.5
1983-06-02    0.5
1987-05-26    0.5
1994-05-16    0.5
```

```r
# Download split-adjusted dividend data from Yahoo Finance
dividends <- getDividends("GE")
head(dividends, n = 4)
```

```
            GE.div?1970-03-03 0.00677
1970-06-11 0.00677
1970-09-21 0.00677
1970-12-07 0.00677
```

Download unadjusted dividends

```
# Download unadjusted dividend data from Yahoo Finance
dividends_raw <- getDividends("GE", split.adjust = FALSE)

# Compare adjusted and unadjusted dividends
head(merge(dividends, dividends_raw))
```

```
            GE.div GE.div.1
1970-03-03 0.00677  0.64992
1970-06-11 0.00677  0.64992
1970-09-21 0.00677  0.64992
1970-12-07 0.00677  0.64992
1971-03-03 0.00677  0.64992
1971-06-17 0.00729  0.34992
```

adjRatios()

- Back-adjust any series for splits, dividends, or both
- Has 3 arguments:
  - `splits`
  - `dividends`
  - `close`
- Returns `xts` object with 2 columns: `Split` and Div

Adjust univariate series for splits and dividends

```
getSymbols("GE", from = "2000-01-01")
```

```
"GE"
```

```
close <- Cl(GE)
splits <- getSplits("GE")
dividends_raw <- getDividends("GE", split.adjust = FALSE)
```

```
# Pass splits, unadjusted dividends, and unadjusted close
ratios <- adjRatios(splits = splits,
                    dividends = dividends_raw,
                    close = close)
```

Adjust univariate series for splits and dividends

```
# Multiply unadjusted close by split and dividend ratios
close_adjusted <- close * ratios[, "Split"] * ratios[, "Div"]

head(merge(close, close_adjusted, Ad(GE)), n = 4)
```

```
           GE.Close GE.Close.1 GE.Adjusted
2000-01-03 150.0000   29.50422    29.44630
2000-01-04 144.0000   28.32405    28.26845
2000-01-05 143.7500   28.27488    28.21937
2000-01-06 145.6718   28.65289    28.59664
```

Let's practice!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Congratulations!

IMPORTING AND MANAGING FINANCIAL DATA IN R

Joshua Ulrich
Instructor

Congratulations!

IMPORTING AND MANAGING FINANCIAL DATA IN R