

What is Tidy Finance?

A transparent, open-source approach to research in financial economics, featuring multiple programming languages

tidy-finance.org offers tools to:

- Learn about empirical applications using tidy principles
- Learn to work with financial data in a tidy manner
- Teach students the importance of reproducible research
- Contribute to reproducible finance research via our blog

Why Tidy?

Code should not just be correct, but also follow principles:

- 1. Design so that code is easy to read for humans
- 2. Compose simple functions to solve complex problems
- 3. Embrace functional programming for reproducible results
- 4. Reuse data structures across applications

Focus of this talk: tidy data

Recap: what is tidy data?

our columns are VARIABLES & our rows are The standard structure of OBSERVATIONS! tidy data means that "tidy datasets are all alike..." variables in a single column. don't even "...but every messy dataset is 4/1. MONY messy in its own way." -HADLEY WICKHAM any column's my values and have variables in Columns AND IN rows

llustrations from the Openscapes blog Tidy Data for reproducibility, efficiency, and collaboration by Julia Lowndes and Allison Horst

Example chunks with tidy code

R

Python

```
1 # Load packages
 2 library(tidyverse)
 3 library(tidyquant)
 5 # Download symbols of DOW index
 6 symbols \leftarrow tq index(x = "DOW") \mid >
     filter(company != "US DOLLAR")
 9 # Download prices of DOW index constituents
   prices <- tq get(x = symbols, get = "stock.prices",</pre>
                     from = "2000-01-01", to = "2022-12-31")
11
12
13 # Calculate returns
14 returns <- prices |>
15
     group by(symbol) |>
     mutate(ret = adjusted / lag(adjusted) - 1) |>
16
     select(symbol, date, ret) |>
17
18
     drop na(ret)
```

Welcoming contributions on our blog



Tidy Finance Blog

Experimental and external contributions based on Tidy Finance with R. Contribute your ideas!



Tidy Market Microstructure 76 min

MARKET MICROSTRUCTURE | R | DATA.TABLE

A beginner's guide to market quality measurement in high-frequency data using R.

BJÖRN HAGSTRÖMER, NIKLAS JAN 4, LANDSBERG 2024



Using DuckDB with WRDS Data 10 min

DATA R

Demonstrate the power of DuckDB and dbplyr with WRDS data.

IAN GOW

DEC 22, 2023



Comparing Fama-French Three vs Five Factors

7 min

DATA REPLICATIONS R

An explanation for the difference in the size factors of Fama and French 3 and 5 factor data

CHRISTOPH SCHEUCH

OCT 2, 2023

Maintainers of tidy-finance.org



Christoph Scheuch

Head of Artificial
Intelligence at
wikifolio.com



Stefan Voigt

Assistant Professor of Finance at University of Copenhagen



Patrick Weiss

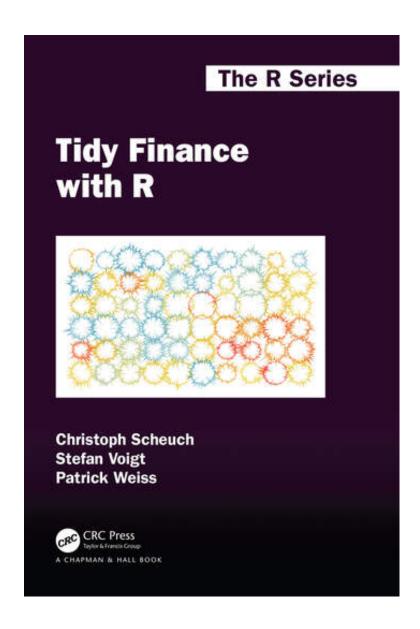
Assistant Professor of Finance at Reykjavik University

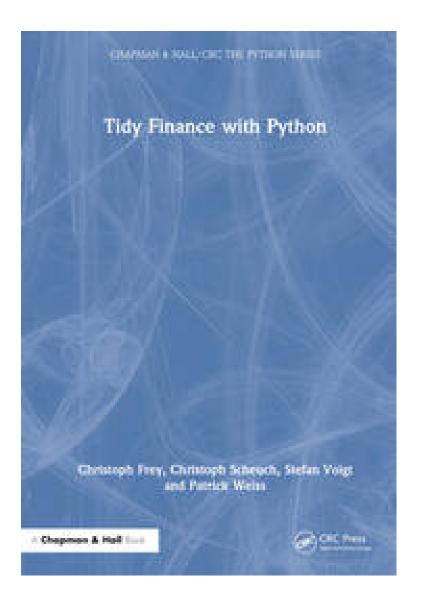


Christoph Frey

Quantitative Researcher at Pinechip Capital

We also wrote books





Accessing & Managing Financial Data

Importance of organizing data efficiently

• Challenge: ensure consistency across various data sources

• Solution:

- Use R to import, prepare & store data
- Use SQLite to organize data in a database

• R Packages:

- Manipulation: tidyverse
- Import: tidyquant, frenchdata, readxl
- Storage: RSQLite

Fama-French factors & portfolios

Most popular data for asset pricing tests since Fama and French (1993)

```
1 library(frenchdata)
 2
 3 factors ff3 monthly raw <- download french data("Fama/French 3 Factors")</pre>
 4 factors ff3 monthly <- factors ff3 monthly raw$subsets$data[[1]] |>
     mutate(
       month = floor date(ymd(str c(date, "01")), "month"),
       across(c(RF, `Mkt-RF`, SMB, HML), ~as.numeric(.) / 100),
       .keep = "none"
     ) |>
10
     rename with(str to lower) |>
     rename(mkt excess = `mkt-rf`) |>
11
     select(month, everything())
12
13
14 print(factors ff3 monthly, n = 5)
```

```
# A tibble: 1,170 \times 5
 month
            mkt excess
                                  hml
                                          rf
                           smb
                 <dbl>
                         <dbl>
 <date>
                                <dbl> <dbl>
1 1926-07-01 0.0296 -0.0256 -0.0243 0.0022
2 1926-08-01 0.0264 -0.0117 0.0382 0.0025
              0.0036 -0.014 0.0013 0.0023
3 1926-09-01
4 1926-10-01 -0.0324 -0.0009 0.007 0.0032
5 1926-11-01
              0.0253 - 0.001 - 0.0051 0.0031
# i 1,165 more rows
```

q-Factors

Alternative to Fama-French data by Hou, Xue, and Zhang (2014)

```
1 library(readr)
          2
          3 factors g monthly link <-</pre>
              "https://global-q.org/uploads/1/2/2/6/122679606/q5 factors monthly 2022.csv"
            factors q monthly <- read csv(factors q monthly link) |>
              mutate(month = ymd(str c(year, month, "01", sep = "-"))) |>
              select(-R F, -R MKT, -year) |>
              rename with(~ str remove(., "R ")) |>
              rename with(~ str to lower(.)) |>
         10
              mutate(across(-month, ~ . / 100))
         11
         12
         13 print(factors q monthly, n = 5)
# A tibble: 672 × 5
 month
                           ia
                                  roe
                                            eg
                  me
  <date>
             <dbl>
                        <dbl>
                                <dbl>
                                         <dbl>
1 1967-01-01 0.0683 -0.0297 0.0192 -0.0218
2 1967-02-01 0.0165 -0.00227 0.0354 0.0222
3 1967-03-01 0.0200 -0.0178 0.0184 -0.0104
4 1967-04-01 -0.00690 -0.0288 0.0106 -0.0173
5 1967-05-01 0.0285
                     0.0252 0.00692 0.00158
# i 667 more rows
```

Macroeconomic predictors

Collection of variables for equity premium prediction (Welch & Goyal, 2008)

```
1 library(readxl)
          3 download.file(
              url = "https://docs.google.com/spreadsheets/d/1g4LOaRj4TvwJr9RIaA nwrXXWTOy46bP/expo
              destfile = "macro predictors.xlsx",
              mode = "wb"
            macro predictors <- read xlsx("macro predictors.xlsx", sheet = "Monthly") |>
         10
              mutate(
                # Several cleaning steps & variable transformations...
         12
# A tibble: 1,152 × 15
  month
           rp div
                                           de
                                                  svar
                        dр
                               dy
                                    ер
                                                          bm
                                                               ntis
                                                                       tbl
             <dbl> <dbl> <dbl> <dbl> <dbl>
  <date>
                                                 <dbl> <dbl> <dbl> <dbl>
1 1926-12-01 -0.0220 -2.97 -2.96 -2.39 -0.586 0.000465 0.441 0.0509 0.0307
2 1927-01-01 0.0422 -2.94 -2.96 -2.37 -0.568 0.000470 0.444 0.0508 0.0323
3 1927-02-01 0.00363 -2.98 -2.93 -2.43 -0.549 0.000287 0.429 0.0517 0.0329
4 1927-03-01 0.0142 -2.98 -2.97 -2.45 -0.531 0.000924 0.470 0.0464 0.032
5\ 1927-04-01\ 0.0459\ -2.98\ -2.97\ -2.47\ -0.513\ 0.000603\ 0.457\ 0.0505\ 0.0339
# i 1,147 more rows
# i 5 more variables: lty <dbl>, ltr <dbl>, tms <dbl>, dfy <dbl>, infl <dbl>
```

Other macroeconomic data

10K data sets available via Federal Reserve Economic Data (FRED) database

```
1 library(tidyquant)
2
3 # Example: consumer price index (CPI)
4 cpi_monthly <- tq_get("CPIAUCNS", get = "economic.data") |>
5 mutate(
6 month = floor_date(date, "month"),
7 cpi = price / price[month == max(month)],
8 .keep = "none"
9 )
10 print(cpi_monthly, n = 5)
```

Use SQLite database for storage

```
1 library(RSQLite)
 2 library(dbplyr)
 4 # Create database
 5 tidy finance <- dbConnect(</pre>
     SQLite(), "tidy finance r.sqlite", extended types = TRUE
 9 # Write data to database
10 dbWriteTable(
11 conn = tidy finance,
12   name = "factors ff3 monthly",
value = factors ff3 monthly,
     overwrite = TRUE
14
15 )
16
17 # Load data from database
18 factors ff3 monthly <- tbl(tidy finance, "factors ff3 monthly") |>
     collect()
19
```

Why SQLite?

Pros:

- Lightweight, self-contained, serverless database engine
- Great for education purposes or prototyping

Cons:

- Limitations with respect to very large data & concurrency
- Transfer to other languages cumbersome (e.g. Python)

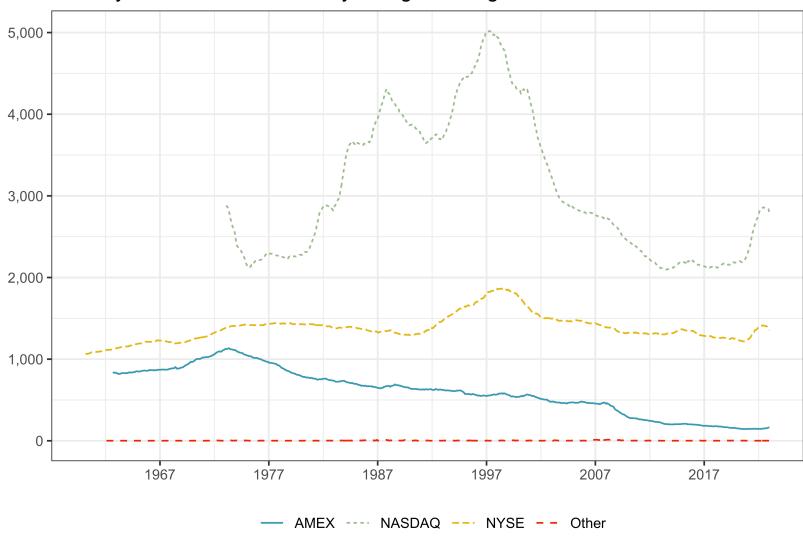
WRDS & Other Data Providers

Wharton Research Data Services (WRDS)

- Popular provider of financial & economic data
- Focus on academic audience & research applications
- Access via RPostgres package
- Main data used in Tidy Finance
 - CRSP: historical monthly & daily returns for US stocks
 - Compustat: historical accounting data for US companies
 - Mergent FISD: characteristics of US corporate bonds
 - TRACE: detailed US corporate bond transactions

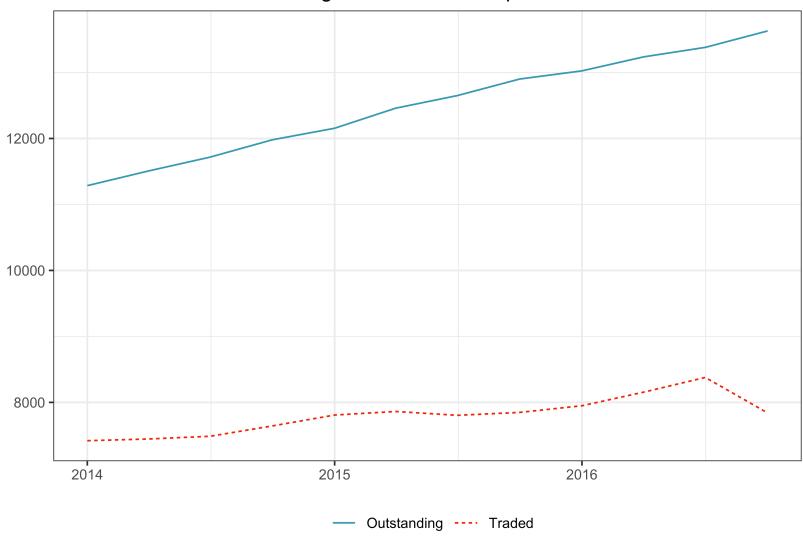
Glimpse at historical stock data

Monthly number of securities by listing exchange



Glimpse at historical bond data

Number of bonds outstanding and traded each quarter



Other data providers

Large ecosystem of alternative data providers

- Extensive list of R packages on tidy-finance.org
- Examples: fredr, ecb, Rblpapi, Quandl, edgarWebR, etc.

Are we missing an important package?

please reach out via contact@tidy-finance.org

Wrap-up

Tidy approach to teaching & research

tidyfinance R package to access financial data in a tidy way:

```
install.packages("tidyfinance")

tidyfinance::download_data(
    type = "wrds_crsp_monthly",
    start_date = "1960-01-01", end_date = "2020-12-31"

)
```

- Detailed open source material at tidy-finance.org
- Get in touch for teaching materials & to contribute to blog
- Follow me for news linkedin.com/in/christophscheuch