

Statistical Programming and Open Science Methods

Debugging tools

Joachim Gassen
Humboldt-Universität zu Berlin

September 02, 2022



SFB/Transregio 266

ACCOUNTING FOR
TRANSPARENCY

Time table October 11

When?	What?
09:00	Writing readable and reusable code
10:30	Coffee
11:00	Debugging tools
12:30	Lunch and coffee
13:30	Relational databases and the concept of normalized data
14:30	Data wrangling and visualization fundamentals
15:30	Assignments and wrap up
16:00	End of event

Disclaimer

Some of the following is borrowed from chapter 22 of Hadley Wickham (2019): Advanced R, <https://adv-r.hadley.nz>

Principle

Finding your bug is a process of confirming the many things that you believe are true — until you find one which is not true.

— Norm Matloff

- ▶ Google the error message
- ▶ Make the bug repeatable by creating a reproducible example
- ▶ Figure out where it is
- ▶ Fix it and test it

Most bugs are simple

- ▶ Typos are paramount and sometimes hard to spot (a good editor with syntax high-lightening helps!)
- ▶ The best way to avoid bugs is do adopt a readable coding style
- ▶ Debugging and unit testing go hand in hand (we will talk about unit testing in the February block)
- ▶ Do not ignore messages and warnings that your code throws at you. They are there for a reason,

Error messages are not always self-explanatory but almost always informative

```
count_distinct_obs <- function(df, ...) {  
  as_tibble(df) %>%  
    select(...) %>%  
    distinct() %>%  
    nrow()  
}  
  
read_csv("../data/sub.csv") %>%  
  count_distinct_obs[cik, name] -> count_sec_reg  
## Error in `[.tbl_df`(., count_distinct_obs, cik, name): object 'cik'
```

Debugging tools

- ▶ Traceback
- ▶ Debug on error
- ▶ Breakpoints
- ▶ Logging
- ▶ Dump analysis

Let's do a case: Take a look at
`code/calc_state_of_incorp_distance_bugged.R`

Corner cases

*# This function returns a p % sample of a vector v . It always
chooses at least q observations to avoid small samples*

```
sample_my_sample <- function(v, p, q) {  
  n <- ceiling(max(p * length(v), q))  
  do_replace <- (n > length(v))  
  sample(v, n, do_replace)  
}
```

```
for (i in 1:10) {  
  print(sample_my_sample(i:10, 0.5, 3))  
}
```

```
## [1] 3 7 10 1 2
```

```
## [1] 5 9 8 4 3
```

```
## [1] 5 7 6 9
```

```
## [1] 8 10 9 7
```

```
## [1] 7 6 8
```

```
## [1] 9 8 7
```

```
## [1] 9 7 8
```

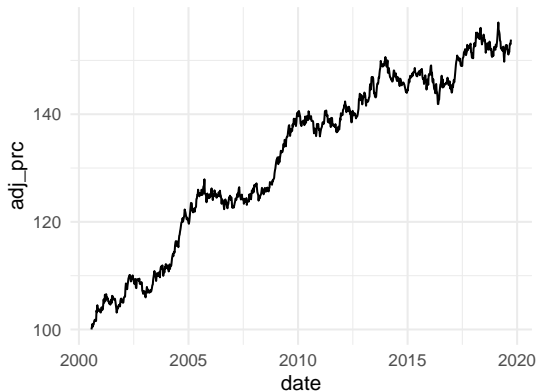
```
## [1] 8 10 9
```

```
## [1] 9 10 10
```

```
## [1] 6 8 8
```


Code meets data

```
read_csv("../raw_data/stock_price.csv") %>%  
  ggplot(aes(x = date, y = adj_prc)) +  
  geom_line() + theme_minimal()
```



Calculate weekly mean return and weekly return variance

```
read_csv("../raw_data/stock_price.csv") %>%  
  mutate(return = (adj_prc - lag(adj_prc))/lag(adj_prc)) %>%  
  summarise(mn_return = mean(return, na.rm = TRUE),  
            var_return = sd(return, na.rm = TRUE)^2)  
  
## # A tibble: 1 x 2  
##   mn_return var_return  
##   <dbl>      <dbl>  
## 1  0.000467  0.0000262  
  
## Correct values  
## # A tibble: 1 x 2  
##   mn_return var_return  
##   <dbl>      <dbl>  
## 1  0.000434  0.0000247
```