Once again, I'm dipping outside of the tidyverse, but this package and its functions have been really useful in getting data quickly in (and out) of R.

For work, I have to pull in data from a few different sources, and manipulate and work with them to give me the final dataset that I use for much of my analysis. So that I don't have to go through all of that joining, recoding, and calculating each time, I created a final merged dataset as a CSV file that I can load when I need to continue my analysis. The problem is that the most recent version of that file, which contains 13 million+ records, was so large, writing it (and subsequently reading it in later) took forever and sometimes timed out.

That's when I discovered the data.table library, and its fread and fwrite functions. Tidyverse is great for working with CSV files, but a lot of the memory and loading time is used for formatting. fread and fwrite are leaner and get the job done a bit faster. For regular-sized CSV files (like my reads2019 set), the time difference is pretty minimal. But for a 5GB datafile, it makes a huge difference.

```
library(tidyverse)
## -- Attaching packages ------ tidyverse
1.3.0 --
## ggplot2 3.2.1
                   purrr 0.3.3
## tibble 2.1.3 dplyr 0.8.3
## tidyr 1.0.0 stringr 1.4.0
## readr 1.3.1 forcats 0.4.0
## -- Conflicts ------
tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
system.time(reads2019 <- read csv("~/Downloads/Blogging A to
Z/SaraReads2019 allchanges.csv",
                     col names = TRUE))
## Parsed with column specification:
## cols(
##
    Title = col character(),
   Pages = col double(),
##
##
    date_started = col_character(),
    date read = col character(),
##
##
    Book.ID = col double(),
##
    Author = col character(),
##
    Additional Authors = col character(),
##
    AverageRating = col double(),
##
    OriginalPublicationYear = col double(),
##
    read time = col double(),
    MyRating = col double(),
##
##
    Gender = col double(),
    Fiction = col_double(),
##
##
    Childrens = col double(),
##
    Fantasy = col double(),
##
    SciFi = col double(),
##
    Mystery = col_double(),
##
    SelfHelp = col double()
##)
##
    user system elapsed
```

0.00 0.10 0.14

```
rm(reads2019)
library(data.table)
## Attaching package: 'data.table'
## The following objects are masked from 'package:dplyr':
##
##
       between, first, last
## The following object is masked from 'package:purrr':
##
##
       transpose
system.time(reads2019 <- fread("~/Downloads/Blogging A to
Z/SaraReads2019_allchanges.csv"))
      user system elapsed
##
        0
               0
```

But let's show how long it took to read my work datafile. Here's the elapsed time from the system.time output.

```
read_csv:
user system elapsed
61.14 11.72 90.56
fread:
user system elapsed
57.97 16.40 57.19
```

But the real win is in how quickly this package writes CSV data. Using a package called wakefield, I'll randomly generate 10,000,000 records of survey data, then see how it takes to write the data to file using both write_csv and fwrite.

```
library(wakefield)
## Warning: package 'wakefield' was built under R version 3.6.3
## Attaching package: 'wakefield'
## The following objects are masked from 'package:data.table':
##
##
       hour, minute, month, second, year
## The following object is masked from 'package:dplyr':
##
##
       id
set.seed(42)
reallybigshew <- r_data_frame(n = 10000000,</pre>
                               id,
                               race,
                               age,
                               smokes,
                               marital,
                               Start = hour,
                               End = hour,
```

iq,
height,
died)

```
system.time(write_csv(reallybigshew, "~/Downloads/Blogging A to
Z/bigdata1.csv"))

## user system elapsed
## 134.22 2.52 137.80

system.time(fwrite(reallybigshew, "~/Downloads/Blogging A to Z/bigdata2.csv"))

## user system elapsed
## 8.65 0.32 2.77
```