Stat 260, Lecture 5, Reading Data

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Load packages

```
library(tidyverse)
library(nycflights13)
```

Reading

- Workflow: scripts: Chapter 4 of printed text, Chapter 6 of online text
- ▶ Introduction to data wrangling: Part II introduction of printed text, Chapter 9 of online text
- ▶ Tibbles: Chapter 7 of printed text, Chapter 10 of online text
- Reading data with readr: Chapter 8 of printed text, Chapter 11 of online text
- Data import (readr/tidyr) cheatsheet at [https://github.com/rstudio/cheatsheets/raw/master/dataimport.pdf]

Tibbles

- In base R, the data structure used to hold data sets is the data frame.
- We can make a data frame from vectors as follows:

The tidyverse authors find the default behaviour of data frames to be odd, and so implemented an improvement called tibbles:

data frames to tibbles and back

▶ data frames can be coerced to tibbles and *vice versa*.

```
as_tibble(dd)
## # A tibble: 3 x 2
##
        x y
##
    <dbl> <fct>
## 1
      NA one
     10 t.wo
    1 three
## 3
as.data.frame(tt)
##
## 1 NA
         one
## 2 10
         two
## 3 1 three
```

tibble printing

- One difference between data frames and tibbles is how they are printed.
- Printing a data frame: all rows and columns, up to your R session's max.print.
- Printing a tibble: the first 10 rows, as many columns as fit the screen, and the column data types.

```
flights
## # A tibble: 336,776 x 19
       year month
                    day dep_time sched_dep_time dep_delay arr_time
##
      <int> <int> <int>
                            <int>
                                           <int>
                                                      <dh1>
                                                               <int>
      2013
                              517
                                             515
                                                                 830
       2013
                             533
                                             529
                                                                 850
      2013
                                                                 923
                              542
                                             540
   4 2013
                             544
                                             545
                                                         -1
                                                                1004
##
   5 2013
                             554
                                                                 812
                                             600
      2013
                             554
                                             558
                                                         -4
                                                                 740
       2013
                              555
                                             600
                                                         -5
                                                                 913
       2013
                              557
                                             600
                                                                 709
       2013
                              557
                                             600
                                                                 838
       2013
                              558
                                             600
                                                         -2
                                                                 753
    ... with 336.766 more rows, and 12 more variables: sched arr time <int>.
       arr_delay <dbl>, carrier <chr>, flight <int>, tailnum <chr>,
       origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>, hour <dbl>,
## #
       minute <dbl>, time hour <dttm>
```

Control printing of tibbles

- ► To see all rows/columns of a tibble, best to View() it.
- But you can also print all rows and columns by setting options(dplyr.print_min=Inf) and options(tibble.width=Inf).
 for your all rows
 and columns

Extracting columns as vectors

► Use the basic tools \$ and [[to extract a variable from a tibble or data frame:

```
dd$x
## [1] NA 10  1
tt$x
## [1] NA 10  1
dd[["x"]]
## [1] NA 10  1
tt[["x"]]
## [1] NA 10  1
```

Subsetting: columns

 Using select() is the preferred method to subset columns of a data frame or tibble, but we can also use the more basic tool

```
[; e.g.,
tt[,"x"]
   # A tibble: 3 x 1
    <db1>
      10
tt[.c("x"."v")]
## # A tibble: 3 x 2
         x y
     <dhl> <chr>>
       NA one
     10 two
      1 three
dd[,"x"] # returns a vector
## [1] NA 10 1
dd[,c("x","y")]
          one
```

Subsetting: rows

Using filter() is the preferred method to extract rows of a data frame or tibble, but we can also use [.

```
tt[2,]
## # A tibble: 1 x 2
   <dbl> <chr>
     10 two
tt[1:2,]
## # A tibble: 2 x 2
        x y
   <dhl> <chr>>
      NA one
## 2 10 two
dd[2.] # returns a vector
## 2 10 two
dd[1:2,]
```

x y ## 1 NA one ## 2 10 two

Exercise

- Create a data frame myd and tibble myt that each have columns named cat, dog and mouse. Each column should be of length three, but the values in each column are up to you.
- What do names(myd) and names(myt) return?
- Create the variable a1 <- c("cat","dog","bird","fish") and the variable a2 <- c("cat","tiger"). We can combine logicals with [to subset. What do the following return?
 - myd[,names(myd) %in% a1]
 - myd[,names(myd) %in% a2]
 - myd[,names(myt) %in% a1]
 - myd[,names(myt) %in% a2]

Importing data

- ► We read in the HIV prevalence data with the base R function read.csv(), which returned a data frame.
- We will now discuss the tidyverse equivalent, read_csv(), which returns a tibble.

```
hiv <- read_csv("../Labs/HIVprev.csv")

## Parsed with column specification:
## cols(
## Country = col_character(),
## year = col_double(),
## prevalence = col_double()
## )</pre>
```

Why use read_csv() instead of read.csv()?

- read_csv() reports how each column of the CSV file was "parsed" (more on this later),
- returns a tibble,
- ▶ uses stringsAsFactors = FALSE as the default, (recall hiv <read.csv("../Labs/HIVprev.csv",stringsAsFactors = FALSE))
- is faster, and
- is more consistent across operating systems.

Other read_ functions

- CSV stands for comma-separated files, aka comma-delimited files
- read_csv() reads semicolon-delimited files,
- read_tsv() reads tab-delimited files,
- read_delim() reads files with user-specified delimiter.
- Exercise: A file called "chicken.C" contains the following data on two chickens, with IDs 22 and 33, who laid 2 and 1 eggs, respectively. (Reference: https://isotropic.org/papers/chicken.pdf) How would you read this data file into R?

IDCeggs 22C2 33C1

Skip and comments

- Some files contain a header that describes the data, aka meta-data, that we should skip when reading.
- ► Some files include comments that start with common characters, such as "#".
- Example file

```
This is a header
that you should skip
# this is a comment
A,B,C
1,2,3
4,5,6 # another comment
```

```
read_csv("lec05exfile.csv",skip=2,comment="#")
## Parsed with column specification:
## cols(
## A = col_double(),
## B = col_double(),
## C = col_date(format = "")
## )
## # A tibble: 2 x 3
## A B C
## <dbl> <dbl> <date>
## 1 1 2.2 1999-05-10
```

2 4 5.5 2001-04-04

Parsing a vector

- read_csv() returns a message that described how each column of the input file was parsed.
- ▶ Parsing a file depends on the parse_* functions, such as parse_number(), that parse vectors.
- The parse_* functions take a vector of character strings as input and return a vector of a given mode, handling missing values as specified by the user.

```
parse_number(c("$10.55","33%","Number is 44","."),na=".")
```

```
## [1] 10.55 33.00 44.00 NA
```

- ► The parse functions are designed to handle data formats and character sets from around the world.
- ▶ In this course we assume North American data formats and character set.
- See the text if you need other formats.

Other parsing functions

- parse_logical(), parse_integer(), parse_double(),
 parse_character(), parse_factor(), parse_datetime(),
 parse_date() and parse_time().
- ▶ Use the str() function to see the mode of an object:

```
str(parse_logical(c("TRUE", "FALSE")))
## logi [1:2] TRUE FALSE
str(parse_logical(c("1","0")))
## logi [1:2] TRUE FALSE
str(parse_integer(c("1","0")))
## int [1:2] 1 0
str(parse_double(c("1","0")))
## num [1:2] 1 0
str(parse factor(c("1","0")))
## Factor w/ 2 levels "1", "0": 1 2
```

Dates and times

- ► These parsers have default formats for dates and times, but your best bet is to specify the format yourself.
- ▶ The formatting rules are described in help(strptime).

```
dd <- c("05/14/1966/12/34/56","04/02/2002/07/43/00","08/17/2005/07/22/00","08/1
dd <- parse_datetime(dd,format = "%m/%d/%Y/%H/%M/%S")
str(dd)

## POSIXct[1:4], format: "1966-05-14 12:34:56" "2002-04-02 07:43:00" ...
mean(dd)

## [1] "1995-09-18 22:59:59 UTC"
diff(dd)</pre>
```

```
## Time differences in days
## [1] 13106.797 1232.985 1091.374
```

Parsing files

- read_csv() and other read functions guess at the format of each column. Sometimes this works, sometimes not.
- ▶ You can read about how these functions guess in the text.
- ▶ Here we'll focus on manually specifying the format.

```
dat <- read_csv("lec05exfile.csv",skip=2,comment="#")

## Parsed with column specification:
## cols(
## A = col_double(),
## B = col_double(),
## C = col_date(format = "")
## ## )</pre>
```

Cut-and-paste the guess and replace parsers as necessary

int [1:2] 1 4

▶ For reproducibility your R scripts should have a manual specification of the parsing of each column, rather than relying on guesses that can change as your data changes.

Exercise

get wd ()

Copy the following data to a file and read it in to R. Specify the column types yourself, based on the descriptions in the header of the file. Hint: read about col_factor().

```
# Variable fert is a factor that records the
# type of fertilizer used in the experiment,
# date records the date and time of the experiment
# and yield is the yield of corn.
fert,date,yield
F1A2,2018/04/01/12/30,22.56
F1A1,2018/04/02/12/00,26.06
F2A2,2018/04/01/12/45,32.03
F2A1,2018/04/02/12/00,33.21
```