# Batch loading data

Daniel Andersor

Week 4, Class 1

# Agenda

- Discuss the midterm
  - Canvas quiz (10 points; please don't stress)
  - Take home (40 points)
- Review Lab 1
- map\_dfr and batch—loading data
- Introduce list columns
  - This will mostly be an intro to Wednesday's lecture

# Learning objectives

- Understand when map\_dfr can and should be applied
- Better understand file paths, and how {fs} can help
- Be able to batch load data of a specific type within a mixed-type directory
- Use filenames to pull data

# Midterm

Questions?

Let's look at the take-home portion

# Review Lab 1

# map\_dfr

 If each iteration returns a data frame, you can use map\_dfr to automatically bind all the data frames together.

# Example

 Create a function that simulates data (please copy the code and follow along)

```
\# A tibble: 10 x 2
##
     sample id sample
##
         <int> <dbl>
## 1
             1 - 0.9965824
##
             2 0.7218241
##
             3 - 0.6172088
##
             4 2.029392
##
   5
             5 1.065416
##
             6 0.9872197
##
             7 0.02745393
##
             8 0.6728723
## 9
             9 0.5720665
            10 0.9036777
```

# Two more quick examples

### simulate(3, 100, 10)

### simulate(5, -10, 1.5)

# Simulation

- Assume we want to vary the sample size from 10 to 150 by increments of 5
- mean stays constant at 100, sd is constant at 10

Try with purrr::map

02:00

```
library(tidyverse)
sims <- map(seq(10, 150, 5), simulate, 100, 10)</pre>
```

### sims[1]

```
## [[1]]
## # A tibble: 10 x 2
##
      sample id sample
##
          <int> <dbl>
##
              1 103.7618
##
              2 111.5353
    3
##
              3 115.7490
##
    4
              4 105.8853
    5
##
              5 93.84955
    6
##
              6 97.71089
    7
##
              7 100.6392
##
    8
              8 96.86526
##
   9
              9 97.51501
## 10
             10 98.46205
```

### sims[2]

```
## [[1]]
## # A tibble: 15 x 2
##
     sample id sample
##
         <int> <dbl>
## 1
                93.64743
##
                99.96206
##
              3 100.4562
## 4
              4 106.8407
## 5
              5 97.47957
## 6
              6 98.48961
##
              7 91.25069
## 8
             8 80.23099
## 9
              9 102.3766
## 10
             10 100.3609
## 11
            11 101.3490
## 12
            12 101.1758
## 13
            13 91.74411
## 14
            14 78.64764
## 15
            15 102.1421
```

# Swap for map\_dfr

### Try it - what happens?

```
sims_df <- map_dfr(seq(10, 150, 5), simulate, 100, 10)
sims_df</pre>
```

```
## # A tibble: 2,320 x 2
     sample id sample
##
         \langle int \rangle \langle dbl \rangle
##
## 1
             1 85.64361
## 2
           2 103.6789
##
           3 94.71782
##
           4 103.1350
##
            5 99.78701
## 6
           6 105.3462
## 7
        7 100.0653
## 8
         8 94.28314
## 9 9 108.8872
## 10
       10 106.0850
## # ... with 2,310 more rows
```

01:00

# Notice a problem here

### sims\_df[1:15, ]

```
# A tibble: 15 x 2
##
      sample id sample
##
          <int> <dbl>
##
              1 85.64361
    1
   2
##
              2 103.6789
   3
##
              3 94.71782
##
              4 103.1350
    5
##
              5 99.78701
##
              6 105.3462
##
   7
              7 100.0653
##
   8
              8 94.28314
##
              9 108.8872
   9
## 10
             10 106.0850
              1 89.49968
## 11
## 12
              2 86.99898
## 13
              3 85.38054
## 14
              4 99.10690
## 15
              5 105.0088
```

# .id argument

```
sims_df2 <- map_dfr(seg(10, 150, 5), simulate, 100, 10,
                   .id = "iteration")
sims df2[1:14, ]
## # A tibble: 14 x 3
##
  iteration sample id sample
## <chr>
                   <int> <dbl>
## 1 1
                       1 112.1250
## 2 1
                       2 88.07056
## 3 1
                       3 108.3908
## 4 1
                       4 100.8193
## 5 1
                       5 102.1545
## 6 1
                       6 113.5398
## 7 1
                       7 101.4171
## 8 1
                       8 99.33668
## 9 1
                       9 100.2855
## 10 1
                      10 90.22043
                      1 91.08882
## 11 2
## 12 2
                       2 107.3664
## 13 2
                       3 101.1745
                       4 96.82053
## 14 2
```

.id: Either a string or NULL. If a string, the output will contain a variable with that name, storing either the name (if .x is named) or the index (if .x is unnamed) of the input. If NULL, the default, no variable will be created.

- {purrr} documentation

### setNames

70

75

##

```
sample_size <- seq(10, 150, 5)</pre>
sample_size
##
        10 15 20 25 30 35 40 45 50 55 60 65 70 75
                                                                  80
                                                                      85
    Γ11
## [22] 115 120 125 130 135 140 145 150
sample_size <- setNames(sample_size,</pre>
                         english::english(seq(10, 150, 5)))
sample_size[1:15]
##
            ten
                     fifteen
                                   twenty twenty-five
                                                              thirty
                                                                      thirty
##
             10
                          15
                                        20
                                                                  30
##
                                             fifty-five
          forty
                 forty-five
                                    fifty
                                                               sixty
                                                                       sixty
##
             40
                          45
                                        50
                                                     55
                                                                  60
##
        seventy seventy-five
                                   eighty
```

80

# Try agair

```
## # A tibble: 14 x 3
##
        sample id sample
   n
##
                <int> <dbl>
  <chr>
##
   1 ten
                    1 98.94914
##
                    2 101.6824
   2 ten
                    3 88.16447
##
   3 ten
## 4 ten
                    4 90.13604
##
                    5 85.53591
   5 ten
##
                  6 90.69977
   6 ten
##
                    7 105.8858
   7 ten
## 8 ten
                   8 89.12978
## 9 ten
                   9 114.4982
## 10 ten
                10 111.6440
                 1 103.2732
## 11 fifteen
                  2 106.8949
## 12 fifteen
## 13 fifteen
                  3 88.83591
## 14 fifteen
                 4 105.5402
```

# Another quick example

### broom::tidy

 The {broom} package helps us extract model output in a tidy format

# Fit separate models by year

### Again - probs not best statistically

```
split(gss_cat, gss_cat$year) %>%
  map_dfr(~lm(tvhours ~ age, .x) %>%
        broom::tidy())
```

```
## # A tibble: 16 x 5
##
  term estimate std.error statistic p.value
## <chr>
                    <dbl>
                              <dbl>
                                       <dbl>
                                                  <dbl>
   1 (Intercept) 2.080163 0.1709061 12.17138 7.995632e-33
##
   2 age
        0.01948584 0.003485199 5.591027 2.599011e- 8
##
##
   3 (Intercept) 2.078999 0.2176829 9.550583 1.191266e-20
##
         0.01963575 0.004400292 4.462375 9.137366e- 6
   4 age
   5 (Intercept) 1.767990 0.2464509 7.173804 1.531756e-12
##
##
         0.02386070 0.005031548 4.742218 2.459650e- 6
   6 age
## 7 (Intercept) 2.096054 0.1496431 14.00702 1.419772e-42
## 8 age 0.01781388 0.002977289 5.983256 2.589482e- 9
   9 (Intercept) 1.855278 0.2156381 8.603668 2.167351e-17
## 10 age
         0.02390720 0.004314567 5.541043 3.628675e- 8
  11 (Intercept) 2.068914 0.2096397
                                    9.868903 2.896085e-22
        0.01989505 0.004086638 4.868317 1.251234e- 6
## 12 age
  13 (Intercept) 1.878070 0.2258400 8.315932 2.280108e-16
## 14 age
          0.02547794 0.004449295 5.726287 1.274840e- 8
## 15 (Intercept) 1.980095 0.1877544 10.54620 3.238043e-25
## 16 age 0.02049066 0.003611900 5.673098 1.650822e- 8
```

### .io

In cases like the preceding, .id becomes invaluable

```
split(gss_cat, gss_cat$year) %>%
    map_dfr(~lm(tvhours ~ age, .x) %>%
        broom::tidy(),
        .id = "year")
```

```
## # A tibble: 16 x 6
##
   year term
                       estimate std.error statistic p.value
##
     <chr> <chr>
                           <dbl>
                                       <dbl>
                                                 <dbl>
                                                             <dbl>
          (Intercept) 2.080163 0.1709061 12.17138 7.995632e-33
##
   1 2000
##
   2 2000 age
                       0.01948584 0.003485199 5.591027 2.599011e- 8
##
          (Intercept) 2.078999 0.2176829 9.550583 1.191266e-20
   3 2002
                       0.01963575 0.004400292 4.462375 9.137366e- 6
   4 2002 age
##
##
   5 2004
           (Intercept) 1.767990 0.2464509 7.173804 1.531756e-12
##
   6 2004
                       0.02386070 0.005031548
                                              4.742218 2.459650e- 6
           age
   7 2006
           (Intercept) 2.096054
##
                                 0.1496431
                                             14.00702
                                                       1.419772e-42
##
   8 2006
                       0.01781388 0.002977289
                                              5.983256 2.589482e- 9
           age
##
           (Intercept) 1.855278
   9 2008
                                 0.2156381
                                              8.603668 2.167351e-17
                                              5.541043 3.628675e- 8
##
  10 2008
                       0.02390720 0.004314567
           age
           (Intercept) 2.068914
##
  11 2010
                                 0.2096397
                                              9.868903 2.896085e-22
  12 2010
                       0.01989505 0.004086638 4.868317 1.251234e- 6
##
           age
           (Intercept) 1.878070 0.2258400 8.315932 2.280108e-16
  13 2012
##
                                              5.726287 1.274840e- 8
## 14 2012
                       0.02547794 0.004449295
           age
```

# Batchloading data

Please follow along

# $\{fS\}$

• note – there are base equivalents. **{fs}** is just a a bit better across platforms and has better defaults.

### Could we apply map\_dfr here?

```
# install.packages("fs")
library(fs)
dir ls(here::here("data"))
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
dir_ls(here::here("data", "pfiles_sim"))
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/r
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/r
```

## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p

### Limit files

- We really only want the .csv
  - That happens to be the only thing that's in there but that's regularly not the case

```
dir_ls(here::here("data", "pfiles_sim"), glob = "*.csv")
```

```
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/r
## /Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/data/p
```

### Batch load

Loop through the directories and import or read\_csv

```
files <- dir_ls(
  here::here("data", "pfiles_sim"),
  glob = "*.csv"
)
batch <- map_dfr(files, read_csv)
batch</pre>
```

```
## # A tibble: 15,945 x 22
##
    Entry Theta Status Count RawScore SE Infit Infit Z Outfit Outf
##
    <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> 
## 1 123 1.2687 1
                       36
                              23 0.3713 0.93 -0.34 0.82
##
  2 88 1.5541 1 36 25 0.3852 0.95 -0.37 0.81
                              33 0.6187 0.9 -0.04 1.63000
  3 105 3.2773 1 36
##
                            35 1.0234 0.93 0.23 0.35
  4 153 4.4752 1 36
##
##
   5 437 2.6655
                   1 36 31 0.5008 0.92 -0.18 0.88
##
   6 307 5.7137 0 36 36 1.8371 1
##
  7 305 3.7326 1 36 34 0.7408 1.06 0.31 0.86
## 8 42 0.609
                1 36
                             18 0.36 1.55 2.56 1.74
## 9 59 -2.623
                   1 36
                             3 1.0344 0.85 0.06 0.17
                       36
## 10
     304 5.7137
                   0
                              36 1.8371 1
## # ... with 15,935 more rows, and 11 more variables: PointMeasureCorr <dbl>
## #
     ObservMatch <dbl>, ExpectMatch <dbl>, PointMeasureExpected <dbl>, RN
```

# Problem

• We've lost a lot of info – no way to identify which file is which

Try to fix it!



### Add id

## #

####

####

```
## # A tibble: 15,945 x 23
## # ... with 15,935 more rows, and 23 more variables: file <chr>, Entry <dbl
## # Status <dbl>, Count <dbl>, RawScore <dbl>, SE <dbl>, Infit <dbl>, Ir
## # Outfit <dbl>, Outfit Z <dbl>, Displacement <dbl>, PointMeasureCorr
```

Weight <dbl>, ObservMatch <dbl>, ExpectMatch <dbl>, PointMeasureExpe

RMSR <dbl>, WMLE <dbl>, testeventid <dbl>, ssid <dbl>, asmtprmrydsbl

Note – the **file** column contains the full path, which is so long it makes no rows print

batch2 <- map dfr(files, read csv, .id = "file")</pre>

asmtscndrvdsbltvcd <dbl>

### batch2 %>% count(file)

```
## # A tibble: 31 x 2 ## # ... with 21 more rows, and 2 more variables: file <chr>, n <int>
```

• Still not terrifically useful. What can we do?

# Step 1

• Remove the here::here path from string

```
## # A tibble: 31 x 2
##
  file
                                      n
##
  <chr>
                                  <int>
##
   1 /g11ELApfiles18 sim.csv
                                    453
##
   2 /q11Mathpfiles18 sim.csv
                                    460
   3 /g11Rdgpfiles18 sim.csv
##
                                    453
   4 /gllSciencepfiles18 sim.csv
##
                                    438
   5 /g11Wripfiles18 sim.csv
##
                                    453
   6 /g3ELApfiles18 sim.csv
##
                                    540
   7 /q3Mathpfiles18 sim.csv
##
                                    536
```

# Pull out pieces you need

- Regular expressions are most powerful here
  - We haven't talked about them much
- Try RegExplain

# Pull grade

 Note – I'm not expecting you to just suddenly be able to do this. This is more for illustration. There's also other ways you could extract the same info

# parse\_number

 In this case parse\_number also works – but note that it would not work to extract the year

```
batch2 %>%
mutate(grade = parse_number(file)) %>%
    select(file, grade)
```

```
## # A tibble: 15,945 x 2
##
     file
                              grade
## <chr>
                              <dbl>
   1 /q11ELApfiles18 sim.csv
                                 11
## 2 /q11ELApfiles18 sim.csv
                                 11
##
   3 /q11ELApfiles18 sim.csv
                                 11
##
   4 /q11ELApfiles18 sim.csv
                                 11
##
    5 /q11ELApfiles18 sim.csv
                                 11
##
   6 /q11ELApfiles18 sim.csv
                                 11
## 7 /q11ELApfiles18 sim.csv
                                 11
## 8 /q11ELApfiles18 sim.csv
                                 11
    9 /q11ELApfiles18 sim.csv
                                 11
## 10 /q11ELApfiles18 sim.csv
                                 11
## # ... with 15,935 more rows
```

# Extract year

 In this case parse\_number also works – but note that it would not work to extract the year

```
batch2 %>%
    mutate(
        grade = str_replace_all(
            file, "/g(\\d?\\d).+", "\\1"
        ),

    year = str_replace_all(
        file, ".+files(\\d\\d)_sim.+", "\\1"
        )

        %>%
        select(file, grade, year)
```

```
## # A tibble: 15,945 x 3
##
      file
                              grade year
##
   <chr>
                              <chr> <chr>
    1 /q11ELApfiles18 sim.csv 11
                                     18
##
   2 /q11ELApfiles18 sim.csv 11
                                    18
##
   3 /g11ELApfiles18 sim.csv 11
                                    18
## 4 /q11ELApfiles18 sim.csv 11
                                    18
    5 /q11ELApfiles18 sim.csv 11
                                    18
    6 /q11ELApfiles18 sim.csv 11
##
                                     18
```

### Extract Content Area

```
## # A tibble: 15,945 x 4
##
     file
                              grade year content
##
   <chr>
                              <chr> <chr> <chr>
##
   1 /q11ELApfiles18 sim.csv 11 18
                                          ELA
##
   2 /q11ELApfiles18 sim.csv 11
                                 18
                                          ELA
##
                                   18
   3 /q11ELApfiles18 sim.csv 11
                                          ELA
##
   4 /q11ELApfiles18 sim.csv 11
                                   18
                                          ELA
   5 /q11ELApfiles18 sim.csv 11
##
                                    18
                                          ELA
##
                                    18
   6 /q11ELApfiles18 sim.csv 11
                                          ELA
## 7 /q11ELApfiles18 sim.csv 11
                                   18
                                          ELA
   8 /g11ELApfiles18 sim.csv 11
##
                                   18
                                          ELA
   9 /q11ELApfiles18 sim.csv 11
                                    18
                                          ELA
## 10 /q11ELApfiles18 sim.csv 11
                                    18
                                          ELA
```

# Double checks: grade

```
## # A tibble: 7 x 2
## grade n
## <chr> <int>
## 1 11 2257
## 2 3 2156
## 3 4 2341
## 4 5 2632
## 5 6 2216
## 6 7 1962
## 7 8 2381
```

# Double checks: year

```
## # A tibble: 1 x 2
## year n
## <chr> <int>
## 1 18 15945
```

### Double checks: content

```
## # A tibble: 5 x 2
## content n
## <chr> <int>
## 1 ELA 3627
## 2 Math 3629
## 3 Rdg 3627
## 4 Science 1435
## 5 Wri 3627
```

# Finalize

```
d <- batch2 %>%
  mutate(grade = str_replace_all(file, "/g(\\d?\\d).+", "\\1")
        grade = as.integer(grade),
        year = str_replace_all(file, ".+files(\\d\\d)_sim.+",
        year = as.integer(grade),
        content = str_replace_all(file, "/g\\d?\\d(.+)pfiles.-
        select(-file) %>%
        select(ssid, grade, year, content, testeventid, asmtprmrydsbiasmtscndrydsbltycd, Entry:WMLE)
```

### Final product

- In this case, we basically have a tidy data frame already!
- We've reduced our problem from 31 files to a single file

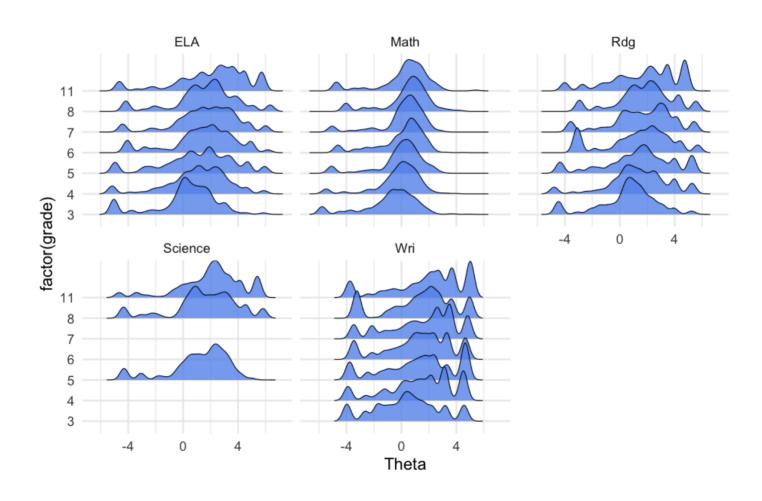
d

## #

WMLE <dbl>

```
## # A tibble: 15,945 x 25
##
         ssid grade year content testeventid asmtprmrydsbltycd asmtscndry
##
    <dbl> <int> <int> <chr>
                                      <dbl>
                                                      <dbl>
## 1 9466908
                11
                      11 ELA
                                     148933
                                                          0
##
   2 7683685 11 11 ELA
                                     147875
                                                         10
  3 9025693 11 11 ELA
##
                                    143699
                                                         40
##
  4 10099824 11 11 ELA
                                                         82
                                    143962
##
   5 18886078 11 11 ELA
                                    150680
                                                         10
   6 10606750 11 11 ELA
##
                                    144583
                                                         80
## 7 10541306 11 11 ELA
## 8 7632967 11 11 ELA
                                    145204
                                                         50
                                    148926
                                                         10
##
   9 7661118 11
                      11 ELA
                                    148893
                                                         50
             11
                                                         82
## 10 10547177
                      11 ELA
                                     144583
## # ... with 15,935 more rows, and 17 more variables: Theta <dbl>, Status <d
## # Count <dbl>, RawScore <dbl>, SE <dbl>, Infit <dbl>, Infit Z <dbl>, C
## # Outfit Z <dbl>, Displacement <dbl>, PointMeasureCorr <dbl>, Weight <
## # ObservMatch <dbl>, ExpectMatch <dbl>, PointMeasureExpected <dbl>, RM
```

### Quick look at distributions



## Summary stats

```
## # Groups:
             grade [7]
##
     grade asmtprmrydsbltycd
                                                        Rdq
                                                                   Wri
                                   ELA
                                             Math
##
     <int>
                      <dbl>
                                 <dbl>
                                             <dbl>
                                                       <dbl>
                                                                 <dbl>
##
                          0 -0.073610 -1.21055
                                                   1.010455
                                                             1.612308
                         10 0.3700416 -0.8182091 0.5184354 0.3206475
##
   3
##
                         20 -0.06335000 -1.251400 1.52
                                                        -0.5775
         3
3
3
3
3
##
                         40 -1.877683 -3.56365 -1.761667 -0.7514286
##
   5
                         50 0.9462857 -0.09186957 0.9791176 1.191481
##
   6
                         60 0.8407750 1.040375 2.181111 1.067
##
   7
                         70 -1.104049
                                       -1.517955 -0.8454839 -1.005625
##
   8
                         74 0.996 0.02083750 0.6
                                                        1.2925
         3
##
                         80 -0.144304 -0.5325596 0.6791667 0.2686301
## 10
                         82 0.3708244 -1.080988 0.5676650
                                                             0.3440741
## # ... with 67 more rows
```

# Backing up a bit

What if we wanted only math files?

```
dir_ls(here::here("data", "pfiles_sim"), regexp = "Math")

## /Users/daniel/Teaching/data_sci_specialization/2020-21/c3-fp-2021/data/p
```

# Only Grade 5

#### You try

```
dir_ls(here::here("data", "pfiles_sim"), regexp = "g5")

## /Users/daniel/Teaching/data_sci_specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data_sci_specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data_sci_specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data_sci_specialization/2020-21/c3-fp-2021/data/p
## /Users/daniel/Teaching/data_sci_specialization/2020-21/c3-fp-2021/data/p
```

### Base equivalents

#### list.files(here::here("data", "pfiles\_sim"))

```
[1] "g11ELApfiles18 sim.csv"
                                     "q11Mathpfiles18 sim.csv"
## [3] "q11Rdqpfiles18 sim.csv"
                                     "gl1Sciencepfiles18 sim.csv"
   [5] "q11Wripfiles18 sim.csv"
                                     "q3ELApfiles18 sim.csv"
    [7] "g3Mathpfiles18 sim.csv"
                                      "q3Rdqpfiles18 sim.csv"
## [9] "g3Wripfiles18 sim.csv"
                                      "g4ELApfiles18 sim.csv"
## [11] "g4Mathpfiles18 sim.csv"
                                      "q4Rdqpfiles18 sim.csv"
## [13] "g4Wripfiles18 sim.csv"
                                     "g5ELApfiles18 sim.csv"
  [15] "g5Mathpfiles18 sim.csv"
                                     "q5Rdqpfiles18 sim.csv"
  [17] "g5Sciencepfiles18 sim.csv"
                                      "q5Wripfiles18 sim.csv"
## [19] "g6ELApfiles18 sim.csv"
                                      "g6Mathpfiles18 sim.csv"
## [21] "g6Rdgpfiles18 sim.csv"
                                     "g6Wripfiles18 sim.csv"
                                     "q7Mathpfiles18 sim.csv"
## [23] "g7ELApfiles18 sim.csv"
                                     "g7Wripfiles18 sim.csv"
## [25] "g7Rdgpfiles18 sim.csv"
## [27] "g8ELApfiles18 sim.csv"
                                     "g8Mathpfiles18 sim.csv"
                                      "g8Sciencepfiles18 sim.csv"
## [29] "g8Rdgpfiles18 sim.csv"
## [31] "g8Wripfiles18 sim.csv"
```

### Full path

list.files(here::here("data", "pfiles\_sim"), full.names = TRUE)

```
[1] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
##
    [2] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
    [3] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
##
    [4] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
    [5] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
    [6] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
##
    [7] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
    [8] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
##
    [9] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
   [10] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [11] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [12] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
  [13] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
  [14] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
##
  [15] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [16] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [17] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
  [18] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
## [19] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
  [20] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [21] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [22] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [23] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [24] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
```

### Only csvs

```
list.files(here::here("data", "pfiles_sim"),
            full.names = TRUE,
            pattern = "*.csv")
    [1] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
##
    [2] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
    [3] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
    [4] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
##
    [5] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
    [6] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
    [7] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
    [8] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
    [9] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
   [10] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
##
## [11] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [12] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [13] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [14] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [15] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [16] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [17] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [18] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
## [19] "/Users/daniel/Teaching/data sci specialization/2020-21/c3-fp-2021/
```

## [20] "/Users/daniel/Teaching/data\_sci\_specialization/2020-21/c3-fp-2021/
## [21] "/Users/daniel/Teaching/data\_sci\_specialization/2020-21/c3-fp-2021/
## [22] "/Users/daniel/Teaching/data\_sci\_specialization/2020-21/c3-fp-2021/

# Why not use base?

We could, but {fs} plays a little nicer with {purrr}

```
files <- list.files(
  here::here("data", "pfiles_sim"),
  pattern = "*.csv"
)
batch3 <- map_dfr(files, read_csv, .id = "file")</pre>
```

## Error: 'g11ELApfiles18\_sim.csv' does not exist in current working direct

Need to return full names

#### files

```
## [1] "g11ELApfiles18 sim.csv"
                                     "g11Mathpfiles18 sim.csv"
                                     "g11Sciencepfiles18_sim.csv"
## [3] "g11Rdgpfiles18 sim.csv"
## [5] "g11Wripfiles18 sim.csv"
                                     "g3ELApfiles18 sim.csv"
                                     "g3Rdgpfiles18 sim.csv"
## [7] "q3Mathpfiles18 sim.csv"
## [9] "g3Wripfiles18_sim.csv"
                                     "g4ELApfiles18 sim.csv"
## [11] "g4Mathpfiles18_sim.csv"
                                     "g4Rdgpfiles18 sim.csv"
                                     "g5ELApfiles18_sim.csv"
## [13] "g4Wripfiles18 sim.csv"
## [15] "g5Mathpfiles18 sim.csv"
                                     "g5Rdgpfiles18 sim.csv"
```

# Iry again

```
files <- list.files(here::here("data", "pfiles_sim"),</pre>
                  pattern = "*.csv",
                  full.names = TRUE)
batch3 <- map_dfr(files, read_csv, .id = "file")</pre>
batch3
## # A tibble: 15,945 x 23
##
     file Entry Theta Status Count RawScore SE Infit Infit Z Outfi
##
     <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> 
##
   1 1
       123 1.2687
                           1 36 23 0.3713 0.93 -0.34 0.82
                           1 36
##
   2 1
          88 1.5541
                                       25 0.3852 0.95 -0.37 0.81
       105 3.2773 1 36 33 0.6187 0.9 -0.04 1.6300
## 3 1
```

## 4 1 153 4.4752 1 36 35 1.0234 0.93 0.23 0.35 ## 5 1 437 2.6655 1 36 31 0.5008 0.92 -0.18 0.88 ## 6 1 307 5.7137 0 36 36 1.8371 1 0 1 1 36 34 0.7408 1.06 0.31 0.86 ## 7 1 305 3.7326 18 0.36 1.55 2.56 1.74 1 36 ## 8 1 42 0.609 59 -2.623 ## 9 1 1 36 3 1.0344 0.85 0.06 0.17 ## 10 1 304 5.7137 0 36 36 1.8371 1 0 ## # ... with 15,935 more rows, and 12 more variables: Displacement <dbl>, ## # PointMeasureCorr <dbl>, Weight <dbl>, ObservMatch <dbl>, ExpectMatch ## # PointMeasureExpected <dbl>, RMSR <dbl>, WMLE <dbl>, testeventid <dbl ## # asmtprmrydsbltycd <dbl>, asmtscndrydsbltycd <dbl>

<db]

#### indexes

• The prior example gave us indexes, rather than the file path. Why?

#### No names

#### names(files)

## NULL

• We **need** the file path! An index isn't nearly as useful.

#### Base method that works

## My recommendation

- If you're working interactively, no reason not to use {fs}
- If you are building functions that take generic paths, might be worth considering skipping the dependency

#### Note

I am **not** saying skip it, but rather that you should **consider** whether it is really needed or not.

# List columns

# Comparing models

Let's say we wanted to fit/compare a set of models for each content area

1. lm(Theta ~ asmtprmrydsbltycd)
2. lm(Theta ~ asmtprmrydsbltycd +
 asmtscndrydsbltycd)
3. lm(Theta ~ asmtprmrydsbltycd +
 asmtscndrydsbltycd +
 asmtprmrydsbltycd:asmtscndrydsbltycd)

### Split the data

##

##

##

..\$ Outfit

..\$ Outfit Z

..\$ Displacement

The base method we've been using...

```
splt_content <- split(d, d$content)</pre>
str(splt content)
## List of 5
##
    $ ELA : tibble[,25] [3,627 \times 25] (S3: tbl df/tbl/data.frame)
##
     ..$ ssid
                              : num [1:3627] 9466908 7683685 9025693 1009982
##
                             : int [1:3627] 11 11 11 11 11 11 11 11 11 11 .
    ..$ grade
##
                              : int [1:3627] 11 11 11 11 11 11 11 11 11 11 .
    ..$ year
##
    ..$ content
                              : chr [1:3627] "ELA" "ELA" "ELA" "ELA" ...
##
                              : num [1:3627] 148933 147875 143699 143962 150
    ..$ testeventid
                              : num [1:3627] 0 10 40 82 10 80 50 10 50 82 ..
##
    ..$ asmtprmrydsbltycd
##
     ..$ asmtscndrydsbltycd
                             : num [1:3627] 0 0 20 0 0 80 0 0 0 0 ...
                              : num [1:3627] 123 88 105 153 437 307 305 42 5
##
     ..$ Entry
##
    ..$ Theta
                              : num [1:3627] 1.27 1.55 3.28 4.48 2.67 ...
##
    ..$ Status
                              : num [1:3627] 1 1 1 1 1 0 1 1 1 0 ...
##
                              : num [1:3627] 36 36 36 36 36 36 36 36 36 36 .
    ..$ Count
##
     ..$ RawScore
                              : num [1:3627] 23 25 33 35 31 36 34 18 3 36 ...
##
                              : num [1:3627] 0.371 0.385 0.619 1.023 0.501 .
     ..$ SE
##
     ..$ Infit
                              : num [1:3627] 0.93 0.95 0.9 0.93 0.92 1 1.06
##
                               num [1:3627] -0.34 -0.37 -0.04 0.23 -0.18 0
     ..$ Infit Z
```

: num [1:3627] 0.82 0.81 1.63 0.35 0.88 1 0.86

: num [1:3627] -0.62 -0.56 1.03 -0.16 -0.12 0

: num [1:3627] 0.0018 0.0019 0.0022 0.0023 0.0

#### We could use this method

• We could then go through and conduct tests to see which model had better fit indices, etc.

#### Alternative

Create a data frame with a list column.

```
d %>%
nest(-content)
```

#### Add model list column

```
mods <- d %>%
    nest(-content) %>%
    mutate(
      m1 = map(
        data,
        ~lm(Theta ~ asmtprmrydsbltycd,data = .x)
    ),
      m2 = map(
        data,
        ~lm(Theta ~ asmtprmrydsbltycd + asmtscndrydsbltycd,
            data = .x)
      m3 = map(
        data, ~lm(Theta ~ asmtprmrydsbltycd * asmtscndrydsbltycd
                  data = .x)
```

#### mods

#### Part of the benefit

#### It's a normal data frame!

```
mods %>%
    pivot_longer(
        m1:m3,
        names_to = "model",
        values_to = "output"
)
```

```
## # A tibble: 15 x 4
## content data
                                        model output
##
  <chr> <chr>
                                        <chr> <list>
## 1 ELA <tibble[,24] [3,627 × 24]> m1
                                              <1m>
## 2 ELA <tibble[,24] [3,627 × 24] > m2 <lm>
   3 ELA <tibble[,24] [3,627 \times 24]> m3
##
                                              <1m>
\#\# 4 Math <tibble[,24] [3,629 × 24]> m1
                                              <1m>
##
   5 Math <tibble[,24] [3,629 \times 24]> m2
                                              < lm >
##
   6 Math <tibble[,24] [3,629 \times 24]> m3
                                              <1m>
   7 Rdg <tibble[,24] [3,627 \times 24]> m1
##
                                              <1m>
   8 Rdq <tibble[,24] [3,627 \times 24]> m2
##
                                              <1m>
## 9 Rdg <tibble[,24] [3,627 × 24]> m3
                                              <1m>
## 10 Science <tibble[,24] [1,435 × 24] > m1
                                              < lm >
  11 Science <tibble[,24] [1,435 × 24] > m2
                                              <1m>
## 12 Science <tibble[,24] [1,435 \times 24] > m3
                                              <lm>
```

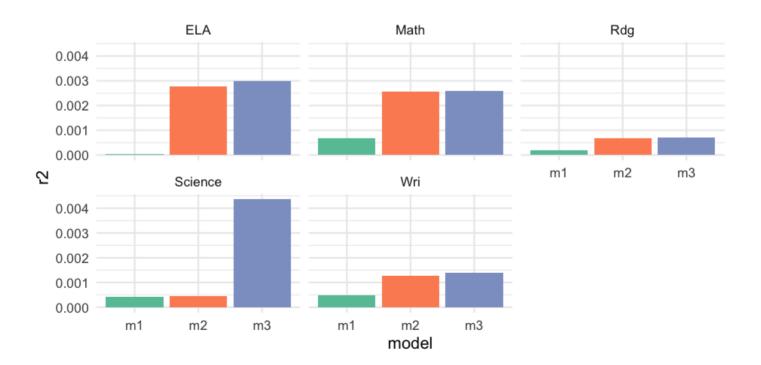
### Extract all $R^2$

```
r2 <- mods %>%
    pivot_longer(
        m1:m3,
        names_to = "model",
        values_to = "output"
) %>%
    mutate(r2 = map_dbl(output, ~summary(.x)$r.squared))
r2
```

```
## # A tibble: 15 x 5
##
   content data
                                             model output
                                                                        r2
                                             <chr> <list>
##
     <chr> <chr> 
                                                                     <dbl>
##
    1 ELA <tibble[,24] [3,627 × 24] > m1
                                                   <lm> 0.00002742625
##
    2 ELA <tibble[,24] [3,627 × 24] > m2 <lm> 0.002784211
           <tibble[,24] [3,627 \times 24]> m3
##
    3 ELA
                                                   < lm >
                                                           0.002994548
            <tibble[,24] [3,629 × 24]> m1
                                                            0.0006718361
##
    4 Math
                                                    < lm >
              <tibble[,24] [3,629 × 24]> m2
##
    5 Math
                                                    < lm >
                                                            0.002575408
##
    6 Math <tibble[,24] [3,629 \times 24]> m3
                                                    < lm >
                                                            0.002586228
##
    7 Rdq <tibble[,24] [3,627 \times 24]> m1
                                                    < lm >
                                                            0.0001925962
    8 Rdg <tibble[,24] [3,627 \times 24]> m2
                                                    < lm >
##
                                                            0.0006773540
           <tibble[,24] [3,627 × 24]> m3
##
    9 Rdq
                                                    < lm >
                                                           0.0007050212
  10 Science <tibble[,24] [1,435 × 24]> m1
                                                    < lm >
                                                           0.0004085780
##
   11 Science \langle \text{tibble}[,24] [1,435 \times 24] \rangle \text{ m2}
                                                    <1m>
                                                           0.0004520424
   12 Science \langle \text{tibble}[,24] [1,435 \times 24] \rangle \text{ m}
                                                    <1m>
                                                            0.004354060
##
  13 Wri
               \langle \text{tibble}[,24] [3,627 \times 24] \rangle \text{ m1}
                                                    < lm >
                                                            0.0004902093
```

#### Plot

```
ggplot(r2, aes(model, r2)) +
    geom_col(aes(fill = model)) +
    facet_wrap(~content) +
    guides(fill = "none") +
    scale_fill_brewer(palette = "Set2")
```



# Summary

- Batch processing is really powerful
- Much of the tools we've learned in the past can be applied once we get the data in a more workable format
- List columns are also **really** nice for organization and using our data frame toolkit

# Next time

- We'll talk more about list columns
- We'll also talk more about the new rowwise() and nest\_by() functions