Applied Data Science with R: Data Transformation

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Last Weeks Homework

Quiz question: Symptoms of messy data

Which of the following is not a symptom of messy data?

Multiple values are stored in one column

Tidy up dataset: Read data

Tidy up dataset: Look at data

A tibble: 6 x 59

head(mbta)

```
##
     mode `2007-01` `2007-02` `2007-03` `2007-04` `2007-05` `2007-06`
## <chr> <chr> <chr> <chr> <dbl> <chr> <dbl> <chr>
## 1 All ~ NA
                        NA 1188. NA
                                                            NA
                                                                             1246.
## 2 Boat 4 3.6 40 4.3 4.9 5.8
## 3 Bus 335.819 338.675 340.352.162 354.367 351.
## 4 Comm~ 142.2 138.5 138. 139.5 139 143
## 5 Heav~ 435.294 448.271 459. 472.201 474.579 477.
## 6 Ligh~ 227.231 240.262 241. 255.557 248.262 246.
## # ... with 52 more variables: `2007-07` <chr>, `2007-08` <chr>,
## # '2007-09' <dbl>, '2007-10' <chr>, '2007-11' <chr>, '2007-11' <chr>, '2007-12' <db
## # '2008-01' <chr>, '2008-02' <chr>, '2008-03' <dbl>, '2008-04' <ch
## #
       2008-05 <chr>, 2008-06 <dbl>, 2008-07 <chr>, 2008-08 <chr>, 2008-08 <chr>, 2008-08 <chr>, 2008-08 <chr>, 2008-08 <chr>, 2008-09 <chr>, 2008-08 <chr>, 2008-09 <chr
## #
       `2008-09` <dbl>, `2008-10` <chr>, `2008-11` <chr>, `2008-12` <db
## # '2009-01' <chr>, '2009-02' <chr>, '2009-03' <dbl>, '2009-04' <ch
## #
       `2009-05` <chr>, `2009-06` <dbl>, `2009-07` <chr>, `2009-08` &ch
```

Tidy up dataset: Gather years

```
## # A tibble: 638 x 3
## mode
                    year passengers
## <chr>
                   <chr> <chr>
   1 All Modes by Qtr 2007-01 NA
##
   2 Boat
                 2007-01 4
##
## 3 Bus
                 2007-01 335 819
## 4 Commuter Rail 2007-01 142.2
   5 Heavy Rail 2007-01 435.294
##
   6 Light Rail 2007-01 227.231
##
## 7 Pct Chg / Yr 2007-01 0.02
## 8 Private Bus 2007-01 4.772
                    2007-01 4.9
##
   9 RIDE
```

Tidy up dataset: Separate year

```
mbta_tidy <- mbta_tidy %>%
   tidyr::separate(year, into = c("year", "month"))
mbta_tidy
```

```
## # A tibble: 638 x 4
##
    mode
                    year month passengers
## <chr>
                    <chr> <chr> <chr>
##
   1 All Modes by Qtr 2007 01
                              NΑ
##
   2 Boat
                    2007 01
                              4
##
   3 Bus
                    2007 01 335.819
## 4 Commuter Rail 2007 01 142.2
##
   5 Heavy Rail 2007 01 435.294
   6 Light Rail 2007 01 227.231
##
   7 Pct Chg / Yr 2007
                              0.02
##
                         01
   8 Private Bus
                    2007
                         01 4.772
##
   9 RIDE
                    2007 01
                              4.9
##
## 10 Trackless Trolley 2007 01 12.757
## # ... with 628 more rows
```

Tidy up dataset: Spread mode of transportation

7 2007

8 2007

07

08

NA

NA

```
mbta_tidy <- mbta_tidy %>%
  tidyr::spread(mode, passengers)
mbta_tidy
```

```
## # A tibble: 58 x 13
##
     year month `All Modes by Q~ Boat Bus `Commuter Rail` `Heavy R
     <chr> <chr> <chr>
##
                                <chr> <chr> <chr>
                                                         <chr>
   1 2007 01
                                     335.~ 142.2
                                                         435.294
##
                NA
                                4
##
   2 2007 02 NA
                                3.6 338.~ 138.5
                                                         448.271
##
   3 2007 03 1187.653
                                40
                                     339.~ 137.7
                                                         458.583
                                4.3 352.~ 139.5
##
   4 2007 04
               NA
                                                         472,201
##
   5 2007 05
                NΑ
                                4.9 354.~ 139
                                                         474.579
             1245.959
                                5.8 350.~ 143
                                                         477.032
##
   6 2007 06
```

9 2007 09 1256.571 5.469 372.~ 143.051 499.566 ## 10 2007 10 NA 5.145 368.~ 146.542 457.741 ## # ... with 48 more rows, and 6 more variables: `Light Rail` <chr>, 8 \cdot \cdot

6.521 357.~ 142.391

6.572 355.~ 142.364

471.735

461.605

Tidy up dataset: Keep wanted columns

```
mbta_tidy <- mbta_tidy %>%
    .[,c(1:2,6:8)]
mbta_tidy
```

```
## # A tibble: 58 x 5
##
           month 'Commuter Rail' 'Heavy Rail' 'Light Rail'
##
     <chr> <chr> <chr>
                                <chr>
                                             <chr>>
   1 2007 01
                142.2
                                435.294
                                             227.231
##
##
   2 2007 02 138.5
                                448.271
                                             240.262
##
   3 2007 03 137.7
                                458.583
                                             241,444
##
   4 2007 04 139.5
                                472,201
                                             255.557
##
   5 2007 05
                139
                                474.579
                                             248.262
                                477.032
                                             246.108
##
   6 2007 06
              143
   7 2007
           07
                142.391
                                471.735
                                             243.286
##
##
   8 2007 08
                 142.364
                                461.605
                                             234,907
   9 2007
                143.051
                                499.566
                                             265.748
##
           09
##
  10 2007 10
                 146.542
                                457.741
                                             241.434
## # ... with 48 more rows
```

Tidy up dataset: Gather rail modes

```
## # A tibble: 174 \times 4
##
     year month rail_type passengers
     <chr> <chr> <chr>
##
                           <chr>
##
   1 2007 01 Commuter Rail 142.2
##
   2 2007 02 Commuter Rail 138.5
##
   3 2007 03
               Commuter Rail 137.7
##
   4 2007 04
                 Commuter Rail 139.5
##
   5 2007 05
                 Commuter Rail 139
   6 2007 06
                 Commuter Rail 143
##
   7 2007 07
                 Commuter Rail 142.391
##
   8 2007 08
                 Commuter Rail 142.364
##
##
   9 2007 09
                 Commuter Rail 143.051
  10 2007
           10
                 Commuter Rail 146.542
```

Tidy up dataset: Compute sum

<dbl>

49859.

##

1

```
mbta_tidy <- mbta_tidy %>%
  dplyr::mutate(passengers = as.numeric(passengers)) %>%
  dplyr::summarise(sum(passengers))
mbta_tidy

## # A tibble: 1 x 1
## `sum(passengers)`
```

Week 4: Data transformation

Prerequisites

Packages

library(tidyverse)

Data

336,776 flights that departed from New York City in 2013

```
# install.packages("nycflights13")
library(nycflights13)
```

year	month	day	dep_time	sched_dep_time	dep_delay
2013	1	1	517	515	2
2013	1	1	533	529	4
2013	1	1	542	540	2
2013	1	1	544	545	-1

Data transformation with dplyr

Variable types

- int: integers
- dbl: doubles, or real numbers
- chr: character vectors, or strings
- dttm: date-times (a date + a time)
- Igl: logical, vectors that contain only TRUE or FALSE
- fctr: factors
- date: dates

dplyr core functions

- filter(): select rows by their values
- arrange(): order rows
- select(): select columns by their names
- mutate(): create new variables
- summarize(): collapse many values down to a single summary
- group_by(): operate on it group-by-group
- rename(): rename columns
- distinct(): find distinct rows

Command structure (for all dplyr verbs):

- first argument is a data frame
- return value is a data frame
- nothing is modified in place

filter()

filter() allows to subset observations based on their values. The function takes logical expressions and returns the rows for which all are TRUE.



filter()

Let's select all flights on January 1st:

```
filter(flights, month == 1, day == 1)
```

year	month	day	dep_time	sched_dep_time	dep_delay
2013	1	1	517	515	2
2013	1	1	533	529	4
2013	1	1	542	540	2
2013	1	1	544	545	-1
2013	1	1	554	600	-6
2013	1	1	554	558	-4

filter()

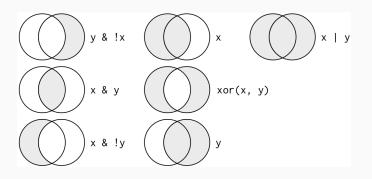
filter() revolves around using comparison operators: >, >=, <, <=,
!= (not equal), and == (equal).</pre>

dplyr functions like filter() never modify inputs but instead return a new data frame that needs to be assigned to an object if you want to save the result.

```
jan1 <- filter(flights, month == 1, day == 1)</pre>
```

Boolean operators

filter() also supports the Boolean operators & ("and"), | ("or"),
! (is "not"), and xor (exclusive "or".



De Morgan's law: !(x & y) = !x | !y & !(x | y) = !x & !y

Boolean operators

Why does this not work?

```
filter(flights, month == 11 | 12)
```

Generally a good idea to use x %in% y, which will select every row where x is part of the values of y.

```
filter(flights, month %in% c(11, 12))
```

between condition

Another useful dplyr filtering helper is between(). between(x, left, right) is equivalent to $x \ge 1$ left & $x \le 1$ right.

To filter() all flights that departed between midnight and 6am (inclusive):

```
filter(flights, between(dep_time, 0, 600))
```

filter() exclusion

filter() by default excludes FALSE and NA values.

filter() exclusion

If you want to preserve missing values, you have to explicitly state it.

filter() exercises

1. Find all flights that

- 1.1 Had an arrival delay of two or more hours.
- 1.2 Arrived more than two hours late, but didn't leave late.
- 1.3 Flew to Houston (IAH or HOU).
- 1.4 Were operated by United, American, or Delta.
- 1.5 Departed in summer (July, August, and September).

filter() exercises solutions

Arrival delay of two or more hours

```
flights %>%
filter(arr_delay > 120)
```

Arrived more than two hours late, but didn't leave late

```
flights %>%
filter(!is.na(dep_delay), dep_delay <= 0, arr_delay > 120)
```

Flew to Houston (IAH or HOU)

```
flights %>%
filter(dest %in% c("IAH", "HOU"))
```

filter() exercises solutions

Were operated by United, American, or Delta

```
flights %>%
filter(carrier %in% c("AA", "DL", "UA"))
```

Departed in summer (July, August, and September)

```
flights %>%
filter(between(month, 7, 9))
```

arrange()

arrange() takes a data frame and a set of column names to order the rows by. Multiple column names are evaluated subsequently.

arrange(flights, year, month, day)

year	month	day	dep_time	sched_dep_time	dep_delay
2013	1	1	517	515	2
2013	1	1	533	529	4
2013	1	1	542	540	2
2013	1	1	544	545	-1
2013	1	1	554	600	-6
2013	1	1	554	558	-4

arrange() in descending order

By dafault arrange() sorts values in ascending order. Use desc() to re-order by a column in descending order.

arrange(flights, desc(arr_delay))

year	month	day	dep_time	sched_dep_time	dep_delay
2013	1	9	641	900	1301
2013	6	15	1432	1935	1137
2013	1	10	1121	1635	1126
2013	9	20	1139	1845	1014
2013	7	22	845	1600	1005
2013	4	10	1100	1900	960

arrange() exercises

2. Sort flights to

- 2.1 find the flight that departed the earliest (earlier than scheduled)
- 2.2 find the most delayed flight.
- 2.3 find the flight that travelled the longest and that travelled the shortest distance.

arrange() exercises solutions

Flight B6 97 left the earliest (43 minutes early).

```
flights %>%
arrange(dep_delay)
```

Flight HA 51 is the most delayed flight (1301 minutes delay).

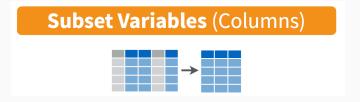
```
flights %>%
arrange(desc(dep_delay))
```

Flight HA 51 is the longest (4,983 miles), flight US 1632 is the shortest (17 miles).

```
flights %>%
arrange(desc(distance))
```

select()

select() is used to select a subset of variables from a dataset.



select(flights, year, month, day)

year	month	day
2013	1	1
2013	1	1
2013	1	1
2013	1	1

select()

select() has various helper functions:

- everything(): selects all variables.
- starts_with("abc"): matches names that begin with "abc".
- ends_with("xyz"): matches names that end with "xyz".
- contains("ijk"): matches names that contain "ijk".
- matches("(.)\\1"): selects variables that match a regular expression.
- num_range("x", 1:3) matches x1, x2 and x3.

See ?select for more details.

select()

You can use select() to rename variables

```
select(flights, tail_num = tailnum)
```

which will drop all of the variables not explicitly mentioned. Therefore it's better to use rename() instead:

```
rename(flights, tail_num = tailnum)
```

select() exercises

- 3.1 What are three distinct ways to select dep_time, dep_delay, arr_time, and arr_delay from flights.
- 3.2 What does the one_of() function do? Why might it be helpful in conjunction with this vector?

select() exercise solutions

Three distinct ways to select variables.

```
select(flights, dep time, dep delay, arr time, arr delay)
select(flights, starts_with("dep_"), starts_with("arr "))
select(flights, matches("^(dep|arr)_(time|delay)$"))
What does the one of () function do?
vars <- c("year", "month", "day", "dep delay",</pre>
          "arr delay")
select(flights, one_of(vars))
```

mutate()

mutate() allows to add new columns to the end of your dataset that are functions of existing columns.



mutate()

dep_delay	arr_delay	distance	air_time	gain	speed
2	11	1400	227	9	370.0441
4	20	1416	227	16	374.2731
2	33	1089	160	31	408.3750
-1	-18	1576	183	-17	516.7213
-6	-25	762	116	-19	394.1379
-4	12	719	150	16	287.6000

transmute()

Use transmute() to only keep the new variables:

```
transmute(flights,
  gain = arr_delay - dep_delay,
  hours = air_time / 60,
  gain_per_hour = gain / hours
)
```

Functions to use with mutate()

There are many functions for creating new variables with mutate():

- Arithmetic operators: +, -, *, /, ^ (e.g. air_time / 60).
- Aggregate functions: sum(x) mean(y) (e.g. mean(dep_delay)).
- Modular arithmetic: %/% (integer division) and %% (remainder), where x == y * (x %/% y) + (x %% y).
- Logs: log(), log2(), log10().
- Offsets: lead() and lag() (e.g. x lag(x)).
- Cumulative and rolling aggregates: cumsum(), cumprod(), cummin(), cummax(), cummean().
- Logical comparisons, <, <=, >, >=, !=.
- Ranking: min_rank(), row_number(), dense_rank(),
 percent_rank(), cume_dist(), ntile().

mutate() exercises

4. Use mutate() to

- 4.1 Create new variables for dep_time and sched_dep_time that measure time in the number of minutes after midnight.
- 4.2 Compare air_time with arr_time dep_time. What do you see? What do you need to do to fix it?
- 4.3 Find the 10 most delayed flights using a ranking function. How do you want to handle ties?

mutate() exercise solutions

New variables for dep_time and sched_dep_time

mutate() exercise solutions

arr_time and dep_time may be in different time zones.

10 most delayed flights

```
flights %>%
  mutate(dep_delay_rank = min_rank(-dep_delay)) %>%
  arrange(dep_delay_rank) %>%
  filter(dep_delay_rank <= 10)</pre>
```

summarize()

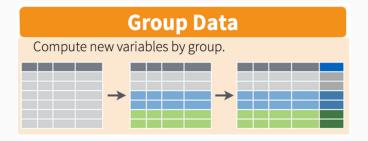
summarize() collapses a data frame to a single row.



```
summarise(flights, delay = mean(dep_delay, na.rm = TRUE))
## # A tibble: 1 x 1
## delay
## <dbl>
## 1 12.6
```

summarize() with group_by()

summarize() is most effectively used with group_by(), which changes the unit of analysis from the complete dataset to individual groups.



Grouping is most useful in conjunction with summarise(), but you can also do convenient operations with mutate() and filter().

summarize() with group_by()

For example, to get the average delay per date

```
flights %>%
  group_by(year, month, day) %>%
  summarise(delay = mean(dep_delay, na.rm = TRUE))
```

summarize() count

For aggregations it is generally a good idea to include a count n(). For example, let's look at the (not cancelled) planes that have the highest average delays:

```
flights %>%
  filter(!is.na(dep_delay), !is.na(arr_delay))
  group_by(tailnum) %>%
  summarise(delay = mean(arr_delay)) %>%
  arrange(delay)
```

summarize() useful functions

There are a number of useful summary functions:

- Measures of location: mean(x), sum(x), median(x).
- Measures of spread: sd(x), IQR(x), mad(x).
- Measures of rank: min(x), quantile(x, 0.25), max(x).
- Measures of position: first(x), nth(x, 2), last(x).
- Counts: n(), sum(!is.na(x)), n_distinct(x).
- Counts and proportions of logical values: sum(x > 10), mean(y == 0).

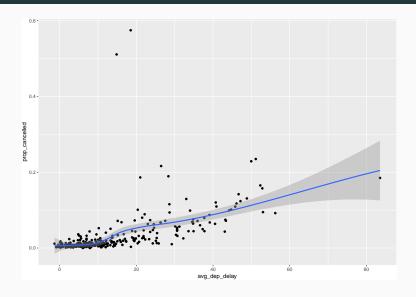
summarize() exercises

- 5. Use summarize() to
 - 5.1 Look at the number of cancelled flights per day. Is there a pattern? Is the proportion of cancelled flights related to the average delay?
 - 5.2 Find the carrier with the worst delays.

summarize() exercises solutions

Number of cancelled flights per day

summarize() exercises solutions



summarize() exercises solutions

Frontier Airlines (FL) has the worst delays.

```
flights %>%
  group_by(carrier) %>%
  summarise(arr_delay = mean(arr_delay, na.rm = TRUE)) %>%
  arrange(desc(arr_delay))
```

Homework Exercises

Homework Exercises

For this week's homework exersises go to Moodle and answer the Quiz posted in the Week 4: Data Transformation section.

Deadline: Tuesday, March 5.

That's it for today. Questions?