Manipulating Data in R

Introduction to R for Public Health Researchers

Reshaping Data

In this module, we will show you how to:

- 1. Reshaping data from wide (fat) to long (tall)
- 2. Reshaping data from long (tall) to wide (fat)
- 3. Merging Data
- 4. Perform operations by a grouping variable

Setup

We will show you how to do each operation in base R then show you how to use the dplyr or tidyr package to do the same operation (if applicable).

See the "Data Wrangling Cheat Sheet using dplyr and tidyr":

 https://www.rstudio.com/wp-content/uploads/2015/02/data-wranglingcheatsheet.pdf

What is wide/long data?

See http://www.cookbook-
r.com/Manipulating_data/Converting_data_between_wide_and_long_format/

- · Wide multiple columns per observation
 - e.g. visit1, visit2, visit3

Long - multiple rows per observation

| | id | visit | value |
|---|----|-------|-------|
| 1 | 1 | 1 | 10 |
| 2 | 1 | 2 | 4 |
| 3 | 1 | 3 | 3 |
| 4 | 2 | 1 | 5 |
| 5 | 2 | 2 | 6 |

What is wide/long data?

More accurately, data is wide or long with respect to certain variables.

Data used: Charm City Circulator

http://johnmuschelli.com/intro_to_r/data/Charm_City_Circulator_Ridership.csv

```
circ = read csv(
 paste0("http://johnmuschelli.com/intro to r/",
         "data/Charm City Circulator Ridership.csv"))
head(circ, 2)
# A tibble: 2 x 15
  day date orangeBoardings orangeAlightings orangeAverage
                  <dbl>
  <chr> <chr>
                                         <dbl>
                                                       <db1>
1 Mond... 01/1...
                          877
                                          1027
                                                          952
2 Tues... 01/1...
                          777
                                           815
                                                         796
# ... with 10 more variables: purpleBoardings <dbl>, purpleAlightings <dbl>,
  purpleAverage <dbl>, greenBoardings <dbl>, greenAlightings <dbl>,
 greenAverage <dbl>, bannerBoardings <dbl>, bannerAlightings <dbl>,
   bannerAverage <dbl>, daily <dbl>
class(circ$date)
[1] "character"
```

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Creating a Date class from a character date

```
library(lubridate) # great for dates!
sum(is.na(circ$date))
[1] 0
sum( circ$date == "")
[1] 0
circ = mutate(circ, date = mdy(date))
sum( is.na(circ$date) ) # all converted correctly
[1] 0
head(circ$date, 3)
[1] "2010-01-11" "2010-01-12" "2010-01-13"
class(circ$date)
[1] "Date"
```

Reshaping data from wide (fat) to long (tall): base R

The reshape command exists. It is a confusing function. Don't use it.

tidyr package

tidyr allows you to "tidy" your data. We will be talking about:

- · gather make multiple columns into variables, (wide to long)
- spread make a variable into multiple columns, (long to wide)
- separate string into multiple columns
- · unite multiple columns into one string

tidyr::gather - puts column data into rows.

We want the column names into "var" variable in the output dataset and the value in "number" variable. We then describe which columns we want to "gather:"

```
long = gather(circ, key = "var", value = "number",
            -dav, -date, -daily)
head(long, 4)
# A tibble: 4 x 5
 day date daily var
                                        number
 <chr> <date> <dbl> <chr>
                                         <dbl>>
1 Monday 2010-01-11 952 orangeBoardings
                                          877
2 Tuesday 2010-01-12 796 orangeBoardings
                                        777
3 Wednesday 2010-01-13 1212. orangeBoardings
                                         1203
4 Thursday 2010-01-14 1214. orangeBoardings
                                         1194
```

Could be explicit on what we want to gather

table(long\$var)

| bannerAlightings | bannerAverage | bannerBoardings | greenAlightings |
|------------------|------------------|------------------|-----------------|
| 1146 | 1146 | 1146 | 1146 |
| greenAverage | greenBoardings | orangeAlightings | orangeAverage |
| 1146 | 1146 | 1146 | 1146 |
| orangeBoardings | purpleAlightings | purpleAverage | purpleBoardings |
| 1146 | 1146 | 1146 | 1146 |

Lab Part 1

Website

Making a separator

We will use str_replace from stringr to put periods in the names (periods are **not** special when in a replacement)

```
long = long %>% mutate(
  var = var %>% str replace("Board", ".Board") %>%
    str replace ("Alight", ".Alight") %>%
    str replace("Average", ".Average")
table(long$var)
banner.Alightings
                    banner.Average banner.Boardings
                                                       green.Alightings
             1146
                               1146
                                                 1146
                                                                   1146
                    green.Boardings orange.Alightings orange.Average
    green.Average
             1146
                               1146
                                                 1146
                                                                   1146
 orange.Boardings purple.Alightings purple.Average
                                                       purple.Boardings
             1146
                               1146
                                                 1146
                                                                   1146
```

Now each var is boardings, averages, or alightings. We want to separate these so we can have these by line. Remember "." is special character:

```
long = separate(long, var, into = c("line", "type"),
            sep = "[.]")
head(long, 2)
# A tibble: 2 x 6
 day date daily line type number
 1 Monday 2010-01-11 952 orange Boardings 877
unique (long$line)
[1] "orange" "purple" "green" "banner"
unique (long$type)
[1] "Boardings" "Alightings" "Average"
```

Re-uniting all the lines

If we had the opposite problem, we could use the unite function:

We could also use paste/paste0.

Making column names a little more separated

Alternative: We could have replaced the column names first **then** reshaped:

```
cn = colnames(circ)
cn = cn %>%
  str_replace("Board", ".Board") %>%
  str_replace("Alight", ".Alight") %>%
  str_replace("Average", ".Average")
colnames(circ) = cn # then reshape using gather!
```

or (with some string working (see data cleaning)):

Lab Part 2

Website

Reshaping data from long (tall) to wide (fat): tidyr

In tidyr, the spread function spreads rows into columns. Now we have a long data set, but we want to separate the Average, Alightings and Boardings into different columns:

```
# have to remove missing days
wide = filter(long, !is.na(date))
wide = spread(wide, type, number)
head (wide)
# A tibble: 6 x 7
 day date daily line Alightings Average Boardings
 <chr> <date> <dbl> <chr>
                                                   <dbl>
                                  <dbl>
                                        <dbl>
1 Friday 2010-01-15 1644 banner
                                     NA
                                            NA
                                                     NA
2 Friday 2010-01-15 1644 green
                                  NA
                                            NA
                                                     NA
                                 1643 1644
3 Friday 2010-01-15 1644 orange
                                                   1645
4 Friday 2010-01-15 1644 purple
                                     NA
                                            NA
                                                     NA
5 Friday 2010-01-22 1394. banner
                                     NA
                                            NA
                                                     NA
6 Friday 2010-01-22 1394. green
                                     NA
                                            NA
                                                     NA
```

Lab Part 3

Website

Merging: Simple Data

Joining in dplyr

- Merging/joining data sets together usually on key variables, usually "id"
- · ?join see different types of joining for dplyr
- Let's look at https://www.rstudio.com/wp-content/uploads/2015/02/data-wrangling-cheatsheet.pdf
- inner join (x, y) only rows that match for x and y are kept
- full_join(x, y) all rows of x and y are kept
- left_join(x, y) all rows of x are kept even if not merged with y
- right_join(x, y) all rows of y are kept even if not merged with x

Inner Join

```
ij = inner_join(base, visits)
Joining, by = "id"
dim(ij)
[1] 24 4
tail(ij)
        Age visit Outcome
    id
19 7 58.33333 1 20.00000
20 7 58.33333 3 33.33333

      21
      7
      58.33333
      2
      46.66667

      22
      8
      58.88889
      2
      21.66667

23 8 58.88889 1 35.00000
24 8 58.88889 3 48.33333
```

Left Join

26 10 60.00000

NA NA

```
lj = left_join(base, visits)
Joining, by = "id"
dim(lj)
[1] 26 4
tail(lj)
        Age visit Outcome
    id
21 7 58.33333 2 46.66667
22 8 58.88889 2 21.66667

      23
      8
      58.88889
      1
      35.00000

      24
      8
      58.88889
      3
      48.33333

25 9 59.44444
                     NA
                                     NA
```

Right Join

24 8 58.88889 3 48.33333 25 11 NA 3 50.00000

```
rj = right_join(base, visits)

Joining, by = "id"

dim(rj)

[1] 25  4

tail(rj)

id     Age visit Outcome
20  4 56.66667   2 41.66667
21  5 57.22222   3 43.33333
22  6 57.77778   1 45.00000
23  7 58.33333   2 46.66667
```

Right Join: Switching arguments

```
rj2 = right join(visits, base)
Joining, by = "id"
dim(rj2)
[1] 26 4
tail(rj2)
   id visit Outcome Age
21 7 2 46.66667 58.33333
22 8 2 21.66667 58.88889

      23
      8
      1
      35.00000
      58.88889

      24
      8
      3
      48.33333
      58.88889

25 9 NA NA 59.44444
26 10
       NA NA 60.00000
identical(rj2, lj) ## after some rearranging
[1] TRUE
```

Full Join

27 11 NA

```
fj = full_join(base, visits)
Joining, by = "id"
dim(fj)
[1] 27 4
tail(fj)
      Age visit Outcome
   id
22 8 58.88889 2 21.66667
23 8 58.88889 1 35.00000
24 8 58.88889 3 48.33333
25 9 59.44444 NA NA
26 10 60.00000
```

NA NA

3 50.00000

Duplicated

• The duplicated command can give you indications if there are duplications in a **vector**e:

```
duplicated(1:5)

[1] FALSE FALSE FALSE FALSE
duplicated(c(1:5, 1))

[1] FALSE FALSE FALSE FALSE TRUE
```

Lab Part 4

Website

Reshaping data from long (tall) to wide (fat): tidyr

We can use rowsums to see if any values in the row is NA and keep if the row, which is a combination of date and line type has any non-missing data.

```
head (wide, 3)
# A tibble: 3 x 7
 day date daily line Alightings Average Boardings
 <chr> <date> <dbl> <chr>
                               <dbl> <dbl>
                                              <dbl>
1 Friday 2010-01-15 1644 banner
                           NA
                                        NA
                                                 NA
2 Friday 2010-01-15 1644 green NA NA
                                               NA
3 Friday 2010-01-15 1644 orange 1643 1644 1645
not namat = !is.na(select(wide, Alightings, Average, Boardings))
head (not namat, 2)
    Alightings Average Boardings
[1,]
    FALSE FALSE
                       FALSE
[2,] FALSE FALSE FALSE
wide$good = rowSums(not namat) > 0
```

Reshaping data from long (tall) to wide (fat): tidyr

Now we can filter only the good rows and delete the good column.

```
wide = wide %>% filter(good) %>% select(-good)
head (wide)
# A tibble: 6 x 7
       date daily line Alightings Average Boardings
 day
 <chr> <date> <dbl> <chr>
                               <dbl> <dbl>
                                              <dbl>
1 Friday 2010-01-15 1644 orange
                               1643 1644
                                              1645
2 Friday 2010-01-22 1394. orange 1388 1394. 1401
3 Friday 2010-01-29 1332 orange
                           1322 1332 1342
4 Friday 2010-02-05 1218. orange 1204 1218. 1231
                           678 671
5 Friday 2010-02-12 671 orange
                                              664
6 Friday 2010-02-19 1642 orange
                           1647 1642
                                               1637
```

Finding the First (or Last) record

slice allows you to select records (compared to first/last on a vector)

Finding the First (or Last) record

pivot_longer and pivot_wider are new (as of 2019) tidyr functions.

See link below:

https://tidyr.tidyverse.org/dev/articles/pivot.html

Website

Website

Merging in base R (not covered)

Data Merging/Append in Base R

- merge() is the most common way to do this with data sets
 - we will use the "join" functions from dplyr
- rbind/cbind row/column bind, respectively
 - rbind is the equivalent of "appending" in Stata or "setting" in SAS
 - cbind allows you to add columns in addition to the previous ways
- t() can transpose data but doesn't make it a data.frame

Merging

Merging