# 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 <a href="http://www.cookbook-r.com/Manipulating\_data/Converting\_data\_between\_wide\_and\_long\_format/">http://www.cookbook-r.com/Manipulating\_data/Converting\_data\_between\_wide\_and\_long\_format/</a>

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

Long - multiple rows per observation

```
# A tibble: 5 x 3
      id visit value
      <dbl> <int> <dbl>
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

```
long = gather(circ, key = "var", value = "number",
              starts with ("orange"), starts with ("purple"),
              starts with ("green"), starts with ("banner"))
long
# A tibble: 13,752 x 5
  day date daily var <chr> <chr>
                                   number
                                             <dbl>
 1 Monday 2010-01-11 952 orangeBoardings 877
 2 Tuesday 2010-01-12 796 orangeBoardings 777
                                              1203
 3 Wednesday 2010-01-13 1212. orangeBoardings
 4 Thursday 2010-01-14 1214. orangeBoardings
                                               1194
 5 Friday 2010-01-15 1644 orangeBoardings
                                               1645
 6 Saturday 2010-01-16 1490. orangeBoardings
                                               1457
7 Sunday 2010-01-17 888. orangeBoardings 8 Monday 2010-01-18 1000. orangeBoardings
                                               839
                                               999
 9 Tuesday 2010-01-19 1035 orangeBoardings
                                               1023
10 Wednesday 2010-01-20 1396. orangeBoardings
                                                1375
# ... with 13,742 more rows
```

```
# A tibble: 12 x 2
  var
                   <int>
  <chr>
 1 bannerAlightings 1146
 2 bannerAverage
                 1146
 3 bannerBoardings
                  1146
 4 greenAlightings
                  1146
 5 greenAverage
                 1146
 6 greenBoardings 1146
 7 orangeAlightings 1146
                  1146
 8 orangeAverage
 9 orangeBoardings
                   1146
10 purpleAlightings 1146
11 purpleAverage
                1146
12 purpleBoardings
                   1146
```

long %>% count(var)

# Lab Part 1

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#### 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")
long %>% count(var)
# A tibble: 12 x 2
  var
   <chr>
                     \langle int \rangle
 1 banner.Alightings
                      1146
 2 banner.Average
                    1146
 3 banner.Boardings 1146
 4 green. Alightings 1146
 5 green. Average
                   1146
 6 green.Boardings 1146
 7 orange. Alightings 1146
 8 orange. Average
                      1146
 9 orange.Boardings
                     1146
10 purple. Alightings 1146
11 purple. Average
                    1146
12 purple.Boardings
                      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.

# Lab Part 2

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#### 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 = long %>% filter(!is.na(date))
wide = wide %>% spread(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

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#### Merging: Simple Data

```
base \leftarrow tibble(id = 1:10, Age = seq(55,60, length=10))
head (base, 2)
# A tibble: 2 x 2
    id Age
  <int> <dbl>
   1 55
  2 55.6
visits \leftarrow tibble (id = c(rep(1:8, 3), 11), visit= c(rep(1:3, 8), 3),
                    Outcome = seq(10,50, length=25))
tail(visits, 2)
# A tibble: 2 x 3
    id visit Outcome
  <dbl> <dbl> <dbl>
1 8 3 48.3
2 11 3 50
```

#### 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 <a href="https://www.rstudio.com/wp-content/uploads/2015/02/data-wrangling-cheatsheet.pdf">https://www.rstudio.com/wp-content/uploads/2015/02/data-wrangling-cheatsheet.pdf</a>
- 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)
# A tibble: 6 x 4
      id Age visit Outcome
  <dbl> <dbl> <dbl> <dbl> <
  7 58.3 1 20
7 58.3 3 33.3
7 58.3 2 46.7
8 58.9 2 21.7
8 58.9 1 35
8 58 9
2
       8 58.9 3 48.3
```

#### Left Join

```
lj = left_join(base, visits)
Joining, by = "id"
dim(lj)
[1] 26 4
tail(lj)
# A tibble: 6 x 4
     id Age visit Outcome
  <dbl> <dbl> <dbl> <dbl> <
   7 58.3 2 46.7
8 58.9 2 21.7
8 58.9 1 35
8 58.9 3 48.3
1
2
3
4
5
     9 59.4 NA
                         NA
     10 60
              NA
                          NA
```

#### Right Join

```
rj = right_join(base, visits)
Joining, by = "id"
dim(rj)
[1] 25 4
tail(rj)
# A tibble: 6 x 4
       id Age visit Outcome
  <dbl> <dbl> <dbl> <dbl> <
    4 56.7 2 41.7

5 57.2 3 43.3

6 57.8 1 45

7 58.3 2 46.7

8 58.9 3 48.3

11 NA 3 50
1 2 3
```

# Right Join: Switching arguments

```
rj2 = right join(visits, base)
Joining, by = "id"
dim(rj2)
[1] 26 4
tail(rj2)
# A tibble: 6 x 4
     id visit Outcome Age
  <dbl> <dbl> <dbl> <dbl> <
            2 46.7 58.3
 2 40.7 58.3
8 2 21.7 58.9
8 1 35 58.9
8 3 48.3 58 0
2
3
4
         NA NA 59.4
     10
        NA NA 60
identical(rj2, lj) ## after some rearranging
[1] TRUE
```

#### Full Join

5

10 60 NA

11 NA

NA

3 50

```
fj = full join(base, visits)
Joining, by = "id"
dim(fj)
[1] 27 4
tail(fj)
# A tibble: 6 x 4
        id Age visit Outcome
   <dbl> <dbl> <dbl> <dbl> <

      8
      58.9
      2
      21.7

      8
      58.9
      1
      35

      8
      58.9
      3
      48.3

1
2
3
4
     9 59.4 NA NA
```

#### **Duplicated**

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

```
duplicated (1:5)
[1] FALSE FALSE FALSE FALSE
duplicated(c(1:5, 1))
[1] FALSE FALSE FALSE FALSE
                                         TRUE
fi %>%
  mutate(dup id = duplicated(id))
# A tibble: 27 x 5
       id Age visit Outcome dup id
   \langle dbl \rangle \langle dbl \rangle \langle dbl \rangle \langle dbl \rangle \langle dgl \rangle
        1 55
                      1 10
                                  FALSE
                 3 23.3 TRUE
2 36.7 TRUE
2 11.7 FALSE
 2 3
        1 55
       1 55
        2 55.6
                 1 25 TRUE
3 38.3 TRUE
3 13.3 FALSE
       2 55.6
        2 55.6
        3 56.1
       3 56.1
                  2 26.7 TRUE
                                                                                    27/37
        3 56.1
                          40
                                  TRUE
```

# Lab Part 4

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#### 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

#### 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
                            NA
2 Friday 2010-01-15 1644 green
                                       NA
                                                NA
3 Friday 2010-01-15 1644 orange 1643 1644 1645
not namat = wide %>% select(Alightings, Average, Boardings)
not namat = !is.na(not namat)
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)

# 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