

# 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-wrangling-cheatsheet.pdf>

# What is wide/long data?

See [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/)

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

```
# A tibble: 2 x 4
  id visit1 visit2 visit3
<int> <dbl> <dbl> <dbl>
1     1     10     4      3
2     2      5     6     NA
```

- Long - multiple rows per observation

```
# A tibble: 5 x 3
  id visit value
<dbl> <int> <dbl>
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](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 purpleBoardings  
  <chr> <chr>          <dbl>          <dbl>          <dbl>          <dbl>  
1 Mond.. 01/1...          877          1027          952             NA  
2 Tues.. 01/1...          777           815          796             NA  
# ... with 9 more variables: purpleAlightings <dbl>, purpleAverage <dbl>,  
#   greenBoardings <dbl>, greenAlightings <dbl>, greenAverage <dbl>,  
#   bannerBoardings <dbl>, bannerAlightings <dbl>, bannerAverage <dbl>,  
#   daily <dbl>
```

```
class(circ$date)
```

```
[1] "character"
```

## 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

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

`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",  
              -day, -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
```

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

- 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	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
5	Friday	2010-01-15	1644	orangeBoardings	1645
6	Saturday	2010-01-16	1490.	orangeBoardings	1457
7	Sunday	2010-01-17	888.	orangeBoardings	839
8	Monday	2010-01-18	999.	orangeBoardings	999
9	Tuesday	2010-01-19	1035	orangeBoardings	1023
10	Wednesday	2010-01-20	1396.	orangeBoardings	1375

```
# ... with 13,742 more rows
```

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

```
long %>% count(var)
```

```
# A tibble: 12 x 2
```

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

# Lab Part 1

[Website](#)

## Making a separator

We will use `str_replace` from `stringr` to put `_` in the names

```
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                n  
  <chr>            <int>  
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
```

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

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
<chr> <date>    <dbl> <chr> <chr>    <dbl>
1 Monday 2010-01-11    952 orange Boardings    877
2 Tuesday 2010-01-12    796 orange Boardings    777
```

```
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:

```
reunited = long %>%  
  unite(col = var, line, type, sep = "_")  
reunited %>% select(day, var) %>% head(3) %>% print
```

```
# A tibble: 3 x 2  
  day      var  
  <chr>    <chr>  
1 Monday  orange_Boardings  
2 Tuesday orange_Boardings  
3 Wednesday orange_Boardings
```

We could also use `paste/paste0`.



# 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 = 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
3 Friday 2010-01-15 1644 orange 1643    1644    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

base has baseline data for ids 1 to 10 and Age

```
base <- tibble(id = 1:10, Age = seq(55, 60, length=10))  
head(base, 2)
```

```
# A tibble: 2 x 2  
  id   Age  
  <int> <dbl>  
1     1  55  
2     2 55.6
```

visits has ids 1 to 8, then 11 (new id), and 3 visits and outcome

```
visits <- 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 <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`
- `anti_join(x, y)` - all rows from `x` not in `y` keeping just columns from `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>
1     7  58.3     1     20
2     7  58.3     3    33.3
3     7  58.3     2    46.7
4     8  58.9     2    21.7
5     8  58.9     1     35
6     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>
1     7  58.3     2    46.7
2     8  58.9     2    21.7
3     8  58.9     1     35
4     8  58.9     3    48.3
5     9  59.4    NA     NA
6    10   60    NA     NA
```

## Logging the joins

The `tidylog` package can show you log outputs from `dplyr` (newly added). You will need to install to use.

```
library(tidylog)
left_join(base, visits)
```

```
Joining, by = "id"left_join: added 2 columns (visit, Outcome)
  > rows only in x      2
  > rows only in y    ( 1)
  > matched rows      24      (includes duplicates)
  >                      ====
  > rows total        26
```

```
# A tibble: 26 x 4
   id   Age visit Outcome
  <dbl> <dbl> <dbl>   <dbl>
1     1   55     1     10
2     1   55     3    23.3
3     1   55     2    36.7
4     2  55.6     2    11.7
5     2  55.6     1     25
6     2  55.6     3    38.3
7     3  56.1     3    13.3
8     3  56.1     2    26.7
9     3  56.1     1     40
10    4  56.7     1     15
# ... with 16 more rows
```



# Right Join

```
rj = right_join(base, visits)
```

```
Joining, by = "id"right_join: added 2 columns (visit, Outcome)
> rows only in x ( 2)
> rows only in y 1
> matched rows 24
>
> rows total 25
```

```
tail(rj, 3)
```

```
# A tibble: 3 x 4
  id    Age visit Outcome
  <dbl> <dbl> <dbl>   <dbl>
1     7  58.3     2    46.7
2     8  58.9     3    48.3
3    11   NA     3     50
```

## Right Join: Switching arguments

```
rj2 = right_join(visits, base)
```

```
Joining, by = "id"right_join: added one column (Age)
> rows only in x    ( 1)
> rows only in y      2
> matched rows      24    (includes duplicates)
>                      ====
> rows total         26
```

```
tail(rj2, 3)
```

```
# A tibble: 3 x 4
   id visit Outcome  Age
  <dbl> <dbl>   <dbl> <dbl>
1     8     3    48.3  58.9
2     9    NA     NA   59.4
3    10    NA     NA   60
```

```
identical(rj2, lj) ## after some rearranging
```

```
[1] TRUE
```

# Full Join

```
fj = full_join(base, visits)
```

```
Joining, by = "id"full_join: added 2 columns (visit, Outcome)
  > rows only in x      2
  > rows only in y      1
  > matched rows       24      (includes duplicates)
  >                      ====
  > rows total          27
```

```
tail(fj, 3)
```

```
# A tibble: 3 x 4
   id    Age visit Outcome
  <dbl> <dbl> <dbl>   <dbl>
1     9  59.4   NA      NA
2    10   60   NA      NA
3    11   NA     3     50
```

# Duplicated

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

```
duplicated(1:5)
```

```
[1] FALSE FALSE FALSE FALSE FALSE
```

```
duplicated(c(1:5, 1))
```

```
[1] FALSE FALSE FALSE FALSE FALSE TRUE
```

```
fj %>% mutate(dup_id = duplicated(id))
```

mutate: new variable 'dup\_id' with 2 unique values and 0% NA

```
# A tibble: 27 x 5
```

	id	Age	visit	Outcome	dup_id
	<dbl>	<dbl>	<dbl>	<dbl>	<lgl>
1	1	55	1	10	FALSE
2	1	55	3	23.3	TRUE
3	1	55	2	36.7	TRUE
4	2	55.6	2	11.7	FALSE
5	2	55.6	1	25	TRUE
6	2	55.6	3	38.3	TRUE
7	3	56.1	3	13.3	FALSE
8	3	56.1	2	26.7	TRUE
9	3	56.1	1	40	TRUE

# Lab Part 4

[Website](#)

## 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
2 Friday 2010-01-15  1644 green          NA         NA         NA
3 Friday 2010-01-15  1644 orange      1643      1644      1645
```

```
not_namat = wide %>% select(Alightings, Average, Boardings)
```

```
select: dropped 4 variables (day, date, daily, line)
```

```
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)
```

```
filter: removed 1,700 rows (37%), 2,884 rows remaining
```

```
select: dropped one variable (good)
```

```
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	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
5	Friday	2010-02-12	671	orange	678	671	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**)

```
long = long %>% filter(!is.na(number) & number > 0)
```

```
filter: removed 5,364 rows (39%), 8,388 rows remaining
```

```
first_and_last = long %>% arrange(date) %>% # arrange by date
  filter(type == "Boardings") %>% # keep boardings only
  group_by(line) %>% # group by line
  slice(c(1, n())) # select ("slice") first and last (n() command) lines
```

```
filter: removed 5,630 rows (67%), 2,758 rows remaining
```

```
group_by: one grouping variable (line)
```

```
slice (grouped): removed 2,750 rows (>99%), 8 rows remaining
```

```
first_and_last %>% head(4)
```

```
# A tibble: 4 x 6
# Groups:   line [2]
  day      date      daily line  type      number
<chr> <date>    <dbl> <chr> <chr>    <dbl>
1 Monday 2012-06-04 13342. banner Boardings 520
2 Friday 2013-03-01    NA  banner Boardings 817
3 Tuesday 2011-11-01 8873  green  Boardings 887
4 Friday 2013-03-01    NA  green  Boardings 2592
```

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

```
merged.data <- merge(base, visits, by = "id")  
head(merged.data, 5)
```

	id	Age	visit	Outcome
1	1	55.00000	1	10.00000
2	1	55.00000	3	23.33333
3	1	55.00000	2	36.66667
4	2	55.55556	2	11.66667
5	2	55.55556	1	25.00000

```
dim(merged.data)
```

```
[1] 24  4
```

# Merging

```
all.data <- merge(base, visits, by = "id", all = TRUE)
tail(all.data)
```

	id	Age	visit	Outcome
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
27	11	NA	3	50.00000

```
dim(all.data)
```

```
[1] 27  4
```