

# 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 long (tall) to wide (fat)
2. Reshaping data from wide (fat) to long (tall)
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>

Data used: Charm City Circulator

[http://www.aejaffe.com/summerR\\_2016/data/Charm\\_City\\_Circulator\\_Ridership.csv](http://www.aejaffe.com/summerR_2016/data/Charm_City_Circulator_Ridership.csv)

```
circ = read.csv("http://www.aejaffe.com/summerR_2016/data/CircData.csv")
head(circ, 2)
```

	day	date	orangeBoardings	orangeAlightings	orangeAverage
1	Monday	01/11/2010	877	1027	952
2	Tuesday	01/12/2010	777	815	796
			purpleBoardings	purpleAlightings	purpleAverage
1			NA	NA	NA
2			NA	NA	NA
			greenBoardings	greenAlightings	greenAverage
1			NA	NA	NA
2			NA	NA	NA
			bannerBoardings	bannerAlightings	bannerAverage
1			NA	NA	NA
2			NA	NA	NA
			daily		
1			NA	952	
2			NA	796	

## Creating a Date class from a character date

```
library(lubridate) # great for dates!  
library(dplyr) # mutate/summarise functions  
circ = mutate(circ, date = mdy(date))  
sum( is.na(circ$date) ) # all converted correctly
```

```
[1] 0
```

```
head(circ$date)
```

```
[1] "2010-01-11" "2010-01-12" "2010-01-13" "2010-01-14" "20  
[6] "2010-01-16"
```

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

```
[1] "Date"
```

## Making column names a little more separated

We will use `str_replace` from `stringr` to put periods in the column names.

```
library(stringr)
cn = colnames(circ)
cn = cn %>%
  str_replace("Board", ".Board") %>%
  str_replace("Alight", ".Alight") %>%
  str_replace("Average", ".Average")
colnames(circ) = cn
cn
```

[1]	"day"	"date"	"orange.Boarding"
[4]	"orange.Alightings"	"orange.Average"	"purple.Boarding"
[7]	"purple.Alightings"	"purple.Average"	"green.Boarding"
[10]	"green.Alightings"	"green.Average"	"banner.Boarding"
[13]	"banner.Alightings"	"banner.Average"	"daily"

## Removing the daily ridership

We want to look at each ridership, and will remove the daily column:

```
circ$daily = NULL
```

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

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

	id	visit1	visit2	visit3
1	1	10	4	3
2	2	5	6	NA

- ▶ 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



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

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

## 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:”

```
library(tidyr)
long = gather(circ, key = "var", value = "number",
              starts_with("orange"),
              starts_with("purple"),
              starts_with("green"),
              starts_with("banner"))
head(long, 2)
```

	day	date	var	number
1	Monday	2010-01-11	orange.Boardings	877
2	Tuesday	2010-01-12	orange.Boardings	777

```
table(long$var)
```

## 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.

```
long = separate_(long, "var",  
                  into = c("line", "type"),  
                  sep = "[.]")  
head(long, 3)
```

	day	date	line	type	number
1	Monday	2010-01-11	orange	Boardings	877
2	Tuesday	2010-01-12	orange	Boardings	777
3	Wednesday	2010-01-13	orange	Boardings	1203

```
unique(long$line)
```

```
[1] "orange" "purple" "green"  "banner"
```

```
unique(long$type)
```

## Finding the First (or Last) record

```
long = long %>% filter(!is.na(number) & number > 0)
first_and_last = long %>% arrange(date) %>% # arrange by date
  filter(type %in% "Boardings") %>% # keep boardings only
  group_by(line) %>% # group by line
  slice( c(1, n())) # select ("slice") first and last (n())
first_and_last %>% head(4)
```

Source: local data frame [4 x 5]

Groups: line [2]

	day	date	line	type	number
	<chr>	<date>	<chr>	<chr>	<dbl>
1	Monday	2012-06-04	banner	Boardings	520
2	Friday	2013-03-01	banner	Boardings	817
3	Tuesday	2011-11-01	green	Boardings	887
4	Friday	2013-03-01	green	Boardings	2592

## 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)
```

	day	date	line	Alightings	Average	Boardings
1	Friday	2010-01-15	orange	1643	1644.0	1645
2	Friday	2010-01-22	orange	1388	1394.5	1401
3	Friday	2010-01-29	orange	1322	1332.0	1342
4	Friday	2010-02-05	orange	1204	1217.5	1231
5	Friday	2010-02-12	orange	678	671.0	664
6	Friday	2010-02-19	orange	1647	1642.0	1637

## 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.

```
# wide = wide %>%  
#   select(Alightings, Average, Boardings) %>%  
#   mutate(good = rowSums(is.na(.)) > 0)  
namat = !is.na(select(wide, Alightings, Average, Boardings))  
head(namat)
```

	Alightings	Average	Boardings
1	TRUE	TRUE	TRUE
2	TRUE	TRUE	TRUE
3	TRUE	TRUE	TRUE
4	TRUE	TRUE	TRUE
5	TRUE	TRUE	TRUE
6	TRUE	TRUE	TRUE

```
wide$good = rowSums(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 = filter(wide, good) %>% select(-good)
head(wide)
```

	day	date	line	Alightings	Average	Boardings
1	Friday	2010-01-15	orange	1643	1644.0	1645
2	Friday	2010-01-22	orange	1388	1394.5	1401
3	Friday	2010-01-29	orange	1322	1332.0	1342
4	Friday	2010-02-05	orange	1204	1217.5	1231
5	Friday	2010-02-12	orange	678	671.0	664
6	Friday	2010-02-19	orange	1647	1642.0	1637

# Data Merging/Append in Base R

- ▶ Merging - joining data sets together - usually on key variables, usually “id”
- ▶ `merge()` is the most common way to do this with data sets
- ▶ `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()` is a function that will transpose the data



## Merging

```
base <- data.frame(id = 1:10, Age= seq(55,60, length=10))  
base[1:2,]
```

	id	Age
1	1	55.00000
2	2	55.55556

```
visits <- data.frame(id = rep(1:8, 3), visit= rep(1:3, 8),  
                     Outcome = seq(10,50, length=24))  
visits[1:2,]
```

	id	visit	Outcome
1	1	1	10.00000
2	2	2	11.73913

## Merging

```
merged.data <- merge(base, visits, by="id")  
merged.data[1:5,]
```

	id	Age	visit	Outcome
1	1	55.00000	1	10.00000
2	1	55.00000	3	23.91304
3	1	55.00000	2	37.82609
4	2	55.55556	2	11.73913
5	2	55.55556	1	25.65217

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

```
[1] 24 4
```

## Merging

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

	id	Age	visit	Outcome
21	7	58.33333	2	48.26087
22	8	58.88889	2	22.17391
23	8	58.88889	1	36.08696
24	8	58.88889	3	50.00000
25	9	59.44444	NA	NA
26	10	60.00000	NA	NA

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

```
[1] 26  4
```

# Joining in dplyr

- ▶ `?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>

## Left Join

```
lj = left_join(base, visits)
```

Joining, by = "id"

```
dim(lj)
```

```
[1] 26  4
```

```
tail(lj)
```

	id	Age	visit	Outcome
21	7	58.33333	2	48.26087
22	8	58.88889	2	22.17391
23	8	58.88889	1	36.08696
24	8	58.88889	3	50.00000
25	9	59.44444	NA	NA
26	10	60.00000	NA	NA

## Right Join

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

Joining, by = "id"

```
dim(rj)
```

```
[1] 24  4
```

```
tail(rj)
```

	id	Age	visit	Outcome
19	3	56.11111	1	41.30435
20	4	56.66667	2	43.04348
21	5	57.22222	3	44.78261
22	6	57.77778	1	46.52174
23	7	58.33333	2	48.26087
24	8	58.88889	3	50.00000

## Full Join

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

Joining, by = "id"

```
dim(fj)
```

```
[1] 26  4
```

```
tail(fj)
```

	id	Age	visit	Outcome
21	7	58.33333	2	48.26087
22	8	58.88889	2	22.17391
23	8	58.88889	1	36.08696
24	8	58.88889	3	50.00000
25	9	59.44444	NA	NA
26	10	60.00000	NA	NA

## Perform Operations By Groups: dplyr

`group_by` is a form of replacement for `tapply` (not a complete replacement).

We will use `group_by` to group the data by line, then use `summarize` (or `summarise`) to get the mean Average ridership:

```
gb = group_by(wide, line)
summarize(gb, mean_avg = mean(Average))
```

```
# A tibble: 4 × 2
  line  mean_avg
  <chr>    <dbl>
1 banner  836.5637
2 green  1969.9668
3 orange 3041.1924
4 purple 4029.1071
```



## Perform Operations By Groups: dplyr with piping

Using piping, this is:

```
wide %>%  
  group_by(line) %>%  
  summarise(mean_avg = mean(Average))
```

```
# A tibble: 4 × 2  
  line mean_avg  
  <chr>      <dbl>  
1 banner  836.5637  
2  green 1969.9668  
3 orange 3041.1924  
4 purple 4029.1071
```

## Perform Operations By Multiple Groups: dplyr

This can easily be extended using `group_by` with multiple groups. Let's define the year of riding:

```
wide = wide %>% mutate(year = year(date),  
                        month = month(date))  
wide %>%  
  group_by(line, year) %>%  
  summarise(mean_avg = mean(Average))
```

Source: local data frame [13 x 3]

Groups: line [?]

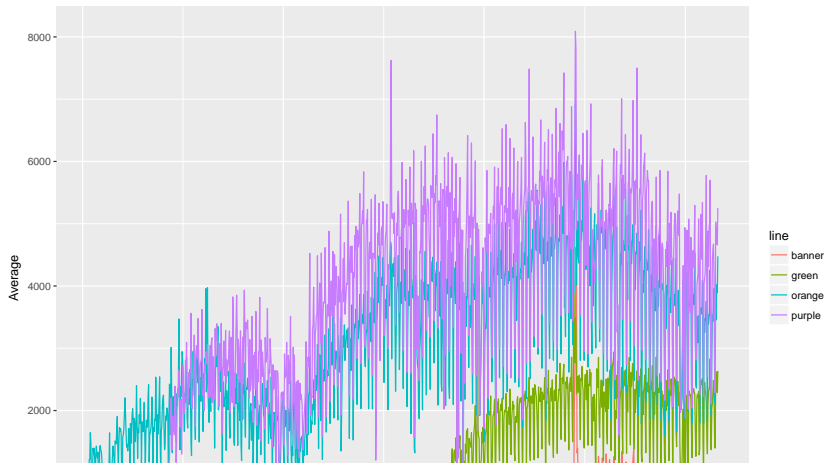
	line	year	mean_avg
	<chr>	<dbl>	<dbl>
1	banner	2012	894.8768
2	banner	2013	635.3833
3	green	2011	1455.1667
4	green	2012	2045.5870
5		2013	2022.5250

Bonus slides - explore after visualization!

# Perform Operations By Multiple Groups: dplyr

We can then easily plot each day over time:

```
library(ggplot2)
ggplot(aes(x = date, y = Average,
           colour = line), data = wide) + geom_line()
```



## Perform Operations By Multiple Groups: dplyr

Let's create the middle of the month (the 15th for example), and name it mon.

```
mon = wide %>%  
  dplyr::group_by(line, month, year) %>%  
  dplyr::summarise(mean_avg = mean(Average))  
mon = mutate(mon,  
              mid_month = dmy(paste0("15-", month, "-", year)),  
              head(mon))
```

Source: local data frame [6 x 5]

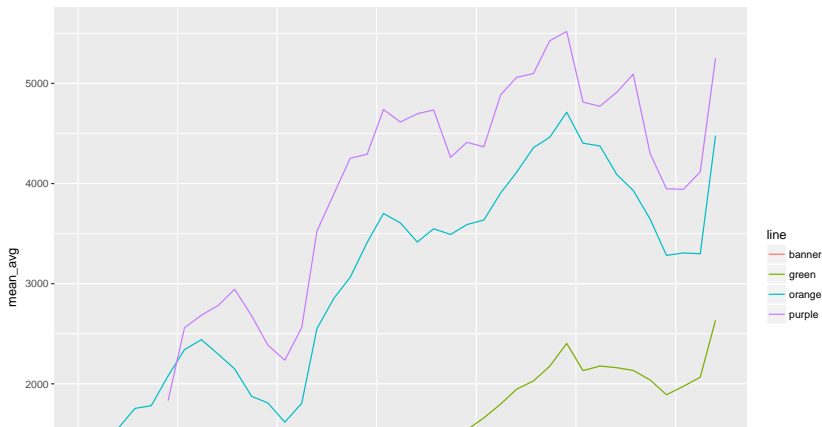
Groups: line, month [6]

	line	month	year	mean_avg	mid_month
	<chr>	<dbl>	<dbl>	<dbl>	<date>
1	banner	1	2013	610.3226	2013-01-15
2	banner	2	2013	656.4643	2013-02-15
3	banner	3	2013	822.0000	2013-03-15
4	banner	6	2013	1000.1000	2013-06-15

## Perform Operations By Multiple Groups: dplyr

We can then easily plot the mean of each month to see a smoother output:

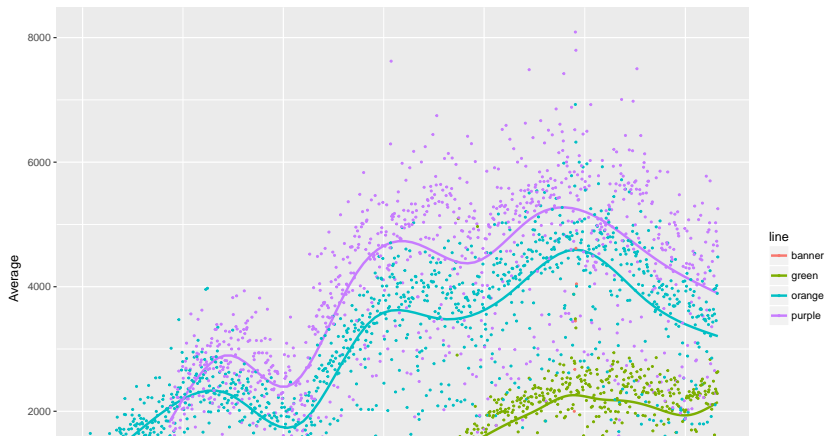
```
ggplot(aes(x = mid_month,  
           y = mean_avg,  
           colour = line), data = mon) + geom_line()
```



## Bonus! Points with a smoother!

```
ggplot(aes(x = date, y = Average, colour = line),  
       data = wide) + geom_smooth(se = FALSE) +  
       geom_point(size = .5)
```

`geom\_smooth()` using method = 'gam'



Extra group\_by examples



## group\_by

group\_by is a form of replacement for tapply (not a complete replacement).

Example using Bike Lanes: [http:](http://www.aejaffe.com/summerR_2016/data/Bike_Lanes.csv)

[//www.aejaffe.com/summerR\\_2016/data/Bike\\_Lanes.csv](http://www.aejaffe.com/summerR_2016/data/Bike_Lanes.csv)

```
bike = read.csv(  
  "http://www.aejaffe.com/summerR_2016/data/Bike_Lanes.csv"  
  as.is = TRUE)
```

## Summarizing data with group\_by and summarize

Average bike length BY project:

```
bike %>%  
  group_by(project) %>%  
  summarise(mean(length)) # get the average length
```

```
# A tibble: 13 × 2
```

	project	`mean(length)`
	<chr>	<dbl>
1		214.3288
2	CHARM CITY CIRCULATOR	276.6658
3	COLLEGETOWN	320.6836
4	COLLEGETOWN NETWORK	213.6373
5	ENGINEERING CONSTRUCTION	512.0976
6	GUILFORD AVE BIKE BLVD	197.2782
7	MAINTENANCE	1942.1523
8	OPERATION ORANGE CONE	250.0784
9	PARK HEIGHTS BIKE NETWORK	283.2252

## Naming columns in output in summarize

Using `summarise/summarize(my_new_column_name = output)` allows you to name the column in the output:

```
bike %>%  
  group_by(project) %>%  
  summarize(mean_length = mean(length)) %>%  
  head(4) # head ONLY for slide printing
```

```
# A tibble: 4 × 2
```

	project	mean_length
	<chr>	<dbl>
1		214.3288
2	CHARM CITY CIRCULATOR	276.6658
3	COLLEGETOWN	320.6836
4	COLLEGETOWN NETWORK	213.6373