

# Data Summarization

Introduction to R for Public Health Researchers

# Data Summarization

- ▶ Basic statistical summarization
  - ▶ `mean(x)`: takes the mean of `x`
  - ▶ `sd(x)`: takes the standard deviation of `x`
  - ▶ `median(x)`: takes the median of `x`
  - ▶ `quantile(x)`: displays sample quantities of `x`. Default is min, IQR, max
  - ▶ `range(x)`: displays the range. Same as `c(min(x), max(x))`

## Some examples

We can use the `mtcars` and Charm City Circulator datasets to explore different ways of summarizing data.

```
head(mtcars)
```

	mpg	cyl	disp	hp	drat	wt	qsec	vs	am
Mazda RX4	21.0	6	160	110	3.90	2.620	16.46	0	1
Mazda RX4 Wag	21.0	6	160	110	3.90	2.875	17.02	0	1
Datsun 710	22.8	4	108	93	3.85	2.320	18.61	1	1
Hornet 4 Drive	21.4	6	258	110	3.08	3.215	19.44	1	0
Hornet Sportabout	18.7	8	360	175	3.15	3.440	17.02	0	0
Valiant	18.1	6	225	105	2.76	3.460	20.22	1	0

## Statistical summarization

```
mean(mtcars$hp)
```

```
[1] 146.6875
```

```
quantile(mtcars$hp)
```

0%	25%	50%	75%	100%
52.0	96.5	123.0	180.0	335.0

## Statistical summarization

```
median(mtcars$wt)
```

```
[1] 3.325
```

```
quantile(mtcars$wt, probs = 0.6)
```

```
60%
```

```
3.44
```

## Statistical summarization

`t.test` will be covered more in detail later, gives a 95% CI:

```
t.test(mtcars$wt)
```

One Sample t-test

```
data:  mtcars$wt
```

```
t = 18.6, df = 31, p-value < 2.2e-16
```

```
alternative hypothesis: true mean is not equal to 0
```

```
95 percent confidence interval:
```

```
 2.864478 3.570022
```

```
sample estimates:
```

```
mean of x
```

```
 3.21725
```

## Statistical summarization

Note that many of these functions have additional inputs regarding missing data, typically requiring the `na.rm` argument.

```
x = c(1,5,7,NA,4,2, 8,10,45,42)
mean(x)
```

```
[1] NA
```

```
mean(x,na.rm=TRUE)
```

```
[1] 13.77778
```

```
quantile(x,na.rm=TRUE)
```

0%	25%	50%	75%	100%
1	4	7	10	45

# Data Summarization on matrices/data frames

- ▶ Basic statistical summarization

- ▶ `rowMeans(x)`: takes the means of each row of `x`
- ▶ `colMeans(x)`: takes the means of each column of `x`
- ▶ `rowSums(x)`: takes the sum of each row of `x`
- ▶ `colSums(x)`: takes the sum of each column of `x`
- ▶ `summary(x)`: for data frames, displays the quantile information



## Charm City Circulator data

Please download the Charm City Circulator data:

[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/C  
              header=TRUE,as.is=TRUE)
```

## Subsetting to specific columns

Let's just take columns that represent average ridership:

```
library(dplyr)
circ2 = select(circ, date, day, ends_with("Average"))
head(circ2, 4)
```

	date	day	orangeAverage	purpleAverage	greenAverage
1	01/11/2010	Monday	952.0	NA	
2	01/12/2010	Tuesday	796.0	NA	
3	01/13/2010	Wednesday	1211.5	NA	
4	01/14/2010	Thursday	1213.5	NA	

	bannerAverage
1	NA
2	NA
3	NA
4	NA

## column and row means

```
avgs = select(circ2, ends_with("Average"))  
colMeans(avgs, na.rm = TRUE)
```

orangeAverage	purpleAverage	greenAverage	bannerAverage
3033.1611	4016.9345	1957.7814	827.2685

```
circ2$daily = rowMeans(avgs, na.rm=TRUE)  
head(circ2$daily)
```

```
[1] 952.0 796.0 1211.5 1213.5 1644.0 1490.5
```

# Summary

```
summary(circ2)
```

date	day	orangeAverage	purp
Length:1146	Length:1146	Min. : 0	Min.
Class :character	Class :character	1st Qu.:2001	1st Q
Mode :character	Mode :character	Median :2968	Media
		Mean :3033	Mean
		3rd Qu.:4020	3rd Q
		Max. :6926	Max.
		NA's :10	NA's

greenAverage	bannerAverage	daily
Min. : 0	Min. : 0.0	Min. : 0
1st Qu.:1491	1st Qu.: 632.5	1st Qu.:2097
Median :2079	Median : 763.0	Median :2846
Mean :1958	Mean : 827.3	Mean :2878
3rd Qu.:2340	3rd Qu.: 945.9	3rd Qu.:3646
Max. :5094	Max. :4617.0	Max. :6123
NA's :661	NA's :876	NA's :10

## Apply statements

You can apply more general functions to the rows or columns of a matrix or data frame, beyond the mean and sum.

```
apply(X, MARGIN, FUN, ...)
```

*X : an array, including a matrix.*

*MARGIN : a vector giving the subscripts which the function will be applied over. E.g., for a matrix 1 indicates rows, 2 indicates columns, c(1, 2) indicates rows and columns. Where X has named dimnames, it can be a character vector selecting dimension names.*

*FUN : the function to be applied: see 'Details'.*

*... : optional arguments to FUN.*

## Apply statements

```
apply(avgs,2,mean,na.rm=TRUE) # column means
```

orangeAverage	purpleAverage	greenAverage	bannerAverage
3033.1611	4016.9345	1957.7814	827.2685

```
apply(avgs,2,sd,na.rm=TRUE) # columns sds
```

orangeAverage	purpleAverage	greenAverage	bannerAverage
1227.5779	1406.6544	592.8969	436.0487

```
apply(avgs,2,max,na.rm=TRUE) # column maxs
```

orangeAverage	purpleAverage	greenAverage	bannerAverage
6926.5	8089.5	5094.0	4617.0

## Other Apply Statements

- ▶ `tapply()`: 'grouping' apply
- ▶ `lapply()`: 'list' apply [tomorrow]
- ▶ `sapply()`: 'simple' apply [tomorrow]
- ▶ Other less used ones...

See more details here: <http://nsaunders.wordpress.com/2010/08/20/a-brief-introduction-to-apply-in-r/>

## tapply()

From the help file: “Apply a function to each cell of a ragged array, that is to each (non-empty) group of values given by a unique combination of the levels of certain factors.”

```
tapply(X, INDEX, FUN = NULL, ..., simplify = TRUE)
```

Simply put, you can apply function FUN to X within each categorical level of INDEX. It is very useful for assessing properties of continuous data by levels of categorical data.



## tapply()

For example, we can estimate the highest average daily ridership for each day of the week in 1 line in the Circulator dataset.

```
tapply(circ2$daily, circ2$day, max, na.rm = TRUE)
```

Friday	Monday	Saturday	Sunday	Thursday	Tuesday
5600.75	5002.25	6123.00	3980.25	4820.50	4855.25

# Data Summarization/Visualization

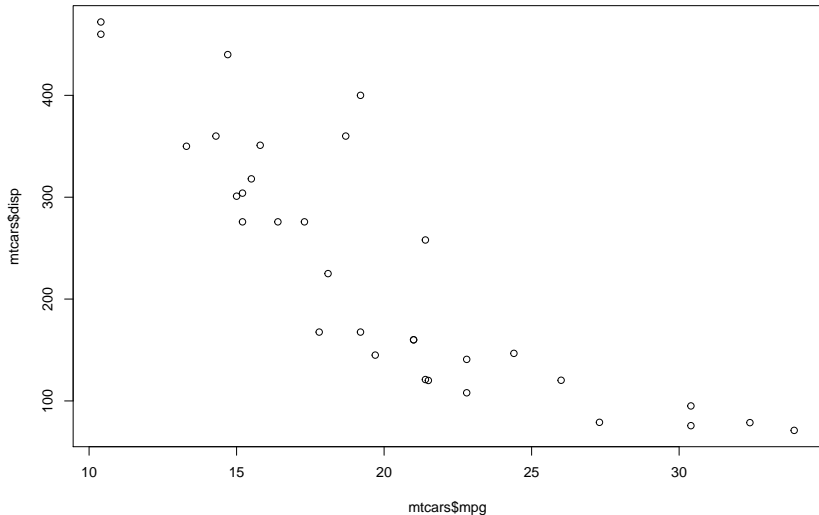
- ▶ Basic summarization plots
  - ▶ `plot(x,y)`: scatterplot of  $x$  and  $y$
  - ▶ `boxplot(y~x)`: boxplot of  $y$  against levels of  $x$
  - ▶ `hist(x)`: histogram of  $x$
  - ▶ `density(x)`: kernel density plot of  $x$

# Basic Plots

Plotting is an important component of exploratory data analysis. We will review some of the more useful and informative plots here. We will go over formatting and making plots look nicer in additional lectures.

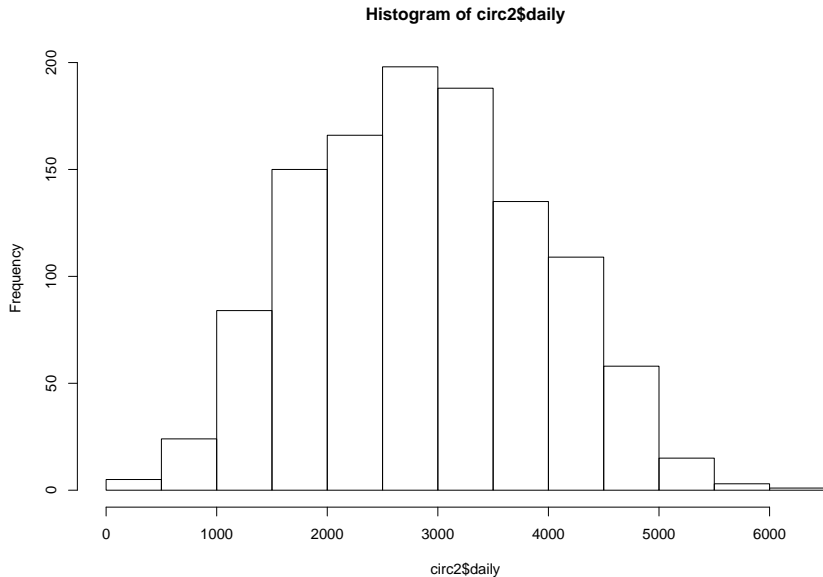
# Scatterplot

```
plot(mtcars$mpg, mtcars$displacement)
```



# Histograms

```
hist(circ2$daily)
```



## Plot with a line

`type = "l"` means a line

```
library(lubridate)
```

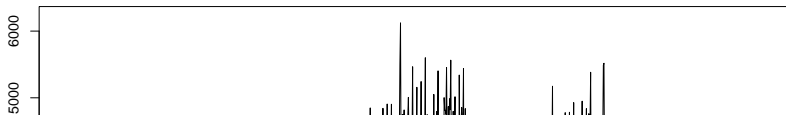
Loading required package: methods

Attaching package: 'lubridate'

The following object is masked from 'package:base':

date

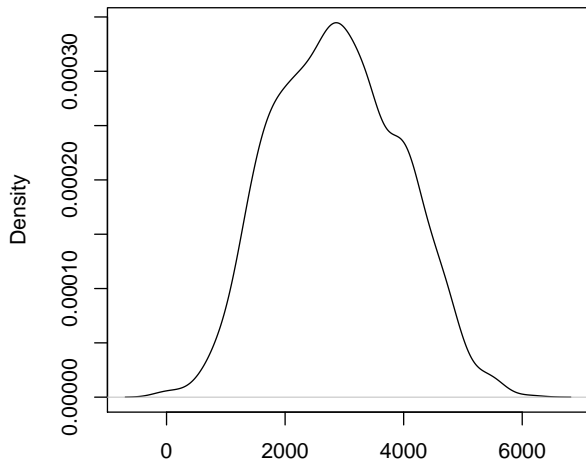
```
circ2$date = mdy(circ2$date)  
plot(circ2$date, circ2$daily, type = "l")
```



# Density

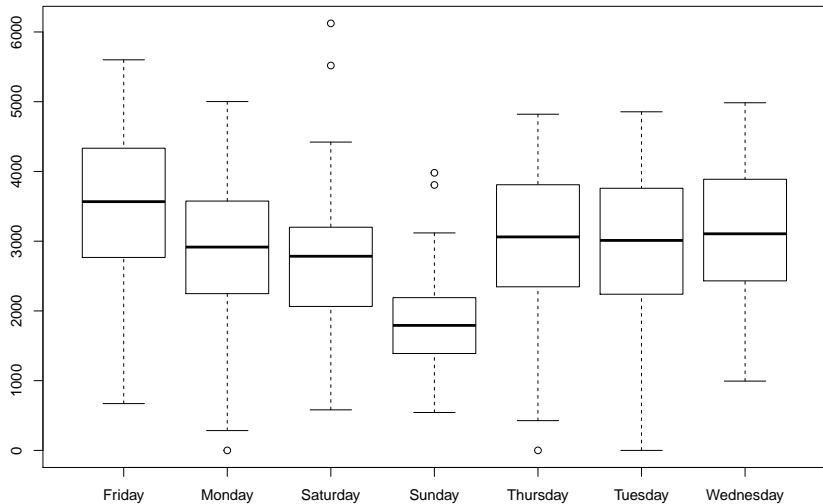
```
## plot(density(circ2$daily))  
plot(density(circ2$daily, na.rm=TRUE))
```

**density.default(x = circ2\$daily, na.rm = TRUE)**



# Boxplots

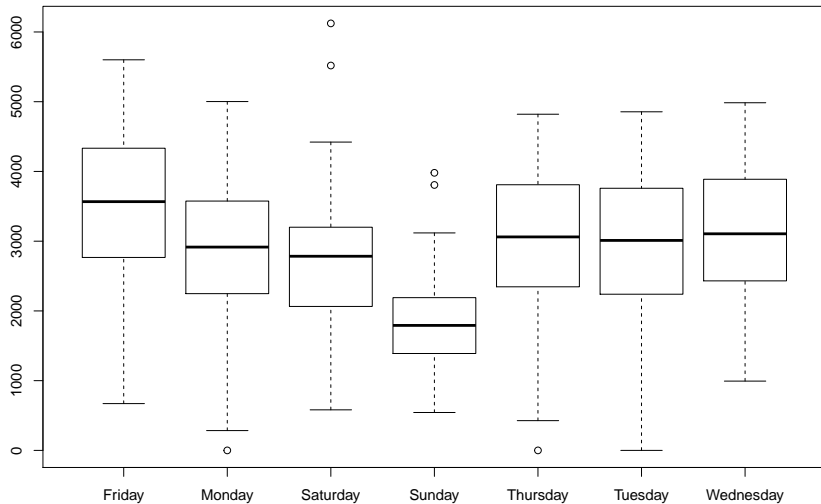
```
boxplot(circ2$daily ~ circ2$day)
```





# Boxplots

```
boxplot(daily ~ day, data=circ2)
```



# Data Summarization for data.frames

- ▶ Basic summarization plots
  - ▶ `matplot(x,y)`: scatterplot of two matrices, x and y
  - ▶ `pairs(x,y)`: plots pairwise scatter plots of matrices x and y, column by column

# Matrix plot

```
pairs(avgs)
```

