

<https://intro2r.com/> Chapter 3

CSCI 297b, Spring 2023

April 21, 2023

# R basic data types

**Numeric** data are numbers that contain a decimal.

**Integers** are whole numbers.

**Logical data** take on the value of either TRUE or FALSE. There's also another special type of logical called NA to represent missing values.

**Character** data are used to represent string values. You can think of character strings as something like a word (or multiple words). A special type of character string is a factor, which is a string but with additional attributes (like levels or an order). We'll cover factors later.

## R basic data types

```
1 num <- 2.2
2 class(num)
3 ## [1] "numeric"
4
5 char <- "hello"
6 class(char)
7 ## [1] "character"
8
9 logi <- TRUE
10 class(logi)
11 ## [1] "logical"
```

## R basic data types

```
1 is.numeric(num)
2 ## [1] TRUE
3
4 is.character(num)
5 ## [1] FALSE
6
7 is.character(char)
8 ## [1] TRUE
9
10 is.logical(logi)
11 ## [1] TRUE
```

## R type conversion

```
1 # coerce numeric to character
2 class(num)
3 ## [1] "numeric"
4 num_char <- as.character(num)
5 num_char
6 ## [1] "2.2"
7 class(num_char)
8 ## [1] "character"
9
10 # coerce character to numeric!
11 class(char)
12 ## [1] "character"
13 char_num <- as.numeric(char)
14 ## Warning: NAs introduced by coercion
```

# Scalars and Vectors

- A vector with a single element is called a scalar.
- Vectors can contain any single type.
- You can't mix types in a vector.
- NA can mix with any type.



scalar



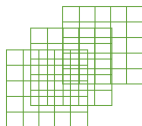
vector

# Matrices and arrays

- A matrix is a vector with additional attributes called *dimensions*.
- Arrays are multidimensional matrices.
- Matrices and arrays can contain only a single type.
- They may also contain NAs.



matrix



array

## Creating matrices and arrays

```
1 my_mat <- matrix(1:16, nrow = 4, byrow = TRUE)
2 my_mat
3 ##           [,1] [,2] [,3] [,4]
4 ## [1,]         1     2     3     4
5 ## [2,]         5     6     7     8
6 ## [3,]         9    10    11    12
7 ## [4,]        13    14    15    16
8
9 my_array <- array(1:16, dim = c(2, 4, 2))
10 my_array
11 ## , , 1
12 ##
13 ##           [,1] [,2] [,3] [,4]
14 ## [1,]         1     3     5     7
15 ## [2,]         2     4     6     8
16 ##
17 ## , , 2
18 ##
19 ##           [,1] [,2] [,3] [,4]
20 ## [1,]         9    11    13    15
21 ## [2,]        10    12    14    16
```



## Optional row and column names

```
1 rownames(my_mat) <- c("A", "B", "C", "D")
2 colnames(my_mat) <- c("a", "b", "c", "d")
3 my_mat
4 ##      a  b  c  d
5 ## A   1  2  3  4
6 ## B   5  6  7  8
7 ## C   9 10 11 12
8 ## D  13 14 15 16
```

## Transpose a matrix

```
1 my_mat_t <- t(my_mat)
2 my_mat_t
3 ##      A B  C  D
4 ## a  1  5  9 13
5 ## b  2  6 10 14
6 ## c  3  7 11 15
7 ## d  4  8 12 16
```

# Diagonal elements

```
1 my_mat_diag <- diag(my_mat)
2 my_mat_diag
3 ## [1]  1  6 11 16
```

# Matrix arithmetic

```
1 mat.1 <- matrix(c(2, 0, 1, 1), nrow = 2)
2                               # notice that the matrix has been filled
3                               # column-wise by default
4 mat.1
5 ##      [,1] [,2]
6 ## [1,]    2    1
7 ## [2,]    0    1
8
9 mat.2 <- matrix(c(1, 1, 0, 2), nrow = 2)
10 mat.2
11 ##      [,1] [,2]
12 ## [1,]    1    0
13 ## [2,]    1    2
14
15 mat.1 + mat.2                # matrix addition
16 ##      [,1] [,2]
17 ## [1,]    3    1
18 ## [2,]    1    3
19 mat.1 * mat.2                # element by element products
20 ##      [,1] [,2]
21 ## [1,]    2    0
22 ## [2,]    0    2
```

# Matrix multiplication

```
1 mat.1 <- matrix(c(2, 0, 1, 1), nrow = 2)
2 mat.1
3 ##      [,1] [,2]
4 ## [1,]    2    1
5 ## [2,]    0    1
6
7 mat.2 <- matrix(c(1, 1, 0, 2), nrow = 2)
8 mat.2
9 ##      [,1] [,2]
10 ## [1,]    1    0
11 ## [2,]    1    2
12
13
14 mat.1 %*% mat.2          # matrix multiplication
15 ##      [,1] [,2]
16 ## [1,]    3    2
17 ## [2,]    1    2
```

# Lists

- Notice the double bracket `[[ ]]` for list items.

```
1 list_1 <- list(c("black", "yellow", "orange"),
2               c(TRUE, TRUE, FALSE, TRUE, FALSE, FALSE),
3               matrix(1:6, nrow = 3))
4 list_1
5 ## [[1]]
6 ## [1] "black" "yellow" "orange"
7 ##
8 ## [[2]]
9 ## [1] TRUE TRUE FALSE TRUE FALSE FALSE
10 ##
11 ## [[3]]
12 ##      [,1] [,2]
13 ## [1,]    1    4
14 ## [2,]    2    5
15 ## [3,]    3    6
```

## List elements can be named

```
1 list_2 <- list(colours = c("black", "yellow", "orange"),
2               evaluation = c(TRUE, TRUE, FALSE,
3                             TRUE, FALSE, FALSE),
4               time = matrix(1:6, nrow = 3))
5 list_2
6 ## $colours
7 ## [1] "black" "yellow" "orange"
8 ##
9 ## $evaluation
10 ## [1] TRUE TRUE FALSE TRUE FALSE FALSE
11 ##
12 ## $time
13 ##      [,1] [,2]
14 ## [1,]    1    4
15 ## [2,]    2    5
16 ## [3,]    3    6
```

## List elements can be renamed using `names`

```
1 names(list_1) <- c("colours", "evaluation", "time")
2 list_1
3 ## $colours
4 ## [1] "black" "yellow" "orange"
5 ##
6 ## $evaluation
7 ## [1] TRUE TRUE FALSE TRUE FALSE FALSE
8 ##
9 ## $time
10 ##      [,1] [,2]
11 ## [1,]    1    4
12 ## [2,]    2    5
13 ## [3,]    3    6
```



## Data frames

treat	nitrogen	block	height	weight	leafarea	shootarea	flowers
tip	medium	1	7.5	7.62	11.7	31.9	1
tip	medium	1	10.7	12.14	14.1	46.0	10
tip	medium	1	11.2	12.76	7.1	66.7	10
tip	medium	1	10.4	8.78	11.9	20.3	1
tip	medium	1	10.4	13.58	14.5	26.9	4
tip	medium	1	9.8	10.08	12.2	72.7	9
notip	low	2	3.7	8.10	10.5	60.5	6
notip	low	2	3.2	7.45	14.1	38.1	4
notip	low	2	3.9	9.19	12.4	52.6	9
notip	low	2	3.3	8.92	11.6	55.2	6
notip	low	2	5.5	8.44	13.5	77.6	9
notip	low	2	4.4	10.60	16.2	63.3	6

- Most used data structure for real world data.
- Each row contains an individual **observation**.
- Each column contains a measured **variable**.
- Each column is a vector of a single type.
- Columns can be different types.

## Data frames

treat	nitrogen	block	height	weight	leafarea	shootarea	flowers
tip	medium	1	7.5	7.62	11.7	31.9	1
tip	medium	1	10.7	12.14	14.1	46.0	10
tip	medium	1	11.2	12.76	7.1	66.7	10
tip	medium	1	10.4	8.78	11.9	20.3	1
tip	medium	1	10.4	13.58	14.5	26.9	4
tip	medium	1	9.8	10.08	12.2	72.7	9
notip	low	2	3.7	8.10	10.5	60.5	6
notip	low	2	3.2	7.45	14.1	38.1	4
notip	low	2	3.9	9.19	12.4	52.6	9
notip	low	2	3.3	8.92	11.6	55.2	6
notip	low	2	5.5	8.44	13.5	77.6	9
notip	low	2	4.4	10.60	16.2	63.3	6

- Each row is an individual petunia plant.
- treat and nitrogen are **categorical** variables (factors)
- treat has 2 levels, tip and notip
- nitrogen has 3 levels, low, medium, high
- height, weight, leafarea, and shootarea are numeric
- flowers is an integer
- block uses integers, but should be treated as a factor

## Data frames

treat	nitrogen	block	height	weight	leafarea	shootarea	flowers
tip	medium	1	7.5	7.62	11.7	31.9	1
tip	medium	1	10.7	12.14	14.1	46.0	10
tip	medium	1	11.2	12.76	7.1	66.7	10
tip	medium	1	10.4	8.78	11.9	20.3	1
tip	medium	1	10.4	13.58	14.5	26.9	4
tip	medium	1	9.8	10.08	12.2	72.7	9
notip	low	2	3.7	8.10	10.5	60.5	6
notip	low	2	3.2	7.45	14.1	38.1	4
notip	low	2	3.9	9.19	12.4	52.6	9
notip	low	2	3.3	8.92	11.6	55.2	6
notip	low	2	5.5	8.44	13.5	77.6	9
notip	low	2	4.4	10.60	16.2	63.3	6

- This type of data is known as **rectangular** or **tidy** data.
- Each column must have the same number of observations.
- Missing data must have NAs in their position.
- Spreadsheets are often NOT tidy.

## Constructing data frames

```
1 p.height <- c(180, 155, 160, 167, 181)
2 p.weight <- c(65, 50, 52, 58, 70)
3 p.names <- c("Joanna", "Charlotte", "Helen", "Karen", "Amy")
4
5 dataf <- data.frame(height = p.height,
6                     weight = p.weight,
7                     names = p.names)
8 dataf
9 ##      height weight      names
10 ## 1      180      65      Joanna
11 ## 2      155      50  Charlotte
12 ## 3      160      52        Helen
13 ## 4      167      58        Karen
14 ## 5      181      70         Amy
```

- Column names are taken from the constructor.
- Can be changed with `names`.
- Numbers at left are row names automatically produced by R, not another column.
- If the vectors are not the same length R will (quietly) cycle!

## Structure of data frames

```
1 dim(dataf)    # 5 rows and 3 columns
2 ## [1] 5 3
3
4 str(dataf)
5 ## 'data.frame':    5 obs. of  3 variables:
6 ## $ height: num  180 155 160 167 181
7 ## $ weight: num   65  50  52  58  70
8 ## $ names : chr  "Joanna" "Charlotte" "Helen" "Karen" ...
```

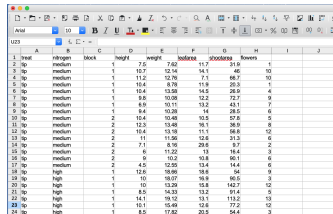
- Gives type, dimensions, column names, types, a few values.
- Convenient to place this in a comment block in your code when dealing with a data frame, for reference.
- R has automatically made names a character vector, not a factor.

## Automatically convert strings to factors

```
1 p.height <- c(180, 155, 160, 167, 181)
2 p.weight <- c(65, 50, 52, 58, 70)
3 p.names <- c("Joanna", "Charlotte", "Helen", "Karen", "Amy")
4
5 dataf <- data.frame(height = p.height,
6                     weight = p.weight,
7                     names = p.names,
8                     stringsAsFactors = TRUE)
9 str(dataf)
10 ## 'data.frame':    5 obs. of  3 variables:
11 ## $ height: num  180 155 160 167 181
12 ## $ weight: num  65 50 52 58 70
13 ## $ names : Factor w/ 5 levels "Amy","Charlotte",...: 4 2 3
    5 1
```

# Preparing data for import

- Easiest way to enter data is use Excel or LibreOffice Calc.
- Save data in tab-delimited or comma-delimited file.
- Keep column headings short.
- No spaces in column headings.
- Avoid special characters, e.g.,  $\text{mm}^2$
- No empty cells! Use NAs.
- Make sure it's tidy!



The screenshot shows a spreadsheet with the following data:

	A	B	C	D	E	F	G	H	I	J
1	test	nitrogen	block	height	weight	lodscore	shootarea	flowers		
2	np	medium	1	7.5	7.62	11.7	31.6	1		
3	np	medium	1	10.7	12.14	14.1	46	10		
4	np	medium	1	11.2	12.76	7.1	68.7	10		
5	np	medium	1	10.4	8.78	11.8	25.5	1		
6	np	medium	1	10.4	13.58	14.5	28.9	4		
7	np	medium	1	9.8	10.08	12.2	72.7	9		
8	np	medium	1	8.9	10.11	13.2	43.1	7		
9	np	medium	1	9.4	10.28	14	28.5	6		
10	np	medium	2	10.4	10.48	10.5	57.8	5		
11	np	medium	2	12.5	13.48	16.1	38.5	8		
12	np	medium	2	10.4	13.18	11.1	68.8	12		
13	np	medium	2	11	11.56	12.6	31.5	6		
14	np	medium	2	7.1	8.16	29.6	9.7	2		
15	np	medium	2	8	11.22	13	18.4	3		
16	np	medium	2	9	10.2	10.8	66.1	6		
17	np	medium	2	4.5	12.93	13.4	14.4	6		
18	np	high	1	12.6	18.86	18.6	54	9		
19	np	high	1	10	18.07	16.9	65.5	3		
20	np	high	1	10	13.29	15.8	142.7	12		
21	np	high	1	8.5	14.33	13.2	91.4	5		
22	np	high	1	14.1	19.12	13.1	113.2	13		
23	np	high	1	10.1	15.49	12.6	77.2	12		
24	np	high	1	8.5	17.82	20.5	54.4	3		

- **Beware!** <https://genomebiology.biomedcentral.com/articles/10.1186/s13059-016-1044-7>

# Importing

```
1 flowers <- read.table(file = 'data/flower.txt',  
2                       header = TRUE, sep = "\t",  
3                       stringsAsFactors = TRUE)
```

- Forward slash works on ALL systems.
- header=TRUE means the first line is variable names.
- sep="\t" for tab-delimited files
- sep=", " for comma-delimited files



# Importing

```
1 > str(flowers)
2 'data.frame':   96 obs. of  8 variables:
3 $ treat      : Factor w/ 2 levels "notip","tip": 2 2 2 2 2 2 2 2 2 2 ...
4 $ nitrogen   : Factor w/ 3 levels "high","low","medium": 3 3 3 3 3 3 3 3 3 3 ...
5 $ block      : int  1 1 1 1 1 1 1 1 2 2 ...
6 $ height     : num  7.5 10.7 11.2 10.4 10.4 9.8 6.9 9.4 10.4 12.3 ...
7 $ weight     : num  7.62 12.14 12.76 8.78 13.58 ...
8 $ leafarea   : num  11.7 14.1 7.1 11.9 14.5 12.2 13.2 14 10.5 16.1 ...
9 $ shootarea  : num  31.9 46 66.7 20.3 26.9 72.7 43.1 28.5 57.8 36.9 ...
10 $ flowers    : int  1 10 10 1 4 9 7 6 5 8 .
```

- After importing check structure of data frame.
- treat and nitrogen have been converted to factors.
- Usually very helpful to include this as a comment block in your script.

## Specialized import functions

```
1 # import .csv file with sep = "\t" and header = FALSE
2 flowers <- read.table(file = 'data/flower.txt')
3
4 # import .csv file with sep = ","
5 # and header = TRUE
6 flowers <- read.csv(file = 'data/flower.csv')
7
8 # import .csv file with dec = "," and sep = ";"
9 # and header = TRUE
10 flowers <- read.csv2(file = 'data/flower.csv')
11
12 # import tab delim file with sep = "\t" and header = TRUE
13 flowers <- read.delim(file = 'data/flower.txt')
```

- Avoid importing from spreadsheet files (.xls etc.).

# Import problems

```
1 Error in file(file, "rt") : cannot open the connection
2 In addition: Warning message:
3 In file(file, "rt") :
4   cannot open file 'flower.txt': No such file or directory
```

- Spelling mistakes.
- Wrong working directory.
- Forgot the extension (`.txt`, `.csv`, *etc.*).

## Forgot header = TRUE

```
1 flowers_bad <- read.table(file = 'data/flower.txt',
2                             sep = "\t")
3 str(flowers_bad)
4 ## 'data.frame':    97 obs. of  8 variables:
5 ## $ V1: chr  "treat" "tip" "tip" "tip" ...
6 ## $ V2: chr  "nitrogen" "medium" "medium" "medium" ...
7 ## $ V3: chr  "block" "1" "1" "1" ...
8 ## $ V4: chr  "height" "7.5" "10.7" "11.2" ...
9 ## $ V5: chr  "weight" "7.62" "12.14" "12.76" ...
10 ## $ V6: chr  "leafarea" "11.7" "14.1" "7.1" ...
11 ## $ V7: chr  "shootarea" "31.9" "46" "66.7" ...
12 ## $ V8: chr  "flowers" "1" "10" "10" ...
```

- All of our variables are character (or factor).
- The first value of each variable is the column name.
- R has provided default names, V1, V2, V3, ...

## Alternative loader functions

```
1 library(readr)
2 # import white space delimited files
3 all_data <- read_table(file = 'data/flower.txt',
4                        col_names = TRUE)
5
6 # import comma delimited files
7 all_data <- read_csv(file = 'data/flower.txt')
8
9 # import tab delimited files
10 all_data <- read_delim(file = 'data/flower.txt',
11                       delim = "\t")
12
13 # or use
14 all_data <- read_tsv(file = 'data/flower.txt')
```

- readr is from the **tidyverse** collection of packages.
- Many of the arguments are the same as read.table
- Returns a tibble, which is very similar to a data frame.

# Packages for large datasets

- `read.table`
- `ff`
- `bigmemory`

Do exercise 3, part 1

# Wrangling data frames

```
1 flowers <- read.table(file = 'data/flower.txt', header = TRUE, sep = "\t")
2 str(flowers)
3 ## 'data.frame':    96 obs. of  8 variables:
4 ## $ treat      : chr  "tip" "tip" "tip" "tip" ...
5 ## $ nitrogen   : chr  "medium" "medium" "medium" "medium" ...
6 ## $ block      : int   1 1 1 1 1 1 1 2 2 ...
7 ## $ height     : num   7.5 10.7 11.2 10.4 10.4 9.8 6.9 9.4 10.4 12.3 ...
8 ## $ weight     : num   7.62 12.14 12.76 8.78 13.58 ...
9 ## $ leafarea   : num   11.7 14.1 7.1 11.9 14.5 12.2 13.2 14 10.5 16.1 ...
10 ## $ shootarea : num   31.9 46 66.7 20.3 26.9 72.7 43.1 28.5 57.8 36.9 ...
11 ## $ flowers    : int    1 10 10 1 4 9 7 6 5 8 ...
```



## The \$ notation

```
1 flowers$height
2 [1] 7.5 10.7 11.2 10.4 10.4 9.8 6.9 9.4 10.4 12.3 10.4
3 [12] 11.0 7.1 6.0 9.0 4.5 12.6 10.0 10.0 8.5 14.1 10.1
4 [23] 8.5 6.5 11.5 7.7 6.4 8.8 9.2 6.2 6.3 17.2 8.0
5 [34] 8.0 6.4 7.6 9.7 12.3 9.1 8.9 7.4 3.1 7.9 8.8
6 [45] 8.5 5.6 11.5 5.8 5.6 5.3 7.5 4.1 3.5 8.5 4.9
7 [56] 2.5 5.4 3.9 5.8 4.5 8.0 1.8 2.2 3.9 8.5 8.5
8 [67] 6.4 1.2 2.6 10.9 7.2 2.1 4.7 5.0 6.5 2.6 6.0
9 [78] 9.3 4.6 5.2 3.9 2.3 5.2 2.2 4.5 1.8 3.0 3.7
10 [89] 2.4 5.7 3.7 3.2 3.9 3.3 5.5 4.4
11
12 f_height <- flowers$height
13 mean(f_height)
14 ## [1] 6.839583
15 summary(f_height)
16 ##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
17 ##      1.200   4.475   6.450   6.840   9.025  17.200
18
19 mean(flowers$height)
20 ## [1] 6.839583
21 summary(flowers$height)
22 ##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
23 ##      1.200   4.475   6.450   6.840   9.025  17.200
```

# Positional indexes

```
1 flowers[1, 4]
2 ## [1] 7.5
3
4 # this would give you the same
5 flowers$height[1]
6 ## [1] 7.5
```

## Positional indexes

```
1 flowers[1:10, 1:4]
2 ##      treat nitrogen block height
3 ## 1      tip    medium      1    7.5
4 ## 2      tip    medium      1   10.7
5 ## 3      tip    medium      1   11.2
6 ## 4      tip    medium      1   10.4
7 ## 5      tip    medium      1   10.4
8 ## 6      tip    medium      1    9.8
9 ## 7      tip    medium      1    6.9
10 ## 8      tip    medium      1    9.4
11 ## 9      tip    medium      2   10.4
12 ## 10     tip    medium      2   12.3
```

## Positional indexes

```
1 flowers[c(1, 5, 12, 30), c(1, 3, 6, 8)]  
2 ##      treat block leafarea flowers  
3 ## 1      tip      1      11.7      1  
4 ## 5      tip      1      14.5      4  
5 ## 12     tip      2      12.6      6  
6 ## 30     tip      2      11.6      5
```

## Empty index means "all of them"

```
1 flowers[1:8, ]
2 ##      treat  nitrogen  block  height  weight  leafarea  shootarea  flowers
3 ## 1    tip    medium    1      7.5    7.62    11.7      31.9      1
4 ## 2    tip    medium    1     10.7   12.14    14.1      46.0     10
5 ## 3    tip    medium    1     11.2   12.76     7.1      66.7     10
6 ## 4    tip    medium    1     10.4    8.78    11.9      20.3      1
7 ## 5    tip    medium    1     10.4   13.58    14.5      26.9      4
8 ## 6    tip    medium    1      9.8   10.08    12.2      72.7      9
9 ## 7    tip    medium    1      6.9   10.11    13.2      43.1      7
10 ## 8    tip    medium    1      9.4   10.28    14.0      28.5      6
```

## Negative numbers mean "not these"

```
1 flowers[-(1:85), -c(4, 7, 8)]
2 ##      treat nitrogen block weight leafarea
3 ## 86 notip      low      1    6.01     17.6
4 ## 87 notip      low      1    9.93     12.0
5 ## 88 notip      low      1    7.03      7.9
6 ## 89 notip      low      2    9.10     14.5
7 ## 90 notip      low      2    9.05      9.6
8 ## 91 notip      low      2    8.10     10.5
9 ## 92 notip      low      2    7.45     14.1
10 ## 93 notip      low      2    9.19     12.4
11 ## 94 notip      low      2    8.92     11.6
12 ## 95 notip      low      2    8.44     13.5
13 ## 96 notip      low      2   10.60     16.2
```

## Can also use column names

```
1 flowers[1:5, c("treat", "nitrogen", "leafarea")]
2 ##      treat nitrogen leafarea
3 ## 1    tip    medium    11.7
4 ## 2    tip    medium    14.1
5 ## 3    tip    medium     7.1
6 ## 4    tip    medium    11.9
7 ## 5    tip    medium    14.5
```

- More readable and portable than  
flowers[1:5, c(1, 2, 6)]

# Logical indexes

```
1 big_flowers <- flowers[flowers$height > 12, ]
2 big_flowers
3 ##      treat nitrogen block height weight leafarea shootarea flowers
4 ## 10     tip    medium     2   12.3   13.48     16.1     36.9         8
5 ## 17     tip      high     1   12.6   18.66     18.6     54.0         9
6 ## 21     tip      high     1   14.1   19.12     13.1    113.2        13
7 ## 32     tip      high     2   17.2   19.20     10.9     89.9        14
8 ## 38     tip      low      1   12.3   11.27     13.7     28.7         5
```



# Logical indexes

```
1 nit_high <- flowers[flowers$nitrogen == "high", ]
2 nit_high
3 ##      treat  nitrogen  block  height  weight  leafarea  shootarea  flowers
4 ## 17     tip      high     1    12.6   18.66     18.6      54.0         9
5 ## 18     tip      high     1    10.0   18.07     16.9      90.5         3
6 ## 19     tip      high     1    10.0   13.29     15.8     142.7        12
7 ## 20     tip      high     1     8.5   14.33     13.2      91.4         5
8 ## 21     tip      high     1    14.1   19.12     13.1     113.2        13
9 ## 22     tip      high     1    10.1   15.49     12.6      77.2        12
10 ....
```

# Logical indexes

```
1 low_notip_high6 <- flowers[flowers$height >= 6 &
2                             flowers$nitrogen == "medium" &
3                             flowers$treat == "notip", ]
4 low_notip_high6
5 ##      treat  nitrogen  block  height  weight  leafarea  shootarea  flowers
6 ##  51  notip    medium     1     7.5   13.60     13.6    122.2      11
7 ##  54  notip    medium     1     8.5   10.04     12.3    113.6       4
8 ##  61  notip    medium     2     8.0   11.43     12.6     43.2      14
```

## Logical indexes using subset

```
1 tip_med_2 <- subset(flowers, treat == "tip" &
2                       nitrogen == "medium" &
3                       block == 2)
4 tip_med_2
5 ##      treat nitrogen block height weight leafarea shootarea flowers
6 ## 9      tip   medium     2  10.4  10.48    10.5     57.8         5
7 ## 10     tip   medium     2  12.3  13.48    16.1     36.9         8
8 ## 11     tip   medium     2  10.4  13.18    11.1     56.8        12
9 ## 12     tip   medium     2  11.0  11.56    12.6     31.3         6
10 ## 13     tip   medium     2   7.1   8.16    29.6      9.7         2
11 ## 14     tip   medium     2   6.0  11.22    13.0     16.4         3
12 ## 15     tip   medium     2   9.0  10.20    10.8     90.1         6
13 ## 16     tip   medium     2   4.5  12.55    13.4     14.4         6
```

- We don't need to use `flowers$treat`

## Logical indexes using subset and select

```
1 tipplants <- subset(flowers ,
2                       treat == "tip" &
3                       nitrogen == "medium" &
4                       block == 2,
5                       select = c("treat", "nitrogen", "leafarea"))
6 tipplants
7 ##      treat  nitrogen  leafarea
8 ##  9      tip    medium     10.5
9 ## 10      tip    medium     16.1
10 ## 11      tip    medium     11.1
11 ## 12      tip    medium     12.6
12 ## 13      tip    medium     29.6
13 ## 14      tip    medium     13.0
14 ## 15      tip    medium     10.8
15 ## 16      tip    medium     13.4
```

- Selecting both rows and columns.

# Ordering data frames

```
1 height_ord <- flowers[order(flowers$height), ]
2 height_ord
3 ##      treat nitrogen block height weight leafarea shootarea flowers
4 ## 68 notip      high     1   1.2  18.24    16.6    148.1      7
5 ## 62 notip    medium     2   1.8  10.47    11.8    120.8      9
6 ## 86 notip      low     1   1.8   6.01    17.6     46.2      4
7 ## 72 notip      high     1   2.1  19.15    15.6    176.7      6
8 ## 63 notip    medium     2   2.2  10.70    15.3     97.1      7
9 ## 84 notip      low     1   2.2   9.97     9.6     63.1      2
10 ## 82 notip      low     1   2.3   7.28    13.8     32.8      6
11 ## 89 notip      low     2   2.4   9.10    14.5     78.7      8
12 ## 56 notip    medium     1   2.5  14.85    17.5     77.8     10
13 ## 69 notip      high     1   2.6  16.57    17.1    141.1      3
14 ## 76 notip      high     2   2.6  18.88    16.4    181.5     14
15 ## 87 notip      low     1   3.0   9.93    12.0     56.6      6
16 ## 42  tip      low     2   3.1   8.74    16.1     39.1      3
17 ...
```

# Ordering data frames

```
1 leafarea_ord <- flowers[order(flowers$leafarea , decreasing = TRUE), ]
2 leafarea_ord
3 ##      treat nitrogen block height weight leafarea shootarea flowers
4 ## 70 notip      high     1  10.9  17.22    49.2    189.6      17
5 ## 13  tip      medium    2   7.1   8.16    29.6     9.7       2
6 ## 24  tip      high     1   6.5  17.13    24.1    147.4       6
7 ## 65 notip      high     1   8.5  22.53    20.8    166.9      16
8 ## 23  tip      high     1   8.5  17.82    20.5     54.4       3
9 ## 66 notip      high     1   8.5  17.33    19.8    184.4      12
10 ## 73 notip      high     2   4.7  13.42    19.8    124.7       5
11 ## 80 notip      high     2   5.2  17.70    19.1    181.1       8
12 ## 17  tip      high     1  12.6  18.66    18.6     54.0       9
13 ## 49 notip      medium    1   5.6  11.03    18.6     49.9       8
14 ## 78 notip      high     2   9.3  18.75    18.4    181.1      16
15 ...
```

# Ordering data frames

```
1 block_height_ord <- flowers[order(flowers$block, flowers$height), ]
2 block_height_ord
3 ##      treat  nitrogen  block  height  weight  leafarea  shootarea  flowers
4 ## 68 notip      high     1     1.2  18.24    16.6    148.1      7
5 ## 86 notip      low      1     1.8   6.01    17.6     46.2      4
6 ## 72 notip      high     1     2.1  19.15    15.6    176.7      6
7 ## 84 notip      low      1     2.2   9.97     9.6     63.1      2
8 ## 82 notip      low      1     2.3   7.28    13.8     32.8      6
9 ## 56 notip    medium     1     2.5  14.85    17.5     77.8     10
10 ## 69 notip      high     1     2.6  16.57    17.1    141.1      3
11 ....
12 ## 38  tip      low      1    12.3  11.27    13.7     28.7      5
13 ## 17  tip      high     1    12.6  18.66    18.6     54.0      9
14 ## 21  tip      high     1    14.1  19.12    13.1    113.2     13
15 ## 62 notip    medium     2     1.8  10.47    11.8    120.8      9
16 ## 63 notip    medium     2     2.2  10.70    15.3     97.1      7
17 ## 89 notip      low      2     2.4   9.10    14.5     78.7      8
18 ## 76 notip      high     2     2.6  18.88    16.4    181.5     14
19 ## 42  tip      low      2     3.1   8.74    16.1     39.1      3
20 ## 92 notip      low      2     3.2   7.45    14.1     38.1      4
21 ....
```

# Ordering data frames

```
1 block_revheight_ord <- flowers[order(flowers$block, -flowers$height), ]
2 block_revheight_ord
3 ##      treat  nitrogen  block  height  weight  leafarea  shootarea  flowers
4 ## 21      tip        high     1   14.1   19.12     13.1    113.2      13
5 ## 17      tip        high     1   12.6   18.66     18.6     54.0      9
6 ## 38      tip        low      1   12.3   11.27     13.7     28.7      5
7 ## 3       tip       medium    1   11.2   12.76      7.1     66.7     10
8 ## 70    notip       high     1   10.9   17.22     49.2    189.6     17
9 ## 2       tip       medium    1   10.7   12.14     14.1     46.0     10
10 ## 4       tip       medium    1   10.4    8.78     11.9     20.3      1
11 ## 5       tip       medium    1   10.4   13.58     14.5     26.9      4
12 ## 22      tip        high     1   10.1   15.49     12.6     77.2     12
13 ....
14 ## 72    notip       high     1    2.1   19.15     15.6    176.7      6
15 ## 86    notip       low      1    1.8    6.01     17.6     46.2      4
16 ## 68    notip       high     1    1.2   18.24     16.6    148.1      7
17 ## 32      tip        high     2   17.2   19.20     10.9     89.9     14
18 ## 10      tip       medium    2   12.3   13.48     16.1     36.9      8
19 ## 25      tip        high     2   11.5   23.89     14.3    101.5     12
20 ## 47      tip        low      2   11.5    8.72     10.2     28.3      6
21 ## 12      tip       medium    2   11.0   11.56     12.6     31.3      6
22 .....
```



# Ordering data frames

```
1 revheight_ord <- flowers[order(-xtfrm(flowers$nitrogen), flowers$height), ]
2 revheight_ord
3 ##      treat  nitrogen  block  height  weight  leafarea  shootarea  flowers
4 ## 62 notip    medium     2     1.8   10.47    11.8     120.8        9
5 ## 63 notip    medium     2     2.2   10.70    15.3     97.1        7
6 ## 56 notip    medium     1     2.5   14.85    17.5     77.8       10
7 ## 53 notip    medium     1     3.5   12.93    16.6    109.3        3
8 ## 58 notip    medium     2     3.9    9.07     9.6     90.4        7
9 ## 64 notip    medium     2     3.9   12.97    17.0     97.5        5
10 ## 52 notip    medium     1     4.1   12.58    13.9    136.6       11
11 ....
12 ## 2      tip    medium     1    10.7   12.14    14.1     46.0       10
13 ## 12     tip    medium     2    11.0   11.56    12.6     31.3        6
14 ## 3      tip    medium     1    11.2   12.76     7.1     66.7       10
15 ## 10     tip    medium     2    12.3   13.48    16.1     36.9        8
16 ## 86 notip     low      1     1.8    6.01    17.6     46.2        4
17 ## 84 notip     low      1     2.2    9.97     9.6     63.1        2
18 ## 82 notip     low      1     2.3    7.28    13.8     32.8        6
19 ## 89 notip     low      2     2.4    9.10    14.5     78.7        8
20 ....
21 ## 39     tip     low      1     9.1    8.96     9.7     23.8        3
22 ## 37     tip     low      1     9.7    6.49     8.1     18.0        3
23 ## 47     tip     low      2    11.5    8.72    10.2     28.3        6
24 ## 38     tip     low      1    12.3   11.27    13.7     28.7        5
25 ## 68 notip     high     1     1.2   18.24    16.6    148.1        7
26 ## 72 notip     high     1     2.1   19.15    15.6    176.7        6
27 ## 69 notip     high     1     2.6   16.57    17.1    141.1        3
28 ....
```

# Ordering data frames

```
1 flowers$nitrogen <- factor(flowers$nitrogen ,
2                             levels = c("low", "medium", "high"))
3 nit_ord <- flowers[order(flowers$nitrogen),]
4 nit_ord
5 ##      treat  nitrogen  block  height  weight  leafarea  shootarea  flowers
6 ## 33    tip      low      1      8.0    6.88      9.3      16.1      4
7 ## 34    tip      low      1      8.0   10.23     11.9     88.1      4
8 ## 35    tip      low      1      6.4    5.97      8.7      7.3      2
9 ## 36    tip      low      1      7.6   13.05      7.2     47.2      8
10 ## 37    tip      low      1      9.7    6.49      8.1     18.0      3
11 ## 38    tip      low      1     12.3   11.27     13.7     28.7      5
12 ....
13 ## 93 notip      low      2      3.9    9.19     12.4     52.6      9
14 ## 94 notip      low      2      3.3    8.92     11.6     55.2      6
15 ## 95 notip      low      2      5.5    8.44     13.5     77.6      9
16 ## 96 notip      low      2      4.4   10.60     16.2     63.3      6
17 ## 1     tip     medium     1      7.5    7.62     11.7     31.9      1
18 ## 2     tip     medium     1     10.7   12.14     14.1     46.0     10
19 ## 3     tip     medium     1     11.2   12.76      7.1     66.7     10
20 ....
21 ## 61 notip     medium     2      8.0   11.43     12.6     43.2     14
22 ## 62 notip     medium     2      1.8   10.47     11.8    120.8      9
23 ## 63 notip     medium     2      2.2   10.70     15.3     97.1      7
24 ## 64 notip     medium     2      3.9   12.97     17.0     97.5      5
25 ## 17    tip      high      1     12.6   18.66     18.6     54.0      9
26 ## 18    tip      high      1     10.0   18.07     16.9     90.5      3
27 ## 19    tip      high      1     10.0   13.29     15.8    142.7     12
28 ## 20    tip      high      1      8.5   14.33     13.2     91.4      5
29 .....
```

## Adding rows to a data frame

```
1 # rbind for rows
2 df1 <- data.frame(id = 1:4, height = c(120, 150, 132, 122),
3                   weight = c(44, 56, 49, 45))
4 df1
5 ##      id height weight
6 ## 1    1    120     44
7 ## 2    2    150     56
8 ## 3    3    132     49
9 ## 4    4    122     45
10
11 df2 <- data.frame(id = 5:6, height = c(119, 110),
12                  weight = c(39, 35))
13 df2
14 ##      id height weight
15 ## 1    5    119     39
16 ## 2    6    110     35
```

## Adding rows to a data frame

```
1 df_rcomb <- rbind(df1 , df2)
2 df_rcomb
3 ##    id height weight
4 ## 1  1    120     44
5 ## 2  2    150     56
6 ## 3  3    132     49
7 ## 4  4    122     45
8 ## 5  5    119     39
9 ## 6  6    110     35
```

## Adding columns to a data frame

```
1 df3 <- data.frame(id = 1:4, height = c(120, 150, 132, 122),
2                   weight = c(44, 56, 49, 45))
3 df3
4 ##      id height weight
5 ## 1    1    120     44
6 ## 2    2    150     56
7 ## 3    3    132     49
8 ## 4    4    122     45
9
10 df4 <- data.frame(location = c("UK", "CZ", "CZ", "UK"))
11 df4
12 ##      location
13 ## 1          UK
14 ## 2          CZ
15 ## 3          CZ
16 ## 4          UK
```

## Adding columns to a data frame

```
1 df_ccomb <- cbind(df3 , df4)
2 df_ccomb
3 ##      id height weight location
4 ## 1    1    120     44         UK
5 ## 2    2    150     56         CZ
6 ## 3    3    132     49         CZ
7 ## 4    4    122     45         UK
```

## Adding computed columns to a data frame

```
1 df_rcomb$height_log10 <- log10(df_rcomb$height)
2 df_rcomb
3 ##      id height weight height_log10
4 ## 1    1   120    44      2.079181
5 ## 2    2   150    56      2.176091
6 ## 3    3   132    49      2.120574
7 ## 4    4   122    45      2.086360
8 ## 5    5   119    39      2.075547
9 ## 6    6   110    35      2.041393
```

## Converting type of a column

```
1 # convert to a factor
2 df_rcomb$Fid <- factor(df_rcomb$id)
3 df_rcomb
4 ##   id height weight height_log10 Fid
5 ## 1  1    120    44    2.079181    1
6 ## 2  2    150    56    2.176091    2
7 ## 3  3    132    49    2.120574    3
8 ## 4  4    122    45    2.086360    4
9 ## 5  5    119    39    2.075547    5
10 ## 6  6    110    35    2.041393    6
11 str(df_rcomb)
12 ## 'data.frame':    6 obs. of  5 variables:
13 ##  $ id          : int  1 2 3 4 5 6
14 ##  $ height      : num  120 150 132 122 119 110
15 ##  $ weight      : num  44 56 49 45 39 35
16 ##  $ height_log10: num  2.08 2.18 2.12 2.09 2.08 ...
17 ##  $ Fid         : Factor w/ 6 levels "1","2","3","4",...: 1 2 3 4 5 6
```



## Merging data frames

- Suppose we have one data frame with information about some common rocky shore invertebrates, called `taxa`
- And we have another data frame with information about where these invertebrates are usually found, called `zone`
- Can we combine these into one data frame with all information about the invertebrates?

# Merging data frames

```
1 taxa <- data.frame(  
2   GENUS = c("Patella", "Littorina", "Halichondria", "Semibalanus"),  
3   species = c("vulgata", "littoria", "panacea", "balanoides"),  
4   family = c("patellidae", "Littorinidae", "Halichondriidae", "  
   Archaeobalanidae"))  
5  
6 taxa  
7 ##      GENUS      species      family  
8 ## 1    Patella    vulgata    patellidae  
9 ## 2    Littorina  littoria    Littorinidae  
10 ## 3 Halichondria  panacea    Halichondriidae  
11 ## 4 Semibalanus  balanoides  Archaeobalanidae  
12  
13 zone <- data.frame(  
14   genus = c("Laminaria", "Halichondria", "Xanthoria", "Littorina",  
15   "Semibalanus", "Fucus"),  
16   species = c("digitata", "panacea", "parietina", "littoria",  
17   "balanoides", "serratus"),  
18   zone = c("v_low", "low", "v_high", "low_mid", "high", "low_mid"))  
19  
20 zone  
21 ##      genus      species      zone  
22 ## 1    Laminaria  digitata    v_low  
23 ## 2 Halichondria  panacea      low  
24 ## 3    Xanthoria  parietina  v_high  
25 ## 4    Littorina  littoria  low_mid  
26 ## 5 Semibalanus  balanoides    high  
27 ## 6      Fucus    serratus  low_mid
```

## Merging data frames

```
1 taxa_zone <- merge(x = taxa, y = zone)
2 taxa_zone
3 ##      species      GENUS      family      genus      zone
4 ## 1 balanoides Semibalanus Archaeobalanidae Semibalanus high
5 ## 2 littoria   Littorina   Littorinidae   Littorina low_mid
6 ## 3 panacea   Halichondria Halichondriidae Halichondria low
```

- Because the two data frames have a column name in common, R assumes you want to join on that column.
- The joined data frame has both GENUS and genus because they are spelled differently.
- The joined data frame has only those rows with information in BOTH original data frames.

## Merging data frames

```
1 taxa_zone <- merge(x = taxa, y = zone, all = TRUE)
2 taxa_zone
3 ##      species      GENUS      family      genus      zone
4 ## 1 balanoides Semibalanus Archaeobalanidae Semibalanus high
5 ## 2 digitata    <NA>        <NA>        Laminaria  v_low
6 ## 3 littoria    Littorina  Littorinidae  Littorina  low_mid
7 ## 4 panacea     Halichondria Halichondriidae Halichondria low
8 ## 5 parietina   <NA>        <NA>        Xanthoria  v_high
9 ## 6 serratus    <NA>        <NA>        Fucus      low_mid
10 ## 7 vulgata     Patella     patellidae    <NA>      <NA>
```

- To include ALL data from BOTH frames use `all = TRUE`
- NAs will be substituted for missing data.

# Merging data frames

- Use `by.x` and `by.y` if the names are different

```
1 taxa_zone <- merge(x = taxa, y = zone,
2                     by.x = "GENUS",
3                     by.y = "genus",
4                     all = TRUE)
5 taxa_zone
6 ##      GENUS species.x      family species.y zone
7 ## 1      Fucus      <NA>      <NA>    serratus low_mid
8 ## 2 Halichondria panacea Halichondriidae panacea low
9 ## 3      Laminaria      <NA>      <NA>    digitata v_low
10 ## 4      Littorina littoria Littorinidae littoria low_mid
11 ## 5      Patella vulgata patellidae      <NA>      <NA>
12 ## 6 Semibalanus balanoides Archaeobalanidae balanoides high
13 ## 7      Xanthoria      <NA>      <NA>    parietina v_high
```

# Merging data frames

- Can use multiple columns

```
1 taxa_zone <- merge(x = taxa, y = zone,
2                     by.x = c("species", "GENUS"),
3                     by.y = c("species", "genus"),
4                     all = TRUE)
5 taxa_zone
6 ##      species      GENUS      family      zone
7 ## 1 balanoides Semibalanus Archaeobalanidae    high
8 ## 2 digitata   Laminaria      <NA>    v_low
9 ## 3 littoria   Littorina    Littorinidae low_mid
10 ## 4 panacea   Halichondria Halichondriidae    low
11 ## 5 parietina Xanthoria      <NA>    v_high
12 ## 6 serratus   Fucus        <NA> low_mid
13 ## 7 vulgata    Patella      patellidae    <NA>
```

# Reshaping data frames

```
1 long_data <- data.frame(  
2   subject = rep(c("A", "B", "C", "D"), each = 3),  
3   sex = rep(c("M", "F", "F", "M"), each = 3),  
4   condition = rep(c("control", "cond1", "cond2"), times = 4),  
5   measurement = c(12.9, 14.2, 8.7, 5.2, 12.6, 10.1, 8.9,  
6                   12.1, 14.2, 10.5, 12.9, 11.9))  
7  
8 long_data  
9 ##      subject sex condition measurement  
10 ## 1         A  M   control      12.9  
11 ## 2         A  M   cond1       14.2  
12 ## 3         A  M   cond2        8.7  
13 ## 4         B  F   control        5.2  
14 ## 5         B  F   cond1       12.6  
15 ## 6         B  F   cond2       10.1  
16 ## 7         C  F   control        8.9  
17 ## 8         C  F   cond1       12.1  
18 ## 9         C  F   cond2       14.2  
19 ## 10        D  M   control       10.5  
20 ## 11        D  M   cond1       12.9  
21 ## 12        D  M   cond2       11.9
```

## Reshaping data frames

```
1 wide_data <- data.frame(subject = c("A", "B", "C", "D"),
2                             sex = c("M", "F", "F", "M"),
3                             control = c(12.9, 5.2, 8.9, 10.5),
4                             cond1 = c(14.2, 12.6, 12.1, 12.9),
5                             cond2 = c(8.7, 10.1, 14.2, 11.9))
6 wide_data
7 ##   subject sex control cond1 cond2
8 ## 1      A  M    12.9   14.2    8.7
9 ## 2      B  F     5.2   12.6   10.1
10 ## 3      C  F     8.9   12.1   14.2
11 ## 4      D  M    10.5   12.9   11.9
```



# Reshaping data frames

```
1 long_data
2 ##      subject sex condition measurement
3 ## 1         A   M   control         12.9
4 ## 2         A   M   cond1         14.2
5 ## 3         A   M   cond2          8.7
6 ## 4         B   F   control          5.2
7 ## 5         B   F   cond1         12.6
8 ## 6         B   F   cond2         10.1
9 ## 7         C   F   control          8.9
10 ## 8         C   F   cond1         12.1
11 ## 9         C   F   cond2         14.2
12 ## 10        D   M   control         10.5
13 ## 11        D   M   cond1         12.9
14 ## 12        D   M   cond2         11.9
15 wide_data
16 ##      subject sex control cond1 cond2
17 ## 1         A   M   12.9   14.2   8.7
18 ## 2         B   F    5.2   12.6  10.1
19 ## 3         C   F    8.9   12.1  14.2
20 ## 4         D   M   10.5   12.9  11.9
```

# Reshaping data frames

- Long data: each measurement on separate row.
- Wide data: each sample on separate row, multiple measurements per row.

```
1 long_data
2 ##      subject sex condition measurement
3 ## 1          A  M   control         12.9
4 ## 2          A  M   cond1          14.2
5 ## 3          A  M   cond2           8.7
6 ## 4          B  F   control          5.2
7 ## 5          B  F   cond1          12.6
8 ## 6          B  F   cond2          10.1
9 ## 7          C  F   control          8.9
10 ## 8          C  F   cond1          12.1
11 ## 9          C  F   cond2          14.2
12 ## 10         D  M   control          10.5
13 ## 11         D  M   cond1          12.9
14 ## 12         D  M   cond2          11.9
15 wide_data
16 ##      subject sex control cond1 cond2
17 ## 1          A  M    12.9  14.2   8.7
18 ## 2          B  F     5.2  12.6  10.1
19 ## 3          C  F     8.9  12.1  14.2
20 ## 4          D  M    10.5  12.9  11.9
```

## melt: convert from wide to long

```
1 library(reshape2)
2 wide_data # remind ourselves what the wide format looks like
3 ##      subject sex control cond1 cond2
4 ## 1      A    M    12.9   14.2    8.7
5 ## 2      B    F     5.2   12.6   10.1
6 ## 3      C    F     8.9   12.1   14.2
7 ## 4      D    M    10.5   12.9   11.9
8
9 # convert wide to long
10 my_long_df <- melt(data = wide_data, id.vars = c("subject", "sex"),
11                   measured.vars = c("control", "cond1", "cond2"),
12                   variable.name = "condition", value.name = "measurement")
13 my_long_df
14 ##      subject sex condition measurement
15 ## 1      A    M    control         12.9
16 ## 2      B    F    control          5.2
17 ## 3      C    F    control          8.9
18 ## 4      D    M    control         10.5
19 ## 5      A    M    cond1          14.2
20 ## 6      B    F    cond1          12.6
21 ## 7      C    F    cond1          12.1
22 ## 8      D    M    cond1          12.9
23 ## 9      A    M    cond2           8.7
24 ## 10     B    F    cond2          10.1
25 ## 11     C    F    cond2          14.2
26 ## 12     D    M    cond2          11.9
```

## Formula notation

- Normally variables refer to the data.

```
1 > x <- 1:5
2 > y <- 11:15
3 > x + y
4 [1] 12 14 16 18 20
```

- Occasionally you want to refer to the variables themselves.

```
1 > ~ x + y
2 ~x + y
3 > x ~ y
4 x ~ y
5 > x ~ y + z
```

- These are used in R to express sets of variables, or relations between sets of variables.

## dcast: convert from long to wide

```
1 long_data # remind ourselves what the long format look like
2 ##      subject sex condition measurement
3 ## 1      A    M    control      12.9
4 ## 2      A    M    cond1       14.2
5 ## 3      A    M    cond2       8.7
6 ## 4      B    F    control      5.2
7 ## 5      B    F    cond1       12.6
8 ## 6      B    F    cond2       10.1
9 ## 7      C    F    control      8.9
10 ## 8      C    F    cond1       12.1
11 ## 9      C    F    cond2       14.2
12 ## 10     D    M    control      10.5
13 ## 11     D    M    cond1       12.9
14 ## 12     D    M    cond2       11.9
15
16 # convert long to wide
17 my_wide_df <- dcast(data = long_data, subject + sex ~ condition,
18                    value.var = "measurement")
19 my_wide_df
20 ##      subject sex cond1 cond2 control
21 ## 1      A    M  14.2   8.7   12.9
22 ## 2      B    F  12.6  10.1    5.2
23 ## 3      C    F  12.1  14.2    8.9
24 ## 4      D    M  12.9  11.9   10.5
```

Do exercise 3, part 2

# Summarizing data frames

```
1 summary(flowers)
2 ##      treat      nitrogen      block      height      weight
3 ## Length:96      low :32      Min.   :1.0      Min.   : 1.200      Min.   : 5.790
4 ## Class :character      medium:32      1st Qu.:1.0      1st Qu.: 4.475      1st Qu.: 9.027
5 ## Mode  :character      high :32      Median :1.5      Median : 6.450      Median :11.395
6 ##
7 ##                               3rd Qu.:2.0      3rd Qu.: 9.025      3rd Qu.:14.537
8 ##                               Max.   :2.0      Max.   :17.200      Max.   :23.890
9 ##      leafarea      shootarea      flowers
10 ## Min.   : 5.80      Min.   : 5.80      Min.   : 1.000
11 ## 1st Qu.:11.07      1st Qu.: 39.05      1st Qu.: 4.000
12 ## Median :13.45      Median : 70.05      Median : 6.000
13 ## Mean   :14.05      Mean   : 79.78      Mean   : 7.062
14 ## 3rd Qu.:16.45      3rd Qu.:113.28      3rd Qu.: 9.000
15 ## Max.   :49.20      Max.   :189.60      Max.   :17.000
```

## Summarizing data frame subsets

```
1 summary(flowers[, 4:7])
2 ##      height      weight      leafarea      shootarea
3 ##  Min.   : 1.200   Min.   : 5.790   Min.   : 5.80   Min.   : 5.80
4 ## 1st Qu.: 4.475   1st Qu.: 9.027   1st Qu.:11.07   1st Qu.: 39.05
5 ##  Median : 6.450   Median :11.395   Median :13.45   Median : 70.05
6 ##  Mean   : 6.840   Mean   :12.155   Mean   :14.05   Mean   : 79.78
7 ## 3rd Qu.: 9.025   3rd Qu.:14.537   3rd Qu.:16.45   3rd Qu.:113.28
8 ##  Max.   :17.200   Max.   :23.890   Max.   :49.20   Max.   :189.60
9
10 # or equivalently
11 # summary(flowers[, c("height", "weight", "leafarea", "shootarea")])
```



# Summarizing data frames subsets

```
1 summary(flowers$leafarea)
2 ##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
3 ##      5.80   11.07   13.45   14.05   16.45   49.20
4
5 # or equivalently
6 # summary(flowers[, 6])
```

## Summarizing data frames with table

```
1 table(flowers$nitrogen)
2 ##
3 ##      low  medium   high
4 ##      32     32     32
```

## Summarizing data frames with 2d table

```
1 table(flowers$nitrogen , flowers$treat)
2 ##
3 ##          notip  tip
4 ##    low         16  16
5 ##  medium         16  16
6 ##    high         16  16
```

## xtabs: tables with formula notation

```
1 xtabs(~ nitrogen + treat, data = flowers)
2 ##           treat
3 ## nitrogen notip tip
4 ##    low         16  16
5 ##  medium         16  16
6 ##    high         16  16
```

- Don't need \$ notation

## Summarizing data frames

```
1 xtabs(~ nitrogen + treat + block, data = flowers)
2 ## , , block = 1
3 ##
4 ##          treat
5 ## nitrogen notip tip
6 ##    low      8   8
7 ##  medium      8   8
8 ##    high      8   8
9 ##
10 ## , , block = 2
11 ##
12 ##          treat
13 ## nitrogen notip tip
14 ##    low      8   8
15 ##  medium      8   8
16 ##    high      8   8
```

- xtabs automatically converted block to a factor

## Flattening tables with ftable

```
1 ftable(xtabs(~ nitrogen + treat + block, data = flowers))
2 ##                               block 1 2
3 ## nitrogen treat
4 ## low      notip      8 8
5 ##          tip        8 8
6 ## medium   notip      8 8
7 ##          tip        8 8
8 ## high     notip      8 8
9 ##          tip        8 8
```

## tapply: table apply

- Separately collect rows for each value of a factor.
- Apply mean to each value of a factor.

```
1 tapply(flowers$height, flowers$nitrogen, mean)
2 ##      low      medium      high
3 ## 5.853125 7.012500 7.653125
```

## Apply sd to each value of a factor

```
1 tapply(flowers$height, flowers$nitrogen, sd)
2 ##          low      medium      high
3 ## 2.828425 3.005345 3.483323
```



## Apply summary to each value of a factor

```
1 tapply(flowers$height, flowers$nitrogen, summary)
2 ## $low
3 ##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
4 ##      1.800   3.600   5.550   5.853   8.000   12.300
5 ##
6 ## $medium
7 ##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
8 ##      1.800   4.500   7.000   7.013   9.950   12.300
9 ##
10 ## $high
11 ##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
12 ##      1.200   5.800   7.450   7.653   9.475   17.200
```

## Including na.rm in each value

```
1 tapply(flowers$height, flowers$nitrogen, mean, na.rm = TRUE)
2 ##          low      medium      high
3 ## 5.853125 7.012500 7.653125
```

- Note that mean is the function that needs this information, but it is a parameter to tapply that is automatically passed on.

## tapply with multiple factors

- Make a different group for each combination of values of two factors

```
1 tapply (flowers$height ,
2         list (flowers$nitrogen , flowers$treat) ,
3         mean)
4 ##           notip      tip
5 ## low      3.66875  8.0375
6 ## medium  4.83750  9.1875
7 ## high    5.70625  9.6000
```

When you get tired of writing `flowers$` all the time

```
1 with(flowers, tapply(height, list(nitrogen, treat), mean))
2 ##           notip      tip
3 ## low      3.66875  8.0375
4 ## medium  4.83750  9.1875
5 ## high    5.70625  9.6000
```

## aggregate: an alternative to tapply

```
1 aggregate(flowers[, 4:7],  
2           by = list(nitrogen = flowers$nitrogen),  
3           FUN = mean)  
4 ##      nitrogen      height      weight leafarea shootarea  
5 ## 1      low 5.853125  8.652812 11.14375  45.1000  
6 ## 2    medium 7.012500 11.164062 13.83125  67.5625  
7 ## 3      high 7.653125 16.646875 17.18125 126.6875
```

## aggregate: an alternative to tapply

```
1 aggregate(flowers[, 4:7],  
2           by = list(nitrogen = flowers$nitrogen ,  
3                     treat = flowers$treat),  
4           FUN = mean)  
5 ##      nitrogen treat   height    weight leafarea shootarea  
6 ## 1      low  notip  3.66875    8.289375  12.32500   59.89375  
7 ## 2    medium notip  4.83750   11.316875  14.17500   94.53125  
8 ## 3      high notip  5.70625   16.604375  18.81875  155.31875  
9 ## 4      low   tip   8.03750    9.016250   9.96250   30.30625  
10 ## 5    medium   tip   9.18750   11.011250  13.48750   40.59375  
11 ## 6      high   tip   9.60000   16.689375  15.54375   98.05625
```

## aggregate also accepts formula notation

- How does height depend on nitrogen and treat?

```
1 aggregate(height ~ nitrogen + treat ,  
2           FUN = mean ,  
3           data = flowers)  
4 ##      nitrogen treat   height  
5 ## 1         low notip 3.66875  
6 ## 2      medium notip 4.83750  
7 ## 3         high notip 5.70625  
8 ## 4         low  tip 8.03750  
9 ## 5      medium  tip 9.18750  
10 ## 6         high  tip 9.60000
```

- Also allows data = flowers to avoid flowers\$

## aggregate also accepts subset

```
1 aggregate(height ~ nitrogen + treat ,  
2           FUN = mean,  
3           subset = flowers < 7,  
4           data = flowers)  
5 ##      nitrogen treat    height  
6 ## 1      low  notip  3.533333  
7 ## 2    medium  notip  5.316667  
8 ## 3      high  notip  3.850000  
9 ## 4      low   tip   8.176923  
10 ## 5    medium   tip  8.570000  
11 ## 6      high   tip  7.900000
```



## aggregate also accepts subset

```
1 aggregate(height ~ nitrogen + treat ,  
2           FUN = mean,  
3           subset = block == "1" ,  
4           data = flowers)  
5 ##      nitrogen treat   height  
6 ## 1         low notip   3.3250  
7 ## 2      medium notip   5.2375  
8 ## 3         high notip   5.9250  
9 ## 4         low  tip    8.7500  
10 ## 5      medium  tip    9.5375  
11 ## 6         high  tip   10.0375
```

## Change data only with scripts!

- Never edit a data file!
- All data changes, transforms, *etc.* should be in a script.
- This documents all changes.
- Allows undoing changes.
- Your analysis is now transparent and reproducible.
- Gone are the days of making undocumented changes in Excel!
- Data is **read only**!

But if you *really* want a new data file ...

```
1 write.table(flowers_df2 ,
2             file = 'data/flowers_04_12.txt' ,
3             col.names = TRUE,
4             row.names = FALSE,
5             sep = "\t")
6
7 write.table(flowers_df2 ,
8             file = 'data/flowers_04_12.csv' ,
9             col.names = TRUE,
10            row.names = FALSE,
11            sep = ",")
12
13 write.csv(flowers_df2 ,
14           file = 'data/flowers_04_12.csv' ,
15           row.names = FALSE)
```

## Other export functions: fwrite

- fwrite is faster for large data frames

```
1 library(read.table)
2 fwrite(flowers_df2 ,
3         file = 'data/flowers_04_12.txt' ,
4         sep = "\t")
5
6 fwrite(flowers_df2 ,
7         file = 'data/flowers_04_12.csv')
```

## Other export functions from the tidyverse

```
1 library(readr)
2 write_tsv(flowers_df2, path = 'data/flowers_04_12.txt')
3
4 write_csv(flowers_df2, path = 'data/flowers_04_12.csv')
```

Do exercise 3, part 3