

# STA130 - Class #3: How R You?

Nathan Taback

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# Today's Class

- RStudio user interface
- R Objects
- R Functions
- R Scripts
- R Packages
- R Lists
- R Notation
- R Missing Data
- dplyr

# Announcements

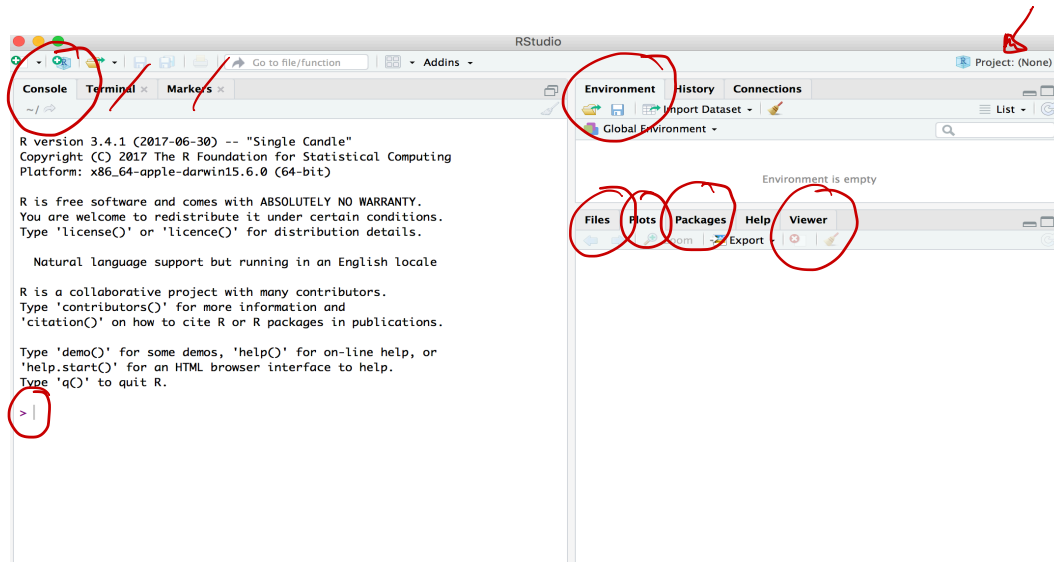
- Tutorial grades will be assigned according to the following marking scheme.

	Mark
Attendance for the entire tutorial	1
Assigned homework completion <sup>a</sup>	1
In-class exercises	4
Total	6

---

- You will learn about the mentorship program in this week's tutorial (3% of final grade).

# RStudio User Interface



# R Objects

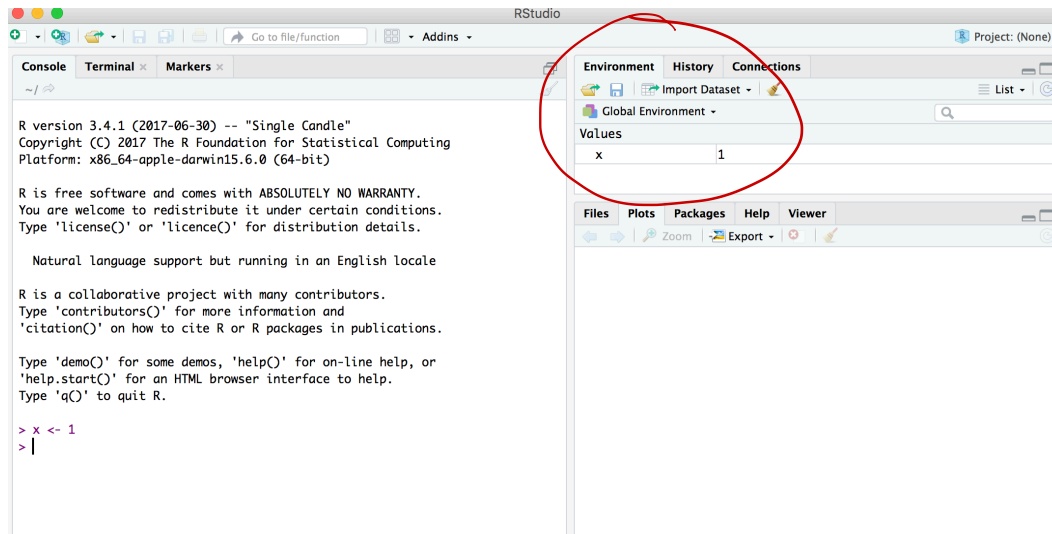
- R lets you save data by storing it inside an R object.
- What's an object? Just a name that you can use to call up stored data.

```
x <- 1  
x
```

```
## [1] 1
```

# Environment Pane in RStudio

- When you create an object, the object will appear in the environment pane of RStudio.



# Functions

$$\text{abs}(x) = |x| = \begin{cases} x, & x \geq 0 \\ -x, & x < 0 \end{cases}$$

- R comes with many functions that you can use to do sophisticated tasks like random sampling.
- For example, you can round a number with the round function `round()`, or calculate its absolute value with `abs()`.
- Write the name of the function and then the data you want the function to operate on in parentheses:

```
round(-2.718282, 2)
```



```
## [1] -2.72
```

```
abs(-5)
```

```
## [1] 5
```

f g

```
abs(round(-2.718282, 2))
```

```
## [1] 2.72
```

→  $\text{abs}(-2.72) = 2.72$

$$(f \circ g)(x) = f(g(x))$$

# Function Constructor

- Every function in R has three basic parts: a name, a body of code, and a set of arguments.
- To make your own function, you need to replicate these parts and store them in an R object, which you can do with the function function.
- To do this, call `function()` and follow it with a pair of braces, `{}`: `my_function <- function() {}`

```
my_function <- function() {  
  x <- 1  
  x  
}
```

`my_function()` — assumed no arguments



# Function Constructor

## Set-Seed (1)

- We can simulate rolling a pair of dice and adding the result with the code:

```
die <- 1:6  
dice <- sample(die, size = 2, replace = TRUE)  
sum(dice)
```

1, 2, 3, 4, 5, 6 die ← c(1, 2, 3, 4, 5, 6)  
→ Chooses two numbers from

```
## [1] 9
```

1, 2, 3, 4, 5, 6 Such that:

1, 1      6, 6

2      -      12

· IS the range of values for sum(dice) -

If the first value is, say,

1, then the second value

is selected from

1, 2, 3, 4, 5, 6

# Function Constructor

- We can create our own function with

```
roll <- function() {  
  die <- 1:6  
  dice <- sample(die, size = 2, replace = TRUE)  
  sum(dice)  
}
```

Call the function `roll()`

```
roll() # call the function. NB: result will differ with every call
```

```
## [1] 4
```

`my_func <- function() {  
 code.  
}`

# Function Arguments

Instead of rolling one die consider rolling four or ten dice then adding the results of all the rolls together.

```
roll2 <- function(numrolls) { # is the argument of the function roll2  
  die <- 1:6  
  dice <- sample(die, size = numrolls, replace = TRUE) # the size of the sample  
  sum(dice) # add up the roll results  
}
```

`numrolls` is called an *argument* of the function `roll2()`.

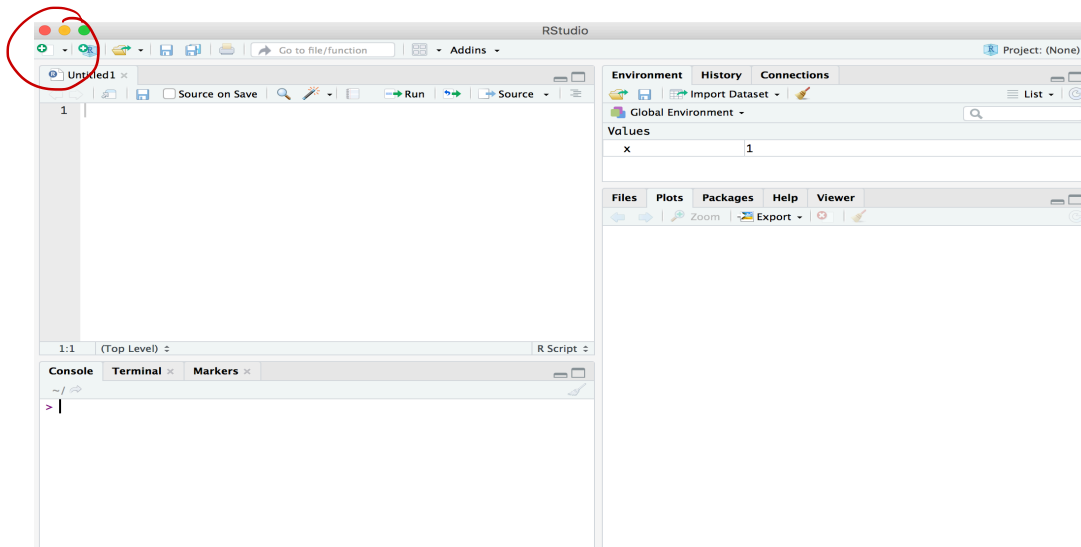
Let's simulate rolling ten dice and adding the results together.

```
roll2(10)
```

```
## [1] 40
```

# Scripts

- If we want to edit the function `roll2()` then we will want to save it in a script.
- To do this in RStudio File > New File > R script in the menu bar.

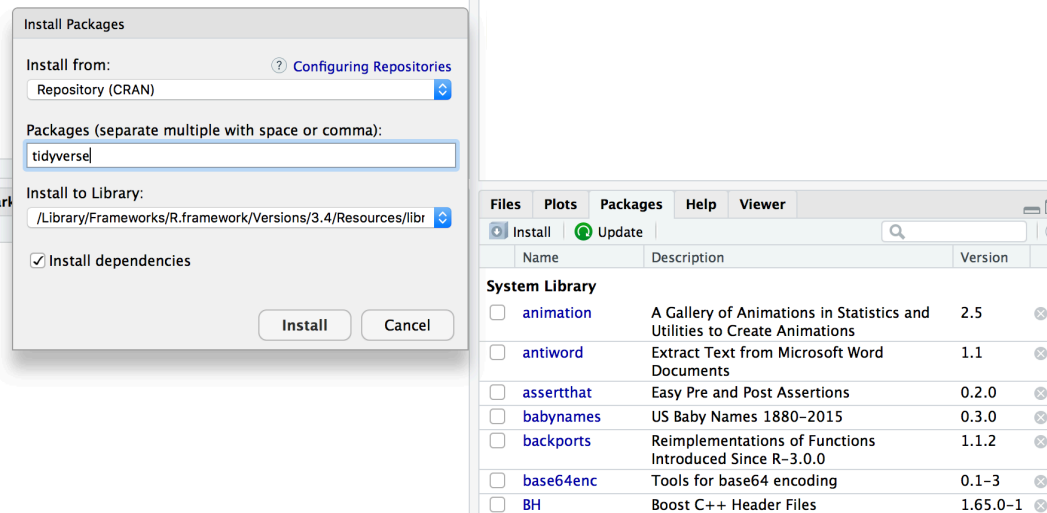


# Packages

- You're not the only person writing your own functions with R.
- Many professors, programmers, and statisticians use R to design tools that can help people analyze data.
- They then make these tools free for anyone to use.
- To use these tools, you just have to download them. They come as preassembled collections of functions and objects called packages.
- We have already used two packages `ggplot2` and `dplyr`.

# Packages

To install the package `tidyverse` in RStudio go to the Packages tab in RStudio and click Install.

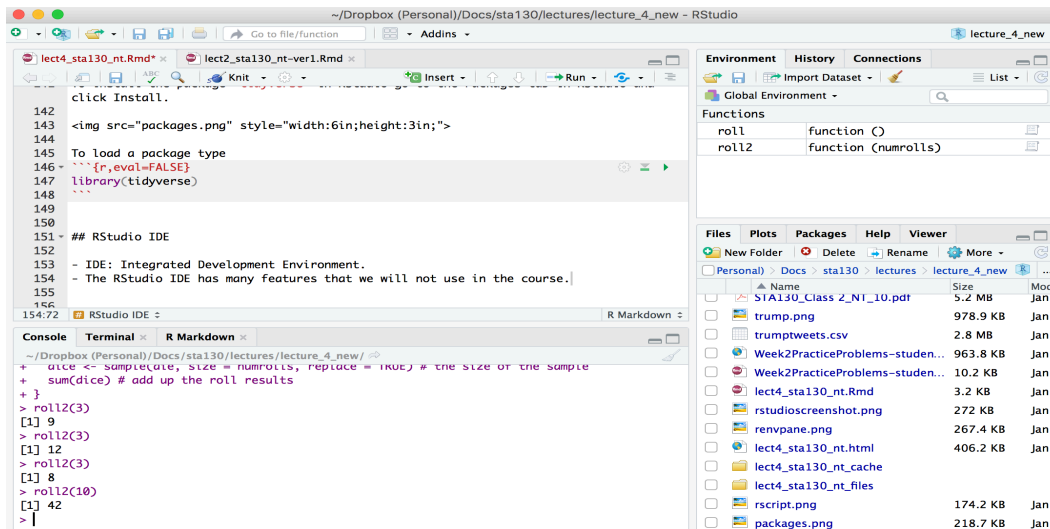


To load a package type

```
library(tidyverse)
```

# RStudio IDE

- IDE: Integrated Development Environment.
- The RStudio IDE has many features that we will not use in the course.



- The **console** is where you can type an R command at the prompt and the result is returned.
- Write code in an R script, R Markdown document, or R Notebook.
- Run a script or R chunks from an R Markdown or R Notebook by pushing the run button in the chunk.

# R Objects

- R stores data in objects such as vectors, arrays, and matrices.
- In most applications we will usually load data from an external file.



# R Objects - Atomic Vectors

You can make an atomic vector by grouping some values of data together with c:

```
die<-c(1,2,3,4,5,6)
```

```
die
```

```
## [1] 1 2 3 4 5 6
```

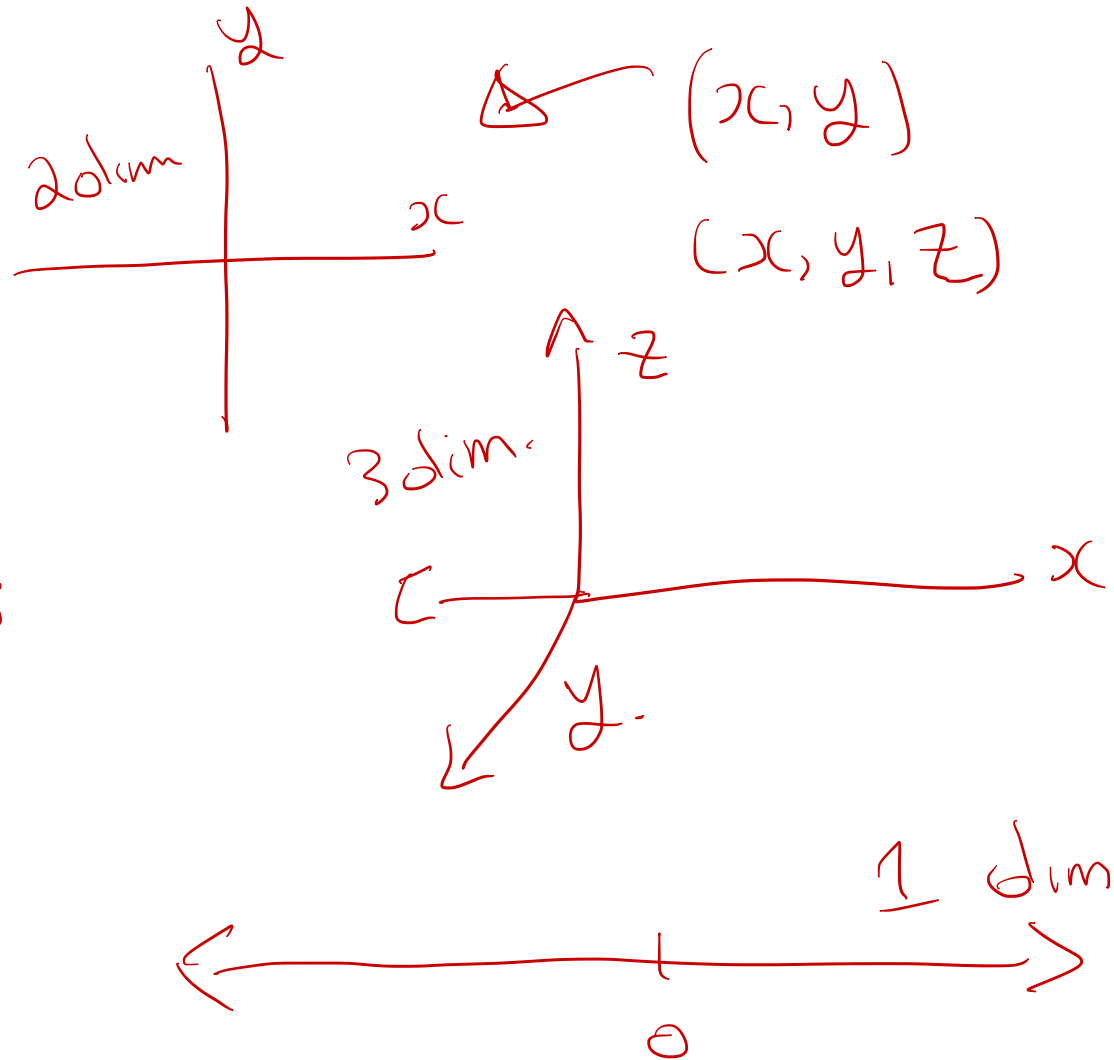
```
is.vector(die)
```

```
## [1] TRUE
```

```
length(die)
```

```
## [1] 6
```

built in  
functions  
in R.



# R Objects - Atomic Vectors

You can also make an atomic vector with just one value. R saves single values as an atomic vector of length 1:

```
two <- 2  
two
```

```
## [1] 2
```

# R Objects - Atomic Vectors: Integer and Character

- Each atomic vector can only store one type of data. You can save different types of data in R by using different types of atomic vectors.
- R recognizes six basic types of atomic vectors: (doubles, integers, characters, logicals) complex, and raw.  $i = \sqrt{-1}$  ~~x~~
- We will not be using complex or raw types in STA130.
- Integer vectors included a capital L with input, and character vectors have input surrounded by quotation marks.

# R Objects - Atomic Vectors: Integer and Character

```
mynums <- c(2L, 3L)
courses <- "STA130"
courses <- c("STA130", "MAT137")
sum(mynums)
```

```
## [1] 5
```

```
sum(courses)
```

```
## Error in sum(courses): invalid 'type' (character) of argument
```

```
sum(courses == "STA130")
```

```
## [1] 1
```

Coercion.

∴  $\text{courses} \equiv \text{"STA130"}$

↳ True  
and TRUE is always 1

# R Objects - Double Vectors

- A double vector stores real numbers. Doubles are often called numerics.

```
die <- c(1,2,3,4,5,6)  
typeof(die)
```

```
## [1] "double"
```

# R Objects - Logical Vectors

- Logical vectors store TRUES and FALSEs, R's form of Boolean data. Logicals are very helpful for doing things like comparisons:

```
3 > 4
```

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

- TRUE or FALSE in capital letters (without quotation marks) will be treated as logical data. R also assumes that T and F are shorthand for TRUE and FALSE.

```
logic <- c(TRUE, FALSE, TRUE)  
logic
```

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

# R Objects - Atomic Vectors: `dim()`

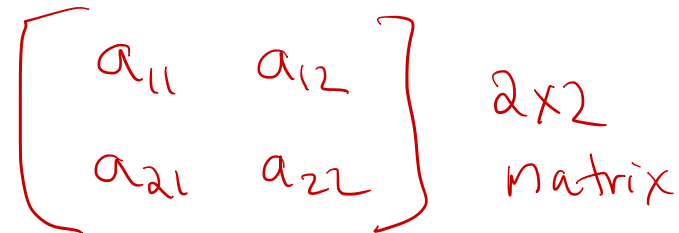
You can transform an atomic vector into an n-dimensional array by giving it a dimensions attribute with `dim`.

```
die <- c(1,2,3,4,5,6)
dim(die) <- c(2,3) # a 2x3 matrix
die
```

```
##      [,1] [,2] [,3]
## [1,]    1    3    5
## [2,]    2    4    6
```

```
die <- c(1,2,3,4,5,6)
dim(die) <- c(3,2) # a 3x2 matrix
die
```

```
##      [,1] [,2]
## [1,]    1    4
## [2,]    2    5
## [3,]    3    6
```



Notation allows for to access values in the matrix

$a_{ij}$  - Value in the  $i^{th}$  row,  $j^{th}$  Column.

`die[,3]`

blank

`die[1,]`

return values of 5, 6.  
return 1, 3, 5

R always fills up each matrix by columns, instead of by rows unless you use `matrix()` or `array()`.

# Factors

- Factors are R's way of storing categorical information, like ethnicity or eye color.
- A factor as something like sex since it can only have certain values.
- Factors very useful for recording the treatment levels of a categorical variable.

```
sex <- factor(c("male", "female", "female", "male"))
typeof(sex)
```

```
## [1] "integer"
```

```
unclass(sex) # shows how R is storing the factor vector
```

```
## [1] 2 1 1 2
## attr(,"levels")
## [1] "female" "male"
```

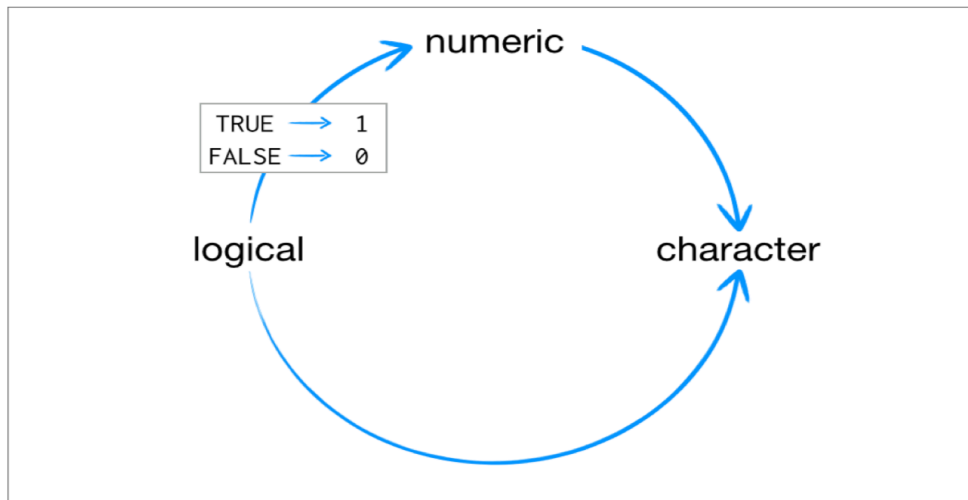
*Handwritten red annotations:*  
A bracket above the first two values (2, 1) is labeled "male".  
A bracket above the last two values (1, 2) is labeled "female".

*Handwritten red definition:*  
Sex = { 2 is "female"  
1 is "male"



# Coercion

R always follows the same rules when it coerces data types. Once you are familiar with these rules, you can use R's coercion behavior to do surprisingly useful things.



1 → "1"

For example `sum(c(TRUE, TRUE, FALSE, FALSE))` will become `sum(c(1, 1, 0, 0))`.

```
sum(c(TRUE, TRUE, FALSE, FALSE))
```

```
## [1] 2
```

# Lists

- Lists are like atomic vectors because they group data into a one-dimensional set.
- Lists do not group together individual values.
- Lists group together R objects, such as atomic vectors and other lists.
- For example, you can make a list that contains a numeric vector of length 31 in its first element, a character vector of length 1 in its second element, and a new list of length 2 in its third element.

```
list1 <- list(1:31, "Prof. Taback", list(TRUE, FALSE))  
list1
```

```
## [[1]]  
## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23  
## [24] 24 25 26 27 28 29 30 31  
##  
## [[2]]  
## [1] "Prof. Taback"  
##  
## [[3]]  
## [[3]][[1]]  
## [1] TRUE  
##  
## [[3]][[2]]  
## [1] FALSE
```

1dim character vector

31 dim vector

2dim logical vector

$list1[[1]]$  first element of  
of list.

$list1[[2]]$  second element of  
list.

$list1[[3]][[3]][[1]]$   
return TRUE

# Data Frames

- Data frames are the two-dimensional version of a list.
- They are the most useful storage structure for data analysis
- A data frame is R's equivalent to the Excel spreadsheet because it stores data in a similar format.

# Data Frames

- Data frames group vectors together into a two-dimensional table.
- Each vector becomes a column in the table.
- As a result, each column of a data frame can contain a different type of data; but within a column, every cell must be the same type of data.

data frame	1	"R"	TRUE
	2	"S"	FALSE
	3	"T"	TRUE
	numeric	character	logical

# Data Frames

```
student_num <- c(1, 2, 3, 4)
name <- c("Nadia", "Shiyi", "Yizhe", "Wei")
mydat <- data.frame(obsnum = student_num, student_name = name)
mydat
```

↳ rename first variable as  
obsnum

```
##  obsnum student_name
## 1      1      Nadia
## 2      2      Shiyi
## 3      3      Yizhe
## 4      4       Wei
```

- Creating a data frame by hand takes a lot of typing, but you can do it with the `data.frame()` function.
- Give `data.frame()` any number of vectors, each separated with a comma.
- Each vector should be set equal to a name that describes the vector.
- `data.frame()` will turn each vector into a column of the new data frame.

# Data Frames

You can view a data frame in RStudio by clicking on the data frame name in the Environment tab

*look in Environment tab*

*Click on this*

obsnum	student_name
1	Nadia
2	Shiyi
3	Yizhe
4	Wei

```
~/Dropbox (Personal)/Docs/sta130/lectures/lecture_4_new/ > view(mydat) > student_num <- c(1, 2, 3, 4) > name <- c("Nadia", "Shiyi", "Yizhe", "Wei") > mydat <- data.frame(student_num, name) > mydat > View(mydat) > student_num <- c(1, 2, 3, 4) > name <- c("Nadia", "Shiyi", "Yizhe", "Wei") > mydat <- data.frame(obsnum = student_num, student_name = name) > mydat > View(mydat) > |
```

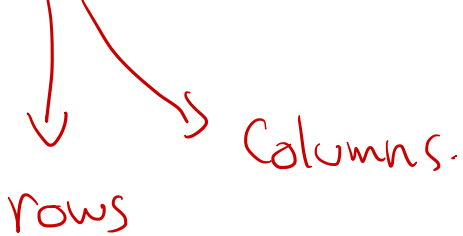
Environment History Connections  
Global Environment  
Data  
mydat 4 obs. of 2 variables  
Values  
name chr [1:4] "Nadia" "Shiyi" "Yizh...  
student\_num num [1:4] 1 2 3 4

Files Plots Packages Help Viewer  
New Folder Delete Rename More  
Personal > Docs > sta130 > lectures > lecture\_4\_new  
Name Size Mod  
lect4\_sta130\_nt.Rmd 8.3 KB Jan  
rstudioscreenshot.png 272 KB Jan  
renvpane.png 267.4 KB Jan  
lect4\_sta130\_nt.html 3.1 MB Jan  
lect4\_sta130\_nt\_cache  
lect4\_sta130\_nt\_files  
rscript.png 174.2 KB Jan  
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rstudioide.png 490.8 KB Jan  
coercion.png 113.8 KB Jan  
dataframe.png 124.2 KB Jan  
Week3\_Testing101.Rmd 20 KB Jan  
dataframeview.png 432.5 KB Jan

# R Notation - [, ]

- To extract a value or set of values from a data frame, write the data frame's name followed by a pair of square brackets with a comma [, ].

mydat[, ]



$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$$A[i, j]$$

# R Notation - [, ]

mydat

```
##  obsnum student_name  
## 1      1      Nadia  
## 2      2      Shiyi  
## 3      3      Yizhe  
## 4      4       Wei
```

first row

```
mydat[1,2] # the value in row 1 and column 2
```

Second Column.

```
## [1] Nadia  
## Levels: Nadia Shiyi Wei Yizhe
```

"Nadia"

```
mydat[c(1,2),2] # all values in rows 1 and 2 in second column
```

mydat[1,2]  
mydat[c(1,2),2]

```
## [1] Nadia Shiyi  
## Levels: Nadia Shiyi Wei Yizhe
```

Student\_name is stored as a factor.



# R Notation - \$

The \$ tells R to return all of the values in a column as a vector.

```
mydat$student_name
```

```
## [1] Nadia Shiyi Yizhe Wei  
## Levels: Nadia Shiyi Wei Yizhe
```

```
vec <- mydat$student_name # assign it to vec  
attributes(vec) # info associated with object vec
```

```
## $levels  
## [1] "Nadia" "Shiyi" "Wei" "Yizhe"  
##  
## $class  
## [1] "factor"
```

```
vec[2] # get second element of vector
```

```
## [1] Shiyi  
## Levels: Nadia Shiyi Wei Yizhe
```

Save as a vector

then access 2<sup>nd</sup>  
element.

# R Notation - combine [,] and \$

did not specify column.

```
mydat[mydat$obsnum == 1,] # first row of data frame and all columns
```

```
##  obsnum student_name  
## 1      1      Nadia
```

```
mydat[mydat$obsnum == 1 | mydat$obsnum == 4 ,] # first and fourth rows of data frame and all columns
```

```
##  obsnum student_name  
## 1      1      Nadia  
## 4      4      Wei
```

or operator.

if  $obsnum = 1$  or  
 $obsnum = 4$  then

TRUE

# Missing Data - **NA**

- Missing information problems happen frequently in data science.
- For example a value is missing because the measurement was lost, corrupted, or never recorded.
- The **NA** character is a special symbol in R. It stands for “not available” and can be used as a placeholder for missing information.

```
1 + NA
```

```
## [1] NA
```

# Missing Data - `na.rm()`

- Suppose you collected the ages of five students, but you forgot to record the fifth student's age.

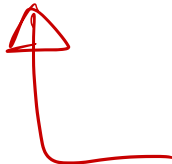
```
age <- c(19, 20, 17, 20, NA)
mean(age) # mean will be NA
```

5<sup>th</sup> element NA missing.

```
## [1] NA
```

```
age <- c(19, 20, 17, 20, NA)
mean(age, na.rm = TRUE) # R will ignore missing values
```

```
## [1] 19
```



# Identify and Set Missing Data - `is.na()`

```
age <- c(19, 20, 17, 20, NA)
is.na(age) # check which elements of age are missing
```

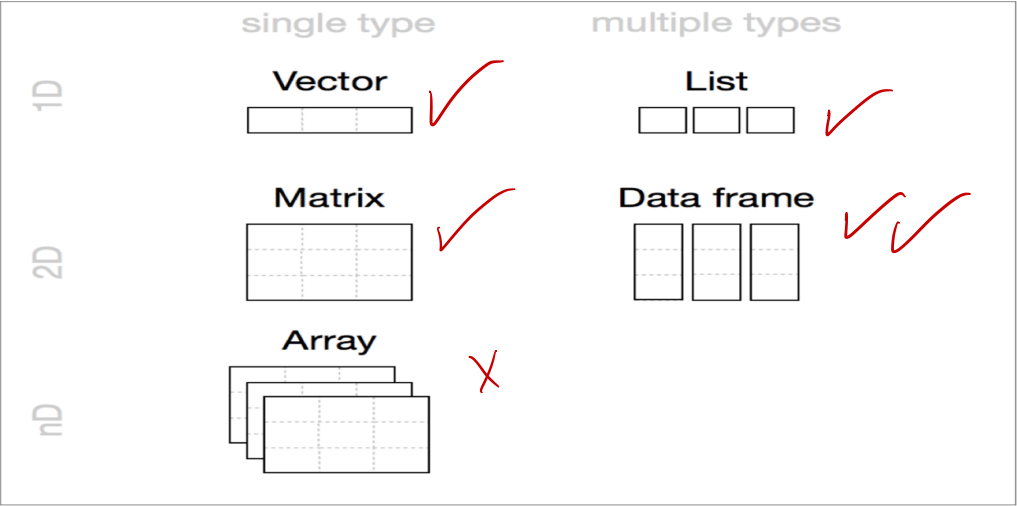
```
## [1] FALSE FALSE FALSE FALSE TRUE
```

```
age[1] <- NA # set the first element of age to NA
age
```

```
## [1] NA 20 17 20 NA
```

*is.na is TRUE if  
the value is missing  
(i.e., the  
value is NA)  
otherwise it  
is FALSE*

# Summary of R Data Structures



# Tidyverse

Tidyverse



## R packages for data science

The tidyverse is an opinionated **collection of R packages** designed for data science. All packages share an underlying philosophy and common APIs.

Install the complete tidyverse with:

```
install.packages("tidyverse")
```

<https://www.tidyverse.org>

# Canadian Flu Rates with **dplyr**

The provincial rates for the week ending January 6, 2018 are in the file `fludat_prov.csv` and the the size of the population in each province is in the file `popdat.csv`. The code below reads the files into R data frames.

```
library(tidyverse)
fludat_prov <- read_csv("fludat_prov.csv") # import data from file
popdat <- read_csv("popdat.csv") # import data from file
```



# Canadian Flu Rates with dplyr

from dplyr lib

```
head(fludat_prov) # head shows the first six rows of a data frame
```

↳ returns first 6 rows.

```
## # A tibble: 6 x 3
##   prov          testpop_size fluA
##   <chr>          <int> <int>
## 1 Newfoundland         96     12
## 2 Prince Edward Island    64     11
## 3 Nova Scotia          144     23
## 4 New Brunswick         347     80
## 5 Province of Québec   6361   1190
## 6 Province of Ontario   2320    344
```

# of people tested for flu

# of people + for fluA.

$fluA / testpop\_size$

e.g.)  $12 / 96$  in Nfld.

```
head(popdat)
```

```
## # A tibble: 6 x 3
##   prov          prov_pop_size region
##   <chr>          <int> <chr>
## 1 Nunavut         35944 Territories
## 2 Alberta        4067175 <NA>
## 3 Saskatchewan   1098352 West
## 4 Yukon           35874 Territories
## 5 Manitoba       1278365 West
## 6 British Columbia 4648055 West
```

# Canadian Flu Rates with `dplyr`

How many Provinces/Territories are in the `fludat_prov` data frame?

```
fludat_prov %>% summarise(numprov = n()) # n() counts the number of rows in the data frame
```

```
## # A tibble: 1 x 1  
##   numprov  
##   <int>  
## 1      13
```

returns the number of  
Rows in a dataframe.

`Summarise(fludat_prov, numprov = n())`.

# Canadian Flu Rates with `dplyr`

Do any variables in `fludat` or `popdat` have missing values?

```
fludat_prov %>% filter(is.na(prov) == TRUE | is.na(testpop_size) == TRUE | is.na(fluA) == TRUE)
```

```
## # A tibble: 0 x 3  
## # ... with 3 variables: prov <chr>, testpop_size <int>, fluA <int>
```

*0 = no missing.*

```
popdat %>% filter(is.na(prov) == TRUE | is.na(prov_pop_size) == TRUE | is.na(region) == TRUE)
```

```
## # A tibble: 2 x 3  
##   prov      prov_pop_size region  
##   <chr>          <int> <chr>  
## 1 Alberta      4067175 <NA>  
## 2 Quebec       8164361 <NA>
```

# Canadian Flu Rates with **dplyr**

Recode specific values using R data frame notation [,] and \$.

```
popdat$region[popdat$prov == "Alberta"] <- "West" #recode only the region value for Alberta
popdat$region[popdat$prov == "Quebec"] <- "East" #recode only the region value for Alberta
popdat$region #print region variable in popdat data
```

```
## [1] "Territories" "West" "West" "Territories" "West"
## [6] "West" "East" "East" "Atlantic" "Atlantic"
## [11] "Territories" "Atlantic" "Atlantic"
```

# Canadian Flu Rates with `dplyr` - Joining Two Tables with `inner_join()`

We can join two data frames with `inner_join(x,y)`: return all rows from x where there are matching values in y, and all columns from x and y. If there are multiple matches between x and y, all combination of the matches are returned.

```
fludat_prov %>% inner_join(popdat, by = "prov")
```

```
## # A tibble: 9 x 5
```

```
##   prov          testpop_size fluA prov_pop_size region
##   <chr>          <int> <int>          <int> <chr>
## 1 Newfoundland      96    12          519716 Atlantic
## 2 Prince Edward Island  64    11          142907 Atlantic
## 3 Nova Scotia       144    23          923598 Atlantic
## 4 New Brunswick     347    80          747101 Atlantic
## 5 Manitoba          849   186         1278365 West
## 6 British Columbia 1078   198         4648055 West
## 7 Yukon              15     1           35874 Territories
## 8 Northwest Territories 28    10           41786 Territories
## 9 Nunavut            18     1            35944 Territories
```

*returns a data frame with 9 rows.*

Why are there only 9 observations when there are 13 Provinces/Territories?

# Canadian Flu Rates with `dplyr` - Joining Two Tables with `inner_join()`

```
fludat_prov$prov
```

```
## [1] "Newfoundland"           "Prince Edward Island"  
## [3] "Nova Scotia"           "New Brunswick"  
## [5] "Province of Québec"    "Province of Ontario"  
## [7] "Manitoba"              "Province of Saskatchewan"  
## [9] "Province of Alberta"    "British Columbia"  
## [11] "Yukon"                 "Northwest Territories"  
## [13] "Nunavut"
```

```
popdat$prov
```

```
## [1] "Nunavut"                "Alberta"  
## [3] "Saskatchewan"          "Yukon"  
## [5] "Manitoba"              "British Columbia"  
## [7] "Ontario"               "Quebec"  
## [9] "Prince Edward Island"  "Newfoundland"  
## [11] "Northwest Territories" "Nova Scotia"  
## [13] "New Brunswick"
```

*not the same*

Province needs to be recoded. Exercise on this week's practice problems.

# Canadian Flu Rates with `dplyr` - Joining Two Tables with `inner_join()`

X			Y		
A	B	C	A	B	D
a	t	1	a	t	3
b	u	2	b	u	2
c	v	3	d	w	1

`inner_join(X, Y)`

A	B	C	D
a	t	1	3
b	u	2	2