

# BASIC R

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# INTRO TO R

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- <https://jabranham.com/learn-r>
- pdf available
- html too for easy copy/paste

*R is "GNU S", a freely available language and environment for statistical computing and graphics which provides a wide variety of statistical and graphical techniques: linear and nonlinear modelling, statistical tests, time series analysis, classification, clustering, etc*

# WHY USE R?

- Free & open source
- Available on nearly every platform
- Extensible (CRAN)
  - We'll be using base R in this workshop
- Documentation & community
- Graphics
- Nerd cred

if you have installed Rstudio, then open it else open R

- File > open new R script
- Top left: editor
- Bottom left: R console
- **Tip:** nearly all your commands should be typed in an R script, then sent to the console for evaluation
  - exceptions: `install.packages()`, help queries

## DATA TYPES

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- What are basic data types?
- logical, numeric, character
  - also complex and raw, but we'll ignore those

- Statement of truth-y-ness

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

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

```
3 != 4
```

```
[1] TRUE
```

```
"Democat" %in% c("Democrat", "Independent", "Republican")
```

```
[1] FALSE
```

## Statements of truthyness

7 == 2

3 != 7

7 >= 2

2 >= 7

## And/or

2 == 2 & 2 > 3

2 == 2 | 2 > 3

```
## Statements of truthyness
```

```
7 == 2 # FALSE
```

```
3 != 7 # TRUE
```

```
7 >= 2 # TRUE
```

```
2 >= 7 # FALSE
```

```
## And/or
```

```
2 == 2 & 2 > 3 # FALSE
```

```
2 == 2 | 2 > 3 # TRUE
```

- numeric is umbrella term for "double" and "integers"
- stores numbers:

```
(x <- c(1, exp(1), pi, 10384.287459))
```

```
[1]      1.000000      2.718282      3.141593 10384.287459
```

```
is.numeric(x)
```

```
[1] TRUE
```

- character type represents letters/words:

```
(myname <- c("My name is Alex"))
```

```
[1] "My name is Alex"
```

```
(myname2 <- c("My", "name", "is", "Alex"))
```

```
[1] "My" "name" "is" "Alex"
```

```
length(myname)
```

```
[1] 1
```

```
length(myname2)
```

- vectors can have only one data type:

```
(x <- c("My name", 3 == 4, 7.27))
```

```
[1] "My name" "FALSE"    "7.27"
```

```
class(x)
```

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

- anything can be coerced to a character
- logicals can be coerced to numeric
  - TRUE is 1, FALSE is 0



- all the above are called **atomic vectors**
- useful to remember this when R yells at you

- sometimes we need to store more than one type of data
- we can do this with a list

```
list(c(1.82, 1940, 93.20, 192.917),  
     c("Beyonce", "Lady Gaga", "Pink"),  
     c(TRUE, FALSE, TRUE, TRUE, FALSE, TRUE, FALSE, FALSE))
```

```
[[1]]
```

```
[1] 1.820 1940.000 93.200 192.917
```

```
[[2]]
```

```
[1] "Beyonce" "Lady Gaga" "Pink"
```

```
[[3]]
```

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

- we can name elements of vectors:

```
list(n = 10,  
     dv = c(1, 3, 5, 7, 9, 19, 92, 4, 10, 4))
```

```
$n  
[1] 10
```

```
$dv  
[1] 1 3 5 7 9 19 92 4 10 4
```

- all the vectors we've worked with so far have been single-dimension
- but we often work with two dimensional data
  - rows are observations
  - columns are variables

```
matrix(c(1, 2, 3, 4, 5, 6), nrow = 2)
```

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

- matrix where columns can be different types:

```
data.frame(x = 1:3,  
           y = c("a", "b", "c"),  
           z = c(TRUE, FALSE, TRUE))
```

	x	y	z
1	1	a	TRUE
2	2	b	FALSE
3	3	c	TRUE

	homogeneous	heterogeneous
1d	atomic vector	list
2d	matrix	data.frame
nd	array	



## SUBSETTING

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- oftentimes, we are interested in subsetting
- how to refer to a specific column or row?

[

- the `[]` function is how we subset

```
x <- 1:10
```

```
x[c(1, 7)]
```

```
[1] 1 7
```

```
x[-c(1, 7)]
```

```
[1] 2 3 4 5 6 8 9 10
```

```
x[x > 3]
```

```
[1] 4 5 6 7 8 9 10
```

```
(dat <- data.frame(x = 1:3,  
                  y = c("a", "b", "c"),  
                  z = c(TRUE, FALSE, TRUE)))
```

	x	y	z
1	1	a	TRUE
2	2	b	FALSE
3	3	c	TRUE

```
dat[1, 3]
```

```
[1] TRUE
```



```
dat[, 1]
```

```
[1] 1 2 3
```

```
dat[3, ]
```

```
  x y   z  
3 3 c TRUE
```

When you have a list, `[` always returns a list:

```
mylist <- list(x = 1:10, y = pi, z = c(TRUE, FALSE))  
mylist[2]
```

```
$y  
[1] 3.141593
```

[[ will return the actual element:

```
mylist[[2]]
```

```
[1] 3.141593
```

- we can subset by name so we don't have to remember/figure out positions

```
dat[, c("x", "y")]
```

```
  x y  
1 1 a  
2 2 b  
3 3 c
```

This is common so \$ provides a quicker way:

```
dat[["x"]]
```

```
[1] 1 2 3
```

```
dat$x
```

```
[1] 1 2 3
```

# DISTRIBUTIONS

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- R has functions dealing with probability distributions built in
- They share common prefixes depending on what you want:

What you want	prefix
cdf	p
quantile (inverse cdf)	q
random draw	r
density	d

## COMMON DISTRIBUTIONS

R's name	name
norm	normal
unif	uniform
t	t
binom	binomial
weibull	weibull
beta	beta
hyper	hypergeometric
nbinom	negative binomial
gamma	gamma



## CONDITIONALS

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- conditional statements:
- **if** (this one thing [condition]), **then** (do this other thing), **else** (do this different other thing)
- In R, need to consider whether (condition) is of length 1 or  $> 1$
- Let's start when (condition) is length one

```
x <- 3
if (x == 7) {
  print("x is 7")
} else {
  print("x is not 7")
}
```

```
[1] "x is not 7"
```

```
x <- TRUE
if (x) {
  print("That's true")
} else {
  print("That's false")
}
```

```
[1] "That's true"
```

## CONDITIONS WITH LENGTH > 1

- remember: if, else only works if condition is of length one

```
x <- 1:10
if (x > 5){
  TRUE
} else {
  FALSE
}
```

```
[1] FALSE
```

```
Warning message:
```

```
In if (x > 5) { :
```

```
the condition has length > 1 and only the first element will be
```

```
x <- 1:10  
ifelse(x > 5, TRUE, FALSE)
```

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

# WRITING FUNCTIONS

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- You can write a function in R quite easily
- Let's say we want to write a function to find the mean



```
my_mean <- function(x){  
  sum(x) / length(x)  
}  
my_mean(0:10)
```

```
[1] 5
```

```
my_mean <- function(x, na.rm = FALSE){  
  if (na.rm) {  
    x <- x[!is.na(x)]  
  }  
  sum(x) / length(x)  
}  
x <- c(1, NA, 3)  
my_mean(x, na.rm = TRUE)
```

```
[1] 2
```

## USING LOOPS

---

- computers are much better than humans at doing repetitive tasks quickly & without error

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- loops are a common way of doing something similar multiple times
- we'll talk about `for`, which loops a prescribed number of times

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- loops are a common way of doing something similar multiple times
- we'll talk about `for`, which loops a prescribed number of times
- R has `while` and `repeat` loops as well, which loop until a logical check fails (returns `FALSE`)

pseudo-code structure of for:

```
output <- vector("numeric", length = 72) ## pre-allocate output!  
for (something in somevector){          ## defined sequence  
  do stuff, referring to each element of ##body  
  somevector sequentially with the  
  placeholder something  
}
```

## FOR LOOPS, EXAMPLE

```
x <- 6:10  
for (i in x) {  
  print(i)  
}
```

```
[1] 6  
[1] 7  
[1] 8  
[1] 9  
[1] 10
```



## FOR LOOPS, EXAMPLE 2

```
x <- 6:10
y <- vector(length = length(x))
for (i in seq_along(x)) {
  if (i > 1){
    y[i] <- x[i] + y[i - 1]
  } else {
    y[i] <- x[i]
  }
}
## what is y?
```

## FOR LOOPS, EXAMPLE 2 ANSWER

```
y
```

```
[1]  6 13 21 30 40
```

```
means <- vector()

for (i in names(mtcars)) {
  means[[i]] <- mean(mtcars[[i]])
}

## What will means be?
```

## FOR LOOPS, EXAMPLE 3 ANSWER

means

mpg	cyl	disp	hp	drat	wt
20.090625	6.187500	230.721875	146.687500	3.596563	3.217250
vs	am	gear	carb		
0.437500	0.406250	3.687500	2.812500		

```
## what does this code do?  
x <- list.files(pattern = "*.csv")  
data <- vector("list", length = length(x))  
for (i in x) {  
  data[[i]] <- read.csv(i)  
}
```

- You may see a lot of advice online against loops
- They used to be slow in R, not the case anymore
- So long as you're smart (pre-allocate output length!)

## THE APPLY FAMILY

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- The apply family of functions make our life easier by applying functions over "stuff"
- Like a pre-built loop
- apply, lapply, sapply, vapply, mapply, rapply, tapply
- We'll look at apply and lapply



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

```
apply(mtcars, 2, mean)
```

mpg	cyl	disp	hp	drat	wt
20.090625	6.187500	230.721875	146.687500	3.596563	3.217250
vs	am	gear	carb		
0.437500	0.406250	3.687500	2.812500		

Note that we can apply our own functions!

```
apply(mtcars, 2, my_mean, na.rm = TRUE)
```

mpg	cyl	disp	hp	drat	wt
20.090625	6.187500	230.721875	146.687500	3.596563	3.217250
vs	am	gear	carb		
0.437500	0.406250	3.687500	2.812500		

```
lapply(X, FUN) # always returns a list
```

```
lapply(mtcars, mean)
```

```
$mpg
```

```
[1] 20.09062
```

```
$cyl
```

```
[1] 6.1875
```

```
$disp
```

```
[1] 230.7219
```

```
$hp
```

```
[1] 146.6875
```

- `lapply` always returns a list
- `sapply` will simplify this (e.g. to a numeric vector) if it can

```
sapply(mtcars, mean)
```

mpg	cyl	disp	hp	drat	wt
20.090625	6.187500	230.721875	146.687500	3.596563	3.217250
vs	am	gear	carb		
0.437500	0.406250	3.687500	2.812500		

## PUTTING IT ALL TOGETHER

---

Create a function to represent this

$$f(x) = \begin{cases} 0 & \text{for } x < 0 \\ \frac{1}{3} & \text{for } 0 \leq x < 1 \\ \frac{2}{3} & \text{for } 1 \leq x < 2 \\ 0 & \text{for } 2 < x \end{cases}$$



```
myfun <- function(x){  
  ifelse(x >= 0 & x < 1, 1 / 3,  
         ifelse(x >= 1 & x < 2, 2 / 3, 0))  
}
```

using the function from the last slide, construct a function that will return a vector of samples using rejection sampling. Make it take one argument  $n$  the number of samples.

```
myreject <- function(n){  
  x <- runif(n, 0, 2)  
  y <- runif(n, 0, 2 / 3)  
  reject <- y > myfun(x)  
  x[!reject]  
}
```

## REJECTION SAMPLING WITH MULTIPLE N'S

- We want to test the effect of varying  $n$  on our rejection sampler.
- Calculate the mean of the samples from our rejection sampler varying  $n$  from 1 to 1,000

```
ns <- seq(1, 1000)
means <- sapply(ns, function(n){mean(myreject(n))})
## plot(means)
## summary(means)
```