

# Base R Cheat Sheet

## Getting Help

### Accessing the help files

?mean

Get help of a particular function.

help.search('weighted mean')

Search the help files for a word or phrase.

help(package = 'dplyr')

Find help for a package.

### More about an object

str(iris)

Get a summary of an object's structure.

class(iris)

Find the class an object belongs to.

## Using Packages

install.packages('dplyr')

Download and install a package from CRAN.

library(dplyr)

Load the package into the session, making all its functions available to use.

dplyr::select

Use a particular function from a package.

data(iris)

Load a built-in dataset into the environment.

## Working Directory

getwd()

Find the current working directory (where inputs are found and outputs are sent).

setwd('C://file/path')

Change the current working directory.

**Use projects in RStudio to set the working directory to the folder you are working in.**

## Vectors

### Creating Vectors

c(2, 4, 6)	2 4 6	Join elements into a vector
2:6	2 3 4 5 6	An integer sequence
seq(2, 3, by=0.5)	2.0 2.5 3.0	A complex sequence
rep(1:2, times=3)	1 2 1 2 1 2	Repeat a vector
rep(1:2, each=3)	1 1 1 2 2 2	Repeat elements of a vector

### Vector Functions

sort(x)

Return x sorted.

rev(x)

Return x reversed.

table(x)

See counts of values.

unique(x)

See unique values.

### Selecting Vector Elements

#### By Position

x[4]

The fourth element.

x[-4]

All but the fourth.

x[2:4]

Elements two to four.

x[!(2:4)]

All elements except two to four.

x[c(1, 5)]

Elements one and five.

#### By Value

x[x == 10]

Elements which are equal to 10.

x[x < 0]

All elements less than zero.

x[x %in% c(1, 2, 5)]

Elements in the set 1, 2, 5.

### Named Vectors

x['apple']

Element with name 'apple'.

## Programming

### For Loop

```
for (variable in sequence){  
  Do something  
}
```

### Example

```
for (i in 1:4){  
  j <- i + 10  
  print(j)  
}
```

### While Loop

```
while (condition){  
  Do something  
}
```

### Example

```
while (i < 5){  
  print(i)  
  i <- i + 1  
}
```

### Functions

```
function_name <- function(var){  
  Do something  
  return(new_variable)  
}
```

### Example

```
square <- function(x){  
  squared <- x*x  
  return(squared)  
}
```

## Reading and Writing Data

Also see the **readr** package.

Input	Output	Description
df <- read.table('file.txt')	write.table(df, 'file.txt')	Read and write a delimited text file.
df <- read.csv('file.csv')	write.csv(df, 'file.csv')	Read and write a comma separated value file. This is a special case of read.table/write.table.
load('file.RData')	save(df, file = 'file.Rdata')	Read and write an R data file, a file type special for R.

Conditions	a == b	Are equal	a > b	Greater than	a >= b	Greater than or equal to	is.na(a)	Is missing
	a != b	Not equal	a < b	Less than	a <= b	Less than or equal to	is.null(a)	Is null

## Types

Converting between common data types in R. Can always go from a higher value in the table to a lower value.

as.logical	TRUE, FALSE, TRUE	Boolean values (TRUE or FALSE).
as.numeric	1, 0, 1	Integers or floating point numbers.
as.character	'1', '0', '1'	Character strings. Generally preferred to factors.
as.factor	'1', '0', '1', levels: '1', '0'	Character strings with preset levels. Needed for some statistical models.

## Maths Functions

log(x)	Natural log.	sum(x)	Sum.
exp(x)	Exponential.	mean(x)	Mean.
max(x)	Largest element.	median(x)	Median.
min(x)	Smallest element.	quantile(x)	Percentage quantiles.
round(x, n)	Round to n decimal places.	rank(x)	Rank of elements.
signif(x, n)	Round to n significant figures.	var(x)	The variance.
cor(x, y)	Correlation.	sd(x)	The standard deviation.

## Variable Assignment

```
> a <- 'apple'  
> a  
[1] 'apple'
```

## The Environment

ls()	List all variables in the environment.
rm(x)	Remove x from the environment.
rm(list = ls())	Remove all variables from the environment.

You can use the environment panel in RStudio to browse variables in your environment.

## Matrices

`m <- matrix(x, nrow = 3, ncol = 3)`  
Create a matrix from x.

	<code>m[2, ]</code> - Select a row	<code>t(m)</code> Transpose
	<code>m[, 1]</code> - Select a column	<code>m %*% n</code> Matrix Multiplication
	<code>m[2, 3]</code> - Select an element	<code>solve(m, n)</code> Find x in: $m \cdot x = n$

## Lists

`l <- list(x = 1:5, y = c('a', 'b'))`  
A list is a collection of elements which can be of different types.

<code>l[[2]]</code>	<code>l[1]</code>	<code>l\$x</code>	<code>l['y']</code>
Second element of l.	New list with only the first element.	Element named x.	New list with only element named y.

Also see the `dplyr` package.

## Data Frames

`df <- data.frame(x = 1:3, y = c('a', 'b', 'c'))`  
A special case of a list where all elements are the same length.

x	y
1	a
2	b
3	c

## Matrix subsetting

<code>df[, 2]</code>	
<code>df[2, ]</code>	
<code>df[2, 2]</code>	

List subsetting	
<code>df\$x</code>	
<code>df[[2]]</code>	

<i>Understanding a data frame</i>
<code>View(df)</code>
See the full data frame.

<code>head(df)</code>
See the first 6 rows.

`nrow(df)`  
Number of rows.

`ncol(df)`  
Number of columns.

`dim(df)`  
Number of columns and rows.

`cbind` - Bind columns.

`rbind` - Bind rows.

## Strings

<code>paste(x, y, sep = ' ')</code>	Join multiple vectors together.
<code>paste(x, collapse = ' ')</code>	Join elements of a vector together.
<code>grep(pattern, x)</code>	Find regular expression matches in x.
<code>gsub(pattern, replace, x)</code>	Replace matches in x with a string.
<code>toupper(x)</code>	Convert to uppercase.
<code>tolower(x)</code>	Convert to lowercase.
<code>nchar(x)</code>	Number of characters in a string.

## Factors

<code>factor(x)</code>	
Turn a vector into a factor. Can set the levels of the factor and the order.	Turn a numeric vector into a factor by 'cutting' into sections.

## Statistics

<code>lm(y ~ x, data=df)</code>	Linear model.
<code>glm(y ~ x, data=df)</code>	Generalised linear model.
<code>summary</code>	Get more detailed information out a model.
<code>pairwise.t.test</code>	Perform a t-test for paired data.

## Distributions

	Random Variates	Density Function	Cumulative Distribution	Quantile
Normal	<code>rnorm</code>	<code>dnorm</code>	<code>pnorm</code>	<code>qnorm</code>
Poisson	<code>rpois</code>	<code>dpois</code>	<code>ppois</code>	<code>qpois</code>
Binomial	<code>rbinom</code>	<code>dbinom</code>	<code>pbinom</code>	<code>qbinom</code>
Uniform	<code>runif</code>	<code>dunif</code>	<code>unif</code>	<code>qunif</code>

## Plotting

<code>plot(x)</code>	Values of x in order.
<code>plot(x, y)</code>	Values of x against y.
<code>hist(x)</code>	Histogram of x.

## Dates

See the `lubridate` package.

# RStudio IDE :: CHEAT SHEET

## Documents and Apps

   Open Shiny, R Markdown, knitr, Sweave, LaTeX, .Rd files and more in Source Pane

Check spelling  Render output  Choose output format  Choose output location  Insert code chunk 

Jump to previous chunk  Jump to next chunk  Run selected lines  Publish to server  Show file outline 

Access markdown guide at **Help > Markdown Quick Reference**

Jump to chunk  Set knitr chunk options  Run this and all previous code chunks  Run this code chunk 

RStudio recognizes that files named **app.R**, **server.R**, **ui.R**, and **global.R** belong to a shiny app

Run app  Choose location to view app  Publish to shinyapps.io or server  Manage publish accounts 

## Debug Mode

Open with **debug()**, **browser()**, or a breakpoint. RStudio will open the debugger mode when it encounters a breakpoint while executing code.

Click next to line number to add/remove a breakpoint.

Highlighted line shows where execution has paused

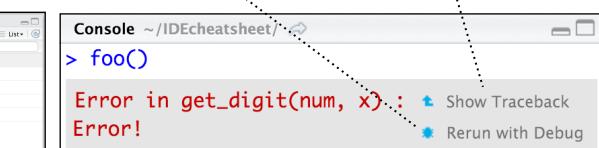
Run commands in environment where execution has paused

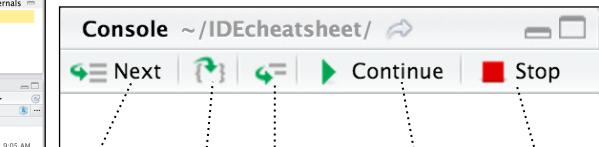
Examine variables in executing environment

Select function in traceback to debug

Launch debugger mode from origin of error

Open traceback to examine the functions that R called before the error occurred

 Error in get\_digit(num, x) :  
Error!

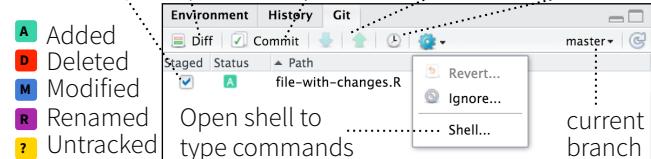
 Step through code one line at a time

## Version Control with Git or SVN



Turn on at **Tools > Project Options > Git/SVN**

Stage files:  Show file diff  Commit staged files to remote  Push/Pull  View History 

 master  
A Added  
D Deleted  
M Modified  
R Renamed  
? Untracked  
B Staged  
S Status  
P Path  
file-with-changes.R  
Revert...  
Ignore...  
Shell...  
Open shell to type commands  
current branch

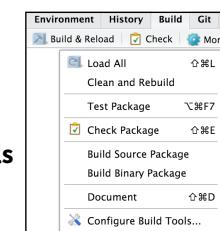
## Package Writing



**File > New Project > New Directory > R Package**

Turn project into package, Enable roxygen documentation with **Tools > Project Options > Build Tools**

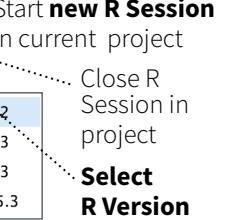
Roxygen guide at **Help > Roxygen Quick Reference**

 Load All  
Clean and Rebuild  
Test Package  
Build Source Package  
Build Binary Package  
Document  
Configure Build Tools...



## Pro Features

**Share Project** Active shared with Collaborators

Start **new R Session** in current project   
Close R Session in project   
**Select R Version** 

### PROJECT SYSTEM

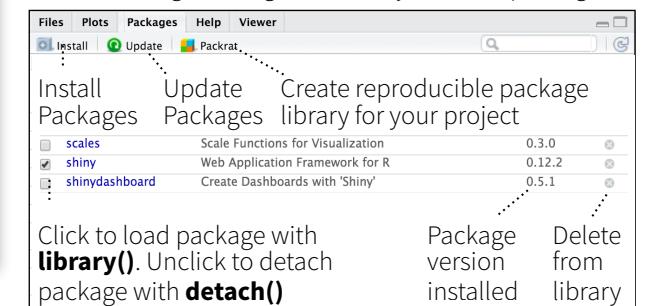
**File > New Project**

RStudio saves the call history, workspace, and working directory associated with a project. It reloads each when you re-open a project.

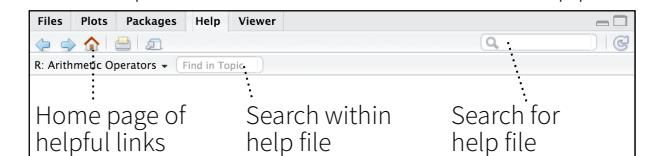
RStudio opens plots in a dedicated Plots pane

 Navigate in recent plots  Open in window  Export plot  Delete plot  Delete all plots 

GUI Package manager lists every installed package

 Install Packages  Update Packages  Create reproducible package library for your project   
scales 0.3.0  
shiny 0.12.2  
shinydashboard 0.5.1  
Click to load package with **library()**. Unclick to detach package with **detach()**  
Package version installed  Delete from library 

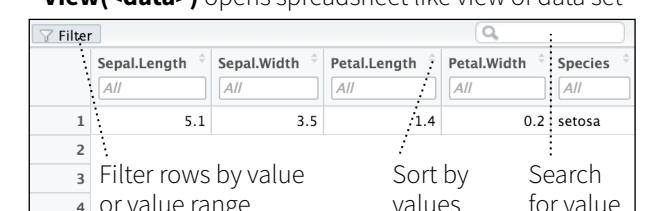
RStudio opens documentation in a dedicated Help pane

 Home page of helpful links  Search within help file  Search for help file 

Viewer Pane displays HTML content, such as Shiny apps, RMarkdown reports, and interactive visualizations

 Stop Shiny app  Publish to shinyapps.io, rpubs, RSConnect, ...  Refresh 

**View(<data>)** opens spreadsheet like view of data set

 Filter  Sepal.Length  Sepal.Width  Petal.Length  Petal.Width  Species   
1 5.1 3.5 1.4 0.2 setosa  
2  
3 Filter rows by value or value range  Sort by values  Search for value 

## 1 LAYOUT

Move focus to Source Editor  
Move focus to Console  
Move focus to Help  
Show History  
Show Files  
Show Plots  
Show Packages  
Show Environment  
Show Git/SVN  
Show Build

## Windows/Linux Mac

Ctrl+1  
Ctrl+2  
Ctrl+3  
Ctrl+4  
Ctrl+5  
Ctrl+6  
Ctrl+7  
Ctrl+8  
Ctrl+9  
Ctrl+0

## 2 RUN CODE

### Search command history

Navigate command history  
Move cursor to start of line  
Move cursor to end of line  
Change working directory

### Interrupt current command

### Clear console

Quit Session (desktop only)

### Restart R Session

Run current (retain cursor)  
Run from current to end  
Run the current function  
Source a file

### Source the current file

Source with echo

## Windows/Linux Mac

**Ctrl+↑**  
**↑/↓**  
Home  
End  
Ctrl+Shift+H

### Esc

### Ctrl+L

Ctrl+Q

### Ctrl+Shift+F10

### Ctrl+Enter

Alt+Enter  
Ctrl+Alt+E  
Ctrl+Alt+F  
Ctrl+Alt+G

### Ctrl+Shift+S

Ctrl+Shift+Enter

## 3 NAVIGATE CODE

### Goto File/Function

Fold Selected  
Unfold Selected  
Fold All  
Unfold All  
Go to line  
Jump to  
Switch to tab  
Previous tab  
Next tab  
First tab  
Last tab  
Navigate back  
Navigate forward  
Jump to Brace  
Select within Braces  
Use Selection for Find  
Find in Files  
Find Next  
Find Previous  
Jump to Word  
Jump to Start/End  
Toggle Outline

## Windows /Linux

### Mac

Ctrl+.  
Alt+L  
Shift+Alt+L  
Alt+O  
Shift+Alt+O  
Shift+Alt+G  
Shift+Alt+J  
Ctrl+Shift+.  
Ctrl+F11  
Ctrl+F12  
Ctrl+Shift+F11  
Ctrl+Shift+F12  
Ctrl+F9  
Ctrl+F10  
Ctrl+P  
Ctrl+Shift+Alt+E  
Ctrl+F3  
Ctrl+Shift+F  
Win: F3, Linux: Ctrl+G  
Cmd+G  
Cmd+Shift+G  
Option+←/→  
Ctrl+↑/↓  
Ctrl+Shift+O

## 4 WRITE CODE

**Attempt completion**  
Navigate candidates  
Accept candidate  
Dismiss candidates  
Undo  
Redo  
Cut  
Copy  
Paste  
Select All  
Delete Line

Select  
Select Word  
Select to Line Start  
Select to Line End  
Select Page Up/Down  
Select to Start/End  
Delete Word Left  
Delete Word Right  
Delete to Line End  
Delete to Line Start  
Indent  
Outdent  
Yank line up to cursor  
Yank line after cursor  
Insert yanked text

### Insert <->

### Insert %>%

Show help for function

Show source code

New document

New document (Chrome)

Open document

Save document

Close document

Close document (Chrome)

Close all documents

Extract function

Extract variable

Reindent lines

### (Un)Comment lines

Reflow Comment

Reformat Selection

Select within braces

Show Diagnostics

Transpose Letters

Move Lines Up/Down

Copy Lines Up/Down

Add New Cursor Above

Add New Cursor Below

Move Active Cursor Up

Move Active Cursor Down

Find and Replace

Use Selection for Find

Replace and Find

## Windows /Linux

### Mac

**Tab or Ctrl+Space**  
**↑/↓**  
Enter, Tab, or →  
Esc  
Ctrl+Z  
Ctrl+Shift+Z  
Ctrl+X  
Ctrl+C  
Ctrl+V  
Ctrl+A  
Ctrl+D

Shift+[Arrow]  
Ctrl+Shift+←/→  
Alt+Shift+←  
Alt+Shift+→  
Shift+PageUp/Down  
Shift+Alt+↑/↓  
Ctrl+Backspace

Ctrl+K

Option+Backspace

Tab (at start of line)

Shift+Tab

Ctrl+U

Ctrl+K

Ctrl+Y

Alt+-

Ctrl+Shift+M

F1

F2

Cmd+Shift+N

Ctrl+Alt+Shift+N

Ctrl+O

Ctrl+S

Ctrl+W

Ctrl+Alt+W

Ctrl+Shift+W

Ctrl+Alt+X

Ctrl+Option+V

Ctrl+I

Ctrl+Shift+C

Ctrl+Shift+/

Ctrl+Shift+A

Ctrl+Shift+E

Ctrl+Shift+E

Cmd+Shift+Opt+P

Ctrl+T

Alt+↑/↓

Shift+Alt+↑/↓

Cmd+Option+↑/↓

Ctrl+Option+Up

Ctrl+Option+Down

Ctrl+Option+Shift+Up

Ctrl+Opt+Shift+Down

Ctrl+F

Ctrl+F3

Ctrl+Shift+J

## Mac

### Tab or Cmd+Space

**↑/↓**

Enter, Tab, or →

Esc

Cmd+Z

Cmd+Shift+Z

Cmd+X

Cmd+C

Cmd+V

Cmd+A

Cmd+D

Shift+[Arrow]

Option+Shift+←/→

Cmd+Shift+←

Cmd+Shift+→

Shift+PageUp/Down

Cmd+Shift+↑/↓

Ctrl+Opt+Backspace

Option+Delete

Ctrl+K

Option+Backspace

Tab (at start of line)

Shift+Tab

Ctrl+U

Ctrl+K

Ctrl+Y

Alt+-

Ctrl+Shift+M

F1

F2

Cmd+Shift+N

Cmd+Shift+Opt+N

Cmd+O

Ctrl+S

Cmd+W

Cmd+Option+W

Cmd+Shift+W

Ctrl+Alt+X

Ctrl+Option+V

Cmd+I

Cmd+Shift+C

Ctrl+Shift+/

Ctrl+Shift+A

Ctrl+Shift+E

Ctrl+Shift+E

Cmd+Shift+Opt+P

Ctrl+T

Alt+↑/↓

Shift+Alt+↑/↓

Cmd+Option+↑/↓

Ctrl+Option+Up

Ctrl+Option+Down

Ctrl+Option+Shift+Up

Ctrl+Opt+Shift+Down

Ctrl+F

Ctrl+F3

Cmd+E

Cmd+Shift+J

## WHY RSTUDIO SERVER PRO?

RSP extends the open source server with a commercial license, support, and more:

- open and run multiple R sessions at once
  - tune your resources to improve performance
  - edit the same project at the same time as others
  - see what you and others are doing on your server
  - switch easily from one version of R to a different version
  - integrate with your authentication, authorization, and audit practices
- Download a free 45 day evaluation at [www.rstudio.com/products/rstudio-server-pro/](http://www.rstudio.com/products/rstudio-server-pro/)



## 5 DEBUG CODE

Toggle Breakpoint	Shift+F9
Execute Next Line	F10
Step Into Function	Shift+F4
Finish Function/Loop	Shift+F6
Continue	Shift+F5
Stop Debugging	Shift+F8

## 6 VERSION CONTROL

Show diff	Ctrl+Alt+D
Commit changes	Ctrl+Alt+M
Scroll diff view	Ctrl+↑/↓
Stage/Unstage (Git)	Spacebar
Stage/Unstage and move to next	Enter

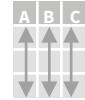
## 7 MAKE PACKAGES

Build and Reload	Ctrl+Shift+B
<b>Load All (devtools)</b>	<b>Cmd+Shift+L</b>
<b>Test Package (Desktop)</b>	<b>Cmd+Shift+T</b>
Test Package (Web)	Ctrl+Alt+F7
Check Package	Ctrl+Shift+E
<b>Document Package</b>	<b>Cmd+Shift+D</b>

# Data Transformation with dplyr :: CHEAT SHEET



dplyr functions work with pipes and expect **tidy data**. In tidy data:



Each **variable** is in its own **column**



Each **observation**, or **case**, is in its own **row**



`x %>% f(y)` becomes `f(x, y)`

## Summarise Cases

These apply **summary functions** to columns to create a new table of summary statistics. Summary functions take vectors as input and return one value (see back).

summary function

`summarise(.data, ...)`  
Compute table of summaries.  
`summarise(mtcars, avg = mean(mpg))`

`count(x, ..., wt = NULL, sort = FALSE)`  
Count number of rows in each group defined by the variables in ... Also **tally()**.  
`count(iris, Species)`

## VARIATIONS

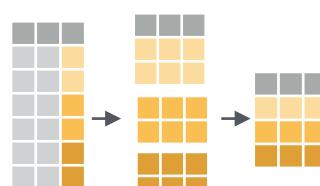
`summarise_all()` - Apply funs to every column.

`summarise_at()` - Apply funs to specific columns.

`summarise_if()` - Apply funs to all cols of one type.

## Group Cases

Use **group\_by()** to create a "grouped" copy of a table. dplyr functions will manipulate each "group" separately and then combine the results.



`mtcars %>%  
group_by(cyl) %>%  
summarise(avg = mean(mpg))`

`group_by(.data, ..., add = FALSE)`  
Returns copy of table grouped by ...  
`g_iris <- group_by(iris, Species)`

`ungroup(x, ...)`  
Returns ungrouped copy of table.  
`ungroup(g_iris)`

## Manipulate Cases

### EXTRACT CASES

Row functions return a subset of rows as a new table.



`filter(.data, ...)` Extract rows that meet logical criteria. `filter(iris, Sepal.Length > 7)`



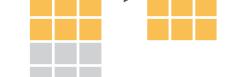
`distinct(.data, ..., .keep_all = FALSE)` Remove rows with duplicate values.  
`distinct(iris, Species)`



`sample_frac(tbl, size = 1, replace = FALSE, weight = NULL, .env = parent.frame())` Randomly select fraction of rows.  
`sample_frac(iris, 0.5, replace = TRUE)`



`slice(.data, ...)` Select rows by position.  
`slice(iris, 10:15)`



`top_n(x, n, wt)` Select and order top n entries (by group if grouped data).  
`top_n(iris, 5, Sepal.Width)`

## Manipulate Variables

### EXTRACT VARIABLES

Column functions return a set of columns as a new vector or table.



`pull(.data, var = -1)` Extract column values as a vector. Choose by name or index.  
`pull(iris, Sepal.Length)`



`select(.data, ...)` Extract columns as a table. Also `select_if()`.  
`select(iris, Sepal.Length, Species)`

Use these helpers with `select()`,  
e.g. `select(iris, starts_with("Sepal"))`

`contains(match)`    `num_range(prefix, range)` : e.g. `mpg:cyl`  
`ends_with(match)`    `one_of(...)`    -, e.g. `-Species`  
`matches(match)`    `starts_with(match)`

### MAKE NEW VARIABLES

These apply **vectorized functions** to columns. Vectorized funs take vectors as input and return vectors of the same length as output (see back).

vectorized function

`mutate(.data, ...)`  
Compute new column(s).  
`mutate(mtcars, gpm = 1/mpg)`

`transmute(.data, ...)`  
Compute new column(s), drop others.  
`transmute(mtcars, gpm = 1/mpg)`

`mutate_all(.tbl, .funs, ...)` Apply funs to every column. Use with `funs()`. Also `mutate_if()`.  
`mutate_all(faithful, funs(log(.), log2(.)))`  
`mutate_if(iris, is.numeric, funs(log(.)))`

`mutate_at(.tbl, .cols, .funs, ...)` Apply funs to specific columns. Use with `funs()`, `vars()` and the helper functions for `select()`.  
`mutate_at(iris, vars(-Species), funs(log(.)))`

`add_column(.data, ..., .before = NULL, .after = NULL)` Add new column(s). Also `add_count()`, `add_tally()`.  
`add_column(mtcars, new = 1:32)`

`rename(.data, ...)` Rename columns.  
`rename(iris, Length = Sepal.Length)`



# Vector Functions

## TO USE WITH MUTATE ()

**mutate()** and **transmute()** apply vectorized functions to columns to create new columns. Vectorized functions take vectors as input and return vectors of the same length as output.

### vectorized function

## OFFSETS

dplyr::lag() - Offset elements by 1  
dplyr::lead() - Offset elements by -1

## CUMULATIVE AGGREGATES

dplyr::cumall() - Cumulative all()  
dplyr::cumany() - Cumulative any()  
    **cummax()** - Cumulative max()  
dplyr::cummean() - Cumulative mean()  
    **cummin()** - Cumulative min()  
    **cumprod()** - Cumulative prod()  
    **cumsum()** - Cumulative sum()

## RANKINGS

dplyr::cume\_dist() - Proportion of all values <=  
dplyr::dense\_rank() - rank with ties = min, no gaps  
dplyr::min\_rank() - rank with ties = min  
dplyr::ntile() - bins into n bins  
dplyr::percent\_rank() - min\_rank scaled to [0,1]  
dplyr::row\_number() - rank with ties = "first"

## MATH

+, -, \*, /, ^, %/%, %% - arithmetic ops  
**log()**, **log2()**, **log10()** - logs  
<, <=, >, >=, !=, == - logical comparisons  
dplyr::between() - x >= left & x <= right  
dplyr::near() - safe == for floating point numbers

## MISC

dplyr::case\_when() - multi-case if\_else()  
dplyr::coalesce() - first non-NA values by element across a set of vectors  
dplyr::if\_else() - element-wise if() + else()  
dplyr::na\_if() - replace specific values with NA  
    **pmax()** - element-wise max()  
    **pmin()** - element-wise min()  
dplyr::recode() - Vectorized switch()  
dplyr::recode\_factor() - Vectorized switch() for factors

# Summary Functions

## TO USE WITH SUMMARISE ()

**summarise()** applies summary functions to columns to create a new table. Summary functions take vectors as input and return single values as output.

### summary function

## COUNTS

dplyr::n() - number of values/rows  
dplyr::n\_distinct() - # of uniques  
    **sum(!is.na())** - # of non-NA's

## LOCATION

**mean()** - mean, also **mean(!is.na())**  
**median()** - median

## LOGICALS

**mean()** - Proportion of TRUE's  
**sum()** - # of TRUE's

## POSITION/ORDER

dplyr::first() - first value  
dplyr::last() - last value  
dplyr::nth() - value in nth location of vector

## RANK

**quantile()** - nth quantile  
**min()** - minimum value  
**max()** - maximum value

## SPREAD

**IQR()** - Inter-Quartile Range  
**mad()** - median absolute deviation  
**sd()** - standard deviation  
**var()** - variance

# Row Names

Tidy data does not use rownames, which store a variable outside of the columns. To work with the rownames, first move them into a column.

A	B
1	a
2	b
3	c

1	a
2	b
3	c

**rownames\_to\_column()**  
Move row names into col.  
a <- rownames\_to\_column(iris, var = "C")

A	B	C
1	a	t
2	b	u
3	c	v

1	a	t
2	b	u
3	c	v

**column\_to\_rownames()**  
Move col in row names.  
column\_to\_rownames(a, var = "C")

Also **has\_rownames()**, **remove\_rownames()**

# Combine Tables

## COMBINE VARIABLES

X	A B C a t 1 b u 2 c v 3	+	y	A B D a t 3 b u 2 d w 1	=	A B C A B D a t 1 a t 3 b u 2 b u 2 c v 3 d w 1
---	----------------------------------	---	---	----------------------------------	---	--

Use **bind\_cols()** to paste tables beside each other as they are.

**bind\_cols(...)** Returns tables placed side by side as a single table.  
BE SURE THAT ROWS ALIGN.

Use a "**Mutating Join**" to join one table to columns from another, matching values with the rows that they correspond to. Each join retains a different combination of values from the tables.

A B C D a t 1 3 b u 2 2 c v 3 NA	<b>left_join(x, y, by = NULL,</b> copy=FALSE, suffix=c("x","y"),...) Join matching values from y to x.
---	--

A B C D a t 1 3 b u 2 2 d w NA 1	<b>right_join(x, y, by = NULL, copy = FALSE,</b> suffix=c("x","y"),...) Join matching values from x to y.
---	---

A B C D a t 1 3 b u 2 2	<b>inner_join(x, y, by = NULL, copy = FALSE,</b> suffix=c("x","y"),...) Join data. Retain only rows with matches.
-------------------------------	---

A B C D a t 1 3 b u 2 2 c v 3 NA	<b>full_join(x, y, by = NULL,</b> copy=FALSE, suffix=c("x","y"),...) Join data. Retain all values, all rows.
---	--

Use **by = c("col1", "col2")** to specify the column(s) to match on.  
**left\_join(x, y, by = "A")**

Use a named vector, **by = c("col1" = "col2")**, to match on columns with different names in each data set.  
**left\_join(x, y, by = c("C" = "D"))**

Use **suffix** to specify suffix to give to duplicate column names.  
**left\_join(x, y, by = c("C" = "D"), suffix = c("1", "2"))**

## COMBINE CASES

X	A B C a t 1 b u 2 c v 3	+	y	A B C C v 3 d w 4
---	----------------------------------	---	---	-------------------------

Use **bind\_rows()** to paste tables below each other as they are.

df A B C x a t 1 x b u 2 x c v 3 z c v 3 z d w 4	<b>bind_rows(..., .id = NULL)</b> Returns tables one on top of the other as a single table. Set .id to a column name to add a column of the original table names (as pictured)
---	---

A B C c v 3	<b>intersect(x, y, ...)</b> Rows that appear in both x and y.
----------------	--

A B C a t 1 b u 2	<b>setdiff(x, y, ...)</b> Rows that appear in x but not y.
-------------------------	---

A B C a t 1 b u 2 c v 3 d w 4	<b>union(x, y, ...)</b> Rows that appear in x or y. (Duplicates removed). <b>union_all()</b> retains duplicates.
---	--

Use **setequal()** to test whether two data sets contain the exact same rows (in any order).

## EXTRACT ROWS

X	A B C a t 1 b u 2 c v 3	+	y	A B D a t 3 b u 2 d w 1	=
---	----------------------------------	---	---	----------------------------------	---

Use a "**Filtering Join**" to filter one table against the rows of another.

A B C a t 1 b u 2	<b>semi_join(x, y, by = NULL, ...)</b> Return rows of x that have a match in y. USEFUL TO SEE WHAT WILL BE JOINED.
-------------------------	--

A B C c v 3	<b>anti_join(x, y, by = NULL, ...)</b> Return rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.
----------------	--



# Work with strings with stringr :: CHEAT SHEET

The `stringr` package provides a set of internally consistent tools for working with character strings, i.e. sequences of characters surrounded by quotation marks.

## Detect Matches

	<code>str_detect(string, pattern)</code> Detect the presence of a pattern match in a string. <code>str_detect(fruit, "a")</code>
	<code>str_which(string, pattern)</code> Find the indexes of strings that contain a pattern match. <code>str_which(fruit, "a")</code>
	<code>str_count(string, pattern)</code> Count the number of matches in a string. <code>str_count(fruit, "a")</code>
	<code>str_locate(string, pattern)</code> Locate the positions of pattern matches in a string. Also <code>str_locate_all</code> . <code>str_locate(fruit, "a")</code>

## Subset Strings

	<code>str_sub(string, start = 1L, end = -1L)</code> Extract substrings from a character vector. <code>str_sub(fruit, 1, 3); str_sub(fruit, -2)</code>
	<code>str_subset(string, pattern)</code> Return only the strings that contain a pattern match. <code>str_subset(fruit, "b")</code>
	<code>str_extract(string, pattern)</code> Return the first pattern match found in each string, as a vector. Also <code>str_extract_all</code> to return every pattern match. <code>str_extract(fruit, "[aeiou]")</code>
	<code>str_match(string, pattern)</code> Return the first pattern match found in each string, as a matrix with a column for each () group in pattern. Also <code>str_match_all</code> . <code>str_match(sentences, "(a the) ([^ ]+)")</code>

## Manage Lengths

	<code>str_length(string)</code> The width of strings (i.e. number of code points, which generally equals the number of characters). <code>str_length(fruit)</code>
	<code>str_pad(string, width, side = c("left", "right", "both"), pad = " ")</code> Pad strings to constant width. <code>str_pad(fruit, 17)</code>
	<code>str_trunc(string, width, side = c("right", "left", "center"), ellipsis = "...")</code> Truncate the width of strings, replacing content with ellipsis. <code>str_trunc(fruit, 3)</code>
	<code>str_trim(string, side = c("both", "left", "right"))</code> Trim whitespace from the start and/or end of a string. <code>str_trim(fruit)</code>

## Mutate Strings

	<code>str_sub()</code> <- value. Replace substrings by identifying the substrings with <code>str_sub()</code> and assigning into the results. <code>str_sub(fruit, 1, 3) &lt;- "str"</code>
	<code>str_replace(string, pattern, replacement)</code> Replace the first matched pattern in each string. <code>str_replace(fruit, "a", "-")</code>
	<code>str_replace_all(string, pattern, replacement)</code> Replace all matched patterns in each string. <code>str_replace_all(fruit, "a", "-")</code>
	<code>str_to_lower(string, locale = "en")<sup>1</sup></code> Convert strings to lower case. <code>str_to_lower(sentences)</code>
	<code>str_to_upper(string, locale = "en")<sup>1</sup></code> Convert strings to upper case. <code>str_to_upper(sentences)</code>
	<code>str_to_title(string, locale = "en")<sup>1</sup></code> Convert strings to title case. <code>str_to_title(sentences)</code>

## Join and Split

	<code>str_c(..., sep = "", collapse = NULL)</code> Join multiple strings into a single string. <code>str_c(letters, LETTERS)</code>
	<code>str_c(..., sep = "", collapse = NULL)</code> Collapse a vector of strings into a single string. <code>str_c(letters, collapse = "")</code>
	<code>str_dup(string, times)</code> Repeat strings times times. <code>str_dup(fruit, times = 2)</code>
	<code>str_split_fixed(string, pattern, n)</code> Split a vector of strings into a matrix of substrings (splitting at occurrences of a pattern match). Also <code>str_split</code> to return a list of substrings. <code>str_split_fixed(fruit, " ", n=2)</code>
	<code>glue::glue(..., .sep = "", .envir = parent.frame(), .open = "{", .close = "}")</code> Create a string from strings and {expressions} to evaluate. <code>glue::glue("Pi is {pi}")</code>
	<code>glue::glue_data(.x, ..., .sep = "", .envir = parent.frame(), .open = "{", .close = "}")</code> Use a data frame, list, or environment to create a string from strings and {expressions} to evaluate. <code>glue::glue_data(mtcars, "{rownames(mtcars)} has {hp} hp")</code>

## Order Strings

	<code>str_order(x, decreasing = FALSE, na_last = TRUE, locale = "en", numeric = FALSE, ...)<sup>1</sup></code> Return the vector of indexes that sorts a character vector. <code>x[str_order(x)]</code>
	<code>str_sort(x, decreasing = FALSE, na_last = TRUE, locale = "en", numeric = FALSE, ...)<sup>1</sup></code> Sort a character vector. <code>str_sort(x)</code>

## Helpers

	<code>str_conv(string, encoding)</code> Override the encoding of a string. <code>str_conv(fruit, "ISO-8859-1")</code>
	<code>str_view(string, pattern, match = NA)</code> View HTML rendering of first regex match in each string. <code>str_view(fruit, "[aeiou]")</code>
	<code>str_view_all(string, pattern, match = NA)</code> View HTML rendering of all regex matches. <code>str_view_all(fruit, "[aeiou]")</code>
	<code>str_wrap(string, width = 80, indent = 0, exdent = 0)</code> Wrap strings into nicely formatted paragraphs. <code>str_wrap(sentences, 20)</code>

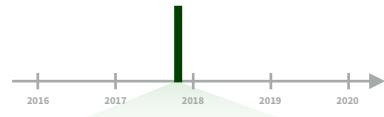
<sup>1</sup> See [bit.ly/ISO639-1](http://bit.ly/ISO639-1) for a complete list of locales.



# Dates and times with lubridate :: CHEAT SHEET



## Date-times



2017-11-28 12:00:00

2017-11-28 12:00:00

A **date-time** is a point on the timeline, stored as the number of seconds since 1970-01-01 00:00:00 UTC

```
dt <- as_datetime(1511870400)
## "2017-11-28 12:00:00 UTC"
```

### PARSE DATE-TIMES (Convert strings or numbers to date-times)

1. Identify the order of the year (**y**), month (**m**), day (**d**), hour (**h**), minute (**m**) and second (**s**) elements in your data.
2. Use the function below whose name replicates the order. Each accepts a wide variety of input formats.

2017-11-28T14:02:00

ymd\_hms(), ymd\_hm(), ymd\_h().  
ymd\_hms("2017-11-28T14:02:00")

2017-22-12 10:00:00

ydm\_hms(), ydm\_hm(), ydm\_h().  
ydm\_hms("2017-22-12 10:00:00")

11/28/2017 1:02:03

mdy\_hms(), mdy\_hm(), mdy\_h().  
mdy\_hms("11/28/2017 1:02:03")

1 Jan 2017 23:59:59

dmy\_hms(), dmy\_hm(), dmy\_h().  
dmy\_hms("1 Jan 2017 23:59:59")

20170131

ymd(), ydm(). ymd(20170131)

July 4th, 2000

mdy(), myd(). mdy("July 4th, 2000")

4th of July '99

dmy(), dym(). dmy("4th of July '99")

2001: Q3

yq() Q for quarter. yq("2001: Q3")

2:01

hms::hms() Also lubridate::hms(), hm() and ms(), which return periods.\* hms::hms(sec = 0, min = 1, hours = 2)

2017.5

date\_decimal(decimal, tz = "UTC")
date\_decimal(2017.5)



now(tzone = "") Current time in tz (defaults to system tz). now()

today(tzone = "") Current date in a tz (defaults to system tz). today()

fast.strptime() Faster strftime.  
fast.strptime('9/1/01', '%y/%m/%d')

parse\_date\_time() Easier strftime.  
parse\_date\_time("9/1/01", "ymd")

2017-11-28

A **date** is a day stored as the number of days since 1970-01-01

```
d <- as_date(17498)
## "2017-11-28"
```

12:00:00

An hms is a **time** stored as the number of seconds since 00:00:00

```
t <- hms::as.hms(85)
## 00:01:25
```

### GET AND SET COMPONENTS

Use an accessor function to get a component. Assign into an accessor function to change a component in place.

```
d ## "2017-11-28"
day(d) ## 28
day(d) <- 1
d ## "2017-11-01"
```

2018-01-31 11:59:59

date(x) Date component. date(dt)

2018-01-31 11:59:59

year(x) Year. year(dt)
isoyear(x) The ISO 8601 year.
epiyear(x) Epidemiological year.

2018-01-31 11:59:59

month(x, label, abbr) Month.

month(dt)

2018-01-31 11:59:59

day(x) Day of month. day(dt)
wday(x, label, abbr) Day of week.
qday(x) Day of quarter.

2018-01-31 11:59:59

hour(x) Hour. hour(dt)

2018-01-31 11:59:59

minute(x) Minutes. minute(dt)

2018-01-31 11:59:59

second(x) Seconds. second(dt)

2018-01-31 11:59:59

week(x) Week of the year. week(dt)
isoweek() ISO 8601 week.
epiweek() Epidemiological week.

2018-01-31 11:59:59

quarter(x, with\_year = FALSE)
Quarter. quarter(dt)

2018-01-31 11:59:59

semester(x, with\_year = FALSE)
Semester. semester(dt)

2018-01-31 11:59:59

am(x) Is it in the am? am(dt)
pm(x) Is it in the pm? pm(dt)

2018-01-31 11:59:59

dst(x) Is it daylight savings? dst(dt)

2018-01-31 11:59:59

leap\_year(x) Is it a leap year?
leap\_year(dt)

2018-01-31 11:59:59

update(object, ..., simple = FALSE)
update(dt, mday = 2, hour = 1)

## Round Date-times



floor\_date(x, unit = "second")  
Round down to nearest unit.  
floor\_date(dt, unit = "month")



round\_date(x, unit = "second")  
Round to nearest unit.  
round\_date(dt, unit = "month")



ceiling\_date(x, unit = "second", change\_on\_boundary = NULL)  
Round up to nearest unit.  
ceiling\_date(dt, unit = "month")

rollback(dates, roll\_to\_first = FALSE, preserve\_hms = TRUE)  
Roll back to last day of previous month. rollback(dt)

## Stamp Date-times

stamp() Derive a template from an example string and return a new function that will apply the template to date-times. Also stamp\_date() and stamp\_time().

1. Derive a template, create a function  
`sf <- stamp("Created Sunday, Jan 17, 1999 3:34")`

**Tip:** use a date with day > 12

2. Apply the template to dates  
`sf(ymd("2010-04-05"))`  
`## [1] "Created Monday, Apr 05, 2010 00:00"`

## Time Zones

R recognizes ~600 time zones. Each encodes the time zone, Daylight Savings Time, and historical calendar variations for an area. R assigns one time zone per vector.

Use the **UTC** time zone to avoid Daylight Savings.

**OlsonNames()** Returns a list of valid time zone names. **OlsonNames()**

5:00 Mountain 6:00 Central  
4:00 Pacific 7:00 Eastern

PT MT CT ET

7:00 Pacific 7:00 Mountain  
7:00 Central

with\_tz(time, tzone = "") Get the same date-time in a new time zone (a new clock time).  
with\_tz(dt, "US/Pacific")

7:00 Pacific 7:00 Mountain  
7:00 Central

force\_tz(time, tzone = "") Get the same clock time in a new time zone (a new date-time).  
force\_tz(dt, "US/Pacific")



# Math with Date-times

Lubridate provides three classes of timespans to facilitate math with dates and date-times

Math with date-times relies on the **timeline**, which behaves inconsistently. Consider how the timeline behaves during:

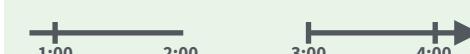
A normal day

```
nor <- ymd_hms("2018-01-01 01:30:00", tz = "US/Eastern")
```



The start of daylight savings (spring forward)

```
gap <- ymd_hms("2018-03-11 01:30:00", tz = "US/Eastern")
```



The end of daylight savings (fall back)

```
lap <- ymd_hms("2018-11-04 00:30:00", tz = "US/Eastern")
```



Leap years and leap seconds

```
leap <- ymd("2019-03-01")
```

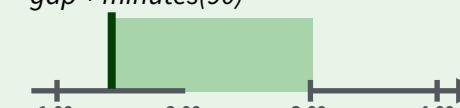


**Periods** track changes in clock times, which ignore time line irregularities.

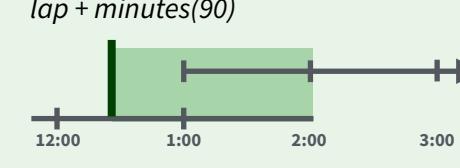
`nor + minutes(90)`



`gap + minutes(90)`



`lap + minutes(90)`

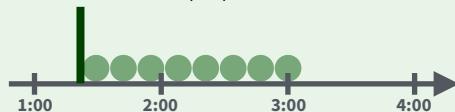


`leap + years(1)`

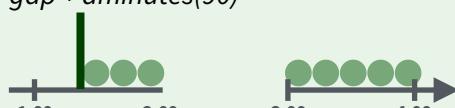


**Durations** track the passage of physical time, which deviates from clock time when irregularities occur.

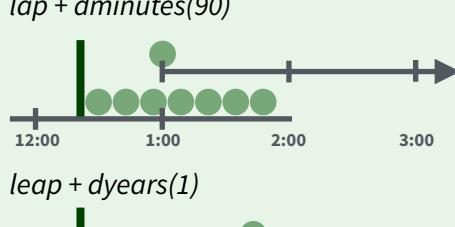
`nor + dminutes(90)`



`gap + dminutes(90)`



`lap + dminutes(90)`



`leap + dyears(1)`



**Intervals** represent specific intervals of the timeline, bounded by start and end date-times.

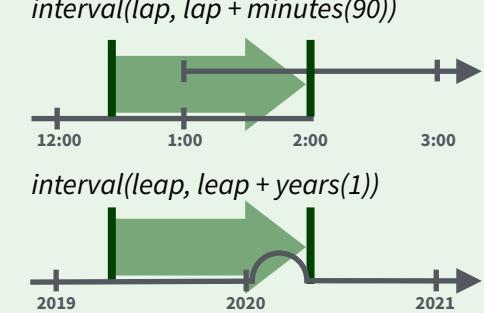
`interval(nor, nor + minutes(90))`



`interval(gap, gap + minutes(90))`



`interval(lap, lap + minutes(90))`



Not all years are 365 days due to **leap days**.

Not all minutes are 60 seconds due to **leap seconds**.

It is possible to create an imaginary date by adding **months**, e.g. February 31st

```
jan31 <- ymd(20180131)
jan31 + months(1)
## NA
```

`%m+%` and `%m-%` will roll imaginary dates to the last day of the previous month.

```
jan31 %m+% months(1)
## "2018-02-28"
```

`add_with_rollback(e1, e2, roll_to_first = TRUE)` will roll imaginary dates to the first day of the new month.

```
add_with_rollback(jan31, months(1),
roll_to_first = TRUE)
## "2018-03-01"
```

## PERIODS

Add or subtract periods to model events that happen at specific clock times, like the NYSE opening bell.

Make a period with the name of a time unit **pluralized**, e.g.

```
p <- months(3) + days(12)
```

"3m 12d 0H 0M 0S"

Number of months Number of days etc.

`years(x = 1) x years.`

`months(x) x months.`

`weeks(x = 1) x weeks.`

`days(x = 1) x days.`

`hours(x = 1) x hours.`

`minutes(x = 1) x minutes.`

`seconds(x = 1) x seconds.`

`milliseconds(x = 1) x milliseconds.`

`microseconds(x = 1) x microseconds`

`nanoseconds(x = 1) x nanoseconds.`

`picoseconds(x = 1) x picoseconds.`

`period(num = NULL, units = "second", ...)`

An automation friendly period constructor.

`period(5, unit = "years")`

`as.period(x, unit)` Coerce a timespan to a period, optionally in the specified units.

Also `is.period()`. `as.period(i)`

`period_to_seconds(x)` Convert a period to the "standard" number of seconds implied by the period. Also `seconds_to_period()`.  
`period_to_seconds(p)`

## DURATIONS

Add or subtract durations to model physical processes, like battery life. Durations are stored as seconds, the only time unit with a consistent length.

**Diftimes** are a class of durations found in base R.

Make a duration with the name of a period prefixed with a **d**, e.g.

```
dd <- ddays(14)
```

"1209600s (~2 weeks)"

Exact length in seconds Equivalent in common units

`dyears(x = 1) 31536000x seconds.`

`dweeks(x = 1) 604800x seconds.`

`ddays(x = 1) 86400x seconds.`

`dhours(x = 1) 3600x seconds.`

`dminutes(x = 1) 60x seconds.`

`dseconds(x = 1) x seconds.`

`dmilliseconds(x = 1) x × 10-3 seconds.`

`dmicroseconds(x = 1) x × 10-6 seconds.`

`dnanoseconds(x = 1) x × 10-9 seconds.`

`dpicoseconds(x = 1) x × 10-12 seconds.`

`duration(num = NULL, units = "second", ...)`

An automation friendly duration constructor. `duration(5, unit = "years")`

`as.duration(x, ...)` Coerce a timespan to a duration. Also `is.duration()`, `is.difftime()`. `as.duration(i)`

`make_difftime(x)` Make difftime with the specified number of units.  
`make_difftime(99999)`

## INTERVALS

Divide an interval by a duration to determine its physical length, divide and interval by a period to determine its implied length in clock time.

Make an interval with **interval()** or `%--%`, e.g.

```
i <- interval(ymd("2017-01-01"), d)
```

```
## 2017-01-01 UTC--2017-11-28 UTC
```

```
j <- d %--% ymd("2017-12-31")
```

```
## 2017-11-28 UTC--2017-12-31 UTC
```

Start Date End Date  
a %within% b Does interval or date-time a fall within interval b? `now() %within% i`

`int_start(int)` Access/set the start date-time of an interval. Also `int_end()`. `int_start(i) < now(); int_start(i)`

`int_aligns(int1, int2)` Do two intervals share a boundary? Also `int_overlaps()`. `int_aligns(i, j)`

`int_diff(times)` Make the intervals that occur between the date-times in a vector.  
`v <- c(dt, dt + 100, dt + 1000); int_diff(v)`

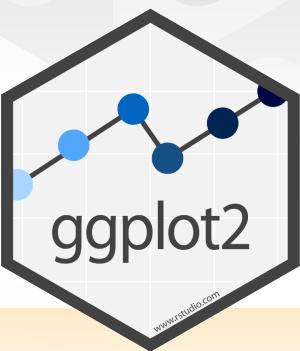
`int_flip(int)` Reverse the direction of an interval. Also `int_standardize()`. `int_flip(i)`

`int_length(int)` Length in seconds. `int_length(i)`

`int_shift(int, by)` Shifts an interval up or down the timeline by a timespan. `int_shift(i, days(-1))`

`as.interval(x, start, ...)` Coerce a timespans to an interval with the start date-time. Also `is.interval()`. `as.interval(days(1), start = now())`

# Data Visualization with ggplot2 :: CHEAT SHEET



## Basics

**ggplot2** is based on the **grammar of graphics**, the idea that you can build every graph from the same components: a **data** set, a **coordinate system**, and geoms—visual marks that represent data points.



To display values, map variables in the data to visual properties of the geom (**aesthetics**) like **size**, **color**, and **x** and **y** locations.



Complete the template below to build a graph.

```
ggplot (data = <DATA>) +
<GEOM_FUNCTION> (mapping = aes(<MAPPINGS>),
stat = <STAT>, position = <POSITION>) +
<COORDINATE_FUNCTION> +
<FACET_FUNCTION> +
<SCALE_FUNCTION> +
<THEME_FUNCTION>
```

required

Not required, sensible defaults supplied

**ggplot**(data = mpg, **aes**(x = cty, y = hwy)) Begins a plot that you finish by adding layers to. Add one geom function per layer.

**aesthetic mappings** **data** **geom**

**qplot**(x = cty, y = hwy, data = mpg, geom = "point") Creates a complete plot with given data, geom, and mappings. Supplies many useful defaults.

**last\_plot()** Returns the last plot

**gsave**("plot.png", **width** = 5, **height** = 5) Saves last plot as 5' x 5' file named "plot.png" in working directory. Matches file type to file extension.

## Geoms

Use a geom function to represent data points, use the geom's aesthetic properties to represent variables. Each function returns a layer.

### GRAPHICAL PRIMITIVES

- a <- ggplot(economics, aes(date, unemploy))  
b <- ggplot(seals, aes(x = long, y = lat))
- a + geom\_blank()**  
(Useful for expanding limits)
- b + geom\_curve(aes(yend = lat + 1, xend = long + 1, curvature = z))** - x, yend, alpha, angle, color, curvature, linetype, size
- a + geom\_path(lineend = "butt", linejoin = "round", linemitre = 1)** - x, y, alpha, color, group, linetype, size
- a + geom\_polygon(aes(group = group))** - x, y, alpha, color, fill, group, linetype, size
- b + geom\_rect(aes(xmin = long, ymin = lat, xmax = long + 1, ymax = lat + 1))** - xmax, xmin, ymax, ymin, alpha, color, fill, linetype, size
- a + geom\_ribbon(aes(ymin = unemploy - 900, ymax = unemploy + 900))** - x, ymax, ymin, alpha, color, fill, group, linetype, size

### LINE SEGMENTS

- common aesthetics: x, y, alpha, color, linetype, size
- b + geom\_abline(aes(intercept = 0, slope = 1))**
  - b + geom\_hline(aes(yintercept = lat))**
  - b + geom\_vline(aes(xintercept = long))**
  - b + geom\_segment(aes(yend = lat + 1, xend = long + 1))**
  - b + geom\_spoke(aes(angle = 1:1155, radius = 1))**

### ONE VARIABLE continuous

- c <- ggplot(mpg, aes(hwy)); c2 <- ggplot(mpg)
- c + geom\_area(stat = "bin")** - x, y, alpha, color, fill, linetype, size
- c + geom\_density(kernel = "gaussian")** - x, y, alpha, color, fill, group, linetype, size, weight
- c + geom\_dotplot()** - x, y, alpha, color, fill
- c + geom\_freqpoly()** - x, y, alpha, color, group, linetype, size
- c + geom\_histogram(binwidth = 5)** - x, y, alpha, color, fill, linetype, size, weight
- c2 + geom\_qq(aes(sample = hwy))** - x, y, alpha, color, fill, linetype, size, weight

### discrete

- d <- ggplot(mpg, aes(f1))
- d + geom\_bar()** - x, alpha, color, fill, linetype, size, weight

### TWO VARIABLES

#### continuous x , continuous y

- e <- ggplot(mpg, aes(cty, hwy))
- e + geom\_label(aes(label = cty), nudge\_x = 1, nudge\_y = 1, check\_overlap = TRUE)** - x, y, label, alpha, angle, color, family, fontface, hjust, lineheight, size, vjust

- e + geom\_jitter(height = 2, width = 2)** - x, y, alpha, color, fill, shape, size

- e + geom\_point()** - x, y, alpha, color, fill, shape, size, stroke

- e + geom\_quantile()** - x, y, alpha, color, group, linetype, size, weight

- e + geom\_rug(sides = "bl")** - x, y, alpha, color, linetype, size

- e + geom\_smooth(method = lm)** - x, y, alpha, color, fill, group, linetype, size, weight

- e + geom\_text(aes(label = cty), nudge\_x = 1, nudge\_y = 1, check\_overlap = TRUE)** - x, y, label, alpha, angle, color, family, fontface, hjust, lineheight, size, vjust

#### discrete x , continuous y

- f <- ggplot(mpg, aes(class, hwy))

- f + geom\_col()** - x, y, alpha, color, fill, group, linetype, size

- f + geom\_boxplot()** - x, y, lower, middle, upper, ymax, ymin, alpha, color, fill, group, linetype, shape, size, weight

- f + geom\_dotplot(binaxis = "y", stackdir = "center")** - x, y, alpha, color, fill, group

- f + geom\_violin(scale = "area")** - x, y, alpha, color, fill, group, linetype, size, weight

#### discrete x , discrete y

- g <- ggplot(diamonds, aes(cut, color))

- g + geom\_count()** - x, y, alpha, color, fill, shape, size, stroke

### THREE VARIABLES

- seals\$z <- with(seals, sqrt(delta\_long^2 + delta\_lat^2))  
l <- ggplot(seals, aes(long, lat))

- l + geom\_contour(aes(z = z))** - x, y, z, alpha, colour, group, linetype, size, weight

#### continuous bivariate distribution

- h <- ggplot(diamonds, aes(carat, price))
- h + geom\_bin2d(binwidth = c(0.25, 500))** - x, y, alpha, color, fill, linetype, size, weight

- h + geom\_density2d()** - x, y, alpha, colour, group, linetype, size

- h + geom\_hex()** - x, y, alpha, colour, fill, size

#### continuous function

- i <- ggplot(economics, aes(date, unemploy))

- i + geom\_area()** - x, y, alpha, color, fill, linetype, size

- i + geom\_line()** - x, y, alpha, color, group, linetype, size

- i + geom\_step(direction = "hv")** - x, y, alpha, color, group, linetype, size

#### visualizing error

- df <- data.frame(grp = c("A", "B"), fit = 4.5, se = 1.2)  
j <- ggplot(df, aes(grp, fit, ymin = fit - se, ymax = fit + se))

- j + geom\_crossbar(fatten = 2)** - x, y, ymax, ymin, alpha, color, fill, group, linetype, size

- j + geom\_errorbar()** - x, ymax, ymin, alpha, color, group, linetype, size, width (also **geom\_errorbarh()**)

- j + geom\_linerange()** - x, ymin, ymax, alpha, color, group, linetype, size

- j + geom\_pointrange()** - x, y, ymin, ymax, alpha, color, fill, group, linetype, shape, size

#### maps

- data <- data.frame(murder = USArrests\$Murder, state = tolower(rownames(USArrests)))  
map <- map\_data("state")  
k <- ggplot(data, aes(fill = murder))

- k + geom\_map(aes(map\_id = state), map = map)**  
**+ expand\_limits(x = map\$long, y = map\$lat)**, map\_id, alpha, color, fill, linetype, size

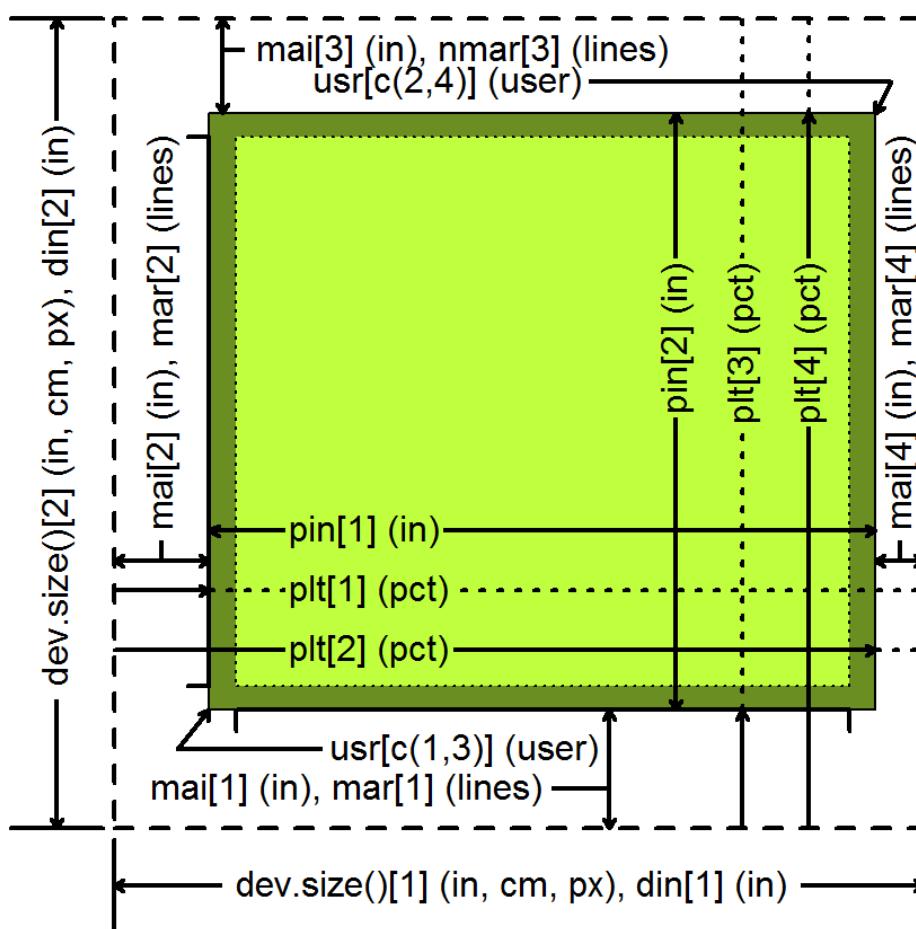


# How Big is Your Graph?

## An R Cheat Sheet

### Introduction

All functions that open a device for graphics will have **height** and **width** arguments to control the size of the graph and a **pointsize** argument to control the relative font size. In **knitr**, you control the size of the graph with the chunk options, **fig.width** and **fig.height**. This sheet will help you with calculating the size of the graph and various parts of the graph within R.



### Your graphics device

**dev.size()** (width, height)  
**par("din")** (r.o.) (width, height) in inches

Both the **dev.size** function and the **din** argument of **par** will tell you the size of the graphics device. The **dev.size** function will report the size in

1. inches (**units="in"**), the default
2. centimeters (**units="cm"**)
3. pixels (**units="px"**)

Like several other **par** arguments, **din** is read only (r.o.) meaning that you can ask its current value (**par("din")**) but you cannot change it (**par(din=c(5,7))** will fail).

### Your plot margins

**par("mai")** (bottom, left, top, right) in inches  
**par("mar")** (bottom, left, top, right) in lines

Margins provide you space for your axes, axis labels, and titles.

A "line" is the amount of vertical space needed for a line of text.

If your graph has no axes or titles, you can remove the margins (and maximize the plotting region) with

```
par(mar=rep(0,4))
```

### Your plotting region

**par("pin")** (width, height) in inches  
**par("plt")** (left, right, bottom, top) in pct

The **pin** argument **par** gives you the size of the plotting region (the size of the device minus the size of the margins) in inches.

The **plt** argument **par** gives you the percentage of the device from the left/bottom edge up to the left edge of the plotting region, the right edge, the bottom edge, and the top edge. The first and third values are equivalent to the percentage of space devoted to the left and bottom margins. Subtract the second and fourth values from 1 to get the percentage of space devoted to the right and top margins.

### Your x-y coordinates

**par("usr")** (xmin, ymin, xmax, ymax)

Your x-y coordinates are the values you use when plotting your data. This normally is not the same as the values you specified with the **xlim** and **ylim** arguments in **plot**. By default, R adds an extra 4% to the plotting range (see the dark green region on the figure) so that points right up on the edges of your plot do not get partially clipped. You can override this by setting **xaxs="i"** and/or the **yaxs="i"** in **par**.

Run **par("usr")** to find the minimum X value, the maximum X value, the minimum Y value, and the maximum Y value. If you assign new values to **usr**, you will update the x-y coordinates to the new values.

### Getting a square graph

**par("pty")**

You can produce a square graph manually by setting the width and height to the same value and setting the margins so that the sum of the top and bottom margins equal the sum of the left and right margins. But a much easier way is to specify **pty="s"**, which adjusts the margins so that the size of the plotting region is always square, even if you resize the graphics window.

### Converting units

For many applications, you need to be able to translate user coordinates to pixels or inches. There are some cryptic shortcuts, but the simplest way is to get the range in user coordinates and measure the proportion of the graphics device devoted to the plotting region.

```
user.range <- par("usr")[c(2,4)] - par("usr")[c(1,3)]
```

```
region.pct <- par("plt")[c(2,4)] - par("plt")[c(1,3)]
```

```
region.px <- dev.size(units="px") * region.pct
```

```
px.per.xy <- region.px / user.range
```

To convert a horizontal or distance from the x-coordinate value to pixels, multiply by **px.per.xy[1]**. To convert a vertical distance, multiply by **region.px.per.xy[2]**. To convert a diagonal distance, you need to invoke Pythagoras.

```
a.px <- x.dist*px.per.xy[1]
b.px <- y.dist*px.per.xy[2]
c.px <- sqrt(a.px^2+b.px^2)
```

To rotate a string to match the slope of a line segment, you need to convert the distances to pixels, calculate the arctangent, and convert from radians to degrees.

```
segments(x0, y0, x1, y1)
delta.x <- (x1 - x0) * px.per.xy[1]
delta.y <- (y1 - y0) * px.per.xy[2]
angle.radians <- atan2(delta.y, delta.x)
angle.degrees <- angle.radians * 180 / pi
text(x1, y1, "TEXT", srt=angle.degrees)
```

## Panels

`par("fig")` (width, height) in pct  
`par("fin")` (width, height) in inches

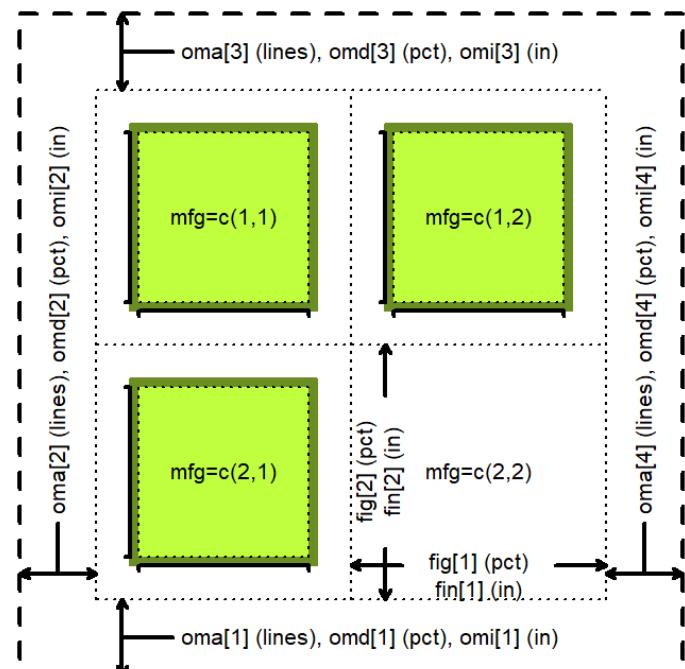
If you display multiple plots within a single graphics window (e.g., with the `mfrow` or `mfcol` arguments of `par` or with the `layout` function), then the `fig` and `fin` arguments will tell you the size of the current subplot window in percent or inches, respectively.

`par("oma")` (bottom, left, top, right) in lines  
`par("omd")` (bottom, left, top, right) in pct  
`par("omi")` (bottom, left, top, right) in inches

Each subplot will have margins specified by `mai` or `mar`, but no outer margin around the entire set of plots, unless you specify them using `oma`, `omd`, or `omi`. You can place text in the outer margins using the `mtext` function with the argument `outer=TRUE`.

`par("mfg")` (r, c) or (r, c, maxr, maxc)

The `mfg` argument of `par` will allow you to jump to a subplot in a particular row and column. If you query with `par("mfg")`, you will get the current row and column followed by the maximum row and column.



## Character and string sizes

### `strheight()`

The `strheight` functions will tell you the height of a specified string in inches (`units="inches"`), x-y user coordinates (`units="user"`) or as a percentage of the graphics device (`units="figure"`).

For a single line of text, `strheight` will give you the height of the letter "M". If you have a string with one or more linebreaks ("n"), the `strheight` function will measure the height of the letter "M" plus the height of one or more additional lines. The height of a line is dependent on the line spacing, set by the `lheight` argument of `par`. The default line height (`lheight=1`), corresponding to single spaced lines, produces a line height roughly 1.5 times the height of "M".

### `strwidth()`

The `strwidth` function will produce different widths to individual characters, representing the proportional spacing used by most fonts ("W" using much more space than an "i"). For the width of a string, the `strwidth` function will sum up the lengths of the individual characters in the string.

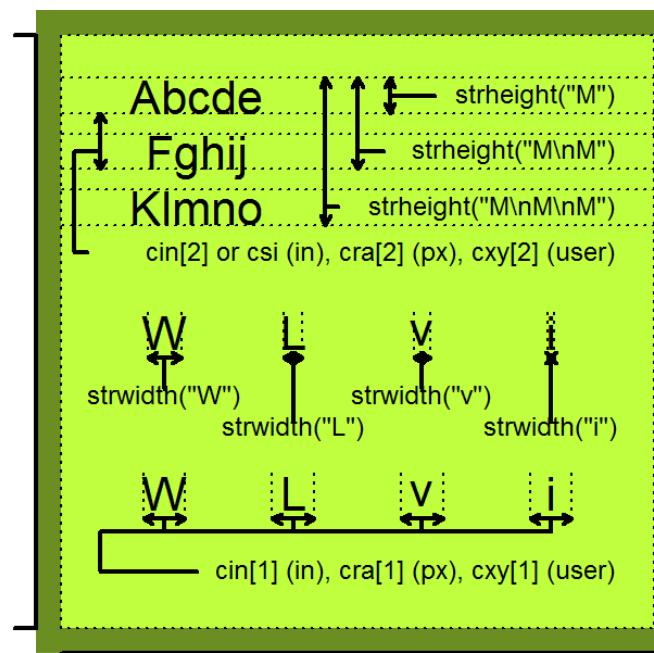
`par("cin")` (r.o.) (width, height) in inches  
`par("csi")` (r.o.) height in inches  
`par("cra")` (r.o.) (width, height) in pixels  
`par("cxy")` (r.o.) (width, height) in xy coordinates

The single value returned by the `csi` argument of `par` gives you the height of a line of text in inches. The second of the two values returned by `cin`, `cra`, and `cxy` gives you the height of a line, in inches, pixels, or xy (user) coordinates.

The first of the two values returned by the `cin`, `cra`, and `cxy` arguments to `par` gives you the approximate width of a single character, in inches, pixels, or xy (user) coordinates. The width, very slightly smaller than the actual width of the letter "W", is a rough estimate at best and ignores the variable width of individual letters.

These values are useful, however, in providing fast ratios of the relative sizes of the differing units of measure

`px.per.in <- par("cra") / par("cin")`  
`px.per.xy <- par("cra") / par("cxy")`  
`xy.per.in <- par("cxy") / par("cin")`



## If your fonts are too big or too small

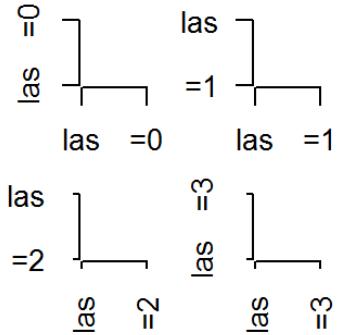
Fixing this takes a bit of trial and error.

- Specify a larger/smaller value for the `pointsize` argument when you open your graphics device.
- Try opening your graphics device with different values for `height` and `width`. Fonts that look too big might be better proportioned in a larger graphics window.
- Use the `cex` argument to increase or decrease the relative size of your fonts.

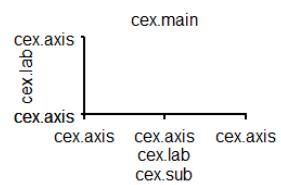
## If your axes don't fit

There are several possible solutions.

- You can assign wider margins using the `mar` or `mai` argument in `par`.
- You can change the orientation of the axis labels with `las`. Choose among
  - `las=0` both axis labels parallel
  - `las=1` both axis labels horizontal
  - `las=2` both axis labels perpendicular
  - `las=3` both axis labels vertical.



- change the relative size of the font
  - `cex.axis` for the tick mark labels.
  - `cex.lab` for `xlab` and `ylab`.
  - `cex.main` for the main title
  - `cex.sub` for the subtitle.

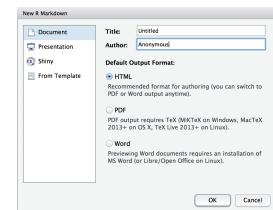


# R Markdown :: CHEAT SHEET

## What is R Markdown?

- .Rmd files** • An R Markdown (.Rmd) file is a record of your research. It contains the code that a scientist needs to reproduce your work along with the narration that a reader needs to understand your work.
- Reproducible Research** • At the click of a button, or the type of a command, you can rerun the code in an R Markdown file to reproduce your work and export the results as a finished report.
- Dynamic Documents** • You can choose to export the finished report in a variety of formats, including html, pdf, MS Word, or RTF documents; html or pdf based slides, Notebooks, and more.

## Workflow



- ① Open a new .Rmd file at File ► New File ► R Markdown. Use the wizard that opens to pre-populate the file with a template
- ② Write document by editing template
- ③ Knit document to create report; use knit button or render() to knit
- ④ Preview Output in IDE window
- ⑤ Publish (optional) to web server
- ⑥ Examine build log in R Markdown console
- ⑦ Use output file that is saved along side .Rmd

The screenshot illustrates the R Markdown workflow within RStudio. It shows the R Markdown editor with code like `summary(cars)` and its resulting output. The preview window shows the rendered HTML page. The file browser shows the generated `report.html` file. The terminal window shows the command used to render the document.

## render

Use `rmarkdown::render()` to render/knit at cmd line. Important args:

**input** - file to render  
**output\_format**

**output\_options** - List of render options (as in YAML)

**output\_file**  
**output\_dir**

**params** - list of params to use

**envir** - environment to evaluate code chunks in

**encoding** - of input file

## Embed code with knitr syntax

### INLINE CODE

Insert with `r <code>`. Results appear as text without code.

Built with `r getRVersion()` ➔ Built with 3.2.3

### CODE CHUNKS

One or more lines surrounded with `{{r}}` and `{{ }}`. Place chunk options within curly braces, after r. Insert with ➔

```{r echo=TRUE}  
getRVersion()  
```

### GLOBAL OPTIONS

Set with `knitr::opts_chunk$set()`, e.g.

```{r include=FALSE}  
knitr::opts\_chunk\$set(echo = TRUE)  
```

### IMPORTANT CHUNK OPTIONS

**cache** - cache results for future knits (default = FALSE)

**cache.path** - directory to save cached results in (default = "cache/")

**child** - file(s) to knit and then include (default = NULL)

**collapse** - collapse all output into single block (default = FALSE)

**comment** - prefix for each line of results (default = "#")

**dependson** - chunk dependencies for caching (default = NULL)

**echo** - Display code in output document (default = TRUE)

**engine** - code language used in chunk (default = 'R')

**error** - Display error messages in doc (TRUE) or stop render when errors occur (FALSE) (default = FALSE)

**eval** - Run code in chunk (default = TRUE)

Options not listed above: `R.options`, `aniopts`, `autodep`, `background`, `cache.comments`, `cache.lazy`, `cache.rebuild`, `cache.vars`, `dev`, `dev.args`, `dpi`, `engine.opts`, `engine.path`, `fig.asp`, `fig.env`, `fig.ext`, `fig.keep`, `fig.lp`, `fig.path`, `fig.pos`, `fig.process`, `fig.retina`, `fig.scap`, `fig.show`, `fig.showtext`, `fig.subcap`, `interval`, `out.extra`, `out.height`, `out.width`, `prompt`, `purl`, `ref.label`, `render`, `size`, `split`, `tidy.opts`

**fig.align** - 'left', 'right', or 'center' (default = 'default')

**fig.cap** - figure caption as character string (default = NULL)

**fig.height**, **fig.width** - Dimensions of plots in inches

**highlight** - highlight source code (default = TRUE)

**include** - Include chunk in doc after running (default = TRUE)

**message** - display code messages in document (default = TRUE)

**results** (default = 'markup')

'asis' - passthrough results

'hide' - do not display results

'hold' - put all results below all code

**tidy** - tidy code for display (default = FALSE)

**warning** - display code warnings in document (default = TRUE)

## .rmd Structure



### YAML Header

Optional section of render (e.g. pandoc) options written as key:value pairs (YAML).

At start of file

Between lines of ---

### Text

Narration formatted with markdown, mixed with:

### Code Chunks

Chunks of embedded code. Each chunk:

Begins with `{{r}}`

ends with `{{ }}`

R Markdown will run the code and append the results to the doc.

It will use the location of the .Rmd file as the **working directory**

## Parameters

Parameterize your documents to reuse with different inputs (e.g., data, values, etc.)

```
---  
params:  
  n: 100  
  d: ! Sys.Date()  
---
```

Today's date is `r params\$d`

ABC 🔎 Knit HTML  
Knit to HTML  
Knit to PDF  
Knit to Word  
Knit with Parameters...

## Interactive Documents

Turn your report into an interactive Shiny document in 4 steps

1. Add runtime: shiny to the YAML header.
2. Call Shiny input functions to embed input objects.
3. Call Shiny render functions to embed reactive output.
4. Render with `rmarkdown::run` or click Run Document in RStudio IDE

```
---  
output: html_document  
runtime: shiny  
---  
{{r, echo = FALSE}}  
numericInput("n",  
  "How many cars?", 5)  
renderTable({  
  head(cars, input$n)  
})
```

How many cars?	
speed	dist
1	4.00 2.00
2	4.00 10.00
3	7.00 4.00
4	7.00 22.00
5	8.00 16.00

Embed a complete app into your document with `shiny::shinyAppDir()`

NOTE: Your report will be rendered as a Shiny app, which means you must choose an html output format, like `html_document`, and serve it with an active R Session.





# Pandoc's Markdown

Write with syntax on the left to create effect on right (after render)

```
Plain text
End a line with two spaces
to start a new paragraph.
*italics* and **bold**
`verbatim` code
sub/superscript22
~~strikethrough~~
escaped: `*` \\
endash: --, emdash: ---
equation: $A = \pi * r^2$
```

```
equation block:
```

```
$$E = mc^2$$
```

```
> block quote
```

```
# Header1 {#anchor}
```

```
## Header 2 {#css_id}
```

```
### Header 3 {.css_class}
```

```
#### Header 4
```

```
##### Header 5
```

```
##### Header 6
```

```
<!--Text comment-->
```

```
\textbf{Text ignored in HTML}
<em>HTML ignored in pdfs</em>
```

```
<http://www.rstudio.com>
[link](www.rstudio.com)
Jump to [Header 1]{#anchor}
image:
```

```
Plain text
End a line with two spaces
to start a new paragraph.
*italics* and **bold**
`verbatim` code
sub/superscript22
~~strikethrough~~
escaped: `*` \\
endash: --, emdash: ---
equation: $A = \pi * r^2$
```

```
equation block:
```

```
$$E = mc^2$$
```

```
> block quote
```

## Header1

## Header 2

### Header 3

#### Header 4

##### Header 5

##### Header 6

HTML ignored in pdfs

<http://www.rstudio.com>

link

Jump to Header 1

image:



Caption

- unordered list
  - sub-item 1
  - sub-item 2
  - sub-sub-item 1
- item 2

Continued (indent 4 spaces)

1. ordered list
2. item 2
  - i. sub-item 1
    - A. sub-sub-item 1

(@) A list whose numbering

continues after

2. an interruption

Term 1

Definition 1

Right	Left	Default	Center
12	12	12	12
123	123	123	123
1	1	1	1

- slide bullet 1
- slide bullet 2

(>- to have bullets appear on click)

horizontal rule/slide break:

\*\*\*

A footnote<sup>[1]</sup>

[^1]: Here is the footnote.

1. Here is the footnote.[^1](#)

# Set render options with YAML

When you render, R Markdown

1. runs the R code, embeds results and text into .md file with knitr
2. then converts the .md file into the finished format with pandoc



Set a document's default output format in the YAML header:

```
---  
output: html_document  
---  
# Body
```

### output value

### creates

html_document	html
pdf_document	pdf (requires Tex)
word_document	Microsoft Word (.docx)
odt_document	OpenDocument Text
rtf_document	Rich Text Format
md_document	Markdown
github_document	Github compatible markdown
ioslides_presentation	ioslides HTML slides
slidy_presentation	slidy HTML slides
beamer_presentation	Beamer pdf slides (requires Tex)

Customize output with sub-options (listed to the right):

```
---  
output: html_document:  
  code_folding: hide  
  toc_float: TRUE  
---  
# Body
```

### html tabs

Use tablet css class to place sub-headers into tabs

```
# Tabset {.tabset .tabset-fade .tabset-pills}  
## Tab 1  
text 1  
## Tab 2  
text 2  
### End tabset
```



## Create a Reusable Template

1. Create a new package with a `inst/rmarkdown/templates` directory

2. In the directory, Place a folder that contains:

`template.yaml` (see below)

`skeleton.Rmd` (contents of the template)

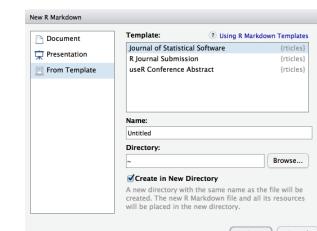
any supporting files

3. Install the package

4. Access template in wizard at File ▶ New File ▶ R Markdown template.yaml

A footnote<sup>1</sup>

1. Here is the footnote.[^1](#)



### sub-option

`citation_package` The LaTeX package to process citations, natbib, biblatex or none

`code_folding` Let readers to toggle the display of R code, "none", "hide", or "show"

`colortheme` Beamer color theme to use

`css` CSS file to use to style document

`dev` Graphics device to use for figure output (e.g. "png")

`duration` Add a countdown timer (in minutes) to footer of slides

`fig_caption` Should figures be rendered with captions?

`fig_height, fig_width` Default figure height and width (in inches) for document

`highlight` Syntax highlighting: "tango", "pygments", "kate", "zenburn", "textmate"

`includes` File of content to place in document (in\_header, before\_body, after\_body)

`incremental` Should bullets appear one at a time (on presenter mouse clicks)?

`keep_md` Save a copy of .md file that contains knitr output

`keep_tex` Save a copy of .tex file that contains knitr output

`latex_engine` Engine to render latex, "pdflatex", "xelatex", or "lualatex"

`lib_dir` Directory of dependency files to use (Bootstrap, MathJax, etc.)

`mathjax` Set to local or a URL to use a local/URL version of MathJax to render equations

`md_extensions` Markdown extensions to add to default definition or R Markdown

`number_sections` Add section numbering to headers

`pandoc_args` Additional arguments to pass to Pandoc

`preserve_yaml` Preserve YAML front matter in final document?

`reference_docx` docx file whose styles should be copied when producing docx output

`self_contained` Embed dependencies into the doc

`slide_level` The lowest heading level that defines individual slides

`smaller` Use the smaller font size in the presentation?

`smart` Convert straight quotes to curly, dashes to em-dashes, ... to ellipses, etc.

`template` Pandoc template to use when rendering file quarterly\_report.html

`theme` Bootswatch or Beamer theme to use for page

`toc` Add a table of contents at start of document

`toc_depth` The lowest level of headings to add to table of contents

`toc_float` Float the table of contents to the left of the main content

	html	pdf	word	odt	rtf	md	gitbook	ioslides	slidify	beamer
X				X						X

## Table Suggestions

Several functions format R data into tables

Table with kable

eruptions	waiting
3.600	79
1.800	54
3.333	74
2.283	62

eruptions waiting

1	3.60	79.00
2	1.80	54.00
3	3.33	74.00
4	2.28	62.00

eruptions waiting

eruptions	waiting
1	3.60
2	1.80
3	3.33
4	2.28

eruptions waiting

eruptions	waiting





<tbl\_r cells="2" ix="5" maxcspan="1" maxrspan="

# R 마크다운

전공쪽지

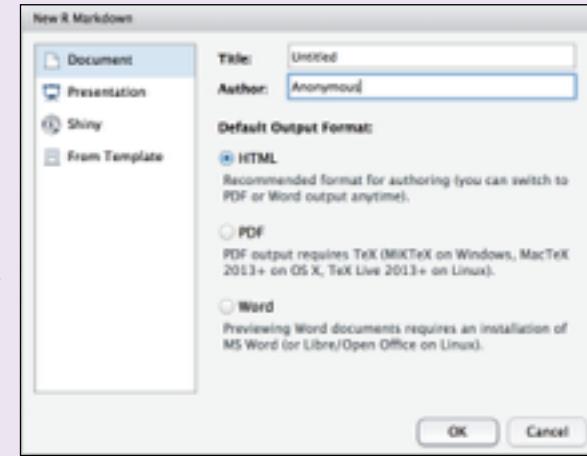
추가 학습 정보 [rmarkdown.rstudio.com](http://rmarkdown.rstudio.com)

rmarkdown 0.2.50 최종갱신일: 8/14



**2. 파일 열기** .Rmd 확장자를 갖는 텍스트 파일로 저장해서 시작하거나, Studio Rmd 템플릿을 열어 시작한다.

- 메뉴막대에서, 다음순으로 클릭한다.  
**File ▶ New File ▶ R Markdown...**
- 윈도우가 열리면, .Rmd 파일로 작성하려는 출력유형을 선택한다.
- 라디오 버튼으로 출력형식을 선택한다 (나중에 출력형식은 변경할 수 있다)
- OK 버튼을 클릭한다.



**4. 출력형식 선정** R 마크다운 파일에서 생성할 문서 유형을 기술하는 YAML 헤더정보를 작성한다.

## YAML

YAML 헤더는 키(key) 집합: 파일 시작지점에 나오는 키-값 쌍. 헤더 시작과 끝은 3개 대쉬를 갖는 라인 (- - -)

```
---  
title: "xwMOOC 보고서"  
author: "무명씨"  
output: html_document  
---
```

보고서의 시작지점으로, YAML 헤더에 메타데이터 정보가 저장되어 있다.

RStudio 템플릿이  
자동으로 YAML 헤더  
정보를 작성해 넣는다.

출력값이 .Rmd 파일에서 어떤 형식 파일을 생성할 것인지 결정한다(단계 6)

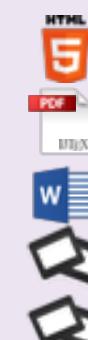
**output: html\_document** ..... html 파일 (웹페이지)

**output: pdf\_document** ..... pdf 문서

**output: word\_document** ..... MS 워드문서 .docx

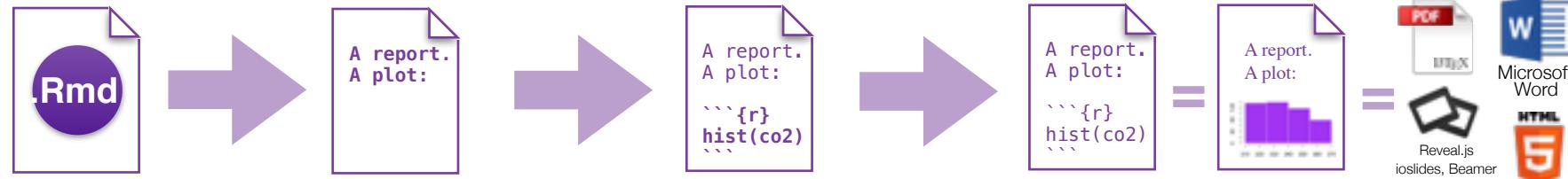
**output: beamer\_presentation** ..... Beamer 발표자료 (pdf)

**output: ioslides\_presentation** ..... 맥 발표자료 (html)



**1. 작업흐름** R 마크다운은 R로 재현가능하고, 동적인 보고서를 작성하는 서식이다. R 마크다운을 사용해서 R 코드와 실행결과를 발표자료, pdf, html, 워드 문서 등에 삽입할 수 있다. 보고서를 작성하려면:

- 파일열기 - .Rmd 확장자를 갖는 파일을 연다.
- 작성하기 - 본문을 작성하기 쉬운 R 마크다운 구문을 사용해서 작성한다.
- 내장하기 - 리포트에 포함될 출력 결과를 생성하는 R 코드를 내장한다.
- 렌더링(Render) - R 코드를 출력형식으로 치환하고 보고서를 발표자료, pdf, html, MS 워드 파일 형식으로 변환한다.



**3. 마크다운** 다음으로, 일반 텍스트로 보고서를 작성한다. 마크다운 구문을 사용해서 최종 보고서에 적용할 텍스트 서식을 기술한다.

## 입력 구문

일반 텍스트  
새로운 단락을 시작하려면 줄 마지막을 공백 2개로 끝낸다.  
\*기울인 글씨\* and \_기울인 글씨\_  
\*\*굵은 글씨\*\* and \_\_굵은 글씨\_\_  
윗첨자^2^  
~~취소선~~  
[링크](www.rstudio.com)

# 제목 1

## 제목 2

### 제목 3

#### 제목 4

##### 제목 5

###### 제목 6

N자 크기 대시 부호: --  
m자 크기 대시 부호: ---  
생략: ...  
즉시 처리하는 수식: \$A = \pi r^2\$  
이미지: 

수평선 (혹은 슬라이드 멈춤):

\*\*\*

> 인용 블록

\* 순서없는 목록  
\* 항목 2  
+ 하위 항목 1  
+ 하위 항목 2

1. 순서있는 목록  
2. 항목 2  
+ 하위 항목 1  
+ 하위 항목 2

표 제목 | 두번째 제목

표 칸 | 칸 2  
칸 3 | 칸 4

## 출력 결과

일반 텍스트  
새로운 단락을 시작하려면 줄 마지막을 공백 2개로 끝낸다.  
기울인 글씨 and 기울인 글씨  
굵은 글씨 and 굵은 글씨  
윗첨자^2^  
취소선  
링크

**제목 1**

**제목 2**

**제목 3**

제목 4

제목 5

제목 6

N자 크기 대시 부호: --  
m자 크기 대시 부호: ---  
생략: ...  
즉시 처리하는 수식:  $A = \pi r^2$   
이미지: R

수평선 (혹은 슬라이드 멈춤):

## 인용 블록

- 순서없는 목록
  - 항목 2
  - 하위 항목 1
  - 하위 항목 2
- 순서있는 목록
  - 항목 2
  - 하위 항목 1
  - 하위 항목 2

표 제목	두번째 제목
표 칸	칸 2
칸 3	칸 4

## 5. 코드내장하기

knitr 구문을 사용해서 R 코드를 보고서에 내장한다.  
R이 코드를 실행하고, 보고서를 렌더링할 때 결과를 포함시킨다.

### 인라인 코드

r 코드를 백틱(`)으로 감싼다.  
R이 인라인 코드를 실행된 결과로 대체한다.

2 더하기 2는 `r 2`  
2`와 같다.



### 화면 출력 선택옵션

knitr 선택옵션을 사용해서 코드 덩어리 출력 스타일을 적용한다.  
코드 상단 팔호 내부에 선택옵션을 지정한다.



### 선택옵션 기본설정 효과

	기본설정	효과
eval	TRUE	코드를 평가하고 실행결과를 포함한다.
echo	TRUE	실행결과와 함께 코드를 출력한다.
warning	TRUE	경고메시지를 출력한다.
error	FALSE	오류메시지를 출력한다.
message	TRUE	메시지를 출력한다.
tidy	FALSE	깔끔한 방식으로 코드 형태를 변형한다.
results	"markup"	"markup", "asis", "hold", "hide"
cache	FALSE	결과값을 캐쉬해서 향후 실행시 건너뛰게 설정한다.
comment	"##"	주석문자로 출력결과에 서두를 붙인다.
fig.width	7	덩어리로 생성되는 그래프에 대한 폭을 인치로 지정한다.
fig.height	7	덩어리로 생성되는 그래프에 대한 높이를 인치로 지정한다.

보다 자세한 사항은 웹사이트를 참조: [yihui.name/knitr/](http://yihui.name/knitr/)

## 6. 렌더링

최종보고서를 생성하는데 .Rmd 파일을 사용하여 청사진을 제작한다.

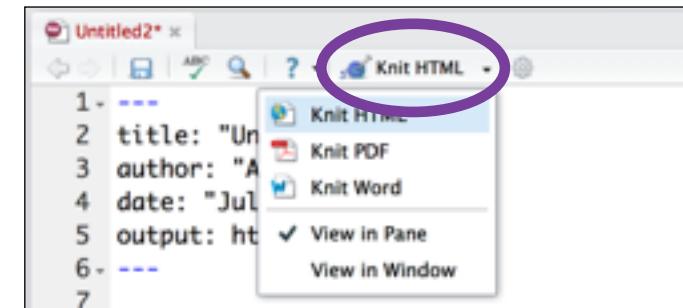
두가지 방식으로 보고서를 렌더링한다.

1. **rmarkdown:::render("파일 경로")** 명령어를 실행한다.

2. RStudio 스크립트 작성창 상단에 **knit HTML** 버튼을 클릭한다.

렌더링 명령을 실행시키면, R은 다음을 수행한다

- 내장된 코드 덩어리를 각각 실행시키고, 실행결과를 보고서에 삽입한다.
- 출력 파일형식에 맞춰 신규 보고서를 생성한다.
- 미리보기로 뷰어창에 출력파일을 연다.
- 작업디렉토리에 출력파일을 저장한다.



## 8. Publish

온라인으로 접속하는 사용자와 보고서를 공유한다.

### Rpubs.com

RStudio 무료 R 마크다운 게시 사이트를 통해 정적 문서를 공유한다.

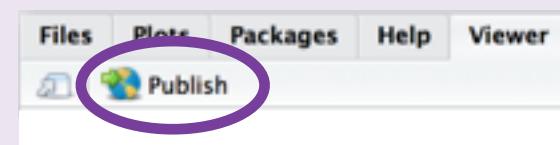
[www.rpubs.com](http://www.rpubs.com)

### ShinyApps.io

Studio 서버에 인터랙티브 문서를 올려 호스팅한다. 무료와 유료 선택옵션이 있다.

[www.shinyapps.io](http://www.shinyapps.io)

RStudio 미리보기 창에 "Publish" 버튼을 클릭하여, [rpubs.com](http://rpubs.com) 사이트에 버튼 한번 클릭으로 바로 올린다.



## 9. 추가 학습

문서와 예제 - [rmarkdown.rstudio.com](http://rmarkdown.rstudio.com)

추가 기사 - [shiny.rstudio.com/articles](http://shiny.rstudio.com/articles)

- [blog.rstudio.com](http://blog.rstudio.com)  
- [@rstudio](http://@rstudio)



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844-448-1212 [rstudio.com](http://rstudio.com)

# Advanced R

## Cheat Sheet

Created by: Arianne Colton and Sean Chen

### Environment Basics

Environment – **Data structure** (with two components below) that powers lexical scoping

Create environment: `env1<-new.env()`

1. **Named list** (“Bag of names”) – each name points to an object stored elsewhere in memory.

If an object has no names pointing to it, it gets automatically deleted by the garbage collector.

- Access with: `ls('env1')`

2. **Parent environment** – used to implement lexical scoping. If a name is not found in an environment, then R will look in its parent (and so on).

- Access with: `parent.env('env1')`

### Four special environments

1. **Empty environment** – ultimate ancestor of all environments
  - Parent: none
  - Access with: `emptyenv()`

2. **Base environment** - environment of the base package
  - Parent: empty environment
  - Access with: `baseenv()`

3. **Global environment** – the interactive workspace that you normally work in
  - Parent: environment of last attached package
  - Access with: `globalenv()`

4. **Current environment** – environment that R is currently working in (may be any of the above and others)
  - Parent: empty environment
  - Access with: `environment()`

## Environments

### Search Path

**Search path** – mechanism to look up objects, particularly functions.

- Access with : `search()` – lists all parents of the global environment (see Figure 1)
- Access any environment on the search path:  
`as.environment('package:base')`

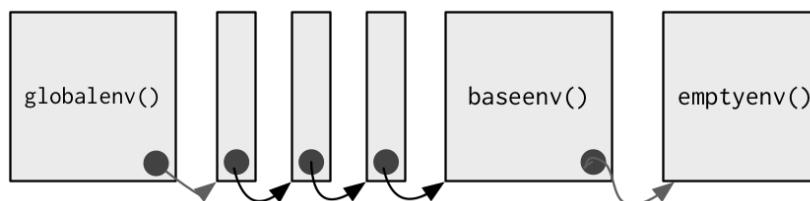


Figure 1 – The Search Path

- Mechanism : always start the search from global environment, then inside the latest attached package environment.
  - New package loading with `library()`/`require()` : new package is attached right after global environment. (See Figure 2)
  - Name conflict in two different package : functions with the same name, latest package function will get called.

`search()`:

```
'.GlobalEnv' ... 'Autoloads' 'package:base'  
library(reshape2); search()  
'.GlobalEnv' 'package:reshape2' ... 'Autoloads' 'package:base'
```

**NOTE:** Autoloads : special environment used for saving memory by only loading package objects (like big datasets) when needed

Figure 2 – Package Attachment

### Binding Names to Values

**Assignment** – act of binding (or rebinding) a name to a value in an environment.

1. `<-` (Regular assignment arrow) – always creates a variable in the current environment
2. `<--` (Deep assignment arrow) - modifies an existing variable found by walking up the parent environments

**Warning:** If `<--` doesn't find an existing variable, it will create one in the global environment.

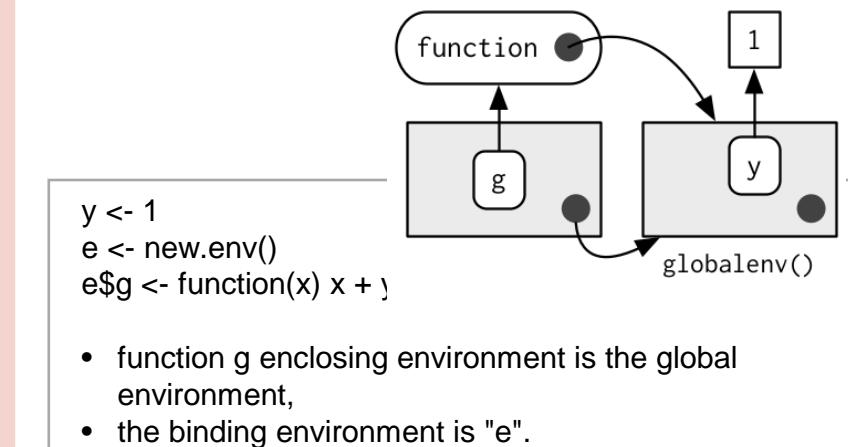
### Function Environments

1. **Enclosing environment** - an environment where the function is created. It determines how function finds value.
  - Enclosing environment never changes, even if the function is moved to a different environment.
  - Access with: `environment('func1')`

2. **Binding environment** - all environments that the function has a binding to. It determines how we find the function.

- Access with: `pryr::where('func1')`

**Example** (for enclosing and binding environment):



- function g enclosing environment is the global environment,
- the binding environment is "e".

3. **Execution environment** - new created environments to host a function call execution.

- Two parents :
  - I. Enclosing environment of the function
  - II. Calling environment of the function
- Execution environment is thrown away once the function has completed.

4. **Calling environment** - environments where the function was called.

- Access with: `parent.frame('func1')`
- Dynamic scoping :
  - About : look up variables in the calling environment rather than in the enclosing environment
  - Usage : most useful for developing functions that aid interactive data analysis

# Data Structures

	Homogeneous	Heterogeneous
1d	Atomic vector	List
2d	Matrix	Data frame
nd	Array	

**Note:** R has no 0-dimensional or scalar types. Individual numbers or strings, are actually vectors of length one, NOT scalars.

Human readable description of any R data structure :

```
str(variable)
```

Every **Object** has a mode and a class

- Mode:** represents how an object is stored in memory
  - 'type' of the object from R's point of view
  - Access with: `typeof()`
- Class:** represents the object's abstract type
  - 'type' of the object from R's object-oriented programming point of view
  - Access with: `class()`

	typeof()	class()
strings or vector of strings	character	character
numbers or vector of numbers	numeric	numeric
list	list	list
data.frame	list	data.frame

## Factors

- Factors are built on top of integer vectors using two attributes :

```
class(x) -> 'factor'
```

```
levels(x) # defines the set of allowed values
```

- Useful when you know the possible values a variable may take, even if you don't see all values in a given dataset.

### Warning on Factor Usage:

- Factors look and often behave like character vectors, they are actually integers. Be careful when treating them like strings.
- Most data loading functions automatically convert character vectors to factors. (Use argument `stringAsFactors = FALSE` to suppress this behavior)

# Object Oriented (OO) Field Guide

## Object Oriented Systems

R has three object oriented systems :

- S3** is a very casual system. It has no formal definition of classes. It implements generic function OO.
  - Generic-function OO** - a special type of function called a generic function decides which method to call.

Example:	<code>drawRect(canvas, 'blue')</code>
Language:	R

- Message-passing OO** - messages (methods) are sent to objects and the object determines which function to call.

Example:	<code>canvas.drawRect('blue')</code>
Language:	Java, C++, and C#

- S4** works similarly to S3, but is more formal. Two major differences to S3 :
  - Formal class definitions** - describe the representation and inheritance for each class, and has special helper functions for defining generics and methods.
  - Multiple dispatch** - generic functions can pick methods based on the class of any number of arguments, not just one.

- Reference classes** are very different from S3 and S4:

- Implements message-passing OO** - methods belong to classes, not functions.
- Notation** - \$ is used to separate objects and methods, so method calls look like `canvas$drawRect('blue')`.

## S3

### 1. About S3 :

- R's first and simplest OO system
- Only OO system used in the base and stats package
- Methods belong to functions, not to objects or classes.

### 2. Notation :

- `generic.class()`

<code>mean.Date()</code>	Date method for the generic - <code>mean()</code>
--------------------------	---

### 3. Useful 'Generic' Operations

- Get all methods that belong to the 'mean' generic:
  - `Methods('mean')`
- List all generics that have a method for the 'Date' class :
  - `methods(class = 'Date')`

### 4. S3 objects

- are usually built on top of lists, or atomic vectors with attributes.
- Factor and data frame are S3 class
  - Useful operations:

Check if object is an S3 object	<code>is.object(x) &amp; !isS4(x) or pryr::oGetType()</code>
Check if object inherits from a specific class	<code>inherits(x, 'classname')</code>
Determine class of any object	<code>class(x)</code>

## Base Type (C Structure)

R base types - the internal C-level types that underlie the above OO systems.

- Includes** : atomic vectors, list, functions, environments, etc.
- Useful operation** : Determine if an object is a base type (Not S3, S4 or RC) `is.object(x)` returns FALSE

- Internal representation** : C structure (or struct) that includes :

- Contents of the object
- Memory Management Information
- Type
  - Access with: `typeof()`

# Functions

## Function Basics

**Functions** – objects in their own right

All R functions have three parts:

body()	code inside the function
formals()	list of arguments which controls how you can call the function
environment()	“map” of the location of the function’s variables (see “Enclosing Environment”)

Every operation is a function call

- +, for, if, [, \$, { ...
- x + y is the same as `+`(x, y)

**Note:** the backtick (`), lets you refer to functions or variables that have otherwise reserved or illegal names.

## Lexical Scoping

### What is Lexical Scoping?

- Looks up value of a symbol. (see “Enclosing Environment”)
- **findGlobals()** - lists all the external dependencies of a function

```
f <- function() x + 1
codetools::findGlobals(f)
> '+' 'x'

environment(f) <- emptyenv()
f()

# error in f(): could not find function "+"
```

- R relies on lexical scoping to find everything, even the + operator.

## Function Arguments

**Arguments** – passed by reference and copied on modify

1. Arguments are matched first by exact name (perfect matching), then by prefix matching, and finally by position.
2. Check if an argument was supplied : **missing()**

```
i <- function(a, b) {
  missing(a) -> # return true or false
}
```

3. Lazy evaluation – since x is not used **stop("This is an error!")** never get evaluated.

```
f <- function(x) {
  10
}
f(stop('This is an error!')) -> 10
```

4. Force evaluation

```
f <- function(x) {
  force(x)
  10
}
```

5. Default arguments evaluation

```
f <- function(x = ls()) {
  a <- 1
  x
}
```

f() -> 'a' 'x'	ls() evaluated inside f
f(ls())	ls() evaluated in global environment

## Return Values

- **Last expression evaluated or explicit return()**. Only use explicit return() when returning early.
- **Return ONLY single object**. Workaround is to return a list containing any number of objects.
- **Invisible return object value** - not printed out by default when you call the function.

```
f1 <- function() invisible(1)
```

## Primitive Functions

### What are Primitive Functions?

1. Call C code directly with **.Primitive()** and contain no R code

```
print(sum) :
> function (... , na.rm = FALSE) .Primitive('sum')
```

2. **formals()**, **body()**, and **environment()** are all NULL

3. Only found in base package

4. More efficient since they operate at a low level

## Influx Functions

### What are Influx Functions?

1. Function name comes in between its arguments, like + or -
2. All user-created infix functions must start and end with %.

```
'%+%' <- function(a, b) paste0(a, b)
'new' %+%'string'
```

3. Useful way of providing a default value in case the output of another function is NULL:

```
'%||%' <- function(a, b) if (!is.null(a)) a else b
function_that_might_return_null() %||% default value
```

## Replacement Functions

### What are Replacement Functions?

1. Act like they modify their arguments in place, and have the special name xxx <-
2. Actually create a modified copy. Can use **pryr::address()** to find the memory address of the underlying object

```
second<- <- function(x, value) {
  x[2] <- value
  x
}
x <- 1:10
second(x) <- 5L
```

# Subsetting

Subsetting returns a copy of the original data, NOT copy-on modified

## Simplifying vs. Preserving Subsetting

### 1. Simplifying subsetting

- Returns the **simplest** possible data structure that can represent the output

### 2. Preserving subsetting

- Keeps the structure of the output the **same** as the input.
- When you use drop = FALSE, it's preserving

	Simplifying*	Preserving
Vector	x[[1]]	x[1]
List	x[[1]]	x[1]
Factor	x[1:4, drop = T]	x[1:4]
Array	x[1, ] or x[, 1]	x[1, , drop = F] or x[, 1, drop = F]
Data frame	x[, 1] or x[[1]]	x[, 1, drop = F] or x[1]

Simplifying behavior varies slightly between different data types:

### 1. Atomic Vector

- x[[1]] is the same as x[1]

### 2. List

- [ ] always returns a list
- Use [[ ]] to get list contents, this returns a single value piece out of a list

### 3. Factor

- Drops any unused levels but it remains a factor class

### 4. Matrix or Array

- If any of the dimensions has length 1, that dimension is dropped

### 5. Data Frame

- If output is a single column, it returns a vector instead of a data frame

## Data Frame Subsetting

**Data Frame** – possesses the **characteristics of both lists and matrices**. If you subset with a single vector, they behave like lists; if you subset with two vectors, they behave like matrices

### 1. Subset with a single vector : Behave like lists

```
df1[c('col1', 'col2')]
```

### 2. Subset with two vectors : Behave like matrices

```
df1[, c('col1', 'col2')]
```

The results are the same in the above examples, however, results are different if subsetting with only one column. (see below)

### 1. Behave like matrices

```
str(df1[, 'col1']) -> int [1:3]
```

- Result: the result is a vector

### 2. Behave like lists

```
str(df1['col1']) -> 'data.frame'
```

- Result: the result remains a data frame of 1 column

## \$ Subsetting Operator

### 1. About Subsetting Operator

- Useful shorthand for [[ combined with character subsetting

```
x$y is equivalent to x[['y', exact = FALSE]]
```

### 2. Difference vs. [[

- \$ does partial matching, [[ does not

```
x <- list(abc = 1)
x$a -> 1      # since "exact = FALSE"
x[['a']] ->   # would be an error
```

### 3. Common mistake with \$

- Using it when you have the name of a column stored in a variable

```
var <- 'cyl'
x$var
# doesn't work, translated to x[['var']]
# Instead use x[[var]]
```

## Examples

### 1. Lookup tables (character subsetting)

```
x <- c('m', 'f', 'u', 'f', 'f', 'm', 'm')
lookup <- c(m = 'Male', f = 'Female', u = NA)
lookup[x]
> m f u f f m m
> 'Male' 'Female' NA 'Female' 'Female' 'Male' 'Male'
unname(lookup[x])
> 'Male' 'Female' NA 'Female' 'Female' 'Male' 'Male'
```

### 2. Matching and merging by hand (integer subsetting)

Lookup table which has multiple columns of information:

```
grades <- c(1, 2, 2, 3, 1)
info <- data.frame(
  grade = 3:1,
  desc = c('Excellent', 'Good', 'Poor'),
  fail = c(F, F, T)
)
```

First Method

```
id <- match(grades, info$grade)
info[id, ]
```

Second Method

```
rownames(info) <- info$grade
info[as.character(grades), ]
```

### 3. Expanding aggregated counts (integer subsetting)

- Problem:** a data frame where identical rows have been collapsed into one and a count column has been added
- Solution:** rep() and integer subsetting make it easy to uncollapse the data by subsetting with a repeated row index:  
rep(x, y) rep replicates the values in x, y times.

```
df1$countCol is c(3, 5, 1)
rep(1:nrow(df1), df1$countCol)
> 1 1 1 2 2 2 2 2 3
```

### 4. Removing columns from data frames (character subsetting)

There are two ways to remove columns from a data frame:

Set individual columns to NULL	df1\$col3 <- NULL
Subset to return only columns you want	df1[c('col1', 'col2')]

### 5. Selecting rows based on a condition (logical subsetting)

- This is the most commonly used technique for extracting rows out of a data frame.

```
df1[df1$col1 == 5 & df1$col2 == 4, ]
```

## Subsetting continued

### Boolean Algebra vs. Sets (Logical and Integer Subsetting)

1. **Using integer subsetting** is more effective when:

- You want to find the first (or last) TRUE.
- You have very few TRUEs and very many FALSEs; a set representation may be faster and require less storage.

2. **which()** - conversion from boolean representation to integer representation

```
which(c(T, F, T F)) -> 1 3
```

- Integer representation length : is always <= boolean representation length
- Common mistakes :
  - I. Use **x[which(y)]** instead of **x[y]**
  - II. **x[-which(y)]** is not equivalent to **x[!y]**

#### Recommendation:

Avoid switching from logical to integer subsetting unless you want, for example, the first or last TRUE value

## Subsetting with Assignment

1. All subsetting operators can be combined with assignment to modify selected values of the input vector.

```
df1$col1[df1$col1 < 8] <- 0
```

2. Subsetting with nothing in conjunction with assignment :

- Why : Preserve original object class and structure

```
df1[] <- lapply(df1, as.integer)
```

## Debugging, Condition Handling and Defensive Programming

### Debugging Methods

#### 1. traceback() or RStudio's error inspector

- Lists the sequence of calls that lead to the error

#### 2. browser() or RStudio's breakpoints tool

- Opens an interactive debug session at an arbitrary location in the code

#### 3. options(error = browser) or RStudio's "Rerun with Debug" tool

- Opens an interactive debug session where the error occurred

#### Error Options:

##### options(error = recover)

- Difference vs. 'browser': can enter environment of any of the calls in the stack

##### options(error = dump\_and\_quit)

- Equivalent to 'recover' for non-interactive mode
- Creates **last.dump.rda** in the current working directory

In batch R process :

```
dump_and_quit <- function() {  
  # Save debugging info to file  
  last.dump.rda  
  dump.frames(to.file = TRUE)  
  # Quit R with error status  
  q(status = 1)  
}  
  
options(error = dump_and_quit)
```

In a later interactive session :

```
load("last.dump.rda")  
debugger()
```

### Condition Handling of Expected Errors

#### 1. Communicating potential problems to users:

##### I. stop()

- Action : raise fatal error and force all execution to terminate
- Example usage : when there is no way for a function to continue

##### II. warning()

- Action : generate warnings to display potential problems
- Example usage : when some of elements of a vectorized input are invalid

##### III. message()

- Action : generate messages to give informative output
- Example usage : when you would like to print the steps of a program execution

#### 2. Handling conditions programmatically:

##### I. try()

- Action : gives you the ability to continue execution even when an error occurs

##### II. tryCatch()

- Action : lets you specify handler functions that control what happens when a condition is signaled

```
result = tryCatch(code,  
  error = function(c) "error",  
  warning = function(c) "warning",  
  message = function(c) "message"  
)
```

Use conditionMessage(c) or c\$message to extract the message associated with the original error.

## Defensive Programming

**Basic principle** : "fail fast", to raise an error as soon as something goes wrong

1. **stopifnot()** or use 'assertthat' package - check inputs are correct

2. **Avoid subset(), transform() and with()** - these are non-standard evaluation, when they fail, often fail with uninformative error messages.

3. **Avoid [ and sapply()** - functions that can return different types of output.

- Recommendation : Whenever subsetting a data frame in a function, you should always use **drop = FALSE**