

Introducción a R

2 parte



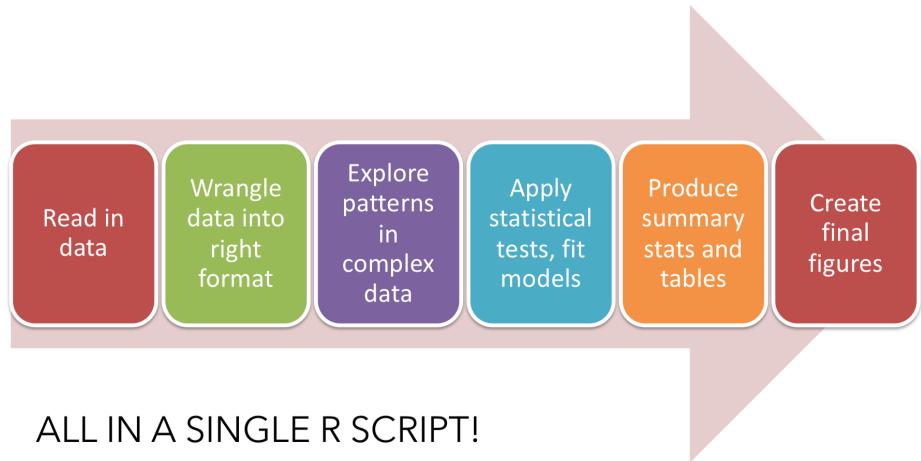
Paulo Izquierdo
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Revisión

Intro a R

- Qué es R?
- Asignar variables
- Vectorizar operaciones
- Funciones básicas
- Bases de datos integrados y otras

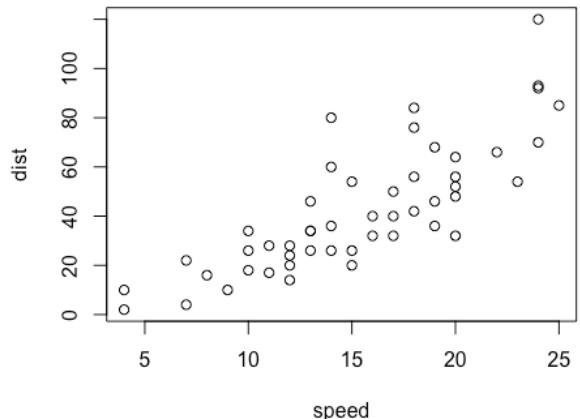


```
mi nombre <- "paulo"
```

```
Error: unexpected symbol in "mi  
nombre"
```

```
> mean(x)
```

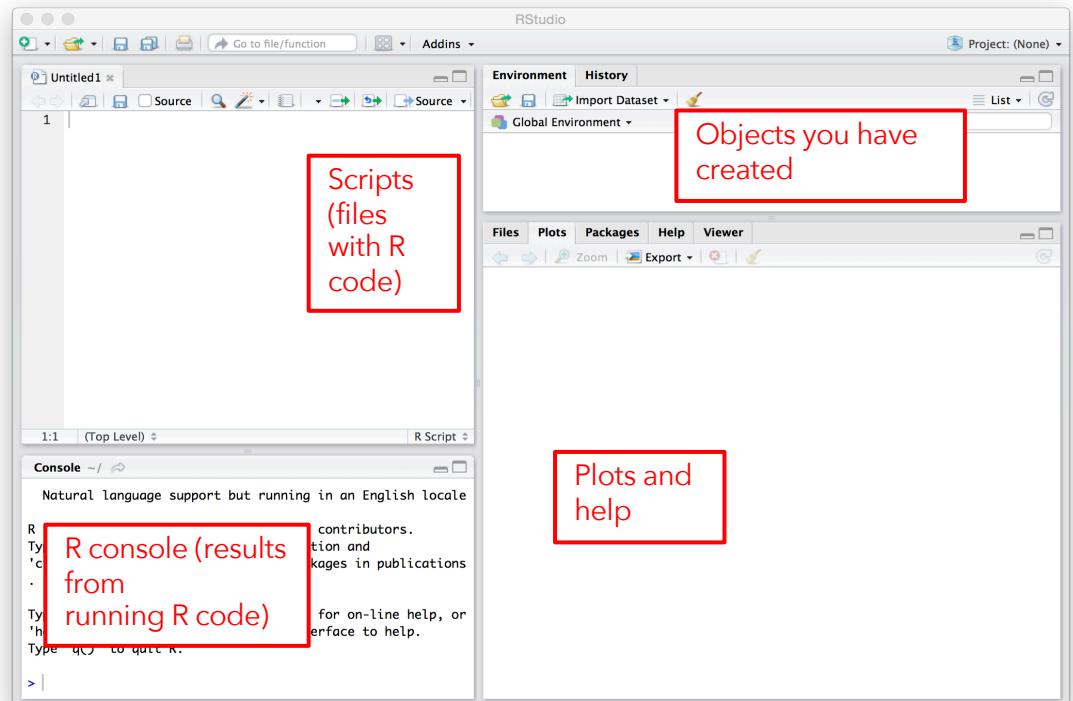
```
> plot(cars)
```



Revisión

Objetos y estructuras

- RStudio!
- Cómo cargar datos en R
- Cuáles son las estructuras de datos más comunes en R
- Cómo crear vectores
- Cómo verificar y cambiar el tipo de datos
- Cómo unir vectores en matrices, dataframes, lists



Revisión

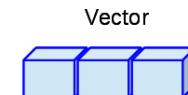
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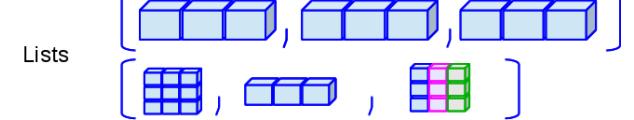
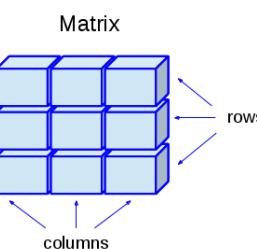
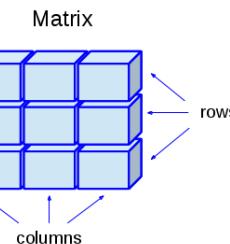
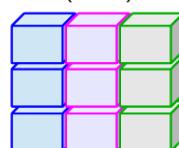


```
getwd ()  
setwd ("~ / Desktop")  
x <- read.csv(file="FileName.csv")
```

```
test <- c("a","b","c")
```



Data Frame
(Table)



Revisión

Manipular datos

- Como indexar diferentes tipos de objetos
- Usar lógica para indexar
- Remover NA para los análisis.
- Cómo ordenar, transponer, eliminar duplicados, etc.



```
> x[1] #primer elemento de un vector
```

```
> x[ ,5] #quinto elemento de una matriz
```

```
> x[[2]] #segundo objeto de una lista
```

```
> x[[2]][1] # primer elemento del segundo objeto de una lista
```

Revisión

Manipular datos

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```
a <- c(2,3,4,)  
[1] 2 3 4
```

```
x <- c(63.33, NA, 64.6, 68.38,  
NA, 79.1, 77.46)
```

```
b <- a>3  
[1] FALSE FALSE TRUE
```

```
mean(x, na.rm=T)  
[1] 70.58
```

Revisión

```
head(iris)
```

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
1	5.1	3.5	1.4	0.2	setosa
2	4.9	3.0	1.4	0.2	setosa
3	4.7	3.2	1.3	0.2	setosa
4	4.6	3.1	1.5	0.2	setosa
5	5.0	3.6	1.4	0.2	setosa
6	5.4	3.9	1.7	0.4	setosa

```
x <- iris[iris$Species=="setosa" &  
          iris$Petal.Length > 1.5, ]
```

```
mean(x$Petal.Width)
```

```
[1] 0.3076923
```

```
mean(iris[iris$Species=="setosa" &  
          iris$Petal.Length > 1.5,  
          Petal.Width])
```

```
[1] 0.3076923
```

Revisión

```
> myDataFrame <-  
  data.frame(a=c(11,13,12,15,17,20),  
             b=c(8,NA, 6, 4,NA,15))  
  
> subset(x=myDataFrame, subset=b>5) #filas  
   a   b  
1 11  8  
3 12  6  
6 20 15  
  
> subset(x=myDataFrame, subset=b>7,  
         select=a) #columnas  
  a  
1 11  
6 20
```

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 Libraries

`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			Programming			While Loop		
Creating Vectors			For Loop			While Loop		
<code>c(2, 4, 6)</code>	2 4 6	Join elements into a vector	<code>for (variable in sequence){</code>	Do something	}	<code>while (condition){</code>	Do something	}
<code>2:6</code>	2 3 4 5 6	An integer sequence						
<code>seq(2, 3, by=0.5)</code>	2.0 2.5 3.0	A complex sequence						
<code>rep(1:2, times=3)</code>	1 2 1 2 1 2	Repeat a vector	<code>for (i in 1:4){</code>	j <- i + 10	print(j)	<code>while (i < 5){</code>	print(i)	i <- i + 1
<code>rep(1:2, each=3)</code>	1 1 1 2 2 2	Repeat elements of a vector						
Vector Functions								
<code>sort(x)</code>	<code>rev(x)</code>	Return x sorted.	<code>if (condition){</code>	Do something		<code>functions_name <- function(var){</code>	Do something	
<code>table(x)</code>	<code>unique(x)</code>	See counts of values.	<code>} else {</code>	Do something different		<code>return(new_variable)</code>		
		See unique values.						
Selecting Vector Elements								
By Position			If Statements			Functions		
<code>x[4]</code>	The fourth element.		<code>if (i > 3){</code>	print('Yes')		<code>function_name <- function(var){</code>	Do something	
<code>x[-4]</code>	All but the fourth.		<code>} else {</code>	print('No')		<code>return(new_variable)</code>		
<code>x[2:4]</code>	Elements two to four.							
<code>x[!(2:4)]</code>	All elements except two to four.							
<code>x[c(1, 5)]</code>	Elements one and five.		Example			Example		
By Value			<code>if (i > 3){</code>	print('Yes')		<code>square <- function(x){</code>	squared <- x*x	
<code>x[x == 10]</code>	Elements which are equal to 10.		<code>} else {</code>	print('No')		<code>return(squared)</code>		
<code>x[x < 0]</code>	All elements less than zero.							
<code>x[x %in% c(1, 2, 5)]</code>	Elements in the set 1, 2, 5.		Reading and Writing Data			Reading and Writing Data		
Named Vectors			<code>df <- read.table('file.txt')</code>	<code>write.table(df, 'file.txt')</code>	Description	<code>Input</code>	<code>Output</code>	
<code>x['apple']</code>	Element with name 'apple'.				Read and write a delimited text file.			
Conditions			<code>df <- read.csv('file.csv')</code>	<code>write.csv(df, 'file.csv')</code>	Read and write a comma separated value file. This is a special case of read.table/write.table.			
			<code>load('file.RData')</code>	<code>save(df, file = 'file.Rdata')</code>	Read and write an R data file, a file type special for R.			
Conditions			<code>a == b</code>	Are equal	<code>a > b</code>	Greater than	<code>a >= b</code>	Greater than or equal to
			<code>a != b</code>	Not equal	<code>a < b</code>	Less than	<code>a <= b</code>	<code>is.na(a)</code>
								Is missing
								<code>is.null(a)</code>
								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.

Matrixes

`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: $m^{-1}x = n$

Lists

`l <- list(x = 1:5, y = c('a', 'b'))`
A list is 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 library](#).

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	df\$x	df[[2]]
1	a		
2	b		
3	c		

Understanding a data frame

`View(df)` See the full data frame.

`head(df)` See the first 6 rows.

Matrix subsetting

<code>df[, 2]</code>		<code>nrow(df)</code> Number of rows.	<code>cbind</code> - Bind columns.
<code>df[2,]</code>		<code>ncol(df)</code> Number of columns.	<code>rbind</code> - Bind rows.
<code>df[2, 2]</code>		<code>dim(df)</code> Number of columns and rows.	

Strings

Also see the [stringr library](#).

`paste(x, y, sep = ' ')`

Join multiple vectors together.

`paste(x, collapse = ' ')`

Join elements of a vector together.

`grep(pattern, x)`

Find regular expression matches in x.

`gsub(pattern, replace, x)`

Replace matches in x with a string.

`toupper(x)`

Convert to uppercase.

`tolower(x)`

Convert to lowercase.

`nchar(x)`

Number of characters in a string.

Factors

`factor(x)`

Turn a vector into a factor. Can set the levels of the factor and the order.

`cut(x, breaks = 4)`

Turn a numeric vector into a factor but 'cutting' into sections.

Statistics

`lm(x ~ y, data=df)`

Linear model.

`glm(x ~ y, data=df)`

Generalised linear model.

`summary`

Get more detailed information out a model.

`prop.test`

Test for a difference between proportions.

`pairwise.t.test`

Perform a t-test for paired data.

`aov`
Analysis of variance.

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

Also see the [ggplot2 library](#).

`plot(x)`

Values of x in order.

`plot(x, y)`

Values of x against y.

`hist(x)`

Histogram of x.

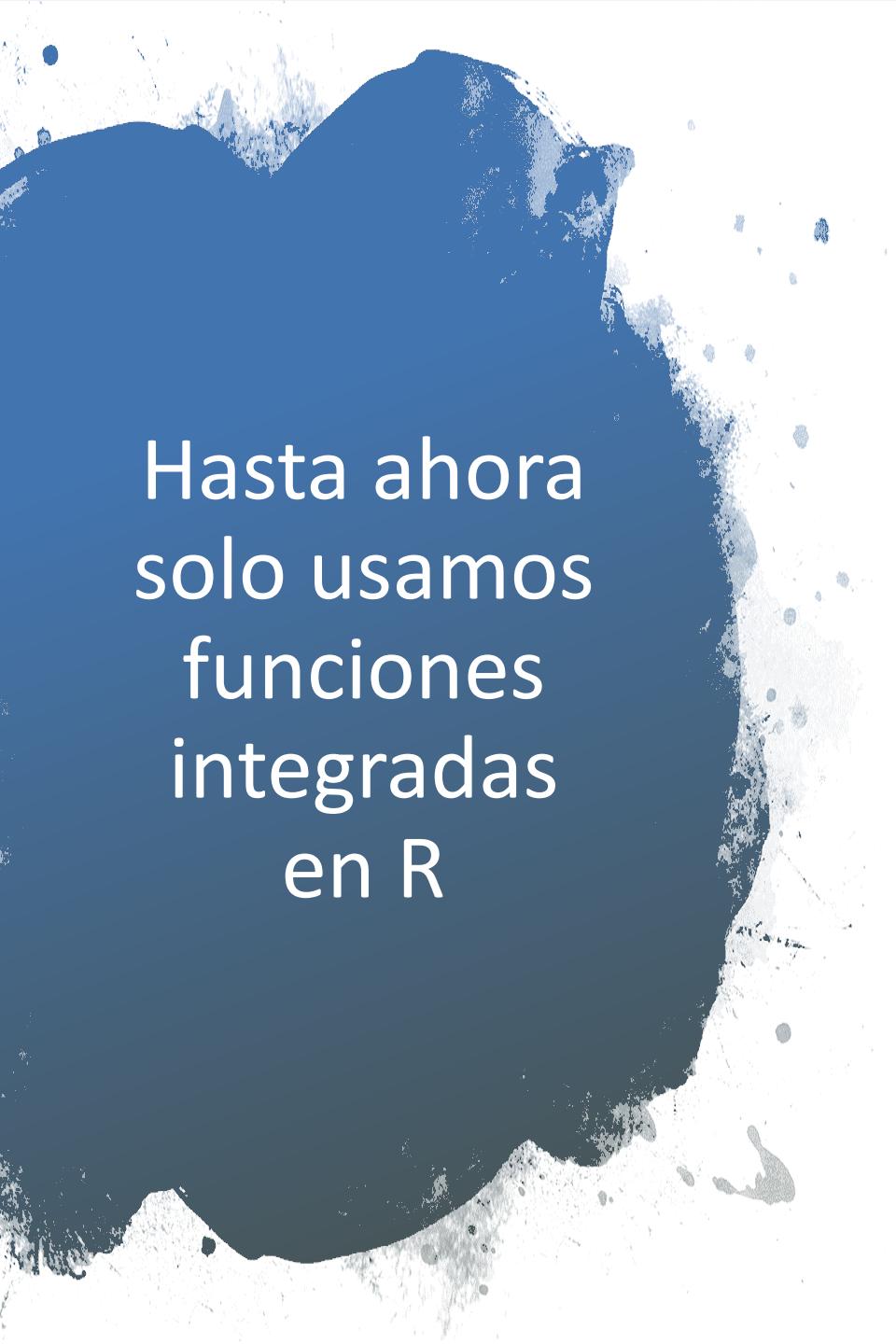
Dates

See the [lubridate library](#).

Ahora vamos a usar : dplyr

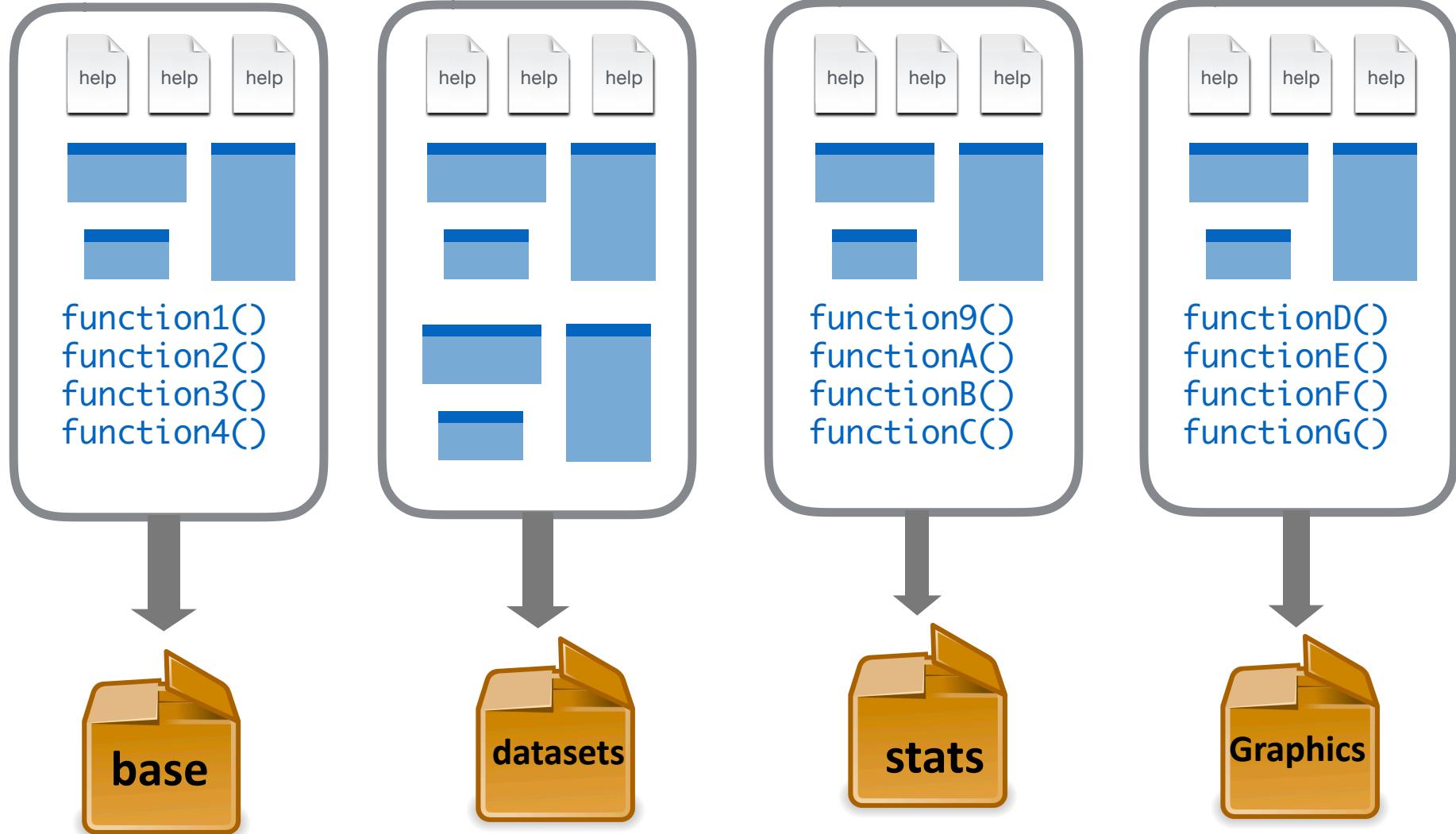
- Instalación de paquetes externos
- Organizar datos con dplyr
- Otros paquetes..





Hasta ahora
solo usamos
funciones
integradas
en R

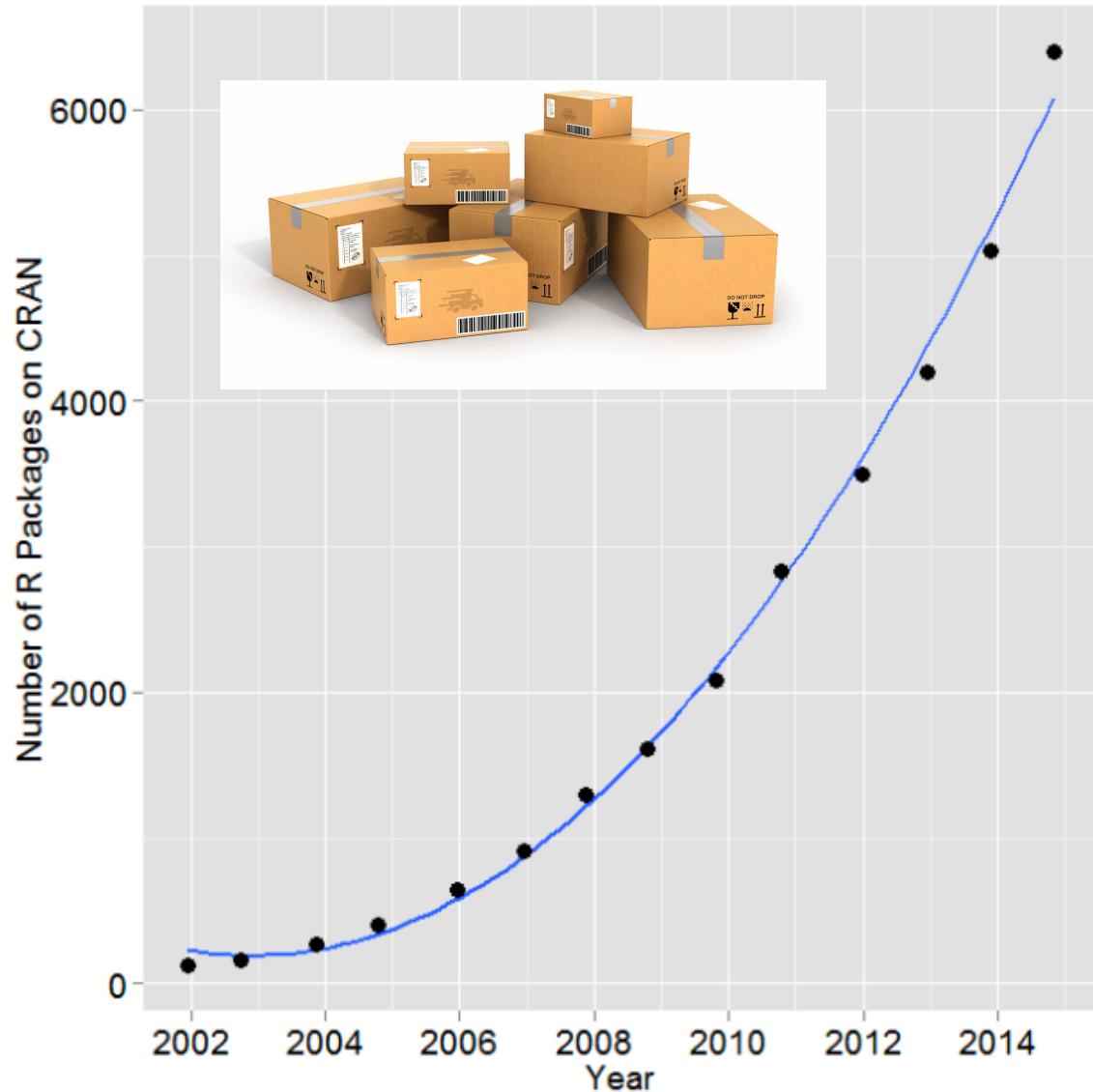
Ya tienes algunos paquetes R cargados automáticamente



Miles de paquetes para explorar

`dim(available.packages()) [1]`

para ver el numero de paquetes disponibles





dplyr

- Paquete con herramientas para fácil manipulación de datos.
- Creado para data frames
- Rápido (usa C++)

The tidyverse

Components



The tidyverse is a collection of R packages that share common philosophies and are designed to work together. This site is a work-in-progress guide to the tidyverse and its packages.

Usar paquetes en R

1

```
install.packages("pkname")
```

Descargar el paquete

1 x por computador

2

```
library("pkname")
```

Cargar el paquete

1 x por sesión en R

```
install.packages("dplyr")
```



Ahora debemos cargarlo

- Cree un nuevo script en R studio

```
# nombre y fecha en la  
parte superior
```

- Limpia tu directorio de trabajo

```
rm (list = ls ())
```

- Cargue su nuevo paquete:

```
library (dplyr)
```



Usen "**vignette**" para leer la guía del paquete
`vignette("dplyr")`



Introduction to dplyr

When working with data you must:

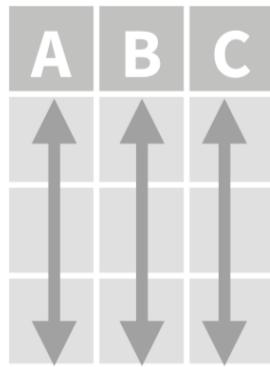
- Figure out what you want to do.
- Describe those tasks in the form of a computer program.
- Execute the program.

The dplyr package makes these steps fast and easy:

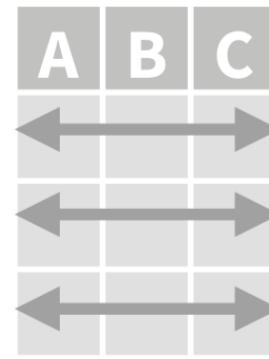
- By constraining your options, it helps you think about your data manipulation

Datos ordenados

dplyr asume que los datos están ordenados



&

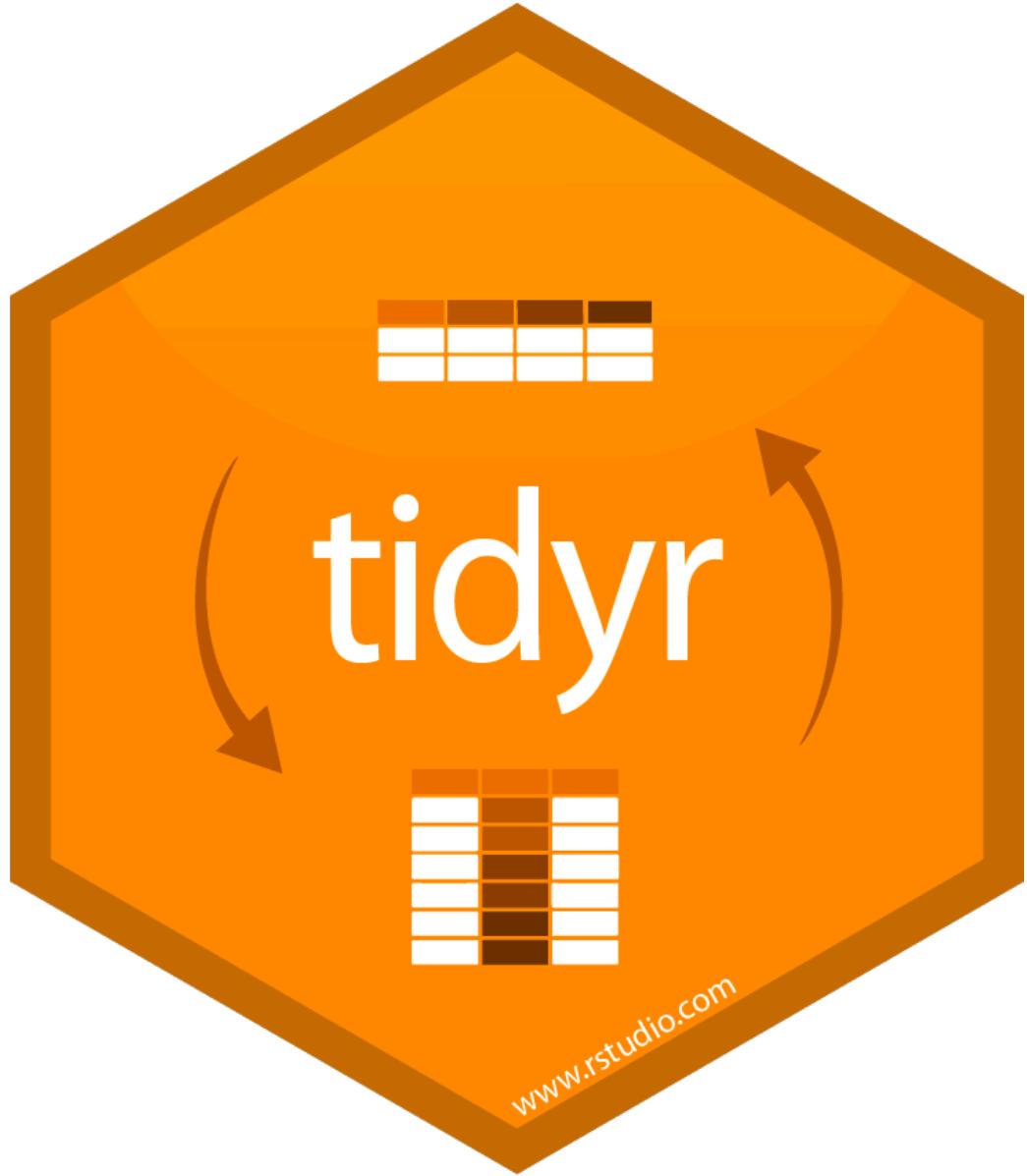


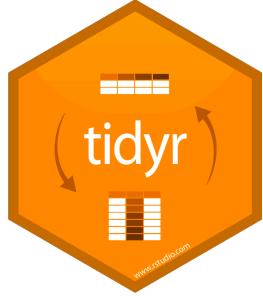
Each **variable** is in
its own **column**

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

Hay un
paquete
adicional
para ordenar
sus datos

```
install.packages("tidyr")  
library("tidyr")
```





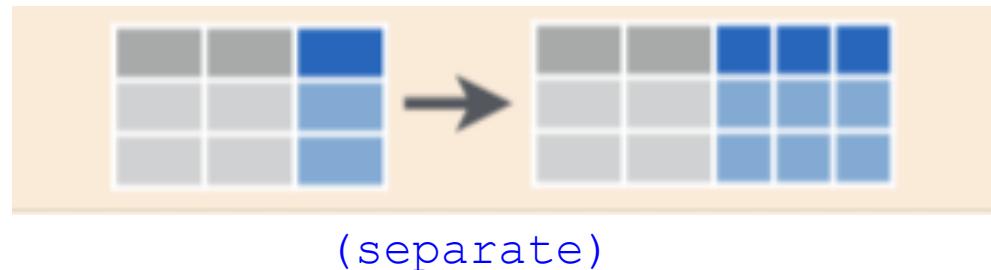
Tiene funciones como:

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

Separar columnas

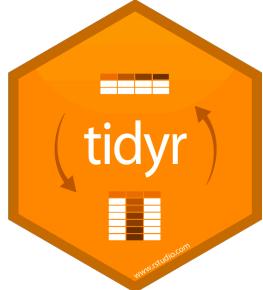
```
df <- data.frame(x = c(11, 21, 13, 41))
```

```
  x  
1 11  
2 21  
3 13  
4 41
```



```
separate(df, x, c("A", "B"), sep=1)
```

```
  A B  
1 1 1  
2 2 1  
3 1 3  
4 4 1
```

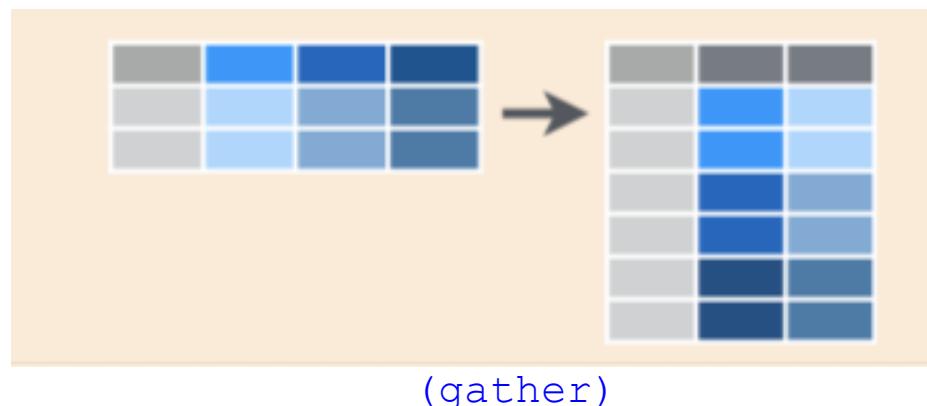


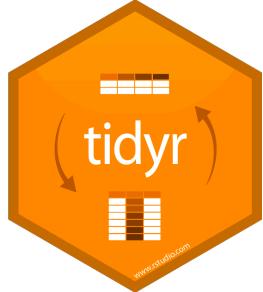
Tiene funciones como:

Cambiar los datos de formato "ancho" a "largo"

```
stocks <- data.frame(  
  time = as.Date('2009-01-01') + 0:9,  
  X = rnorm(10, 0, 1),  
  Y = rnorm(10, 0, 2),  
  Z = rnorm(10, 0, 4)  
)
```

```
> stocks  
# A tibble: 10 x 4  
  time           X     Y     Z  
  <date>     <dbl>   <dbl>   <dbl>  
1 2009-01-01 -1.54  -1.60  0.991  
2 2009-01-02 -1.57  0.737 -4.55  
3 2009-01-03  1.60  -0.326 -1.47  
4 2009-01-04  0.122 -4.91  -4.48  
5 2009-01-05 -1.17   2.31   1.84  
6 2009-01-06  1.19   0.244 -6.20  
7 2009-01-07 -2.28  -0.250 -3.56  
8 2009-01-08  1.12   0.194 -3.25  
9 2009-01-09  0.378 -3.03   7.09  
10 2009-01-10  0.515 -0.334 -7.17
```





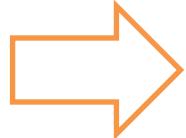
Tiene funciones como:

Cambiar los datos de formato "ancho" a "largo"

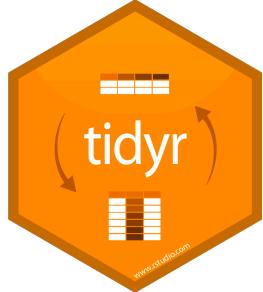


```
gather(stocks, "stock", "price", -time)
```

```
> stocks
# A tibble: 10 x 4
  time          X     Y     Z
  <date>    <dbl> <dbl> <dbl>
1 2009-01-01 -1.54 -1.60  0.991
2 2009-01-02 -1.57  0.737 -4.55
3 2009-01-03  1.60 -0.326 -1.47
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9 2009-01-09  0.378 -3.03   7.09
10 2009-01-10  0.515 -0.334 -7.17
```



```
> gather(stocks, "stock", "price", -time)
# A tibble: 30 x 3
  time      stock  price
  <date>    <chr> <dbl>
1 2009-01-01 X     -1.54
2 2009-01-02 X     -1.57
3 2009-01-03 X      1.60
4 2009-01-04 X      0.122
5 2009-01-05 X     -1.17
6 2009-01-06 X      1.19
7 2009-01-07 X     -2.28
8 2009-01-08 X      1.12
9 2009-01-09 X      0.378
10 2009-01-10 X      0.515
# ... with 20 more rows
```



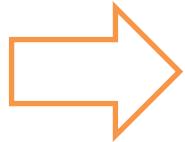
Tiene funciones como:

Cambiar los datos de formato "ancho" a "largo"



```
gather(stocks, "stock", "price", -time)
```

```
> stocks
# A tibble: 10 x 4
  time       X     Y     Z
  <date>   <dbl> <dbl> <dbl>
1 2009-01-01 -1.54 -1.60  0.991
2 2009-01-02 -1.57  0.737 -4.55
3 2009-01-03  1.60 -0.326 -1.47
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```



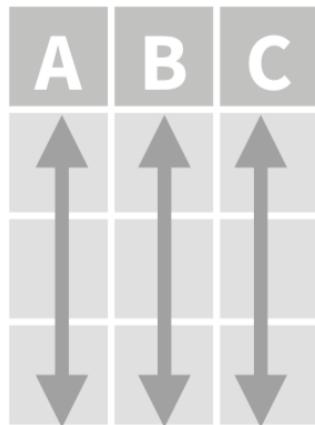
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# A tibble: 30 x 3
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  <date>   <chr>   <dbl>
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2 2009-01-02 X      -1.57
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# ... with 20 more rows
```

```
pivot_longer(stocks, -time, "stock", "price") #nueva función
```

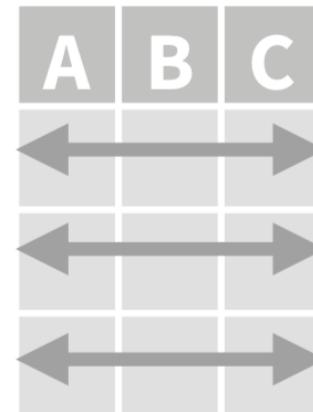
Supongamos que nuestros datos estan ordenados..



iris es un buen ejemplo



&



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

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



Tibbles

dplyr usa data frames, pero los convierte a “tibbles”

“Es parecido, pero se ve mejor en la consola”

Source: local data frame [53,940 x 10]

```
carat      cut color clarity depth table
1  0.23     Ideal   E    SI2  61.5    55
2  0.21   Premium   E    SI1  59.8    61
3  0.23      Good   E    VS1  56.9    65
4  0.29   Premium   I    VS2  62.4    58
5  0.31      Good   J    SI2  63.3    58
6  0.24 Very Good   J    VVS2 62.8    57
7  0.24 Very Good   I    VVS1 62.3    57
8  0.26 Very Good   H    SI1  61.9    55
9  0.22       Fair   E    VS2  65.1    61
10 0.23 Very Good   H    VS1  59.4    61
...
Variables not shown: price (int), x (dbl), y (dbl), z (dbl)
```

tbl

```
967  55.0  2892  6.12  6.14  3.79
968  59.0  2892  6.62  6.55  4.29
969  55.0  2893  5.66  5.71  3.56
970  56.0  2893  5.99  6.04  3.76
971  57.0  2893  5.73  5.75  3.53
972  57.0  2893  5.78  5.83  3.63
973  60.0  2893  5.87  5.78  3.45
974  55.0  2893  5.89  5.92  3.69
975  62.0  2893  6.02  6.04  3.61
976  55.0  2893  6.00  5.93  3.78
977  59.0  2893  6.09  6.06  3.64
978  57.0  2894  5.91  5.99  3.71
979  57.0  2894  5.96  6.00  3.72
980  56.0  2894  5.88  5.92  3.62
981  56.0  2895  5.75  5.78  3.51
982  59.0  2895  5.66  5.76  3.53
983  53.0  2895  5.71  5.75  3.56
984  58.0  2896  5.85  5.89  3.51
985  60.0  2896  5.81  5.91  3.59
986  63.0  2896  6.00  6.05  3.51
987  56.0  2896  5.18  5.24  3.21
988  56.0  2896  5.91  5.96  3.65
989  55.0  2896  5.82  5.86  3.59
990  56.0  2896  5.83  5.89  3.64
991  58.0  2896  5.94  5.88  3.60
992  57.0  2896  6.39  6.35  4.02
993  57.0  2896  6.46  6.45  3.97
994  57.0  2897  5.48  5.51  3.33
995  58.0  2897  5.91  5.85  3.59
996  52.0  2897  5.30  5.34  3.26
997  55.0  2897  5.69  5.74  3.57
998  61.0  2897  5.82  5.89  3.48
999  58.0  2897  5.81  5.77  3.58
1000 59.0  2898  6.68  6.61  4.03
[ reached getOption("max.print") --
omitted 52940 rows ]
```

data.frame

Convertir un data frame a tibble: `as_tibble()`



```
iris_tib <- as_tibble(iris)
```

```
# A tibble: 150 x 5
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species
    <dbl>       <dbl>        <dbl>        <dbl>   <fct>
1       5.1         3.5          1.4         0.2  setosa
2       4.9         3.0          1.4         0.2  setosa
3       4.7         3.2          1.3         0.2  setosa
4       4.6         3.1          1.5         0.2  setosa
5       5.0         3.6          1.4         0.2  setosa
6       5.4         3.9          1.7         0.4  setosa
7       4.6         3.4          1.4         0.3  setosa
8       5.0         3.4          1.5         0.2  setosa
9       4.4         2.9          1.4         0.2  setosa
10      4.9         3.1          1.5         0.1 setosa
# ... with 140 more rows
```



Tiene herramientas para:

Reordenar

Subconjuntos

Resumir

Crear nuevas variables

Combinar

Reordenar

`arrange` ordena las filas por valores de una columna de menor a mayor.

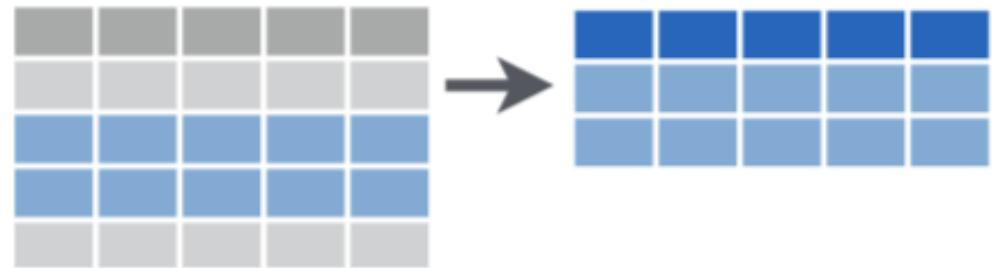
```
arrange(iris, Sepal.Width)
```

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
1	5.0	2.0	3.5	1.0	versicolor
2	6.0	2.2	4.0	1.0	versicolor
3	6.2	2.2	4.5	1.5	versicolor
4	6.0	2.2	5.0	1.5	virginica
5	4.5	2.3	1.3	0.3	setosa
6	5.5	2.3	4.0	1.3	versicolor

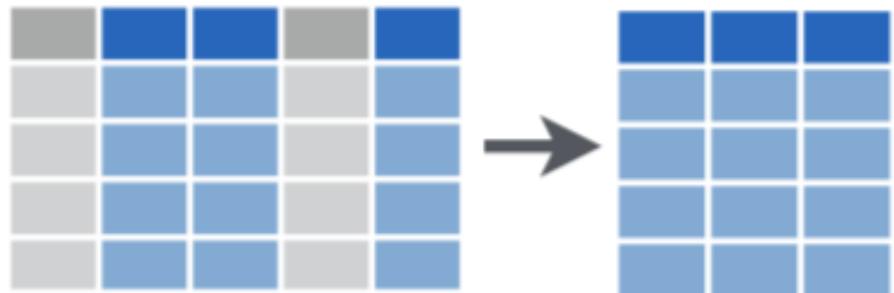
...

Subconjuntos

`filter()` por filas



`select()` por columnas



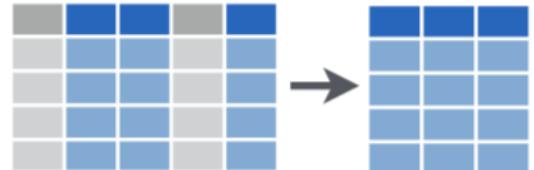
Subconjuntos

Intenten esto:

```
filter(iris, Sepal.Length > 7)
```



```
select(iris, Sepal.Width,  
       Petal.Length, Species)
```



Resumir



Intenten esto: `summarise(iris, mean(Sepal.Length))`

`summarise_all()` lo hace para todas las columnas



Resumir

```
> summarise_all(iris, mean)
Sepal.Length Sepal.Width Petal.Length Petal.Width Species
1 5.843333 3.057333    3.758       1.199333 NA
Warning message:
In mean.default(Species) : argument is not numeric or logical: returning NA
```

```
> str(iris)
data.frame': 150 obs. of 5 variables:
 $ Sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
 $ Sepal.Width : num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
 $ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
 $ Petal.Width : num 0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
 $ Species     : Factor w/ 3 levels "setosa","versicolor",...: 1 1 1 1 1 1 1 1 1 1 ...
```

Resumir

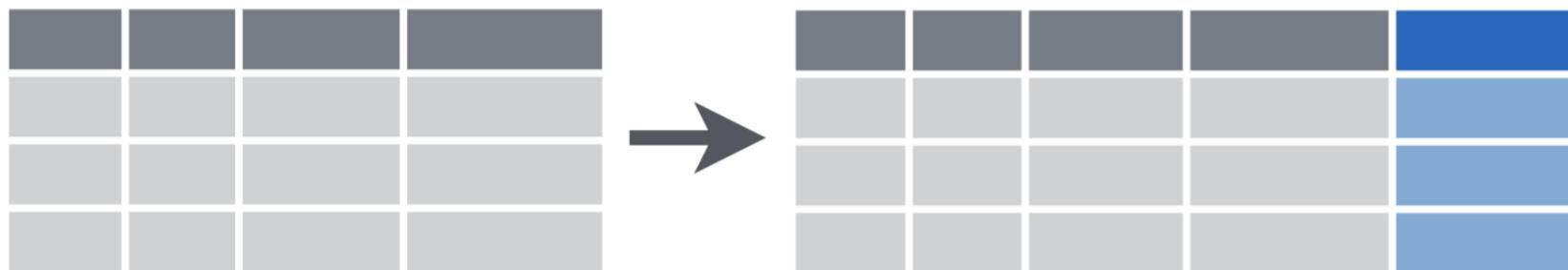


summary
function

- mean ()
- sd ()
- min ()
- max ()
- sum ()

Crear nuevas variables

`mutate()` para crear nuevas columnas



Intenten esto: `mutate(iris, sepal = Sepal.Length + Sepal.Width)`

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species	sepal
1	5.1	3.5	1.4	0.2	setosa	8.6
2	4.9	3.0	1.4	0.2	setosa	7.9
3	4.7	3.2	1.3	0.2	setosa	7.9
4	4.6	3.1	1.5	0.2	setosa	7.7
5	5.0	3.6	1.4	0.2	setosa	8.6
6	5.4	3.9	1.7	0.4	setosa	9.3

Crear nuevas variables

`between()` genera un vector lógico:

Intenten esto:

```
mutate(iris, sepal_m = between(Sepal.Width, 3, 4))
```

**creates new column named Sepal_m of T or F*

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species	sepal_m
1	5.1	3.5	1.4	0.2	setosa	TRUE
2	4.9	3.0	1.4	0.2	setosa	TRUE
3	4.7	3.2	1.3	0.2	setosa	TRUE
4	4.6	3.1	1.5	0.2	setosa	TRUE
5	5.0	3.6	1.4	0.2	setosa	TRUE
6	5.4	3.9	1.7	0.4	setosa	TRUE

Combinar

(columnas)

x1	x2
A	1
B	2
C	3



x1	x3
A	T
B	F
D	T

1

x1	x2	x3
A	1	T
B	2	F
C	3	NA

```
dplyr::left_join(a, b, by = "x1")
```

Join matching rows from b to a.

Combinar

(columnas)

x1	x2
A	1
B	2
C	3



x1	x3
A	T
B	F
D	T



x1	x2	x3
A	1	T
B	2	F
C	3	NA

`dplyr::left_join(a, b, by = "x1")`

Join matching rows from b to a.

x1	x3	x2
A	T	1
B	F	2
D	T	NA

`dplyr::right_join(a, b, by = "x1")`

Prioritize
data in b

Join matching rows from a to b.

Combinar

(columnas)

x1	x2
A	1
B	2
C	3



x1	x3
A	T
B	F
D	T



x1	x2	x3
A	1	T
B	2	F
C	3	NA

`dplyr::left_join(a, b, by = "x1")`

Join matching rows from b to a.

x1	x3	x2
A	T	1
B	F	2
D	T	NA

`dplyr::right_join(a, b, by = "x1")`

Join matching rows from a to b.

x1	x2	x3
A	1	T
B	2	F

`dplyr::inner_join(a, b, by = "x1")`

Join data. Retain only rows in both sets.

Combinar

(columnas)

x1	x2
A	1
B	2
C	3



x1	x3
A	T
B	F
D	T



x1	x2	x3
A	1	T
B	2	F
C	3	NA

`dplyr::left_join(a, b, by = "x1")`

Join matching rows from b to a.

x1	x3	x2
A	T	1
B	F	2
D	T	NA

`dplyr::right_join(a, b, by = "x1")`

Join matching rows from a to b.

x1	x2	x3
A	1	T
B	2	F

`dplyr::inner_join(a, b, by = "x1")`

Join data. Retain only rows in both sets.

x1	x2	x3
A	1	T
B	2	F
C	3	NA
D	NA	T

`dplyr::full_join(a, b, by = "x1")`

Join data. Retain all values, all rows.



Tiene herramientas para:

Reordenar

Subconjuntos

Resumir

Crear nuevas variables

Combinar



Piping

Agrupamiento

$\% > \%$



- Se llama "piping" y hace que su código sea más legible.

$\% > \%$ Pasa el objeto del lado izquierdo para que sea el primer argumento de la función del lado derecho.

```
data %>% function()
```



Intenten esto:

```
summarise(iris, mean(Sepal.Width))
```

```
iris %>% summarise(mean(Sepal.Width))
```



Piping

Agrupamiento



group_by

`dplyr::group_by(iris, Species)`

Group data into rows with the same value of Species.

`iris %>% group_by(Species) %>% summarise(...)`

Compute separate summary row for each group.





Intenten esto:

```
iris %>%
```

```
group_by(Species) %>%
summarise(mean(Sepal.Width))
```

Species	mean(Sepal.Width)
* <fct>	<dbl>
1 setosa	3.43
2 versicolor	2.77
3 virginica	2.97

Data Wrangling with dplyr and tidyverse

Cheat Sheet



Syntax - Helpful conventions for wrangling

`dplyr::tbl_df(iris)`

Converts data to `tbl` class. `tbl`'s are easier to examine than data frames. R displays only the data that fits onscreen:

```
Source: local data frame [150 x 5]
  Sepal.Length Sepal.Width Petal.Length
1          5.1        3.5         1.4
2          4.9        3.0         1.4
3          4.7        3.2         1.3
4          4.6        3.1         1.5
5          5.0        3.6         1.4
..          ...
Variables not shown: Petal.Width (dbl), Species (fctr)
```

`dplyr::glimpse(iris)`

Information dense summary of `tbl` data.

`utils::View(iris)`

View data set in spreadsheet-like display (note capital V).

iris					
		Filter			
Sepal.Length		Sepal.Width		Petal.Length	
1	5.1	3.5	1.4	0.2	setosa
2	4.9	3.0	1.4	0.2	setosa
3	4.7	3.2	1.3	0.2	setosa
4	4.6	3.1	1.5	0.2	setosa
5	5.0	3.6	1.4	0.2	setosa
6	5.4	3.9	1.7	0.4	setosa
7	4.6	3.4	1.4	0.3	setosa
8	5.0	3.4	1.5	0.2	setosa

`dplyr::%>%`

Passes object on left hand side as first argument (or . argument) of function on righthand side.

`x %>% f(y)` is the same as `f(x, y)`

`y %>% f(x, ., z)` is the same as `f(x, y, z)`

"Piping" with `%>%` makes code more readable, e.g.

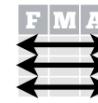
```
iris %>%
  group_by(Species) %>%
  summarise(avg = mean(Sepal.Width)) %>%
  arrange(avg)
```

Tidy Data - A foundation for wrangling in R

In a tidy data set:



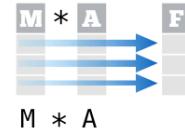
&



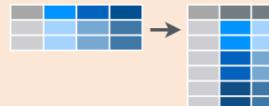
Each **variable** is saved in its own **column**

Each **observation** is saved in its own **row**

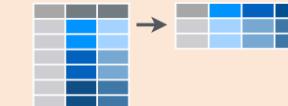
Tidy data complements R's **vectorized operations**. R will automatically preserve observations as you manipulate variables. No other format works as intuitively with R.



Reshaping Data - Change the layout of a data set



`tidyverse::gather(cases, "year", "n", 2:4)`
Gather columns into rows.



`tidyverse::spread(pollution, size, amount)`
Spread rows into columns.

`tidyverse::separate(storms, date, c("y", "m", "d"))`
Separate one column into several.



`tidyverse::unite(data, col, ..., sep)`
Unite several columns into one.

`dplyr::data_frame(a = 1:3, b = 4:6)`
Combine vectors into data frame (optimized).

`dplyr::arrange(mtcars, mpg)`
Order rows by values of a column (low to high).

`dplyr::arrange(mtcars, desc(mpg))`
Order rows by values of a column (high to low).

`dplyr::rename(tb, y = year)`
Rename the columns of a data frame.

Subset Observations (Rows)



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

`dplyr::distinct(iris)`
Remove duplicate rows.

`dplyr::sample_frac(iris, 0.5, replace = TRUE)`
Randomly select fraction of rows.

`dplyr::sample_n(iris, 10, replace = TRUE)`
Randomly select n rows.

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

`dplyr::top_n(storms, 2, date)`
Select and order top n entries (by group if grouped data).

Subset Variables (Columns)



`dplyr::select(iris, Sepal.Width, Petal.Length, Species)`
Select columns by name or helper function.

Helper functions for select - ?select

`select(iris, contains("."))`

Select columns whose name contains a character string.

`select(iris, ends_with("Length"))`

Select columns whose name ends with a character string.

`select(iris, everything())`

Select every column.

`select(iris, matches(".t."))`

Select columns whose name matches a regular expression.

`select(iris, num_range("x1", 1:5))`

Select columns named x1, x2, x3, x4, x5.

`select(iris, one_of(c("Species", "Genus")))`

Select columns whose names are in a group of names.

`select(iris, starts_with("Sepal"))`

Select columns whose name starts with a character string.

`select(iris, Sepal.Length:Petal.Width)`

Select all columns between Sepal.Length and Petal.Width (inclusive).

`select(iris, -Species)`

Select all columns except Species.

Logic in R - Comparison, ?base::Logic

<	Less than	!=	Not equal to
>	Greater than	%in%	Group membership
==	Equal to	is.na	Is NA
<=	Less than or equal to	!is.na	Is not NA
>=	Greater than or equal to	&, , !, xor, any, all	Boolean operators

Summarise Data



`dplyr::summarise(iris, avg = mean(Sepal.Length))`

Summarise data into single row of values.

`dplyr::summarise_each(iris, funs(mean))`

Apply summary function to each column.

`dplyr::count(iris, Species, wt = Sepal.Length)`

Count number of rows with each unique value of variable (with or without weights).



Summarise uses **summary functions**, functions that take a vector of values and return a single value, such as:

`dplyr::first`

First value of a vector.

`dplyr::last`

Last value of a vector.

`dplyr::nth`

Nth value of a vector.

`dplyr::n`

of values in a vector.

`dplyr::n_distinct`

of distinct values in a vector.

`IQR`

IQR of a vector.

`min`

Minimum value in a vector.

`max`

Maximum value in a vector.

`mean`

Mean value of a vector.

`median`

Median value of a vector.

`var`

Variance of a vector.

`sd`

Standard deviation of a vector.

Group Data

`dplyr::group_by(iris, Species)`

Group data into rows with the same value of Species.

`dplyr::ungroup(iris)`

Remove grouping information from data frame.

`iris %>% group_by(Species) %>% summarise(...)`

Compute separate summary row for each group.



Make New Variables



`dplyr::mutate(iris, sepal = Sepal.Length + Sepal.Width)`

Compute and append one or more new columns.

`dplyr::mutate_each(iris, funs(min_rank))`

Apply window function to each column.

`dplyr::transmute(iris, sepal = Sepal.Length + Sepal.Width)`

Compute one or more new columns. Drop original columns.



Mutate uses **window functions**, functions that take a vector of values and return another vector of values, such as:

`dplyr::lead`

Copy with values shifted by 1.

`dplyr::lag`

Copy with values lagged by 1.

`dplyr::dense_rank`

Ranks with no gaps.

`dplyr::min_rank`

Ranks. Ties get min rank.

`dplyr::percent_rank`

Ranks rescaled to [0, 1].

`dplyr::row_number`

Ranks. Ties got to first value.

`dplyr::ntile`

Bin vector into n buckets.

`dplyr::between`

Are values between a and b?

`dplyr::cume_dist`

Cumulative distribution.

`dplyr::cumall`

Cumulative all

`dplyr::cumany`

Cumulative any

`dplyr::cummean`

Cumulative mean

`cumsum`

Cumulative sum

`cummax`

Cumulative max

`cummin`

Cumulative min

`cumprod`

Cumulative prod

`pmax`

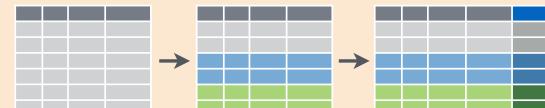
Element-wise max

`pmin`

Element-wise min

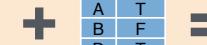
`iris %>% group_by(Species) %>% mutate(...)`

Compute new variables by group.



Combine Data Sets

a	x1	x2	b	x1	x3
A	1	T	B	F	
B	2	F	D	T	
C	3	NA			



Mutating Joins

x1	x2	x3
A	1	T
B	2	F
C	3	NA

x1	x3	x2
A	1	T
B	2	F
D	T	NA

x1	x2	x3
A	1	T
B	2	F
C	3	NA

Filtering Joins

x1	x2
A	1
B	2

x1	x2
C	3

y	x1	x2	z	x1	x2
A	1		B	2	
B	2		C	3	
C	3		D	4	

x1	x2
A	1
B	2
C	3

x1	x2
D	4

x1	x2
A	1
B	2
C	3

x1	x2	x1	x2
A	1	B	2
B	2	C	3
C	3	D	4

x1	x2	x1	x2
A	1	B	2
B	2	C	3
C	3	D	4

x1	x2	x1	x2
A	1	B	2
B	2	C	3
C	3	D	4

Set Operations

`dplyr::intersect(y, z)`
Rows that appear in both y and z.

`dplyr::union(y, z)`
Rows that appear in either or both y and z.

`dplyr::setdiff(y, z)`
Rows that appear in y but not z.

Binding

x1	x2
A	1
B	2
C	3

`dplyr::bind_rows(y, z)`
Append z to y as new rows.

x1	x2
A	1
B	2
C	3

`dplyr::bind_cols(y, z)`
Append z to y as new columns.
Caution: matches rows by position.

Lo que aprendimos!

- Cómo instalar paquetes externos
- Que es tidyverse
- Cómo organizar datos usando dplyr

