

# Data Import :: CHEAT SHEET



R's **tidyverse** is built around **tidy data** stored in **tibbles**, which are enhanced data frames.



The front side of this sheet shows how to read text files into R with **readr**.



The reverse side shows how to create tibbles with **tibble** and to layout tidy data with **tidyr**.

## OTHER TYPES OF DATA

Try one of the following packages to import other types of files

- **haven** - SPSS, Stata, and SAS files
- **readxl** - excel files (.xls and .xlsx)
- **DBI** - databases
- **jsonlite** - json
- **xml2** - XML
- **httr** - Web APIs
- **rvest** - HTML (Web Scraping)

## Save Data

Save **x**, an R object, to **path**, a file path, as:

### Comma delimited file

**write\_csv(x, path, na = "NA", append = FALSE, col\_names = !append)**

### File with arbitrary delimiter

**write\_delim(x, path, delim = " ", na = "NA", append = FALSE, col\_names = !append)**

### CSV for excel

**write\_excel\_csv(x, path, na = "NA", append = FALSE, col\_names = !append)**

### String to file

**write\_file(x, path, append = FALSE)**

### String vector to file, one element per line

**write\_lines(x,path, na = "NA", append = FALSE)**

### Object to RDS file

**write\_rds(x, path, compress = c("none", "gz", "bz2", "xz"), ...)**

### Tab delimited files

**write\_tsv(x, path, na = "NA", append = FALSE, col\_names = !append)**

## Read Tabular Data - These functions share the common arguments:

```
read_*(file, col_names = TRUE, col_types = NULL, locale = default_locale(), na = c("", "NA"),
quoted_na = TRUE, comment = "", trim_ws = TRUE, skip = 0, n_max = Inf, guess_max = min(1000,
n_max), progress = interactive())
```

```
a,b,c
1,2,3
4,5,NA
```

| A | B | C  |
|---|---|----|
| 1 | 2 | 3  |
| 4 | 5 | NA |

### Comma Delimited Files

**read\_csv("file.csv")**

To make file.csv run:

**write\_file(x = "a,b,c\n1,2,3\n4,5,NA", path = "file.csv")**

```
a;b;c
1;2;3
4;5;NA
```

| A | B | C  |
|---|---|----|
| 1 | 2 | 3  |
| 4 | 5 | NA |

### Semi-colon Delimited Files

**read\_csv2("file2.csv")**

**write\_file(x = "a;b;c\n1;2;3\n4;5;NA", path = "file2.csv")**

```
a|b|c
1|2|3
4|5|NA
```

| A | B | C  |
|---|---|----|
| 1 | 2 | 3  |
| 4 | 5 | NA |

### Files with Any Delimiter

**read\_delim("file.txt", delim = "|")**

**write\_file(x = "a|b|c\n1|2|3\n4|5|NA", path = "file.txt")**

```
a b c
1 2 3
4 5 NA
```

| A | B | C  |
|---|---|----|
| 1 | 2 | 3  |
| 4 | 5 | NA |

### Fixed Width Files

**read\_fwf("file.fwf", col\_positions = c(1, 3, 5))**

**write\_file(x = "a b c\n1 2 3\n4 5 NA", path = "file.fwf")**

### Tab Delimited Files

**read\_tsv("file.tsv")** Also **read\_table()**.

**write\_file(x = "a\tb\tc\n1\t2\t3\n4\t5\tNA", path = "file.tsv")**

## USEFUL ARGUMENTS

```
a,b,c
1,2,3
4,5,NA
```

### Example file

**write\_file("a,b,c\n1,2,3\n4,5,NA","file.csv")**  
**f <- "file.csv"**

| 1 | 2 | 3  |
|---|---|----|
| 4 | 5 | NA |

### Skip lines

**read\_csv(f, skip = 1)**

| A | B | C  |
|---|---|----|
| 1 | 2 | 3  |
| 4 | 5 | NA |

### No header

**read\_csv(f, col\_names = FALSE)**

| A | B | C |
|---|---|---|
| 1 | 2 | 3 |

### Read in a subset

**read\_csv(f, n\_max = 1)**

| x | y | z  |
|---|---|----|
| A | B | C  |
| 1 | 2 | 3  |
| 4 | 5 | NA |

### Provide header

**read\_csv(f, col\_names = c("x", "y", "z"))**

| A  | B | C  |
|----|---|----|
| NA | 2 | 3  |
| 4  | 5 | NA |

### Missing Values

**read\_csv(f, na = c("1", "!"))**

## Read Non-Tabular Data

### Read a file into a single string

**read\_file(file, locale = default\_locale())**

### Read each line into its own string

**read\_lines(file, skip = 0, n\_max = -1L, na = character(), locale = default\_locale(), progress = interactive())**

### Read Apache style log files

**read\_log(file, col\_names = FALSE, col\_types = NULL, skip = 0, n\_max = -1, progress = interactive())**

### Read a file into a raw vector

**read\_file\_raw(file)**

### Read each line into a raw vector

**read\_lines\_raw(file, skip = 0, n\_max = -1L, progress = interactive())**

## Data types

readr functions guess the types of each column and convert types when appropriate (but will NOT convert strings to factors automatically).

A message shows the type of each column in the result.

```
## Parsed with column specification:
## cols(
##   age = col_integer(),
##   sex = col_character(),
##   earn = col_double()
## )
```

age is an integer

earn is a double (numeric)

sex is a character

1. Use **problems()** to diagnose problems.

**x <- read\_csv("file.csv"); problems(x)**

2. Use a **col\_** function to guide parsing.

- **col\_guess()** - the default
- **col\_character()**
- **col\_double()**, **col\_euro\_double()**
- **col\_datetime(format = "")** Also **col\_date(format = "")**, **col\_time(format = "")**
- **col\_factor(levels, ordered = FALSE)**
- **col\_integer()**
- **col\_logical()**
- **col\_number()**, **col\_numeric()**
- **col\_skip()**

**x <- read\_csv("file.csv", col\_types = cols(
A = col\_double(),
B = col\_logical(),
C = col\_factor()))**

3. Else, read in as character vectors then parse with a **parse\_** function.

- **parse\_guess()**
- **parse\_character()**
- **parse\_datetime()** Also **parse\_date()** and **parse\_time()**
- **parse\_double()**
- **parse\_factor()**
- **parse\_integer()**
- **parse\_logical()**
- **parse\_number()**

**x\$A <- parse\_number(x\$A)**



# Tibbles - an enhanced data frame



The **tibble** package provides a new S3 class for storing tabular data, the tibble. Tibbles inherit the data frame class, but improve three behaviors:

- **Subsetting** - `[` always returns a new tibble, `[[` and `$` always return a vector.
- **No partial matching** - You must use full column names when subsetting
- **Display** - When you print a tibble, R provides a concise view of the data that fits on one screen

```
# A tibble: 234 × 6
  manufacturer <chr> model <chr> displ <dbl>
1 audi a4 1.8
2 audi a4 2.0
3 audi a4 2.0
4 audi a4 2.8
5 audi a4 3.1
6 audi a4 quattro 1.8
7 audi a4 quattro 1.8
8 ... with 224 more rows, and
9 more variables: year <int>,
10 cyl <int>, trans <chr>
```

**tibble display**

```
156 1999 6 auto(l4)
157 1999 6 auto(l4)
158 2008 6 auto(l4)
159 2008 8 auto(s4)
160 1999 4 manual(m5)
161 1999 4 auto(l4)
162 2008 4 manual(m5)
163 2008 4 manual(m5)
164 2008 4 auto(l4)
165 2008 4 auto(l4)
166 1999 4 auto(l4)
[ reached getOption("max.print")
  -- omitted 68 rows ]
```

**data frame display**

A large table to display

- Control the default appearance with options:
  - `options(tibble.print_max = n, tibble.print_min = m, tibble.width = Inf)`
- View full data set with **View()** or **glimpse()**
- Revert to data frame with **as.data.frame()**

## CONSTRUCT A TIBBLE IN TWO WAYS

**tibble(...)**  
Construct by columns.  
`tibble(x = 1:3, y = c("a", "b", "c"))`

**tribble(...)**  
Construct by rows.  
`tribble(~x, ~y, 1, "a", 2, "b", 3, "c")`

```
A tibble: 3 × 2
  x y
  <int> <chr>
1 1 a
2 2 b
3 3 c
```

Both make this tibble

- **as\_tibble(x, ...)** Convert data frame to tibble.
- **enframe(x, name = "name", value = "value")** Convert named vector to a tibble
- **is\_tibble(x)** Test whether x is a tibble.

# Tidy Data with tidyr

**Tidy data** is a way to organize tabular data. It provides a consistent data structure across packages.

A table is tidy if:



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

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

Tidy data:



Makes variables easy to access as vectors

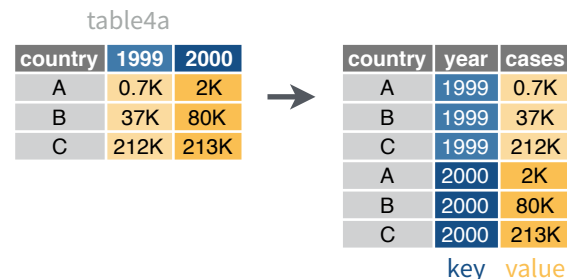
Preserves cases during vectorized operations

## Reshape Data - change the layout of values in a table

Use **gather()** and **spread()** to reorganize the values of a table into a new layout.

**gather(data, key, value, ..., na.rm = FALSE, convert = FALSE, factor\_key = FALSE)**

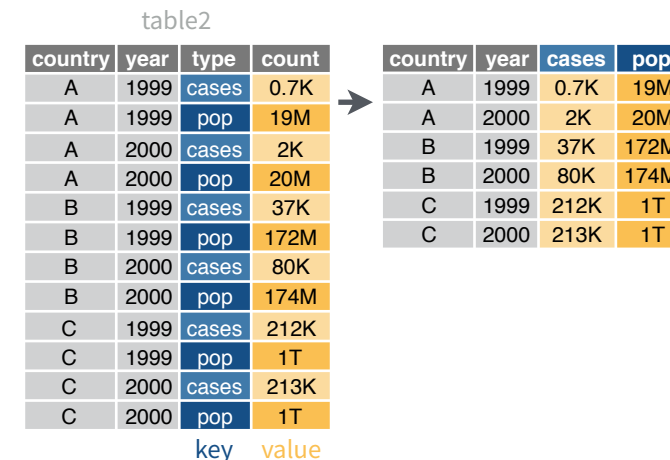
`gather()` moves column names into a **key** column, gathering the column values into a single **value** column.



`gather(table4a, `1999`, `2000`, key = "year", value = "cases")`

**spread(data, key, value, fill = NA, convert = FALSE, drop = TRUE, sep = NULL)**

`spread()` moves the unique values of a **key** column into the column names, spreading the values of a **value** column across the new columns.

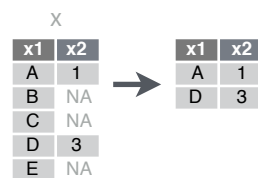


`spread(table2, type, count)`

## Handle Missing Values

**drop\_na(data, ...)**

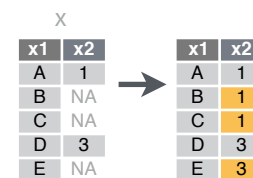
Drop rows containing NA's in ... columns.



`drop_na(x, x2)`

**fill(data, ..., .direction = c("down", "up"))**

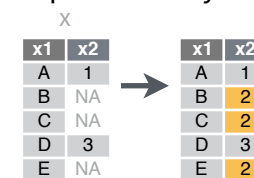
Fill in NA's in ... columns with most recent non-NA values.



`fill(x, x2)`

**replace\_na(data, replace = list(), ...)**

Replace NA's by column.



`replace_na(x, list(x2 = 2))`

## Expand Tables - quickly create tables with combinations of values

**complete(data, ..., fill = list())**

Adds to the data missing combinations of the values of the variables listed in ...  
`complete(mtcars, cyl, gear, carb)`

**expand(data, ...)**

Create new tibble with all possible combinations of the values of the variables listed in ...  
`expand(mtcars, cyl, gear, carb)`

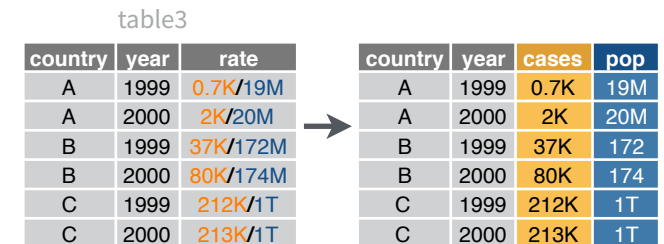
# Split Cells



Use these functions to split or combine cells into individual, isolated values.

**separate(data, col, into, sep = "[^[:alnum:]]+", remove = TRUE, convert = FALSE, extra = "warn", fill = "warn", ...)**

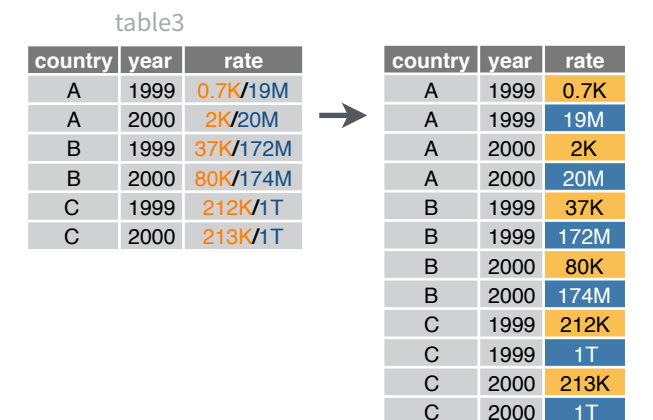
Separate each cell in a column to make several columns.



`separate(table3, rate, into = c("cases", "pop"))`

**separate\_rows(data, ..., sep = "[^[:alnum:]]+", convert = FALSE)**

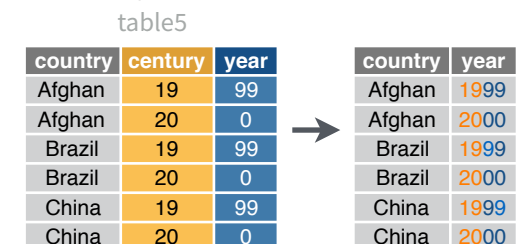
Separate each cell in a column to make several rows. Also **separate\_rows\_()**.



`separate_rows(table3, rate)`

**unite(data, col, ..., sep = "\_", remove = TRUE)**

Collapse cells across several columns to make a single column.



`unite(table5, century, year, col = "year", sep = "")`

