

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")
```

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()  
## )
```

earn is a double (numeric)

age is an integer

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.

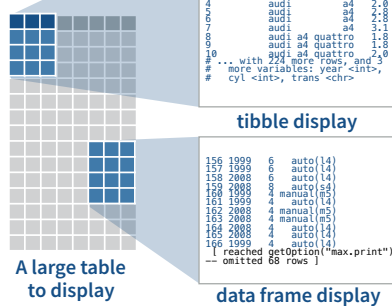
- **parse_guess()**
 - **parse_character()**
 - **parse_datetime()** Also **parse_date()** and **parse_time()**
 - **parse_double()**
 - **parse_factor()**
 - **parse_integer()**
 - **parse_logical()**
 - **parse_number()**
- ```
x$A <- parse_number(x$A)
```



## 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



- 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"))`

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

Both make this tibble

A tibble: 3 x 2

| x | y |
|---|---|
| 1 | a |
| 2 | b |
| 3 | c |

**as\_tibble(x, ...)** Convert data frame to tibble.

**enframe(x, name = "name", value = "value")**  
Convert named vector to a tibble

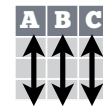
**is\_tibble(x)** Test whether x is a tibble.



## 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:

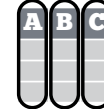


Each **variable** is in its own **column**

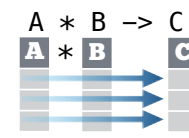


Each **observation**, or **case**, is in its own **row**

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.

table4a

| country | 1999 | 2000 |
|---------|------|------|
| A       | 0.7K | 2K   |
| B       | 37K  | 80K  |
| C       | 212K | 213K |

→

| country | year | cases |
|---------|------|-------|
| A       | 1999 | 0.7K  |
| B       | 1999 | 37K   |
| C       | 1999 | 212K  |
| A       | 2000 | 2K    |
| B       | 2000 | 80K   |
| C       | 2000 | 213K  |

key value

`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.

table2

| country | year | type  | count |
|---------|------|-------|-------|
| A       | 1999 | cases | 0.7K  |
| A       | 1999 | pop   | 19M   |
| A       | 2000 | cases | 2K    |
| A       | 2000 | pop   | 20M   |
| B       | 1999 | cases | 37K   |
| B       | 1999 | pop   | 172M  |
| B       | 2000 | cases | 80K   |
| B       | 2000 | pop   | 174M  |
| C       | 1999 | cases | 212K  |
| C       | 1999 | pop   | 1T    |
| C       | 2000 | cases | 213K  |
| C       | 2000 | pop   | 1T    |

key value

`spread(table2, type, count)`

## Handle Missing Values

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

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

x

| x1 | x2 |
|----|----|
| A  | 1  |
| B  | NA |
| C  | NA |
| D  | 3  |
| E  | NA |

→

| x1 | x2 |
|----|----|
| A  | 1  |
| D  | 3  |

`drop_na(x, x2)`

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

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

x

| x1 | x2 |
|----|----|
| A  | 1  |
| B  | NA |
| C  | NA |
| D  | 3  |
| E  | NA |

→

| x1 | x2 |
|----|----|
| A  | 1  |
| B  | 1  |
| C  | 1  |
| D  | 3  |
| E  | 3  |

`fill(x, x2)`

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

Replace NA's by column.

x

| x1 | x2 |
|----|----|
| A  | 1  |
| B  | NA |
| C  | NA |
| D  | 3  |
| E  | NA |

→

| x1 | x2 |
|----|----|
| A  | 1  |
| B  | 2  |
| C  | 2  |
| D  | 3  |
| E  | 2  |

`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)`

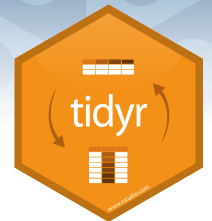
**expand(data, ...)**

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.

table3

| country | year | rate     |
|---------|------|----------|
| A       | 1999 | 0.7K/19M |
| A       | 2000 | 2K/20M   |
| B       | 1999 | 37K/172M |
| B       | 2000 | 80K/174M |
| C       | 1999 | 212K/1T  |
| C       | 2000 | 213K/1T  |

→

| country | year | cases | pop  |
|---------|------|-------|------|
| A       | 1999 | 0.7K  | 19M  |
| A       | 2000 | 2K    | 20M  |
| B       | 1999 | 37K   | 172M |
| B       | 2000 | 80K   | 174M |
| C       | 1999 | 212K  | 1T   |
| C       | 2000 | 213K  | 1T   |

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

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

Separate each cell in a column to make several rows.

table3

| country | year | rate     |
|---------|------|----------|
| A       | 1999 | 0.7K/19M |
| A       | 2000 | 2K/20M   |
| B       | 1999 | 37K/172M |
| B       | 2000 | 80K/174M |
| C       | 1999 | 212K/1T  |
| C       | 2000 | 213K/1T  |

→

| country | year | rate |
|---------|------|------|
| A       | 1999 | 0.7K |
| A       | 1999 | 19M  |
| A       | 2000 | 2K   |
| A       | 2000 | 20M  |
| B       | 1999 | 37K  |
| B       | 1999 | 172M |
| B       | 2000 | 80K  |
| B       | 2000 | 174M |
| C       | 1999 | 212K |
| C       | 1999 | 1T   |
| C       | 2000 | 213K |
| C       | 2000 | 1T   |

`separate_rows(table3, rate, sep = "/")`

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

Collapse cells across several columns to make a single column.

table5

| country | century | year |
|---------|---------|------|
| Afghan  | 19      | 99   |
| Afghan  | 20      | 00   |
| Brazil  | 19      | 99   |
| Brazil  | 20      | 00   |
| China   | 19      | 99   |
| China   | 20      | 00   |

→

| country | year |
|---------|------|
| Afghan  | 1999 |
| Afghan  | 2000 |
| Brazil  | 1999 |
| Brazil  | 2000 |
| China   | 1999 |
| China   | 2000 |

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