

# 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", "!"))`

## 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

sex is a character

earn is a double (numeric)

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)`

## 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())



# 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>
audi a4 1.8
audi a4 2.0
audi a4 2.0
audi a4 2.8
audi a4 3.1
audi a4 3.1
audi a4 quattro 1.8
audi a4 quattro 1.8
audi a4 quattro 2.0
... with 224 more rows, and 3
more variables: year <int>,
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 **pivot\_longer()** and **pivot\_wider()** to reorganize the values of a table into a new layout.

**pivot\_longer()**(data, cols, names\_to = "name", names\_prefix = NULL, names\_sep = NULL, names\_pattern = NULL, names\_ptypes = list(), names\_transform = list(), names\_repair = "check\_unique", values\_to = "value", values\_drop\_na = FALSE, values\_ptypes = list(), values\_transform = list(), ...)

**pivot\_longer()** pivots cols columns, moving column names into a names\_to column, and column values into a values\_to 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

`pivot_longer(table4a, cols = 2:3, names_to = "year", values_to = "cases")`

**pivot\_wider()**(data, id\_cols = NULL, names\_from = name, names\_prefix = "", names\_sep = "\_", names\_glue = NULL, names\_sort = FALSE, names\_repair = "check\_unique", values\_from = value, values\_fill = NULL, values\_fn = NULL, ...)

**pivot\_wider()** pivots a names\_from and a values\_from column into a rectangular field of cells.

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

→

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	NA	NA

`pivot_wider(table2, names_from = type, values_from = 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

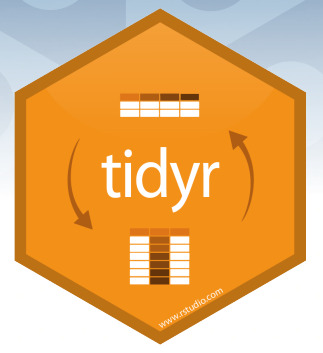
**complete(data, ..., fill = list())**

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	172
B	2000	80K	174
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 = "")`