

# Data tidying with tidyr : : CHEATSHEET



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

A table is tidy if:



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

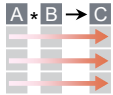
&



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



Access **variables** as **vectors**



Preserve **cases** in vectorized operations

## Tibbles

### AN ENHANCED DATA FRAME

Tibbles are a table format provided by the **tibble** package. They inherit the data frame class, but have improved behaviors:

- **Subset** a new tibble with `[ ]`, a vector with `[ ]` and `$`.
- **No partial matching** when subsetting columns.
- **Display** concise views of the data on one screen.

**options**(`tibble.print_max = n`, `tibble.print_min = m`, `tibble.width = Inf`) Control default display settings.

**View()** or **glimpse()** View the entire data set.

### CONSTRUCT A TIBBLE

**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")

Both make this tibble

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

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

**enframe**(`x`, `name = "name"`, `value = "value"`)

Convert a named vector to a tibble. Also **deframe()**.

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

## Reshape Data - Pivot data to reorganize values into a new layout.

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**(`data`, `cols`, `names_to = "name"`, `values_to = "value"`, `values_drop_na = FALSE`)

"Lengthen" data by collapsing several columns into two. Column names move to a new `names_to` column and values to a new `values_to` column.

**pivot\_longer**(`table4a`, `cols = 2:3`, `names_to = "year"`, `values_to = "cases"`)

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    |

| 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   |

**pivot\_wider**(`data`, `names_from = "name"`, `values_from = "value"`)

The inverse of **pivot\_longer**(). "Widen" data by expanding two columns into several. One column provides the new column names, the other the values.

**pivot\_wider**(`table2`, `names_from = type`, `values_from = count`)

## Split Cells - Use these functions to split or combine cells into individual, isolated values.

table5

| country | century | year |
|---------|---------|------|
| A       | 19      | 99   |
| A       | 20      | 00   |
| B       | 19      | 99   |
| B       | 20      | 00   |

| country | year |
|---------|------|
| A       | 1999 |
| A       | 2000 |
| B       | 1999 |
| B       | 2000 |

**unite**(`data`, `col`, ..., `sep = "_"`, `remove = TRUE`, `na.rm = FALSE`) Collapse cells across several columns into a single column.

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

table3

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

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

**separate\_wider\_delim**(`data`, `cols`, `delim`, ..., `names = NULL`, `names_sep = NULL`, `names_repair = "check_unique"`, `too_few`, `too_many`, `cols_remove = TRUE`) Separate each cell in a column into several columns. Also **separate\_wider\_regex()** and **separate\_wider\_position()**.

**separate\_wider\_delim**(`table3`, `rate`, `delim = "/"`, `into = c("cases", "pop")`)

table3

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

| country | year | rate |
|---------|------|------|
| A       | 1999 | 0.7K |
| A       | 2000 | 2K   |
| A       | 2000 | 20M  |
| B       | 1999 | 37K  |
| B       | 1999 | 172M |
| B       | 2000 | 80K  |
| B       | 2000 | 174M |

**separate\_longer\_delim**(`data`, `cols`, `delim`, ..., `width`, `keep_empty`) Separate each cell in a column into several rows.

**separate\_longer\_delim**(`table3`, `rate`, `delim = "/"`)

## Expand Tables

Create new combinations of variables or identify implicit missing values (combinations of variables not present in the data).

| x1 | x2 | x3 |
|----|----|----|
| A  | 1  | 3  |
| B  | 1  | 4  |
| B  | 2  | 3  |

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

| x1 | x2 | x3 |
|----|----|----|
| A  | 1  | 3  |
| B  | 1  | 4  |
| B  | 2  | 3  |

**complete**(`data`, ..., `fill = list()`) Add missing possible combinations of values of variables listed in ... Fill remaining variables with NA.  
**complete**(`mtcars`, `cyl`, `gear`, `carb`)

## Handle Missing Values

Drop or replace explicit missing values (NA).

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

**drop\_na**(`data`, ...) Drop rows containing NA's in ... columns.  
**drop\_na**(`x`, `x2`)

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

**fill**(`data`, ..., `.direction = "down"`) Fill in NA's in ... columns using the next or previous value.  
**fill**(`x`, `x2`)

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

**replace\_na**(`data`, `replace`) Specify a value to replace NA in selected columns.  
**replace\_na**(`x`, `list(x2 = 2)`)

# Nested Data

A **nested data frame** stores individual tables as a list-column of data frames within a larger organizing data frame. List-columns can also be lists of vectors or lists of varying data types. Use a nested data frame to:

- Preserve relationships between observations and subsets of data. Preserve the type of the variables being nested (factors and datetimes aren't coerced to character).
- Manipulate many sub-tables at once with **purrr** functions like `map()`, `map2()`, or `pmap()` or with **dplyr** `rowwise()` grouping.

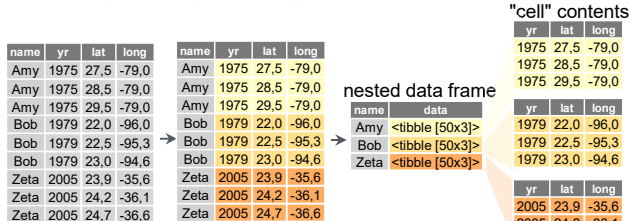
## CREATE NESTED DATA

**nest(data, ...)** Moves groups of cells into a list-column of a data frame. Use alone or with **dplyr::group\_by()**:

1. Group the data frame with **group\_by()** and use **nest()** to move the groups into a list-column.  

```
n_storms <- storms |>
  group_by(name) |>
  nest()
```
2. Use **nest(new\_col = c(x, y))** to specify the columns to group using **dplyr::select()** syntax.  

```
n_storms <- storms |>
  nest(data = c(year:long))
```



Index list-columns with `[[ ]]`. `n_storms$data[[1]]`

## CREATE TIBBLES WITH LIST-COLUMNS

**tibble::tribble(...)** Makes list-columns when needed.  
**tribble(~max, ~seq,**

|    |      |
|----|------|
| 3, | 1:3, |
| 4, | 1:4, |
| 5, | 1:5) |

| max | seq       |
|-----|-----------|
| 3   | <int [3]> |
| 4   | <int [4]> |
| 5   | <int [5]> |

**tibble::tibble(...)** Saves list input as list-columns.  
`tibble(max = c(3, 4, 5), seq = list(1:3, 1:4, 1:5))`

**tibble::enframe(x, name="name", value="value")**  
Converts multi-level list to a tibble with list-cols.  
`enframe(list("3"=1:3, "4"=1:4, "5"=1:5), "max", "seq")`

## OUTPUT LIST-COLUMNS FROM OTHER FUNCTIONS

**dplyr::mutate()**, **transmute()**, and **summarise()** will output list-columns if they return a list.

```
mtcars |>
  group_by(cyl) |>
  summarise(q = list(quantile(mpg)))
```

## RESHAPE NESTED DATA

**unnest(data, cols, ..., keep\_empty = FALSE)** Flatten nested columns back to regular columns. The inverse of `nest()`.  
`n_storms |> unnest(data)`

**unnest\_longer(data, col, values\_to = NULL, indices\_to = NULL)**  
Turn each element of a list-column into a row.

```
starwars |>
  select(name, films) |>
  unnest_longer(films)
```

| name  | films     |
|-------|-----------|
| Luke  | <chr [5]> |
| Luke  | <chr [6]> |
| C-3PO | <chr [7]> |
| R2-D2 | <chr [7]> |

| name  | films               |
|-------|---------------------|
| Luke  | The Empire Strik... |
| Luke  | Revenge of the S... |
| C-3PO | The Empire Strik... |
| C-3PO | Attack of the Cl... |
| C-3PO | The Phantom M...    |
| R2-D2 | The Empire Strik... |
| R2-D2 | Attack of the Cl... |
| R2-D2 | The Phantom M...    |

**unnest\_wider(data, col)** Turn each element of a list-column into a regular column.

```
starwars |>
  select(name, films) |>
  unnest_wider(films, names_sep =
    "_")
```

| name  | films     |
|-------|-----------|
| Luke  | <chr [5]> |
| C-3PO | <chr [6]> |
| R2-D2 | <chr [7]> |

| name  | films_1                         | films_2                | films_3             |
|-------|---------------------------------|------------------------|---------------------|
| Luke  | The Empire... Strik...          | Revenge of... the S... | Return of... Jed... |
| C-3PO | The Empire... Strik...<chr [6]> | Attack of... Cl...     | The Phantom... M... |
| R2-D2 | The Empire... Strik...          | Attack of... Cl...     | The Phantom... M... |

**hoist(data, .col, ..., .remove = TRUE)** Selectively pull list components out into their own top-level columns. Uses **purrr::pluck()** syntax for selecting from lists.

```
starwars |>
  select(name, films) |>
  hoist(films, first_film = 1, second_film = 2)
```

| name  | films     |
|-------|-----------|
| Luke  | <chr [5]> |
| C-3PO | <chr [6]> |
| R2-D2 | <chr [7]> |

| name  | first_film    | second_film   | films     |
|-------|---------------|---------------|-----------|
| Luke  | The Empire... | Revenge of... | <chr [3]> |
| C-3PO | The Empire... | Attack of...  | <chr [4]> |
| R2-D2 | The Empire... | Attack of...  | <chr [5]> |

## TRANSFORM NESTED DATA

A vectorized function takes a vector, transforms each element in parallel, and returns a vector of the same length. By themselves vectorized functions cannot work with lists, such as list-columns.

**dplyr::rowwise(data, ...)** Group data so that each row is one group, and within the groups, elements of list-columns appear directly (accessed with `[[ ]]`, not as lists of length one. **When you use rowwise(), dplyr functions will seem to apply functions to list-columns in a vectorized fashion.**



Apply a function to a list-column and **create a new list-column.**

```
n_storms |>
  rowwise() |>
  mutate(n = list(dim(data)))
```

**dim()** returns two values per row

**wrap with list to tell mutate to create a list-column**

Apply a function to a list-column and **create a regular column.**

```
n_storms |>
  rowwise() |>
  mutate(n = nrow(data))
```

**nrow()** returns one integer per row

Collapse **multiple list-columns** into a single list-column.

```
starwars |>
  rowwise() |>
  mutate(transport = list(append(vehicles, starships)))
```

**append()** returns a list for each row, so col type must be list

Apply a function to **multiple list-columns.**

```
starwars |>
  rowwise() |>
  mutate(n_transports = length(c(vehicles, starships)))
```

**length()** returns one integer per row

See **purrr** package for more list functions.