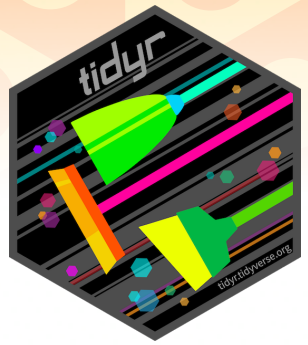
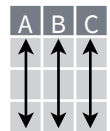


# Tidy Data with tidyr :: CHEAT SHEET



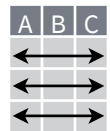
**Tidy data** is a way to organize tabular data in a consistent data structure across packages.

A table is tidy if:



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

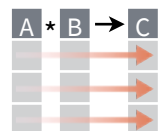
&



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 a tibble

```
A tibble: 3 x 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

→

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**(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
Afghan	19	99
Afghan	20	00
Brazil	19	99
Brazil	20	00

→

country	year
Afghan	1999
Afghan	2000
Brazil	1999
Brazil	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	172
B	2000	80K	174

**separate**(data, col, into, sep = "[^:alnum:]]+", remove = TRUE, convert = FALSE, extra = "warn", fill = "warn", ...) Separate each cell in a column into several columns. Also **extract()**.

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

table3

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

**separate\_rows**(data, ..., sep = "[^:alnum:]]+", convert = FALSE) Separate each cell in a column into several rows.

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

## Expand Tables

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

x

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

→

x1	x2
A	1
A	2
B	1
B	2

**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)`

x

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

→

x1	x2	x3
A	1	3
A	2	NA
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).

x

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

→

x1	x2
A	1
D	3

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

`drop_na(x, x2)`

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

`fill(x, x2)`

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**(data, replace)

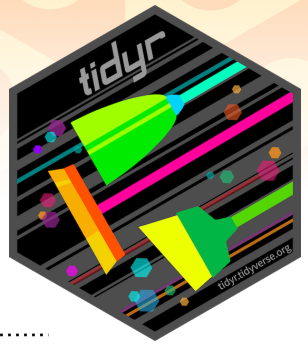
Specify a value to replace NA in selected columns.

`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 date times aren't coerced to character).
- Manipulate many sub-tables at once with **purrr** functions like `map()`, `map2()`, or `pmap()` or with **dplyr** `rowwise()` grouping.



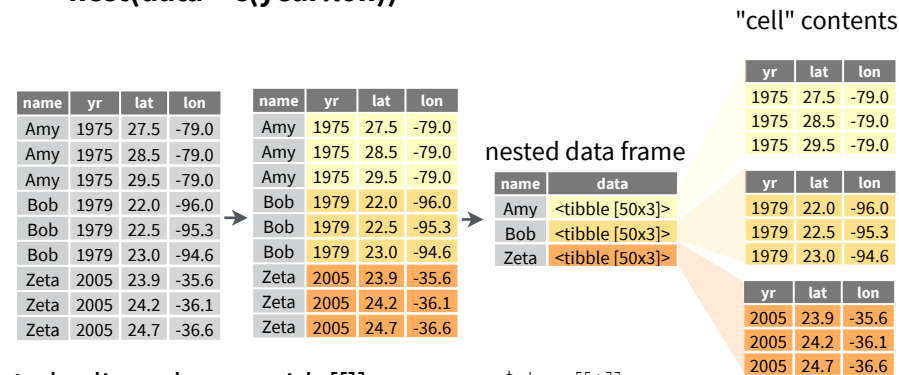
## CREATE NESTED DATA

**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:lon))
```



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 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)))
```

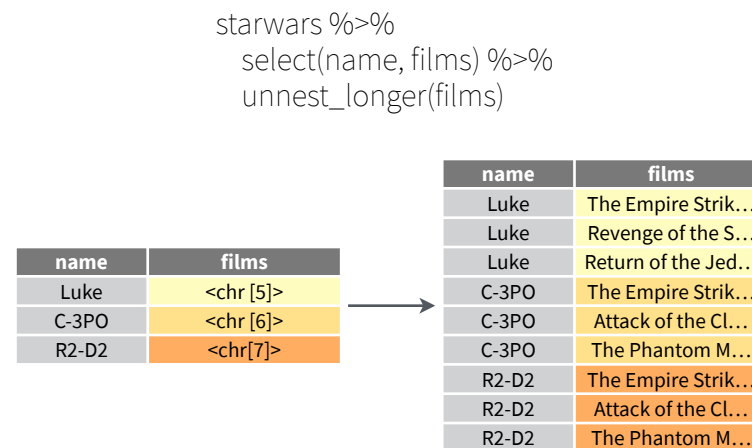
## RESHAPE NESTED DATA

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

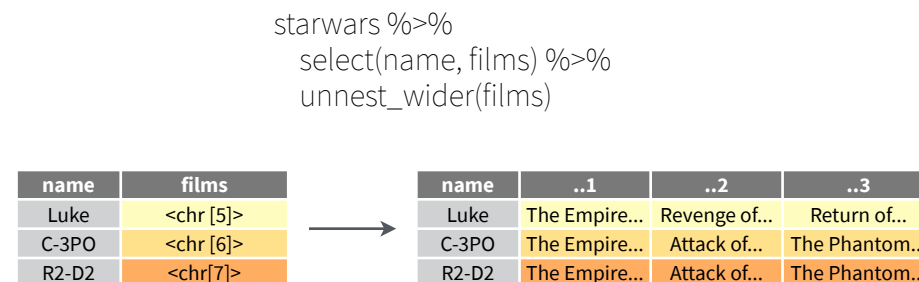
```
n_storms %>% unnest(data)
```

**unnest\_longer(data, col, values\_to = NULL, indices\_to = NULL)**

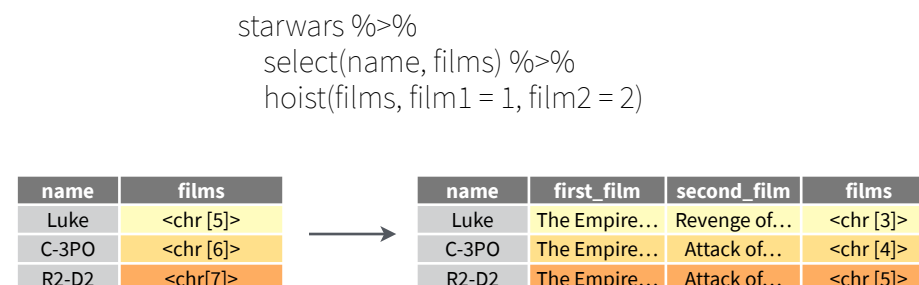
Turn each element of a list-column into a row.



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



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



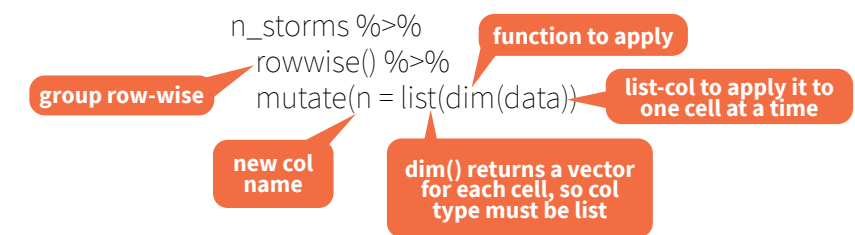
## TRANSFORM NESTED DATA

A vectorized function takes a vector and outputs a vector of the same length (see dplyr cheat sheet). When working with list-columns you don't have access to the usual vectorized functions.

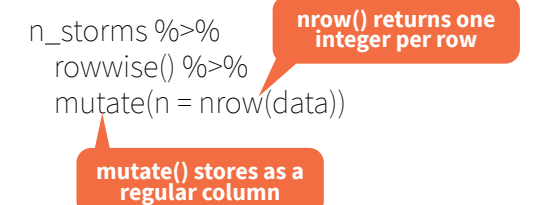
**dplyr::rowwise(.data, ...)** Group data by row. This allows you to specify what you want for one row, dplyr applies the subsequent function to each row. This allows you to work with list-columns using the expected behavior of vectorized functions, and to apply functions element-wise to a list.



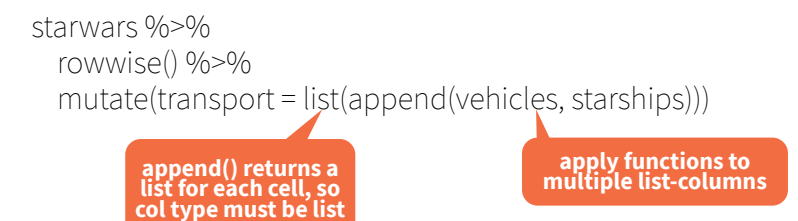
Apply a function to a list-column and **create a new list-column**



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



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



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