

Przetwarzanie i wizualizacja danych

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A data set is **tidy** iff:

1. Each **variable** is in its own **column**
2. Each **case** is in its own **row**
3. Each **value** is in its own **cell**

tame data

baker	cinnamon_1	cardamom_2	nutmeg_3
Emma	1	0	1
Harry	1	1	1
Ruby	1	0	1
Zainab	0	NA	0

trial	Emma	Harry	Ruby	Zainab
cinnamon_1	1	1	1	0
cardamom_2	0	1	0	NA
nutmeg_3	1	1	1	0

tidy data

baker	spice	order	correct
Emma	Cinnamon	1	1
Harry	Cinnamon	1	1
Ruby	Cinnamon	1	1
Zainab	Cinnamon	1	0
Emma	Cardamom	2	0
Harry	Cardamom	2	1
Ruby	Cardamom	2	0
Zainab	Cardamom	2	NA
Emma	Nutmeg	3	1
Harry	Nutmeg	3	1
Ruby	Nutmeg	3	1
Zainab	Nutmeg	3	0

Dane w postaci szerokiej i długiej

wide

id	x	y	z
1	a	c	e
2	b	d	f

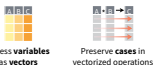
long

id	key	val
1	x	a
2	x	b
1	y	c
2	y	d
1	z	e
2	z	f

Data tidying 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:



Tibbles

AN ENHANCED DATA FRAME

Tibbles are a table format provided by the **tidyverse** 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(...) Convert a data frame to a tibble.

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

Convert a numeric 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.

table1

country	2000	2005
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

table2

country	year	type	count
A	1999	cases	0.7K
A	1999	pop	19M
A	2000	cases	2K
A	2000	pop	30M
B	1999	cases	37K
B	1999	pop	177M
B	2000	cases	80K
B	2000	pop	184M
C	1999	cases	212K
C	1999	pop	11
C	2000	cases	213K
C	2000	pop	11

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

table3

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

table3

country	year	rate
A	1999	0.7K/19M
A	2000	2K/30M
B	1999	37K/177M
B	2000	80K/174M

→

country	year	cases	pop
A	1999	0.7K	19M
A	2000	2K	30M
B	1999	37K	177M
B	2000	80K	174M

table3

country	year	rate
A	1999	0.7K
A	1999	19M
A	2000	2K
A	2000	30M
B	1999	37K/177M
B	1999	177M
B	2000	80K
B	2000	174M

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(table4, cols = 2:3, names_to = "year", values_to = "cases")

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 provides the values.

pivot_wider(table2, names_from = type, values_from = count)

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 = "")

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"))

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

col1	col2	col3
A	1	2
B	1	4
B	1	6

→

col1	col2	col3
A	1	2
A	1	4
A	1	6
A	2	2
A	2	4
A	2	6
B	1	2
B	1	4
B	1	6

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)

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

col1	col2
A	1
A	NA
B	1
B	NA
B	NA

→

col1	col2
A	1
B	1

drop_na(data, ...) Drop rows containing NAs in ... columns. **drop_na**(x, 2)

x

col1	col2
A	1
A	NA
B	1
B	NA
B	NA

→

col1	col2
A	1
B	1

fill(data, ..., direction = "down") Fill in NAs in ... columns using the next or previous value. **fill**(x, 2)

x

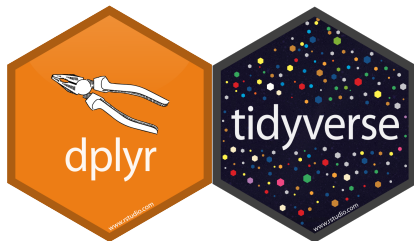
col1	col2
A	1
A	2
B	1
B	2

→

col1	col2
A	1
A	2
B	1
B	2

replace_na(data, replace) Specify a value to replace NA in selected columns. **replace_na**(x, list(x = 2))

dplyr jest pakietem zaprojektowanym do wydajnego (napisany w C++) i efektywnego manipulowania danymi. Dodatkowo wszystko co robimy na ramce danych możemy też zrobić na innych obiektach np. tabelach w bazie danych. Jest on obecnie częścią większej grupy pakietów zwanej tidyverse.





Data Transformation with dplyr : : CHEAT SHEET

dplyr functions work with pipes and expect tidy data. In tidy data:



Each variable is in its own column



Each observation, or case, is in its own row



Summarise Cases

These apply **summary functions** to columns to create a new table of summary statistics. Summary functions take vectors as input and return one value (see back).

summary function



summarise(data, ...)

Compute table of summaries.

`summarise(mtcars, avg = mean(mpg))`



count(x, ..., wt = NULL, sort = FALSE)

Count number of rows in each group defined by the variables in ... Also **tally**!

`count(iris, Species)`

VARIATIONS

summarise_all() - Apply funs to every column.

summarise_at() - Apply funs to specific columns.

summarise_if() - Apply funs to all cols of one type.

Group Cases

Use **group_by()** to create a "grouped" copy of a table. dplyr functions will manipulate each "group" separately and then combine the results.



`mtcars %>%`

`group_by(cyl) %>%`

`summarise(avg = mean(mpg))`



group_by(data, ..., add = FALSE)
Returns copy of table grouped by ...
`g_iris <- group_by(iris, Species)`

ungroup(x, ...)
Returns ungrouped copy of table.
`ungroup(g_iris)`

Manipulate Cases

EXTRACT CASES

Row functions return a subset of rows as a new table.



filter(data, ...) Extract rows that meet logical criteria. `filter(iris, Sepal.Length > 7)`



distinct(data, ..., keep_all = FALSE) Remove rows with duplicate values. `distinct(iris, Species)`



sample_frac(tbl, size = 1, replace = FALSE, weight = NULL, env = parent.frame()) Randomly select fraction of rows.

`sample_frac(iris, 0.5, replace = TRUE)`



sample_n(tbl, size, replace = FALSE, weight = NULL, env = parent.frame()) Randomly select size rows. `sample_n(iris, 10, replace = TRUE)`



slice(data, ...) Select rows by position.

`slice(iris, 10:15)`



top_n(n, n, wt) Select and order top n entries (by group if grouped data). `top_n(iris, 5, Sepal.Width)`

Logical and boolean operators to use with filter()

< <= is.na() %in% | xor()
> >= !is.na() ! &

See ?base:logic and ?Comparison for help.

ARRANGE CASES



arrange(data, ...) Order rows by values of a column or columns (low to high), use with **desc()** to order from high to low.

`arrange(mtcars, mpg)`

`arrange(mtcars, desc(mpg))`

ADD CASES



add_row(data, ..., before = NULL, after = NULL)

Add one or more rows to a table.

`add_row(faithful, eruptions = 1, waiting = 1)`

Manipulate Variables

EXTRACT VARIABLES

Column functions return a set of columns as a new vector or table.



pull(data, var = -1) Extract column values as a vector. Choose by name or index. `pull(iris, Sepal.Length)`



select(data, ...) Extract columns as a table. Also **select_if()**. `select(iris, Sepal.Length, Species)`

Use these helpers with **select()**, e.g. `select(iris, starts_with("Sepal"))`

contains(match) **num_range**(prefix, range) **ends_with**(match) **one_of**(...) **matches**(match) **starts_with**(match)

MAKE NEW VARIABLES

These apply **vectorized functions** to columns. Vectorized funs take vectors as input and return vectors of the same length as output (see back).

vectorized function



mutate(data, ...) Compute new column(s). `mutate(mtcars, gpm = 1/mpg)`



transmute(data, ...) Compute new column(s), drop others. `transmute(mtcars, gpm = 1/mpg)`



mutate_all(tbl, funs, ...) Apply funs to every column. Use with **funs()**. Also **mutate_if()**. `mutate_all(faithful, funs(log))`, `mutate_if(iris, is.numeric, funs(log))`



mutate_at(tbl, cols, funs, ...) Apply funs to specific columns. Use with **funs()**, **vars()** and the helper functions for **select()**. `mutate_at(iris, vars(-Species), funs(log))`



add_column(data, ..., before = NULL, after = NULL) Add new column(s). Also **add_count()**, **add_tally()**. `add_column(mtcars, new = 1:32)`

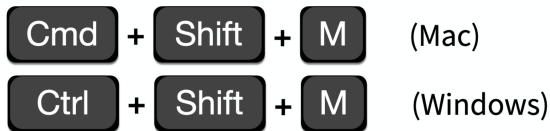


rename(data, ...) Rename columns. `rename(iris, Length = Sepal.Length)`



- **filter**,
- **arrange**,
- **% > %** – strumienie.

Shortcut to type `%>%`



- `select`,
- `mutate`,
- `summarise`,
- `group_by`.