

Przetwarzanie i wizualizacja danych

prof. UAM dr hab. Tomasz Górecki

tomasz.gorecki@amu.edu.pl

Zakład Statystyki Matematycznej i Analizy Danych
Wydział Matematyki i Informatyki
Uniwersytet im. Adama Mickiewicza w Poznaniu



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

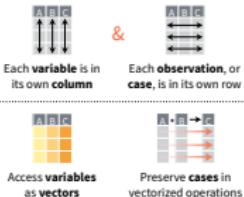
id	x	y	z
1	a	c	e
2	b	d	f

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 tidyverse :: 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

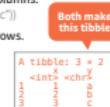
Tibbles are a table format provided by the **tidyverse** package. They inherit the data frame class, but have improved behaviors:

- 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(...)` Construct by rows.

```
tibble(x = c(1, "b", 3), y = c("a", 2, "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	year	cases
A	1999	2K
B	1999	30K
C	2012	213K

table4b

country	year	cases
A	1999	0.7K
B	1999	37K
C	2000	212K
A	2000	8K
B	2000	80K
C	2000	213K

table2

country	year	type	count
A	1999	pop	1K
A	2000	pop	2K
A	2000	2K	10M
B	1999	cases	37K
B	1999	cases	17M
B	2000	cases	80K
C	1999	cases	11
C	2000	cases	213K
C	2000	pop	11

table2

country	year	cases	pop
A	1999	0.7K	10M
A	2000	37K	2K
B	1999	37K	17M
B	2000	80K	8K
C	1999	11	212K
C	2000	213K	11

Split Cells

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

table5

country	century	year
A	20	20
A	20	20
B	19	19
B	20	20

table6

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

table3

country	year	rate
A	1999	0.7K
A	2000	2K
B	1999	37K
B	2000	80K

table4

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

table3

country	year	rate
A	1999	0.7K
A	2000	2K
B	1999	37K
B	2000	80K

table4

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

Expand Tables

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

X

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

X

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

Handle Missing Values

Drop or replace explicit missing values (NA).

X

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

X

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

X

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

X

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

X

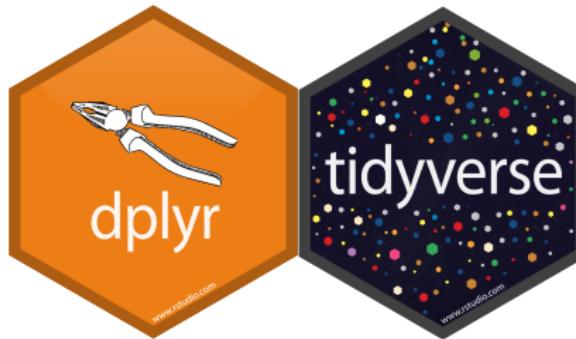
country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K

X

country	year	cases	pop
A	1999	0.7K	12M
A	2000	2K	2M
B	1999	37K	17M
B	2000	80K	8K



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



`x %>% f(y)`
becomes `f(x, y)`

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, 0.1, 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(x, 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::logical 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) t, e.g. mpg:cyl
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(log1, log2, log))`
`mutate_if(iris, is.numeric, funs(log1, 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(log1))`

→ 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)`



dplyr – karta pomocy (2)

Vector Functions

TO USE WITH MUTATE ()

mutate() and **transmute()** apply vectorized functions to columns to create new columns. Vectorized functions take vectors as input and return vectors of the same length as output.

vectorized function

OFFSETS

dplyr::lag() - Offset elements by 1
dplyr::lead() - Offset elements by -1

CUMULATIVE AGGREGATES

dplyr::cumall() - Cumulative all()
dplyr::cumany() - Cumulative any()
dplyr::cummax() - Cumulative max()
dplyr::cummean() - Cumulative mean()
dplyr::cummin() - Cumulative min()
dplyr::cumprod() - Cumulative prod()
dplyr::cumsum() - Cumulative sum()

RANKINGS

dplyr::cume_dist() - Proportion of all values <=
dplyr::dense_rank() - rank with ties = min, no gaps
dplyr::min_rank() - rank with ties = min
dplyr::ntile() - bins into n bins
dplyr::percent_rank() - min_rank scaled to [0,1]
dplyr::row_number() - rank with ties = "first"

MATH

+, -, *, /, ^, %%, %% - arithmetic ops
log(), log2(), log10() - log
*, ^, %% - element-wise comparisons
dplyr::between() - x == left & x <= right
dplyr::near() - safe == for floating point numbers

MISC

dplyr::case_when() - multi-case if...else()
dplyr::coalesce() - first non-NA values by element across a set of vectors
dplyr::if_else() - element-wise if() + else()
dplyr::na_if() - replace specific values with NA
dplyr::pmax() - element-wise max/
pmin() - element-wise min()
dplyr::recode() - Vectorized switch()
dplyr::recode_factor() - Vectorized switch() for factors

Summary Functions

TO USE WITH SUMMARISE ()

summarise() applies summary functions to columns to create a new table. Summary functions take vectors as input and return single values as output.

summary function

COUNTS

dplyr::n() - number of values/rows
dplyr::n_distinct() - # of uniques
sum(is.na()) - # of non-NAs

LOCATION

mean() - mean, also mean(is.na())
median() - median

LOGICALS

mean() - proportion of TRUE's
sum() - # of TRUE's

POSITION/ORDER

dplyr::first() - first value
dplyr::last() - last value
dplyr::nth() - value in nth location of vector

RANK

quantile() - nth quantile
min() - minimum value
max() - maximum value

SPREAD

IQR() - Inter-Quartile Range
mad() - median absolute deviation
sd() - standard deviation
var() - variance

Row Names

Tidy data does not use rownames, which store a variable outside of the columns. To work with the rownames, first move them into a column.

grid::rownames_to_column() - Move row names into col.
a <- rownames_to_column(iris, var = "C")

grid::column_to_rownames() - Move col in row names.
grid::column_to_rownames(a, var = "C")

Also has `rownames()`, `remove_rownames()`

Combine Tables

COMBINE VARIABLES



Use **bind_cols()** to paste tables beside each other as they are.

bind_cols(...)
Returns tables placed side by side as a single table.
BE SURE THAT ROWS ALIGN.

Use a "Mutating Join" to join one table to columns from another, matching values with the rows that they correspond to. Each join retains a different combination of values from the tables.

dplyr::left_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y"), ...) - Join matching values from y to x.

dplyr::right_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y"), ...) - Join matching values from x to y.

dplyr::inner_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y"), ...) - Join data. Retain only rows with matches.

dplyr::full_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y"), ...) - Join all data. Retain all values, all rows.

Use by = c("col1", "col2") to specify the column(s) to match on.
left_join(x, y, by = "A")

Use a named vector, by = c("col1" = "col2"), to match on columns with different names in each data set.
left_join(x, y, by = c("C" = "D"))

Use suffix to specify suffix to give to duplicate column names.
left_join(x, y, by = c("C" = "D"), suffix = c("1", "2"))

COMBINE CASES



Use **bind_rows()** to paste tables below each other as they are.

bind_rows(..., id = NULL)
Returns tables one on top of the other as a single table. Set id to a column name to add a column of the original table names (as pictured).

intersect(x, y, ...)
Rows that appear in both x and y.

setdiff(x, y, ...)
Rows that appear in x but not y.

union(x, y, ...)
Rows that appear in x or y.
(Duplicates removed). union_all() retains duplicates.

Use **setequal()** to test whether two data sets contain the exact same rows (in any order).

EXTRACT ROWS



Use a "Filtering Join" to filter one table against the rows of another.

dplyr::semi_join(x, y, by = NULL, ...) - Returns rows of x that have a match in y.

USEFUL TO SEE WHAT WILL BE JOINED.

dplyr::anti_join(x, y, by = NULL, ...) - Returns rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.



RStudio® is a trademark of RStudio, Inc. • CC BY SA RStudio • info@rstudio.com • 844-448-1232 • rstudio.com • Learn more with browsable vignettes(package = c("dplyr", "tidyverse")) • dplyr 0.7.0 • tibble 1.2.0 • Updated: 2017-03

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

Shortcut to type %>%

Cmd + Shift + M (Mac)

Ctrl + Shift + M (Windows)

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