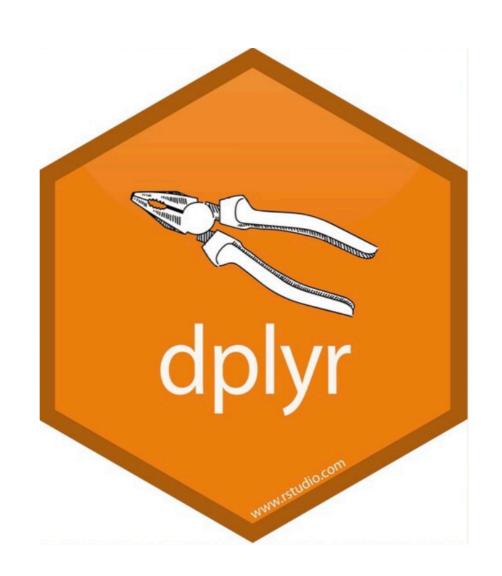
Data Manipulation with dplyr (basics)



The dplyr package

The dplyr package

Tidy data

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Tidy data

The basic functions

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Manipulating rows: filter(), arrange(), slice()

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Manipulating rows: filter(), arrange(), slice()

Manipulating columns: mutate(), select(), summarize()

The dplyr package

Tidy data

The basic functions

Manipulating rows: filter(), arrange(), slice()

Manipulating columns: mutate(), select(), summarize()

Others: group_by()

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First argument is the data (data frame/tibble)

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Subsequent arguments describe your proposed actions

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If you know the querying language SQL, the verbs (functions) are going to sound very similar

dplyr functions all have similar form:

First argument is the data (data frame/tibble)

Subsequent arguments describe your proposed actions

Result is a data frame / tibble

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Every row is a case (person/place/thing being observed)

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Hadley Wickham (creator of *tidyverse*) wrote a paper building up the theory of tidy data: You can find it <u>here</u>

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We will spend a large amount of time on tidying data (tidyr)!

Data Transformation with dplyr:: cheat sheet

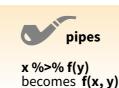


dplyr functions work with pipes and expect **tidy data**. In tidy data:



its own **column**

Each **variable** is in 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))



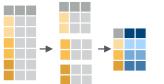
count(x, ..., wt = NULL, sort = FALSE) Count number of rows in each group defined by the variables in ... Also **tally**(). count(iris, Species)

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)



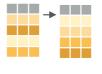
slice(.data, ...) Select rows by position. slice(iris, 10:15)

top_n(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()

%in% xor() is.na() >= !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"))

num_range(prefix, range) :, e.g. mpg:cyl contains(match) ends_with(match) one_of(...) -, e.g, -Species 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



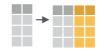
mutate(.data, ...)

Compute new column(s). mutate(mtcars, qpm = 1/mpq)



transmute(.data, ...)

Compute new column(s), drop others. transmute(mtcars, qpm = 1/mpq)



mutate_all(.tbl, .funs, ...**)** Apply funs to every column. Use with funs(). Also mutate_if(). mutate_all(faithful, funs(log(.), log2(.))) 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() manipulates rows by keeping qualifying rows

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data: The data to be manipulated

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```
filter(data, ...)
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What logical test can we use inside of filter?

x < y	Less than
x > y	Greater than
× == y	Equal to
x <= y	Less than or equal to
x >= y	Greater than or equal to
x != y	Not equal to
x %in% y	Group membership
is.na(x)	Is NA
!is.na(x)	Is not NA

#	A tibb	Le: 5 >	x 3
	Χ	У	С
	<int></int>	<int></int>	<chr></chr>
	1	1	а
	2	2	b
	3	2	а
	4	3	b
	5	1	а

```
tb \%>\% filter(x = 1)
```

tb
$$\%$$
> $\%$ filter(x = 1)



tb
$$\%$$
> $\%$ filter(x = 1)





tb
$$\%$$
> $\%$ filter(x = 1)



What's wrong with these function calls?



tb
$$\%$$
> $\%$ filter(x = 1)



What's wrong with these function calls?



tb %>% filter(c == "a")



Filter can take in multiple logical tests!

#	A tibb	Le: 5 >	x 3
	Χ	У	С
	<int></int>	<int></int>	<chr></chr>
	1	1	а
	2	2	b
	3	2	а
	4	3	b
	5	1	а



Filter can take in multiple logical tests!

If you supply them with the comma, they will be combined with "and"



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The same operation can be done with boolean operators!

filter()

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```
tb %>% filter(y == 2, c == "a") = tb %>% filter(y == 2 & c == "a")
```

filter()

Filter can take in multiple logical tests!

If you supply them with the comma, they will be combined with "and"

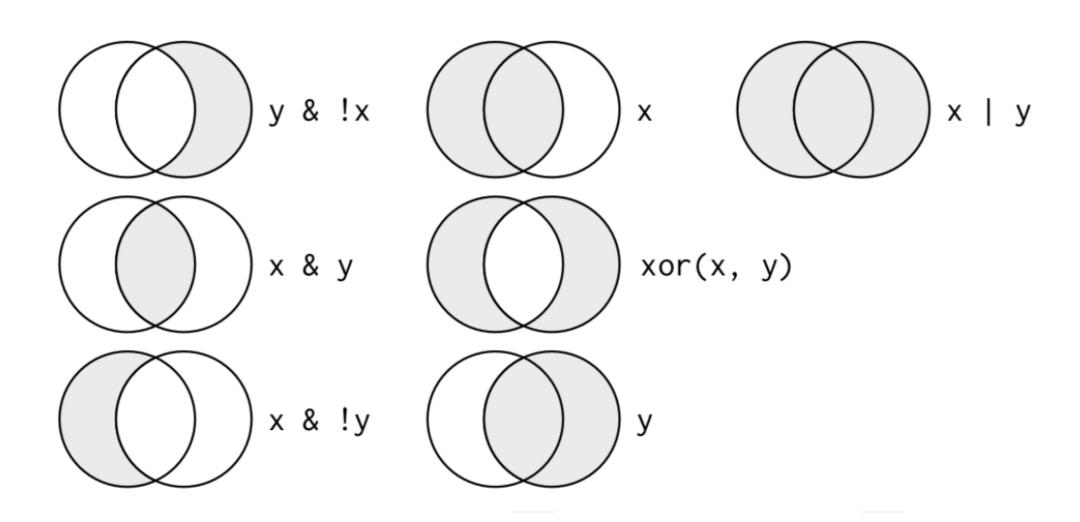
The same operation can be done with boolean operators!

```
tb %>% filter(y == 2, c == "a") = tb %>% filter(y == 2 & c == "a")
```

?base::Logic

a & b	and
a b	or
xor(a, b)	exactly or
!a	not

filter()



Credit: R4DS

```
A tibble: 5 x 3
    x    y    c
<int> <int> <chr>
    1    1    a
    2    2    b
    3    2    a
    4    3    b
    5    1    a
```

arrange() arranges rows! By default, in ascending order

```
# A tibble: 5 x 3
    x    y    c
    <int> <int> <chr>
    1    1    a
    2    2    b
    3    2    a
    4    3    b
    5    1    a
```

arrange() arranges rows! By default, in ascending order

```
tb %>% arrange(y)
```

```
# A tibble: 5 x 3
    x    y    c
    <int> <int> <chr>
    1    1    a
    2    2    b
    3    2    a
    4    3    b
    5    1    a
```

arrange() arranges rows! By default, in ascending order

```
tb %>% arrange(y)
```

```
# A tibble: 5 x 3
    x    y    c
    <int> <int> <chr>
    1    1    a
    2    2    b
    3    2    a
    4    3    b
    5    1    a
```

arrange() arranges rows! By default, in ascending order

```
tb %>% arrange(y)
```

tb %>% arrange(desc(y))

arrange() arranges rows! By default, in ascending order

```
tb %>% arrange(y)
```

tb %>% arrange(desc(y))

arrange() arranges rows! By default, in ascending order

```
tb %>% arrange(y) tb %>% arrange(desc(y))
```

```
# A tibble: 5 x 3
                                       # A tibble: 5 x 3
    X
        у с
                                                  У С
 <int> <int> <chr>
                                        <int> <int> <chr>
                                                  3 b
          1 a
     5
                                                  2 b
          1 a
                                                  2 a
      2 b
       2 a
          3 b
                                                  1 a
```

desc() will sort by descending order

arrange() arranges rows! By default, in ascending order

```
tb %>% arrange(y) tb %>% arrange(desc(y))
```

```
# A tibble: 5 x 3
                                     # A tibble: 5 x 3
    X
       УС
                                               У С
<int> <int> <chr>
                                      <int> <int> <chr>
    1
          1 a
                                               3 b
    5
                                               2 b
          1 a
                                            2 a
      2 b
      2 a
                                             1 a
          3 b
                                             1 a
```

desc() will sort by descending order

A tibble: 5 x 3
 x y c
 <int> <int> <chr>
 1 1 a
 2 2 b
 3 2 a
 4 3 b
 5 1 a

```
# A tibble: 5 x 3
    x    y    c
    <int> <int> <chr>
    1    1    a
    2    2    b
    3    2    a
    4    3    b
    5    1    a
```

```
tb %>% arrange(y,c)
```

```
tb %>% arrange(y,c)
```

```
# A tibble: 5 x 3
    x     y c
<int> <int> <chr>
    1     1     a
    5     1     a
    3     2     a
    2     2     b
    4     3     b
```

```
tb %>% arrange(y,c)
```

```
tb %>% arrange(y,desc(c))
```

```
# A tibble: 5 x 3
    x     y c
<int> <int> <chr>
    1     1     a
    5     1     a
    3     2     a
    2     b
    4     3     b
```

You can also arrange based on multiple variables!

```
tb %>% arrange(y,c)
```

tb %>% arrange(y,desc(c))

You can also arrange based on multiple variables!

Ties are broken from left to right with multiple variables

```
tb %>% slice(1)
```





```
tb %>% slice(1:3)
tb %>% slice(1)
# A tibble: 1 x 3
                                      # A tibble: 3 x 3
      УС
    X
                                              УС
                                           X
<int> <int> <chr>
                                       <int> <int> <chr>
         1 a
                                           1
                                            2 b
                                                2 a
              tb %>% slice(-c(2,4))
                 # A tibble: 3 x 3
                         УС
                      X
                  <int> <int> <chr>
                            1 a
                           2 a
```

A tibble: 5 x 3 x y c <int> <int> <chr> 1 1 a 2 2 b 3 2 a 4 3 b

```
tb \%>\% mutate(z = x+y)
```

mutate() allows you to create and/or modify columns

tb %>% mutate(z = x+y, w = x*y)

```
tb \%>\% mutate(z = x+y)
                        tb %>% mutate(z = x+y, w = x*y)
# A tibble: 5 x 4
                          # A tibble: 5 x 5
      УС
                                УС
                                     Z
 <int> <int> <int><</pre>
                           <int> <int> <int> <int><</pre>
        1 a
                                  1 a
    2 2 b
                                  2 b
    3 2 a
                              3 2 a
    4 3 b
                                  3 b
                                               12
                                                5
      1 a
                                  1 a
```

A tibble: 5 x 3 x y c <int> <int> <chr> 1 1 a 2 2 b 3 2 a 4 3 b

```
tb %>% mutate(z = x+y, z^2 = z^2)
```

mutate() allows you to create and/or modify columns

Rolls the computations over

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mutate() allows you to create and/or modify columns

```
tb %>% mutate(z = x+y, z^2 = z^2) tb %>% mutate(y = x+y)
# A tibble: 5 x 5
                                    # A tibble: 5 x 3
                      z2
    X y C
                                            У С
 <int> <int> <int> <dbl>
                                     <int> <int> <chr>
         1 a
                                            2 a
         2 b
                4 16
                                            4 b
    3 2 a 5 25
        3 b 7 49
                                        4 7 b
                  6 36
```

Rolls the computations over

```
# A tibble: 5 x 3
```

mutate() allows you to create and/or modify columns

```
tb %>% mutate(z = x+y, z^2 = z^2) tb %>% mutate(y = x+y)
# A tibble: 5 x 5
    X Y C Z
                   z2
 <int> <int> <int> <dbl>
        1 a
       2 b
              4 16
    3 2 a 5 25
    4 3 b 7 49
               6 36
```

Rolls the computations over

```
# A tibble: 5 x 3
    X y C
 <int> <int> <chr>
         2a
         4 b
    4 7 b
```

Replaces variables as well

```
# A tibble: 5 x 3
```

mutate() allows you to create and/or modify columns

```
tb %>% mutate(z = x+y, z^2 = z^2) tb %>% mutate(y = x+y)
# A tibble: 5 x 5
    X Y C Z
                   z2
 <int> <int> <chr> <int> <dbl>
        1 a
     2 b 4 16
    3 2 a 5 25
    4 3 b 7 49
            6 36
```

Rolls the computations over

```
# A tibble: 5 x 3
    X y C
 <int> <int> <chr>
         2a
         4 b
    4 7 b
```

Replaces variables as well

There are many mutate() variants!

select() allows you select/deselect certain columns

```
tb %>% select(x,y)
```

select() allows you select/deselect certain columns

Names are preferred if possible for more readable code

There are different helper functions that can make selecting variables easier!

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```
tb %>% select(one_of(c("x", "y")))
```

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Helper functions

```
starts_with("x")
ends_with("x")
contains("x")
one_of(c("x", "y", "z"))
```

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Helper functions

```
starts_with("x")
ends_with("x")
contains("x")
one_of(c("x", "y", "z"))
```

These functions have to do with regular expressions. We will talk about these later!

#	A tibb	Le: 5	x 3
	Χ	У	С
	<int></int>	<int></int>	<chr></chr>
	1	1	а
	2	2	b
	3	2	а
	4	3	b
	5	1	a

```
A tibble: 5 x 3
    x     y     c
<int> <int> <chr>
    1     1     a
    2     2     b
    3     2     a
    4     3     b
    5     1     a
```

summarize() applies a summary function that creates a real-valued statistic

```
A tibble: 5 x 3
    x     y     c
<int> <int> <chr>
    1     1     a
    2     2     b
    3     2     a
    4     3     b
    5     1     a
```

summarize() applies a summary function that creates a real-valued statistic

There are lots of variants of summarize as well!

```
A tibble: 5 x 3

x y c

<int> <int> <chr>

1 1 a

2 2 b

3 2 a

4 3 b

5 1 a
```

summarize() applies a summary function that creates a real-valued statistic

There are lots of variants of summarize as well!

```
A tibble: 5 x 3

x y c

<int> <int> <chr>

1 1 a

2 2 b

3 2 a

4 3 b

5 1 a
```

summarize() applies a summary function that creates a real-valued statistic

```
tb %>% summarize(xbar = mean(x))
```

There are lots of variants of summarize as well!

```
x y c
<int> <int> <chr>
1 1 a
2 2 b
3 2 a
4 3 b
5 1 a
```

summarize() applies a summary function that creates a real-valued statistic

```
tb %>% summarize(xbar = mean(x))

# A tibble: 1 x 1
    xbar
    <dbl>
    3
```

There are lots of variants of summarize as well!

group_by() groups certain variables together to enable conditional computation. We will usually use group_by() in conjunction with summarize() to create statistics for each group

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```
diamonds %>%
group_by(cut) %>%
summarize(mean = mean(price))
```

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There are lots of variants of summarize as well!

```
diamonds %>%
group_by(cut) %>%
summarize(mean = mean(price))
```

```
# A tibble: 5 x 2
cut mean
<ord>
<ord>
<dbl>
1 Fair 4359.
Cood 3929.
Cood 3929.
Cood 3982.
Cood 3983.
Cood 398
```