# Data transformation: dplyr package

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## Introduction

Data transformation is an important step before starting to visualize and analyze your data. Transformation may involve:

- creating new variables and summaries
- renaming some variables
- reordering observations
- format variables
- select variables
- perform analyses by group
- filter/ subset observations
- etc ...

The *dplyr* package is the workhorse for data transformation in R.



# **Installing dplyr**

• as a standalone package:

```
install.packages("dplyr")
```

• which is then called in R by:

library(dplyr)

• as part of the *tidyverse* package

install.packages('tidyverse')

- Note: *tidyverse* is a collection of packages for data science:
  - dplyr
  - ggplot2
  - tidyr
  - **.**..
- here, loading tidyverse automatically loads dplyr plus the rest of the packages
   library(tidyverse)



We will load *dplyr* as part of *tidyverse* package and show some of its uses on the iris data

```
#install.packages('tidyverse') # uncomment to install tidyverse
library(tidyverse)
iris = iris
```

A subset of the iris dataset is shown in the table below:

Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
5.1	3.5	1.4	0.2	setosa
4.9	3.0	1.4	0.2	setosa
4.7	3.2	1.3	0.2	setosa
4.6	3.1	1.5	0.2	setosa
5.0	3.6	1.4	0.2	setosa
5.4	3.9	1.7	0.4	setosa

In the next slides, we will show some key dplyr functions and their examples on the iris data



# **Dplyr functions**



### **Pipes** (%>%)

Pipes are used to combine operations. Pipes are very common in R codes that use functions from dplyr and tidyverse.

- say we want to calculate the means for Sepal.Length and Sepal.Width columns by species for the iris dataset:
- I. One way to do this could be
  - first, group the observations by *Species*
  - o second, calculate the means for the columns

```
by_group = group_by(iris, Species)
meanSepals <- summarise(by_group, mean_Sepal.Length = mean(Sepal.Length, na.rm = TRUE),
    mean_Sepal.Width = mean(Sepal.Width, na.rm = TRUE))</pre>
```

Species	mean_Sepal.Length	mean_Sepal.Width
setosa	5.006	3.428
versicolor	5.936	2.770
virginica	6.588	2.974

II. A second way is to combine the operations through a pipe, annotated by %>% in the code below



```
meanSepals2 <- iris %>% group_by(Species) %>%
  summarise(mean_Sepal.Length = mean(Sepal.Length, na.rm = TRUE),
  mean_Sepal.Width = mean(Sepal.Width, na.rm = TRUE))
```

Species	mean_Sepal.Length	mean_Sepal.Width
setosa	5.006	3.428
versicolor	5.936	2.770
virginica	6.588	2.974

#### • Note:

• the group by() and summarise() functions used here are discussed in later slides



# Select()

Selects columns by name or position

• Select all sepal columns by name

```
# Notes:
## pipes: %>%
## select()
irissubset = iris %>% select(Sepal.Length, Sepal.Width)
```

Sepal.Length	Sepal.Width
5.1	3.5
4.9	3.0
4.7	3.2
4.6	3.1
5.0	3.6
5.4	3.9



#### Alternative ways of specifying columns

- Using *contains* option in *select* function
  - o contains picks out all column names having a Sepal pattern

```
# Notes:
## select(contains())

irissubset = iris %>% select(contains("Sepal"))
```

• By position

```
irissubset = iris %>% select(1:2)
```

Sepal.Length	Sepal.Width
5.1	3.5
4.9	3.0
4.7	3.2
4.6	3.1
5.0	3.6
5.4	3.9



#### Other helper functions within *select()*:

- starts\_with("Sep") ends\_with("")
- num\_range("x", 1:3)
- Use select() together with everything() to reorder columns in a data frame as shown below

#### Initial order of columns

#### irissubset = head(iris)

Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
5.1	3.5	1.4	0.2	setosa
4.9	3.0	1.4	0.2	setosa
4.7	3.2	1.3	0.2	setosa
4.6	3.1	1.5	0.2	setosa
5.0	3.6	1.4	0.2	setosa
5.4	3.9	1.7	0.4	setosa

#### Move Species/Petal columns to start

Species	Petal.Length	Petal.Width	Sepal.Length	Sepal.Width
setosa	1.4	0.2	5.1	3.5
setosa	1.4	0.2	4.9	3.0
setosa	1.3	0.2	4.7	3.2
setosa	1.5	0.2	4.6	3.1
setosa	1.4	0.2	5.0	3.6
setosa	1.7	0.4	5.4	3.9



## **Arrange()**

Sorts rows according to one or more columns

- by default, arrange() sorts in an ascending order
- to change to descending order, place a minus sign (i.e. -) before the ordering column
  - alternative 2: use *desc(column)* function
- Missing values are always sorted to the end



### **Arrange: Ascending**

- Before arrange()
  - For illustration, only the sepal columns are used

```
# Notes
## pipes: %>%
## select()
irissubset = iris %>%
  select(c(Sepal.Length, Sepal.Width))
```

Sepal.Length	Sepal.Width
5.1	3.5
4.9	3.0
4.7	3.2
4.6	3.1
5.0	3.6
5.4	3.9

- After arrange() on Sepal.Length column

```
# Notes:
## pipes: %>%
## arrange()
arranged_iris = irissubset %>%
  arrange(Sepal.Length)
```

Sepal.Length	Sepal.Width
4.3	3.0
4.4	2.9
4.4	3.0
4.4	3.2
4.5	2.3
4.6	3.1



### **Arrange: Descending**

- Before arrange()
  - For illustration, only the sepal columns are used

```
# Notes
## pipes: %>%
## select()

irissubset = iris %>%
  select(c(Sepal.Length, Sepal.Width))
```

Sepal.Length	Sepal.Width
5.1	3.5
4.9	3.0
4.7	3.2
4.6	3.1
5.0	3.6
5.4	3.9

• After arrange()
o on Sepal.Length column

```
# Notes:
## pipes: %>%
## arrange()
arranged_iris2 = irissubset %>%
arrange(-Sepal.Length)
```

Sepal.Length	Sepal.Width	
7.9	3.8	
7.7	3.8	
7.7	2.6	
7.7	2.8	
7.7	3.0	
7.6	3.0	



#### Arrange on two columns

- Before arrange()
  - For illustration, only the sepal columns are used

```
# Notes
## pipes: %>%
## select()

irissubset = iris %>%
  select(c(Sepal.Length, Sepal.Width))
```

Sepal.Length	Sepal.Width
5.1	3.5
4.9	3.0
4.7	3.2
4.6	3.1
5.0	3.6
5.4	3.9

- After arrange()
  - o decreasing Sepal.Length, increasing Sepal.Width

```
# Notes:
## pipes: %>%
## arrange()
arranged_iris3 = irissubset %>%
  arrange(-Sepal.Length, Sepal.Width)
```

Sepal.Length	Sepal.Width
7.9	3.8
7.7	2.6
7.7	2.8
7.7	3.0
7.7	3.8
7.6	3.0



## Mutate()

For creating new variables to a dataset and/or transforming existing variables

- the columns are always added at the end of the dataset
- as an example, if we want to calculate differences between Sepal & Petal lengths and widths

```
irissubset = head(iris %>% select(Species, everything()))
irissubset = irissubset %>%
  mutate(SP_length_diff = Sepal.Length-Petal.Length, SP_width_diff = Sepal.Width - Petal.Width)
```

Species	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	SP_length_diff	SP_width_diff
setosa	5.1	3.5	1.4	0.2	3.7	3.3
setosa	4.9	3.0	1.4	0.2	3.5	2.8
setosa	4.7	3.2	1.3	0.2	3.4	3.0
setosa	4.6	3.1	1.5	0.2	3.1	2.9
setosa	5.0	3.6	1.4	0.2	3.6	3.4
setosa	5.4	3.9	1.7	0.4	3.7	3.5



Similar to *select()*, *mutate()* can use many functions to create new variables:

- arithmetic operators: +, -, \*, /
- logs
- offsets e.g. lead(), lag()
- cumulative and rolling aggregates: cumprod(), cumsum(), cummean() etc

#### mutate\_if()

Checks if a condition is met before applying a transformation to a column

• say we want to format all character variables to be factors in the iris dataset

```
irissubset = head(iris %>% select(Species, everything()))

# calculate the ratios and then format the rel
irissubset = irissubset %>%
  mutate(SP_len_ratio = Sepal.Length/Petal.Length, SP_wid_ratio = Sepal.Width/Petal.Width) %>%
  mutate_if(is.character, as.factor)
```



#### mutate\_at()

Allows to specify specific columns on which an operation should be performed

• say we want to format all ratios between corresponding Sepal and Petal variables to appear with 2 decimal places

```
irissubset = head(iris %>% select(Species, everything()))

# function to format variables to 2 decimal places
format_fn = function(x) formatC(x, format = 'f', digits = 2)

# columns to format
cols_to_format = c("SP_len_ratio", "SP_wid_ratio")

# calculate the ratios and then format the rel
irissubset = irissubset %>%
    mutate(SP_len_ratio = Sepal.Length/Petal.Length, SP_wid_ratio = Sepal.Width/Petal.Width) %>%
    mutate_at(.vars = all_of(cols_to_format), .funs = format_fn)
```

Species	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	SP_len_ratio	SP_wid_ratio
setosa	5.1	3.5	1.4	0.2	3.64	17.50
setosa	4.9	3.0	1.4	0.2	3.50	15.00
setosa	4.7	3.2	1.3	0.2	3.62	16.00



### Filter()

Used to extract rows based on a specified condition

• for instance, to extract all rows of the **Setosa** species

```
irissetosa = iris %>% filter(Species == 'setosa') # keep Setosa Species
flextable(table(irissetosa %>% pull(Species)) %>% as.data.frame())
```

Var1	Freq
setosa	50
versicolor	0
virginica	0

• filter Setosa Species with Sepal length above the median Sepal Length

```
irissetosa2 = iris %>% filter(Species == 'setosa' & Sepal.Width > median(Sepal.Width))
```



## **Summarise()**

Provides summary statistics from a dataset. For instance, if we want to find the median values for Petal columns by species:

```
medianPetals <- iris %>% group_by(Species) %>% select(contains('Petal')) %>%
  summarise(median_Pelal.Length = mean(Petal.Length, na.rm = TRUE),
  median_Petal.Width = mean(Petal.Width, na.rm = TRUE))
```

Species	median_Pelal.Length	median_Petal.Width
setosa	1.462	0.246
versicolor	4.260	1.326
virginica	5.552	2.026

### Group\_by()

It helps other functions perform their operations by groups in a dataset. It does not change the data in itself, but changes how the data is interpreted by other functions.



# Resources

- https://r4ds.had.co.nz/transform.html
   https://cran.r-project.org/web/packages/dplyr/dplyr.pdf