Introduction to R/Tidyverse

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Data Science for Developing Scholars in Down Syndrome Research (DS3) 2025









Links for this session

https://support.rstudio.com/hc/en-us/articles/200526207-Using-RStudio-Projects
https://r4ds.had.co.nz/workflow-projects.html

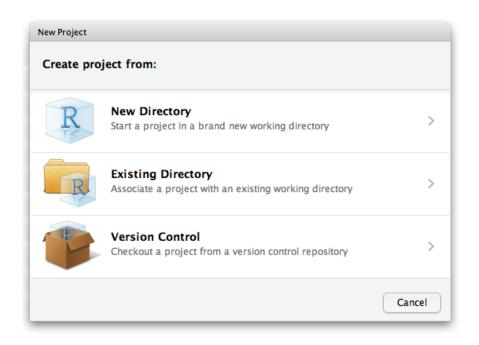
https://github.com/DS3-2025/Rproject_template

https://tidyverse.tidyverse.org/

https://r4ds.hadley.nz/

https://github.com/DS3-2025/tidy_data_exercise

https://ggplot2.tidyverse.org/ https://ggplot2-book.org/



Project_directory

- -/data
- · /results
- -/plots
- /rdata
- analysis_script.R
- helper_functions.R
- project.Rproj

RStudio Projects

- Open existing projects via .Rproj file
- Automatically sets your working directory
- Self-contained set of directories, scripts, and data files (very important for multiple projects)

Organizing your Rstudio Projects

- Only /data and R scripts are required everything else can be recreated (incl. earlier versions)
- Treat /data directory as read-only
- Analysis outputs go to /results or /plots (with version info)
- R workspace and large intermediate files stored in /rdata
- Additional directories added as needed, eg /Archive
- Compatible with manual or other version control

Reproducible data analysis: Package management using renv



- install.packages("renv")
- renv::init() to initialize a new project-local environment with a private R library
- renv::install() to install packages after initialization
- renv::snapshot() to save the state of your project to renv.lock
- renv::restore() to restore the state of your project from renv.lock

Project directory

```
- /data
- /results
- /plots
```

/rdata

/renv

renv.lock

helper_functions.R

· analysis_script.R

- project.Rpro

Allows for fully self-contained R projects (Usually) takes care of installing packages

Into the Tidyverse



https://tidyverse.tidyverse.org/

Base R: install.packages("tidyverse")

renv: renv::install("tidyverse")

Core tidyverse packages:

ggplot2, for data visualisation.

<u>dplyr</u>, for data manipulation.

tidyr, for data tidying.

<u>readr</u>, for data import.

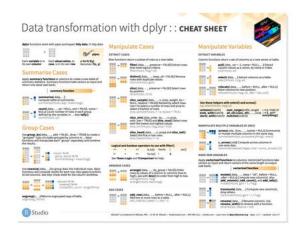
purrr, for functional programming.

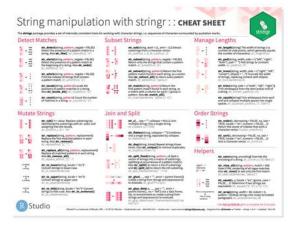
tibble, for tibbles, a modern re-imagining of data frames.

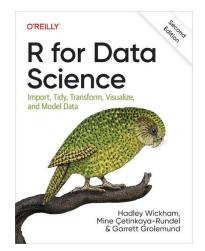
stringr, for strings.

forcats, for factors.

lubridate, for date/times.







Into the Tidyverse: Pipes



Pipe operator %>%

- Avoids nesting
- Minimizes need to create objects and functions
- Structure sequences of operations left-to-right or <u>top-to-bottom</u>
- Easy to inspect and add steps anywhere
- x %>% f is equivalent to f(x)
- x %>% f(y) is equivalent to f(x, y)
- x %>% f %>% g %>% h is equivalent to h (g(f(x)))
- Keyboard shortcut in RStudio: cmd/ctrl+shift+m
- Note: R 4.1.0 introduced a native pipe operator |>
 with some minor differences
 https://www.tidyverse.org/blog/2023/04/base-vs-magrittr-pipe/

Into the Tidyverse: syntax

Base R "dollar sign" syntax

Example - summary statistics:

one continuous variable:

mean (mtcars\$mpg)

one categorical variable:

table (mtcars\$cyl)

two categorical variables:

table (mtcars\$cyl, mtcars\$am)

one continuous, one categorical:

mean (mtcars\$mpg[mtcars\$cyl==4])
mean (mtcars\$mpg[mtcars\$cyl==6])
mean (mtcars\$mpg[mtcars\$cyl==8])

<u>Tidyverse syntax</u>

Example - summary statistics:

one continuous variable:

```
mtcars %>% dplyr::summarize(mean(mpg))
one categorical variable:
```

```
mtcars %>%
    dplyr::group_by(cyl)%>%
    dplyr::summarize(n())
```

two categorical variables:

```
mtcars %>%
    dplyr::group_by(cyl, am) %>%
    dplyr::summarize(n())
```

one continuous, one categorical:

```
mtcars %>%
    dplyr::group_by(cyl)%>%
    dplyr::summarize(mean(mpg))
```

Into the Tidyverse

Z-score calculation with base R:

```
x <- sweep (sweep (t (dat), 1,
apply(t(dat),1,mean,na.rm=T), FUN = "-"),
1, apply(t(dat),1,sd,na.rm=T), FUN = "/")
```

- hard to decipher (learning barrier)
- have to enter target object name in several places

Z-score calc with tidyverse + scale():

```
zscores <- dat |>
      select(LabID, Analyte, Value) |>
      pivot wider() |>
      scale()
```

 Somewhat easier to decipher, but not obvious that this calculates Z-scores, even looking at ?scale defaults (center = TRUE, scale = TRUE)

Manual Z-score calc with tidyverse:

```
zscores <- dat |>
      select(LabID, Analyte, Value) |>
      group by(Analyte) |>
      mutate(
       zscore = (Value - mean(Value, na.rm = TRUE)) / sd(Value, na.rm = TRUE)
      ) |>
      ungroup()
```

- Naming of new variable
- Easier to see how calculation was performed
- Easy to keep both original and transformed values for comparison

Into the Tidyverse

Z-score calculation with base R:

```
x <- sweep(sweep(t(dat), 1,
apply(t(dat),1,mean,na.rm=T), FUN = "-"),
1, apply(t(dat),1,sd,na.rm=T), FUN = "/")</pre>
```

Even more verbose tidyverse version:

```
zscores <- dat |>
    select(LabID, Analyte, Value) |>
    group_by(Analyte) |>
    mutate(
        mean = mean(Value, na.rm = TRUE),
        sd = sd(Value, na.rm = TRUE),
        zscore = (Value - mean) / sd
) |>
    ungroup()
```

- hard to decipher (learning barrier)
- have to enter target object name in several places

- Naming of new variable
- Easier to see how calculation was performed
- Easy to keep both original and transformed values for comparison







Tibbles = enhanced data frames

- Easier preview of data
- Concise summary information including data types

Importing delimited data

- Easy reading in of data from .txt, .csv, .tsv, .xlsx
- Guessing of column types
- Will not convert strings
- Imported as tibble

>	mpg %>% as.data	a.frame()										
	manufacturer	model	displ	year	cyl	trans	drv	cty	hwy	fl	class	
1	audi	a4	1.8	1999	4	auto(l5)	f	18	29	р	compact	
2	audi	a4	1.8	1999	4	manual(m5)	f	21	29	р	compact	
3	audi	a4	2.0	2008	4	manual(m6)	f	20	31	р	compact	
4	audi	a4	2.0	2008	4	auto(av)	f	21	30	р	compact	
5	audi	a4	2.8	1999	6	auto(15)	f	16	26	р	compact	
6	audi	a4	2.8	1999	6	manual(m5)	f	18	26	р	compact	
7	audi	a4	3.1	2008	6	auto(av)	f	18	27	р	compact	
83	ford	explorer 4wd	5.0	1999	8	auto(14)	4	13	17	r	suv	
83 84		explorer 4wd f150 pickup 4wd		1999 1999	8 6	auto(l4) auto(l4)	4 4	13 14	17 17	r r	suv pickup	
	ford		4.2		6							
84	ford ford	f150 pickup 4wd	4.2 4.2	1999	6	auto(l4)	4	14	17	r	pickup	
84 85	ford ford ford	f150 pickup 4wd f150 pickup 4wd	4.2 4.2 4.6	1999 1999	6	auto(14) manual(m5)	4	14 14	17 17	r	pickup pickup	
84 85 86	ford ford ford ford	f150 pickup 4wd f150 pickup 4wd f150 pickup 4wd	4.2 4.2 4.6 4.6	1999 1999 1999	6 6 8	auto(14) manual(m5) manual(m5)	4 4 4	14 14 13	17 17 16	r r r	pickup pickup pickup	
84 85 86 87	ford ford ford ford ford	f150 pickup 4wd f150 pickup 4wd f150 pickup 4wd f150 pickup 4wd	4.2 4.6 4.6 4.6	1999 1999 1999 1999	6 6 8 8	auto(14) manual(m5) manual(m5) auto(14)	4 4 4 4	14 14 13 13	17 17 16 16	r r r	pickup pickup pickup pickup	
84 85 86 87 88	ford ford ford ford ford ford	f150 pickup 4wd f150 pickup 4wd f150 pickup 4wd f150 pickup 4wd f150 pickup 4wd	4.2 4.6 4.6 4.6 5.4	1999 1999 1999 1999 2008	6 6 8 8	auto(14) manual(m5) manual(m5) auto(14) auto(14)	4 4 4 4 4	14 14 13 13	17 17 16 16 17	r r r r	pickup pickup pickup pickup pickup	

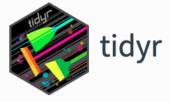
VS.

> mpg										
# A tibble: 234	× 11									
manufacturer	model	displ	year	cyl	trans	drv	cty	hwy	fl	class
<chr></chr>	<chr></chr>	<dbl></dbl>	<int></int>	<int></int>	<chr></chr>	<chr></chr>	<int></int>	<int></int>	<chr></chr>	<chr></chr>
1 audi	a4	1.8	<u>1</u> 999	4	auto(15)	f	18	29	р	compact
2 audi	a4	1.8	<u>1</u> 999	4	manual(m5)	f	21	29	р	compact
3 audi	a4	2	<u>2</u> 008	4	manual(m6)	f	20	31	р	compact
4 audi	a4	2	<u>2</u> 008	4	auto(av)	f	21	30	р	compact
5 audi	a4	2.8	<u>1</u> 999	6	auto(15)	f	16	26	р	compact
6 audi	a4	2.8	<u>1</u> 999	6	manual(m5)	f	18	26	р	compact
7 audi	a4	3.1	<u>2</u> 008	6	auto(av)	f	18	27	р	compact
8 audi	a4 quattro	1.8	<u>1</u> 999	4	manual(m5)	4	18	26	р	compact
9 audi	a4 quattro	1.8	<u>1</u> 999	4	auto(15)	4	16	25	р	compact
10 audi	a4 quattro	2	<u>2</u> 008	4	manual(m6)	4	20	28	р	compact
# with 224 mo	re rows				<u> </u>					



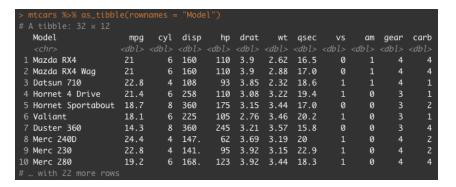
Data manipulation

- mutate() adds new variables that are functions of existing variables.
- select () picks variables based on their names.
- filter() picks rows based on their values.
- summarize () reduces multiple values down to a single summary.
- arrange() changes the ordering of the rows.
- group by () perform group-wise operations.



Reshaping data

- Conversion to/from Tidy data where each column is a variable and each row is an observation.
- pivot longer() converts to Tidy/long format.
- pivot wider() converts to wide format.
- tibble::column_to_rownames (var = "id_col") converts to data frame (required for some functions).



VS.

+ add additional variables



Character string manipulations

- str_detect(x, pattern) looks for match to the pattern;
 commonly used with dplyr::filter()
- str_extract(x, pattern) extracts the text of the match; commonly used with dplyr::mutate()
- str_replace(x, pattern, replacement) replaces the matches with new text; commonly used with dplyr::mutate()



Managing factors

- R uses factors to handle categorical variables
- Often important to control the ordering of factors eg for plotting or modelling
- fct_relevel() changes the order of a factor as specified by
 a character vector
- fct_inorder() changes the order of a factor as specified by current order; commonly used with dplyr::arrange()



Functional programming tools for iterating with functions and vectors

- map() family of functions to replace for loops
- see the <u>Iteration</u> chapter of R for Data Science to learn more <u>https://github.com/rstudio/cheatsheets/blob/master/purrr.pdf</u>





Summarizes key statistical model information in tidy format

- tidy() summarizes information about model components.
- glance() reports information about the entire model.
- augment () adds information about observations to a dataset (eg residuals).
- Works with 100+ model objects.
- Plays well with the nest/unnest functions in tidyr and the map functions in purr https://broom.tidymodels.org/articles/broom_and_dplyr.html



Publication-quality data visualization

- Implements a "grammar of graphics"
- Start by defining the data to be plotted ("aesthetics"):

 ggplot(aes(x, y, color, fill, shape, alpha, linetype))
- Then add layers ("geoms") to specify how data is plotted, eg: + geom point()
- Add additional geom layers, eg:
 + geom boxplot()
- Can split into separate plots, eg male vs. female, by "faceting":
 + facet_wrap(~ Sex)
- Finally add title and modify theme:

```
+ labs(title = "Plot title", subtitle = "plot details"
+ theme(aspect.ratio = 1)
```

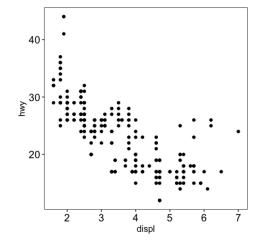


Basics

White the control of persons of the

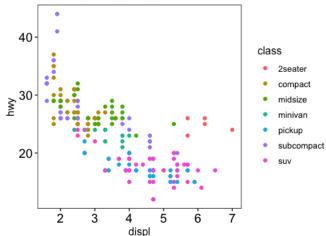


```
mpg %>%
    ggplot(aes(displ, hwy)) +
    geom point()
```



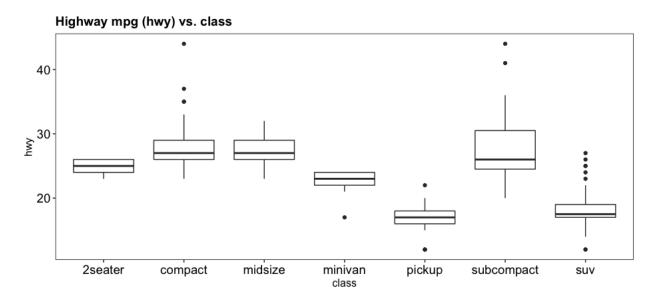
```
mpg %>%
    ggplot(aes(displ, hwy, color = class)) +
    geom_point() +
    theme(aspect.ratio = 1) +
    labs(title = "Highway mpg (hwy) vs. displacement")
```

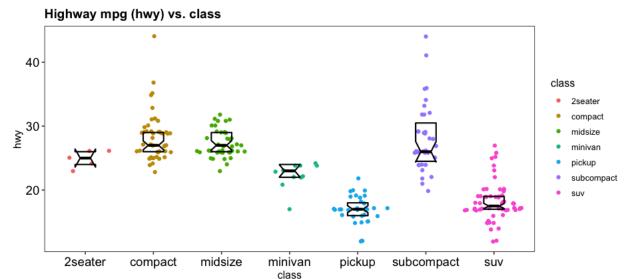
Highway mpg (hwy) vs. displacement





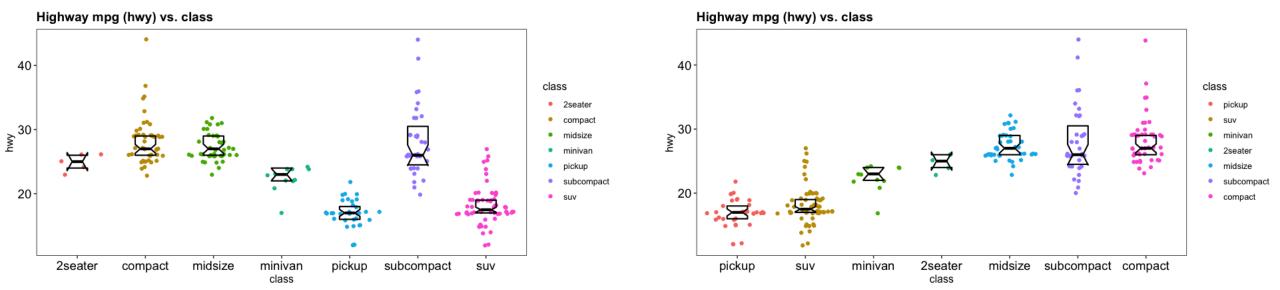
```
mpg %>%
    ggplot(aes(class, hwy)) +
    geom_boxplot() +
    labs(title = "Highway mpg (hwy) vs. class")
```







```
mpg %>%
    group_by(class) %>%
    mutate(mean = mean(hwy)) %>%
    ungroup() %>%
    arrange(mean) %>%
    mutate(class = fct_inorder(class)) %>%
    ggplot(aes(class, hwy, color = class)) +
    ggforce::geom_sina() +
    geom_boxplot(
        notch=TRUE, varwidth=FALSE, outlier.shape=NA, coef=FALSE, width=0.3, color="black", fill="transparent", size=0.75
) +
    labs(title = "Highway mpg (hwy) vs. class")
```



Equivalent plots using base R

