# Data Cleaning with Tidyverse

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#### Resources

- Tidyverse Documentation
- O'Reilly Learning Platform
  - R for Data Science, 2nd Edition
  - R Programming for Statistics and Data Science

```
setwd("path to folder")
```

# Part 1: About Tidyverse

```
install.packages("tidyverse")
```

```
library(tidyverse)
```

Tidyverse is a collection of packages focused on data analysis and data visualizations that share an underlying design philosophy, grammar, and data structures.

Packages Included in Tidyverse

tibble	lighter and more user-friendly version of data frames
<u>tidyr</u>	create tidy and meaningfully arranged data
<u>readr</u>	better importation of data into R
ggplot	data visualization functions
<u>dplyr</u>	data manipulation tools
<u>lubridate</u>	clean dates and times
purr	better functional programming
<u>forcats</u>	handle, clean, and manipulate categorical variables

### Part 2: Loading Data with Tidyverse

#### Readr function: read\_csv

allows you to read a csv file into a tibble data frame

```
sw_df <- read_csv("starwars.csv")</pre>
```

#### Readxl Package: Read Excel Files

allows you to read Excel files in a tibble data frame

```
horror_books <- readxl::read_xlsx("halloween.xlsx", sheet = 1)
horror_movies <- readxl::read_xlsx("halloween.xlsx", sheet = 2)
horror_articles <- readxl::read_xlsx("halloween.xlsx", sheet = 3)</pre>
```

#### Haven package: Read non-proprietary data files

The package Haven allows you to read and export non-proprietary files for SPSS, SAS, and STATA

```
demographics_df <- haven::read_sav("demographics.sav")</pre>
```

### The Pipe Operator

The pipe operator allows you to run commands or operation on a single object based on an order of operations

• let's say you want to see the **name**, **height**, **mass**, and **species** of characters who were born on **Tatooine** 

```
sw_df |> # object we are working on
```

```
filter(homeworld == "Tatooine") |> # first operation
select(name, height, mass, species) # second operation

# order of operations matter

sw_df |> # object we are working on
select(name, height, mass, species) |> # first operation
filter(homeworld == "Tatooine") # second operation

# why did this not work?
```

#### Tibble function: view

view the contents of a data frame in a separate viewer window or in the RStudio viewer pane.

```
view(sw_df)
```

## Tibble function: glimpse

like the str() function in base r, this allow you see the structure of your data but in a more compact manner

```
glimpse(sw_df)
```

# Part 3: Cleaning Data

Main Tidyverse Functions

filter	retains or filters out observations based on variable criteria
select	retains or filters out variables
arrange	sorts variables
mutate	change variable's observations OR create a new variable and observations using observations from another variable
group_by	group observations
<u>summarise</u>	get descriptive statistics about a variable

# Dplyr function: filter

the filter function allows you to select rows in your data frame that meet specific conditions or criteria in a variable

```
sw_df
# let's filter the data frame so we are seeing characters who have blue eyes
sw_eye <- sw_df |>
  filter(eye_color == "blue")
sw_eye
```

#### Boolean operators

boolean operators allows you to build criteria in your code

Boolean operators

&	AND
	OR
==	EQUAL
!=	NOT EQUAL
<	LESS THAN
>	GREATER THAN
<=	LESS THAN OR EQUAL
>=	GREATER THAN OR EQUAL

```
# let's filter the data frame for characters who
# do have blue eyes
# and were born after 50 BBY

sw_eye50 <- sw_df |>
filter(eye_color == "blue" & birth_year < 50)

sw_eye50</pre>
```

#### Dplyr function: select

the **select** function allows you to keep or discard variables

```
# keep variables

sw_select <- sw_df |>
    select(name, height, mass)

sw_select

# remove variables

sw_not_select <- sw_df |>
    select(-height, -mass)

sw_not_select
```

#### Dplyr function: mutate

the **mutate** function *creates* new variables in your data or *change* existing variables by performing calculations or transformations.

NOTE: if you name your variable as an *existing variable*, it will *overwrite* the existing variable. If you give it a *new name*, it will create a *new variable* 

```
demographics_df

demographics_mutate <- demographics_df |>
   mutate(income_new = income/1000) |> # create new variable
   relocate(income_new, .after = income) # relocate variable in data frame

demographics_mutate

# let's overwrite the old variable

demographics_overwrite <- demographics_df |>
   mutate(income = income/1000) # overwrite income variable

demographics_overwrite
```

# Dplyr function: arrange

the **arrange** function allows you to *sort* variables

```
# oldest characters

sw_df |>
    arrange(desc(birth_year))

# characters with the same skin color than the same hair color

sw_df |>
    arrange(desc(skin_color), hair_color)
```

# Dplyr function: group\_by & summarise

the **group\_by** function allows you to *group* common observations in a variable and **summarise** function allows you to get descriptive statistics about the groupings

#### Base Function: as.character

The **as.** function along with **mutate** will allow you to change the data type of a variable. For this example we are going to recode the *character\_id* variable to interpret the data type as a *character* instead of a *double* 

```
sw_df <- sw_df |>
  mutate(character_id = as.character(character_id))
sw_df
```

## Forcats function: as\_factor

The as\_factor function allows you to redefine a variable value as a factor using the mutate function.

```
sw_df
sw_df <- sw_df |>
  mutate(sex = as_factor(sex))

levels(sw_df$sex)
```

### Dyplr Function: recode

we can rename the values of observations within a variable using the **mutate** function in combination with the **recode** or **recode\_factor** functions

# Dplyr function: rename

the **rename** function allows you rename variables in your data frame

```
glimpse(sw_df)

sw_df <- sw_df |>
    rename("sex_label" = sex)

glimpse(sw_df)
```

# Tidyr function: drop\_na

we can remove all missing data from data frames or variables using the drop\_na function

```
# we can see if are data frame has missing NA values using the is.na function.
which(is.na(sw_df$mass))

# because there are missing values we cannot calculate some descriptive statistics
mean(sw_df$mass)

# we can drop all NA values from the data frame

sw_dropNA <- sw_df |>
    drop_na()

mean(sw_dropNA$mass)

# we can also just drop NAs from a variable

sw_dropNA_var <- sw_df |>
    drop_na(mass)

mean(sw_dropNA_var$mass)
```

# Tidyr function: replace\_na

you can also recode the NA values for observations with the replace\_na function

```
# let's replace the NAs the homeworld variable with "unknown"

sw_df <- sw_df |>
  mutate(homeworld = replace na(homeworld, "unknown"))
```

#### Readr function: write csv

the **write\_csv** function allows us to export data frames to a csv file once we are done cleaning it up or when we have done some analysis that we want to export

```
# now that we have this date frame cleaned let's save it
# let's export the file
write_csv(sw_df, "starwars_clean.csv")
```

### Haven function: Export as proprietary file

we can even export files that we have been working on as proprietary files to work on in SPSS, SAS, or STATA

```
# export to SPSS
haven::write_sav(sw_df, "starwars_clean.sav")
```

# Part 4: Explore Your Data

• Psych Package - built-in functions for factor analysis, reliability analysis, descriptive statistics and data visualization.

```
install.packages("psych")
library(psych)
```

```
sw_ds <- describe(sw_df)
write_csv(sw_ds, "starwars_ds.csv")</pre>
```

• SummaryTools Package - simplifies data exploration and descriptive statistics generation for data frames and vectors.

```
install.packages("summarytools")
library(summarytools)
```

```
descr(sw_df)
freq(sw_df$sex)
ctable(sw_df$sex, sw_df$gender)
```

• DataExplorer package - automates and streamlines the process of exploring and visualizing datasets.

```
install.packages("DataExplorer")
library(DataExplorer)
```

```
create_report(reviews_df)
```