

6 Adv Data Wrangling P1

12 Logical Vectors

1. What a Logical Vector Is
 - a logical vector contains only: TRUE, FALSE, or NA
 - you won't always see logical columns in raw data but you create them constantly in analysis
2. Comparisons Create Logicals
 - using comparison operators on numeric or character data returns logical vectors:
<, <=, >, >=, ==, !=
 - is.na() to test for missing values
3. Boolean Algebra Combines Logicals
 - you can combine logical conditions
4. Summaries of Logical Vectors
 - logical vectors behave like 0/1 when summarized
 - any(x): is any value TRUE?
 - all(x): are all values TRUE?
 - sum(x): counts how many TRUEs
 - mean(x): proportion of TRUEs

13 Numbers

1. R Stores Numbers in Floating-Point Form
 - R does not store most numbers exactly.
 - some calculations produce tiny rounding errors
2. Never Check Numeric Equality Directly
 - use a tolerance instead
3. Decimal vs. Scientific Notation
 - R automatically prints very large or very small numbers in scientific notation
4. Integer vs. Double
 - most numbers in R are double-precision (floating-point).
 - add L to force an integer
5. Special Numeric Values
 - Inf = infinity
 - -Inf = negative infinity
 - NaN = "not a number"
 - NA = missing value
6. Summaries With Numbers Must Handle NAs
 - most numeric summaries require na.rm = TRUE:

16 Factors

1. Understanding Factors
 - factors represent categorical data
 - each factor has values and levels
 - levels define the allowed categories and their ordering
2. Why Level Order Matters
 - the order of levels affects how plots display categories
 - the level order influences how models interpret categories
 - reordering levels can clarify relationships in your data
3. Factors vs Characters
 - use characters when categories have no meaningful order
 - use factors when there is a meaningful category structure
 - convert characters to factors when needed
4. Creating And Adjusting Factors
 - use `as.factor()` to convert a variable into a factor
 - use `fct_relevel()` to manually reorder levels
 - use `fct_reorder()` to reorder based on another variable
 - use `fct_lump()` to combine rare categories
5. Using Factors In ggplot
 - ggplot displays categories in the order of factor levels
 - reordering factors before plotting improves readability
 - `fct_reorder()` is especially helpful for bar charts and boxplots

17 Dates and Times

1. Parsing Dates With Lubridate
 - use `ymd()`, `mdy()`, and `dmy()` to parse common date formats
 - use `ymd_hms()` or similar functions to parse date-times
 - the letter order tells lubridate how to interpret the string
2. Distinguishing Dates And Date-Times
 - dates store only year, month, and day
 - date-times store the date plus time and timezone
 - choose the correct type depending on your analysis
3. Extracting Date Components
 - use `year()`, `month()`, and `day()` to pull calendar components
 - use `hour()`, `minute()`, and `second()` for date-times
 - use `wday()` to extract day of week with optional labels
4. Modifying And Shifting Date Components
 - assign new values to components like `year(x) <- 2025`
 - shift dates using `days()`, `weeks()`, `months()`, or `hours()`
 - adding or subtracting time units adjusts the date or time accordingly
5. Working With Time Zones

- every date-time includes a timezone attribute
 - use `with_tz()` to convert to a new timezone
 - use `force_tz()` only when you need to change the label without converting
6. Rounding And Grouping Dates
- use `floor_date()` to round down to the nearest unit
 - use `ceiling_date()` to round up
 - use `round_date()` when general rounding is appropriate
7. Understanding Durations, Periods, And Intervals
- durations measure exact seconds
 - periods use human calendar units like months or years
 - intervals represent a span between two specific date-times
8. Handling Invalid Or Missing Dates
- `lubridate` returns NA for impossible dates like “2020-02-30”
 - parsing functions simplify cleaning messy date data
 - checking for NA is important before analysis