

FROM RAW DATA TO CLEAN DATA IN R

Essential Data Science Operations for Importing, Inspecting, Transforming, and Visualizing Datasets

◆ **GOLDEN RULE:** Keep raw data unchanged; save cleaning decisions in a script so the analysis is reproducible.

① **SETUP**

② **IMPORT & CHECK**

③ **SUBSET & STRUCTURE**

④ **CLEAN & TRANSFORM**

⑤ **AUDIT & PLOT**

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SETUP

INSTALL PACKAGES

R is your car's engine; RStudio (Posit) is the interactive dashboard.
Installing (Once per PC): Downloads packages from CRAN into your hard drive library.
Loading (Every Session): Places tools on your active workbench script.

```
install.packages(c("tidyverse", "readxl", "janitor", "lubridate",  
"stringr", "skimr"))
```

Alternative way: Menu: Tools > Install Packages > type package name > Install

LOAD

Load packages at the start of each R session so their functions are available.

```
library(tidyverse)  
library(readxl)  
library(janitor)  
library(lubridate)  
library(stringr)  
library(skimr)
```

→ Work with data + plots
→ Read Excel files
→ Clean column names
→ Handle dates
→ Clean text
→ Summarise data quickly

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IMPORT
&
CHECK

IMPORT

Importing means reading an external file, such as a CSV, Excel or text file, into R and storing it as a data frame for analysis

```
df <- read_csv("file.csv")
```

→ Read CSV table

```
df <- read_excel("file.xlsx")
```

→ Read Excel
workbook

```
df <- read_delim("file.txt",  
delim = "\t")
```

→ Read text table
(use tab
separator)

INSPECT

Inspection tells you what the data looks like before you change it.

```
head(df)  
View(df)  
glimpse(df)  
str(df)  
summary(df)  
skim(df)
```

→ First rows
→ Spreadsheet view
→ Columns and types
→ Detailed structure
→ Basic summaries
→ Full overview

TYPE CHECKING

Confirm each variable has the right type for analysis.

```
class(df$variable)  
as.numeric(x)  
as.character(x)  
as.factor(x)  
as.logical(x)
```

→ Check current type
→ Make numbers
→ Make text
→ Make groups
→ Make TRUE/FALSE

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SUBSET
&
STRUCTURE

ASSIGNMENT & SUBSETTING

Subsetting keeps the rows and columns needed for a focused analysis.

```
scores <- c(5, 8, 9)  
df <- select(name, age, score)  
df <- filter(score >= "Urban")  
df <- slice(1:10)  
df <- arrange(desc(score))  
df <- distinct()
```

→ Create an object
→ Keep columns
→ Keep matching rows
→ Keep rows
→ Sort rows
→ Remove duplicates

REORGANISING COLUMNS

Reorganisation makes variable names and column order clear and consistent.

```
clean_names(df)  
rename(df, score = total_score)  
mutate(df, pass = score >= 50)  
relocate(df, id, date)  
select(df, id, date, site, score)
```

→ Standardise names
→ Rename a column
→ Create a column
→ Move columns
→ Set final order

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CLEAN
&
TRANSFORM

MISSING DATA

Missing-data checks avoid biased summaries and model errors.

```
is.na(df$score)  
df <-  
fill(df$score)  
sum(is.na(score))  
replace_na(list  
(score = 0))
```

→ Find missing
values
→ Fill down rows
→ Count missing
values
→ Fill values

STRINGS

String cleaning fixes spacing, case and inconsistent labels.

```
str_trim(site)  
str_to_title(site)  
str_to_lower(site)
```

→ Remove spaces
→ Convert to title
case
→ Lower case

DATES

Date cleaning turns text dates into usable time variables.

```
ymd(date)  
mdy(date)  
dmy(date)  
year(date)  
month(date)  
wday(date)
```

→ Year-month-day
→ Month-day-year
→ Day-month-year
→ Extract year
→ Extract month
→ Weekday

PIPE WORKFLOW

A pipe keeps cleaning steps readable from top to bottom.

```
clean_df <- raw_df |>  
mutate(score =  
as.numeric(score)) |>  
filter(!is.na(score))  
|>select(id, date,  
score, site)
```

→ Start clean
object
→ Fix numbers
→ Remove
missing
→ Pick data
→ Make group
→ Make number

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AUDIT
&
PLOT

AUDIT DATA

Auditing catches problems before analysis

```
summary(clean_df)  
sum(duplicated())  
count(df, site)  
count(df, group %in%  
c("A", "B"))  
summarise(mean(score),  
na.rm = TRUE, n =  
n())
```

→ Basic summary
→ Count duplicates
→ Rows per group
→ Group data
→ Mean + count
→ Ignore NAs +
count

VISUALIZATION

Plots show patterns, outliers and relationships.

```
ggplot(clean_df, aes(age,  
score)) +  
geom_point() +  
geom_smooth(method =  
"loess") +  
labs(title = "Score vs  
Age",  
x = "Age", y =  
"Score") +  
theme_minimal()
```

→ Choose
data/aes
→ Colour by group
→ Draw points
→ Add labels
→ Clean layout

PLOT GUIDE

Choose plots by variable type

```
geom_bar()  
geom_boxplot()  
geom_point()  
geom_line()  
facets  
coord.
```

MISTAKES

Small errors can change results.

```
=  
==  
na.rm = TRUE  
library(...)  
set.seed(123)  
raw_df
```

→ Names arguments
→ Tests equality
→ Ignore missing
→ Load tools
→ Fix dates
→ Keep original

MINI
PIPELINE

```
raw_df <- read_csv("survey_data.csv") |>  
clean_df <- raw_df |> clean_names() |> mutate(date = dmy(date), score = as.numeric(score)) |>  
filter(!is.na(score)) |>  
distinct()
```

→ Read raw data
→ Clean names, fix types,
dates, remove missing
→ And duplicates

⚙️ **GENERAL TIPS:** Name files clearly • Check columns before cleaning • Count missing values • Remove duplicates carefully • Check strange values • Save your results



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