

Data Manipulation with *tidyverse* and *dplyr*

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1. Introduction to tidyverse and dplyr

- **tidyverse**: A collection of R packages designed for data science. Includes:
 - ggplot2 → data visualization
 - dplyr → data manipulation
 - tidyr → data tidying
 - readr → data import
 - stringr, forcats, etc.
- **dplyr**: Core package for data manipulation. Provides **verbs** (functions) that are simple and intuitive.

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

# Load
library(dplyr)
```

2. Core dplyr Verbs

2.1 filter() → Select rows (like WHERE in SQL)

```
# Syntax
filter(data, condition)

# Example: keep rows with Age > 25
df %>% filter(Age > 25)
```

2.2 select() → Select columns

```
# Syntax
select(data, col1, col2, ...)

# Example: keep only Name and Score
df %>% select(Name, Score)
```

2.3 arrange() → Sort rows

```
# Syntax
arrange(data, column)

# Example: sort by Score ascending
df %>% arrange(Score)

# Sort by Score descending
df %>% arrange(desc(Score))
```

2.4 mutate() → Create/modify columns

```
# Syntax
mutate(data, new_col = expression)

# Example: create new column Pass/Fail
df %>% mutate(Passed = ifelse(Score >= 80, "Yes", "No"))
```

2.5 summarise() → Summary statistics

```
# Syntax
summarise(data, summary_name = function(column))

# Example: calculate mean score
df %>% summarise(Avg_Score = mean(Score))
```

2.6 group_by() + summarise() → Grouped operations

```
df %>%
  group_by(Passed) %>%
  summarise(Avg_Score = mean(Score), Count = n())
```

3. Data Joins (Merging Tables)

Just like SQL joins:

- `inner_join(x, y, by)` → only matching rows
- `left_join(x, y, by)` → keep all rows from left
- `right_join(x, y, by)` → keep all rows from right
- `full_join(x, y, by)` → keep all rows from both

```
students <- data.frame(ID = 1:3, Name = c("A", "B", "C"))
scores <- data.frame(ID = c(1,2,4), Score = c(90, 85, 70))
students %>% left_join(scores, by = "ID")
```

4. Example Dataset

We'll use a simple dataset for practice:

```
students <- data.frame(
  ID = 1:6,
  Name = c("Ali", "Sara", "John", "Mina", "Ravi", "Lily"),
  Age = c(20, 22, 21, 23, 22, 20),
  Marks = c(85, 90, 70, 60, 95, 88),
  Dept = c("CS", "Math", "CS", "Physics", "Math", "CS")
)
```

5. Practice Problems

Basic

1. Select only the Name and Marks columns.
2. Filter students who scored **above 80**.
3. Filter students in the **CS department**.
4. Arrange students by Marks in **descending order**.
5. Select students with Age > 21 and Marks > 80.

Intermediate

1. Create a new column Grade:

- a. "A" if Marks ≥ 85 , else "B".
2. Add 5 bonus points to all students' Marks.
3. Summarize the **average Marks** of all students.
4. Find the **maximum Marks** in the dataset.
5. Count how many students are in each department.

Grouped operations

1. Compute the **average Marks per department**.
2. Find the **minimum Age** per department.
3. Count how many students passed (Marks ≥ 70) vs failed.
4. Group students by Grade (A/B) and find the **mean Age**.

Joins

1. Create another dataframe sports:

```
sports <- data.frame(ID = c(1,3,5,6), Sport = c("Football",  
"Tennis", "Cricket", "Hockey"))
```

Perform a **left join** with students.

2. Use an **inner join** to keep only students with sports data.
3. Use a **full join** to see all students and all sports.

Challenging

1. Find the **top 3 students** with highest Marks.
2. Compute **department-wise pass percentage** (Marks ≥ 70).
3. Create a summary table: for each Dept \rightarrow show Avg Marks, Max Marks, No. of Students.

6. Suggested Homework (Optional)

1. Import a real dataset (e.g., mtcars or iris).
2. Apply **all the verbs** (**filter**, **select**, **arrange**, **mutate**, **summarise**, **group_by**).
3. Perform at least **one join** with another dataset.