

# Column-wise and Row-wise Operations in dplyr

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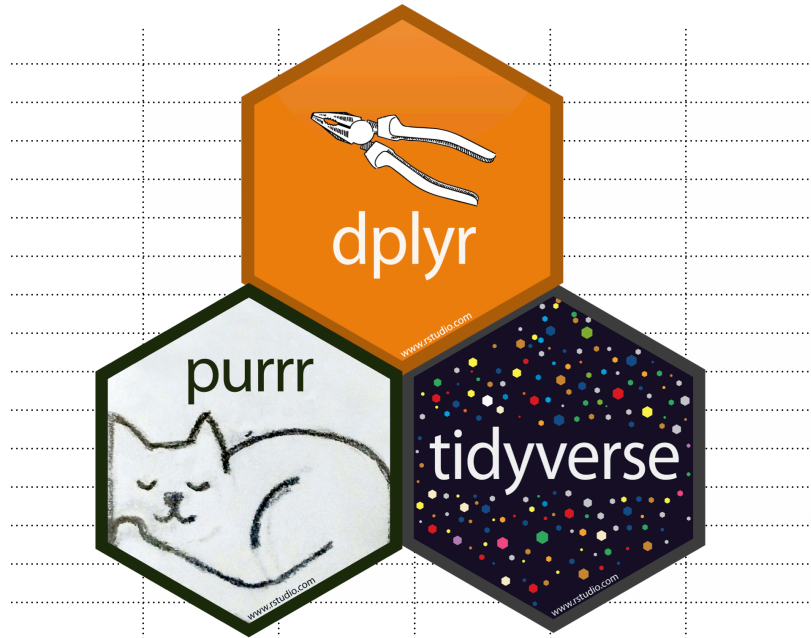


Figure 1: row-wise and column-wise operations

With the development of **dplyr** or its umbrella package **tidyverse**, it becomes quite easy to perform operations over columns or rows in R. These column- or row-wise methods can also be directly integrated with other dplyr verbs like `select`, `mutate`, `filter` and `summarise`, making them more comparable with other functions in `apply` or `map` families. In this blog, I will briefly cover some useful column- or row-wise operations.

## 1 Column-wise operation

**Example 1:** select those string columns with less than 5 levels in the dataset of **starwars**.

```
starwars %>%  
  select_if(~ any(is.character(.x) & length(unique(.x)) <= 5)) %>%  
  head()
```

```
# A tibble: 6 x 2  
  sex    gender  
  <chr>  <chr>  
1 male   masculine  
2 none   masculine
```

```
3 none    masculine
4 male    masculine
5 female  feminine
6 male    masculine
```

We can combine `select_if` and `any` to identify certain columns by certain criterion. **Note:** we are using tilde (`~`) to define an anonymous function, and thus we should use `.x` to refer to the selected columns. See this [link](#) for detailed illustration of tilde (`~`), dot (`.`), and dot x (`.x`) in dplyr.

If you want to calculate the levels of those selected columns, you can try `across` function and `summarise` the number of levels by column.

```
starwars %>%
  summarise(across(where(is.character), ~ length(unique(.x))))
```

```
# A tibble: 1 x 8
  name hair_color skin_color eye_color sex gender homeworld species
<int>   <int>      <int>    <int> <int> <int>    <int>   <int>
1    87        12        31      15    5     3      49     38
```

Alternatively, you can make use of the `map` or `map_dbl` function in **purrr** by the following command. Note that when a `map` function is applied to a data.frame, it will operate over columns by default.

```
# map_dbl returns a double vector, while map returns a list
starwars %>%
  select_if(~ is.character(.x)) %>%
  map_dbl(~length(unique(.x))) %>%
  head()
```

```
name hair_color skin_color eye_color sex gender
87      12      31      15    5     3
```

**Example 2:** select those numeric columns and calculate the means and sds across columns in the dataset of `starwars`.

```
starwars %>%
  summarise(across(where(~ is.numeric(.x)),
    list(Mean = ~ mean(.x, na.rm = TRUE),
         Sd = ~ sd(.x, na.rm = TRUE))))
```

```
# A tibble: 1 x 6
  height_Mean height_Sd mass_Mean mass_Sd birth_year_Mean birth_year_Sd
    <dbl>      <dbl>    <dbl>   <dbl>         <dbl>         <dbl>
1      175.      34.8      97.3    169.          87.6          155.
```

This example provides us a good illustration of the use of `.x` in **dplyr** style syntax, since we have some missing values (NAs) in certain columns. Thus, we need to specify the parameter with `na.rm = TRUE` inside the functions.

There is indeed a more convenient and elegant way of solving this by using the function `summarise_if`. It allows us to select certain columns and operate by columns like this:

```
starwars %>%
  summarise_if(is.numeric,
               list(Sum = sum, Mean = mean, Sd = sd),
               na.rm = TRUE)
```

```
# A tibble: 1 x 9
  height_Sum mass_Sum birth_year_Sum height_Mean mass_Mean birth_year_Mean height_Sd mass_Sd
    <int>    <dbl>         <dbl>      <dbl>    <dbl>         <dbl>    <dbl>    <dbl>
1   14143   5741.         3765.      175.     97.3          87.6    34.8    169.
```

## 2 Row-wise operation

**Example 3:** calculate the **sums**, **means** and **sds** for each row for the dataset of **iris**.

```
iris %>%
  rowwise() %>%
  mutate(
    Rowsum = sum(c_across(Sepal.Length:Petal.Width)),
    Rowsd = sd(c_across(Sepal.Length:Petal.Width)),
    Rowmean = mean(c_across(Sepal.Length:Petal.Width))
  ) %>%
  ungroup() %>%
  head()
```

```
# A tibble: 6 x 8
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species Rowsum Rowsd Rowmean
    <dbl>      <dbl>         <dbl>      <dbl> <fct>    <dbl> <dbl>    <dbl>
1      5.1      3.5          1.4        0.2 setosa    10.2  2.18    2.55
```

|   |     |     |     |     |        |      |      |      |
|---|-----|-----|-----|-----|--------|------|------|------|
| 2 | 4.9 | 3   | 1.4 | 0.2 | setosa | 9.5  | 2.04 | 2.38 |
| 3 | 4.7 | 3.2 | 1.3 | 0.2 | setosa | 9.4  | 2.00 | 2.35 |
| 4 | 4.6 | 3.1 | 1.5 | 0.2 | setosa | 9.4  | 1.91 | 2.35 |
| 5 | 5   | 3.6 | 1.4 | 0.2 | setosa | 10.2 | 2.16 | 2.55 |
| 6 | 5.4 | 3.9 | 1.7 | 0.4 | setosa | 11.4 | 2.23 | 2.85 |

Here the function `c_across` is specifically designed to work with `rowwise` operations. **Note:** `rowwise` groups your data by row (class: `rowwise_df`), and it is best to `ungroup` immediately. Of course, if you are more comfortable with the `apply` function, you can also use the following command:

```
iris %>%
  select(Sepal.Length:Petal.Width) %>%
  apply(., 1, function(x) c(sum(x), sd(x), mean(x))) %>%
  as_tibble() %>%
  t() %>%
  head()
```

|    | [,1] | [,2]     | [,3]  |
|----|------|----------|-------|
| V1 | 10.2 | 2.179449 | 2.550 |
| V2 | 9.5  | 2.036950 | 2.375 |
| V3 | 9.4  | 1.997498 | 2.350 |
| V4 | 9.4  | 1.912241 | 2.350 |
| V5 | 10.2 | 2.156386 | 2.550 |
| V6 | 11.4 | 2.230844 | 2.850 |

```
iris %>%
  rowwise() %>%
  dplyr::mutate(
    Rowsum = sum(c_across(Sepal.Length:Petal.Width)),
    Rowmean = mean(c_across(Sepal.Length:Petal.Width)),
    Rowsd = sd(c_across(Sepal.Length:Petal.Width)),
    .before = "Species"
  ) %>%
  ungroup() %>%
  head()
```

```
# A tibble: 6 x 8
  Sepal.Length Sepal.Width Petal.Length Petal.Width Rowsum Rowmean Rowsd Species
    <dbl>         <dbl>         <dbl>         <dbl>   <dbl>   <dbl> <dbl> <fct>
1 10.2         2.179449         2.550         9.5     2.04    2.38 setosa
2  9.5         2.036950         2.375         9.4     2.00    2.35 setosa
3  9.4         1.997498         2.350         9.4     1.91    2.35 setosa
4  9.4         1.912241         2.350         9.4     1.91    2.35 setosa
5 10.2         2.156386         2.550        10.2     2.16    2.55 setosa
6 11.4         2.230844         2.850        11.4     2.23    2.85 setosa
```

|   |     |     |     |     |      |      |      |        |
|---|-----|-----|-----|-----|------|------|------|--------|
| 1 | 5.1 | 3.5 | 1.4 | 0.2 | 10.2 | 2.55 | 2.18 | setosa |
| 2 | 4.9 | 3   | 1.4 | 0.2 | 9.5  | 2.38 | 2.04 | setosa |
| 3 | 4.7 | 3.2 | 1.3 | 0.2 | 9.4  | 2.35 | 2.00 | setosa |
| 4 | 4.6 | 3.1 | 1.5 | 0.2 | 9.4  | 2.35 | 1.91 | setosa |
| 5 | 5   | 3.6 | 1.4 | 0.2 | 10.2 | 2.55 | 2.16 | setosa |
| 6 | 5.4 | 3.9 | 1.7 | 0.4 | 11.4 | 2.85 | 2.23 | setosa |

```
iris %>%
  as_tibble() %>%
  dplyr::mutate(
    row = pmap(across(1:4), ~ {
      list(rsum = sum, rmean = mean) %>%
        map_dfc(function(f) f(c(...)))
    }),
    .before = "Sepal.Length"
  ) %>%
  unnest(row) %>%
  head()
```

# A tibble: 6 x 7

|   | rsum  | rmean | Sepal.Length | Sepal.Width | Petal.Length | Petal.Width | Species |
|---|-------|-------|--------------|-------------|--------------|-------------|---------|
|   | <dbl> | <dbl> | <dbl>        | <dbl>       | <dbl>        | <dbl>       | <fct>   |
| 1 | 10.2  | 2.55  | 5.1          | 3.5         | 1.4          | 0.2         | setosa  |
| 2 | 9.5   | 2.38  | 4.9          | 3           | 1.4          | 0.2         | setosa  |
| 3 | 9.4   | 2.35  | 4.7          | 3.2         | 1.3          | 0.2         | setosa  |
| 4 | 9.4   | 2.35  | 4.6          | 3.1         | 1.5          | 0.2         | setosa  |
| 5 | 10.2  | 2.55  | 5            | 3.6         | 1.4          | 0.2         | setosa  |
| 6 | 11.4  | 2.85  | 5.4          | 3.9         | 1.7          | 0.4         | setosa  |

```
iris %>%
  as_tibble() %>%
  dplyr::mutate(
    row = pmap(
      across(Sepal.Length:Petal.Width),
      ~ bind_cols(
        rsum = sum(c(...)),
        rmean = mean(c(...)),
        rsd = sd(c(...))
      )
    ),
    .before = "Sepal.Length"
```

```
) %>%
unnest(row) %>%
head()
```

```
# A tibble: 6 x 8
```

|   | rsum  | rmean | rsd   | Sepal.Length | Sepal.Width | Petal.Length | Petal.Width | Species |
|---|-------|-------|-------|--------------|-------------|--------------|-------------|---------|
|   | <dbl> | <dbl> | <dbl> | <dbl>        | <dbl>       | <dbl>        | <dbl>       | <fct>   |
| 1 | 10.2  | 2.55  | 2.18  | 5.1          | 3.5         | 1.4          | 0.2         | setosa  |
| 2 | 9.5   | 2.38  | 2.04  | 4.9          | 3           | 1.4          | 0.2         | setosa  |
| 3 | 9.4   | 2.35  | 2.00  | 4.7          | 3.2         | 1.3          | 0.2         | setosa  |
| 4 | 9.4   | 2.35  | 1.91  | 4.6          | 3.1         | 1.5          | 0.2         | setosa  |
| 5 | 10.2  | 2.55  | 2.16  | 5            | 3.6         | 1.4          | 0.2         | setosa  |
| 6 | 11.4  | 2.85  | 2.23  | 5.4          | 3.9         | 1.7          | 0.4         | setosa  |

### 3 Related links:

- <https://dplyr.tidyverse.org/articles/rowwise.html>
- <https://purrr.tidyverse.org/reference/map.html>