# Applying Vectorized Functions

# Chapter 8

# Stats 20 Lec 2

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## Learning Objectives

After studying this chapter, you should be able to:

- Summarize an R object with str() and summary().
- Understand how to use the apply family of functions: apply(), lapply(), sapply(), tapply().

## 1 Vectorized Summary Functions

Recall that a function in R is vectorized if applying the function to an object will automatically apply the function to individual components of the object. For (atomic) vectors, vector arithmetic is implements operations element-by-element.

```
c(1,2,3) + c(2,3,4)
## [1] 3 5 7
```

For more complex data structures (like matrices, data frames, and lists), we may be interested in applying a function to each row, column, or component. We will start with some generic vectorized functions that provide useful summaries for columns or components of R objects.

#### 1.1 The str() Function

##

##

\$ :List of 2

For a quick overview of any object in R, the **str()** function returns a compact display of the internal **structure** of the input object. As an example, we will apply this function to the **trees** data in the **datasets** package.

```
data(trees) # Load the trees data
str(trees) # Display the structure of the trees object

## 'data.frame': 31 obs. of 3 variables:
## $ Girth : num 8.3 8.6 8.8 10.5 10.7 10.8 11 11 11.1 11.2 ...
## $ Height: num 70 65 63 72 81 83 66 75 80 75 ...
## $ Volume: num 10.3 10.3 10.2 16.4 18.8 19.7 15.6 18.2 22.6 19.9 ...
```

The output of the str(trees) command shows that trees is a data frame with 31 observations and 3 variables. A brief summary of each component (column) in trees is given: Each component of trees is numeric (num), and the first few values from each component are printed.

The str() function is well suited for displaying the contents of nested lists (lists inside lists).

... Name : Factor w/ 3 levels "April", "Leslie", ...: 2 3 1

..\$ Height: num [1:3] 62 71 66 ..\$ Weight: num [1:3] 115 201 66

..\$ Income: num [1:3] 4000 NA 2000

```
## ..$: int [1:5] 1 2 3 4 5
## ..$: int [1:3, 1:3] 1 2 3 4 5 6 7 8 9
```

## 1.2 The summary() Function

We previously used the **summary()** function to compute a few standard summary statistics on numeric vectors. **summary(trees\$Volume)** 

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 10.20 19.40 24.20 30.17 37.30 77.00
```

The summary() function is an example of a **polymorphic** function in that it changes its output based on the type of input. Specifically, the output of summary() will depend on the class of the input object.

For data frames, the summary() function will compute summary statistics for each column in the data frame. If the column is a character or factor vector, the summary() output will adapt and return frequencies. For lists, the summary() function will return the length, class attribute, and mode of each component.

#### summary(trees)

```
##
        Girth
                          Height
                                        Volume
##
    Min.
           : 8.30
                     Min.
                             :63
                                    Min.
                                            :10.20
##
    1st Qu.:11.05
                     1st Qu.:72
                                    1st Qu.:19.40
    Median :12.90
                     Median:76
                                    Median :24.20
##
    Mean
            :13.25
                     Mean
                             :76
                                    Mean
                                            :30.17
                     3rd Qu.:80
##
    3rd Qu.:15.25
                                    3rd Qu.:37.30
    Max.
            :20.60
                     Max.
                             :87
                                    Max.
                                            :77.00
summary(parks.df)
```

```
##
        Name
                     Height
                                      Weight
                                                        Income
                                                           :2000
##
    April:1
                        :62.00
                                          : 66.0
                                                   Min.
                Min.
                                  Min.
##
    Leslie:1
                1st Qu.:64.00
                                  1st Qu.: 90.5
                                                    1st Qu.:2500
    Ron
                Median :66.00
                                  Median :115.0
                                                   Median:3000
##
           : 1
##
                Mean
                        :66.33
                                  Mean
                                          :127.3
                                                   Mean
                                                           :3000
##
                3rd Qu.:68.50
                                  3rd Qu.:158.0
                                                    3rd Qu.:3500
##
                Max.
                        :71.00
                                  Max.
                                          :201.0
                                                    Max.
                                                           :4000
##
                                                    NA's
                                                           :1
```

```
summary(L)
```

```
##
         Length Class
                             Mode
   [1,] 10
                 -none-
                             numeric
   [2,]
          6
                 -none-
                             numeric
   [3,]
          4
                 data.frame list
   [4,]
          2
                 -none-
                             list
```

# 2 The apply Family of Functions

One of the most widely used features of R is the apply family of functions. The apply family consists of vectorized functions that minimize the need to use loops or repetitive code. We will cover the most common apply functions used in statistics, namely apply(), lapply(), sapply(), and tapply(). There are other functions in the same family (mapply(), rapply(), and vapply()), but these will not be covered.

### 2.1 The apply() Function

Suppose we want to compute the mean of each variable in the trees data frame. We could compute these individually, but it would require repetitive code (or a for loop).

```
mean(trees[,1]) # Or mean(trees$Girth)

## [1] 13.24839

mean(trees[,2]) # Or mean(trees$Height)

## [1] 76

mean(trees[,3]) # Or mean(trees$Volume)

## [1] 30.17097
```

For data sets with more than a few variables, using repetitive code is inefficient and cumbersome.

The apply() function is used to apply a function to the rows or columns (the margins) of matrices or data frames.

The syntax of apply() is apply(X,MARGIN,FUN,...), where the arguments are:

- X: A matrix or data frame
- MARGIN: A vector giving the subscript(s) over which the function will be applied over. A 1 indicates rows, 2 indicates columns, and c(1,2) indicates rows and columns.
- FUN: The function to be applied.
- ...: Any optional arguments to be passed to the FUN function (for example, na.rm=TRUE).

Using apply(), we can apply the mean() function to each column in trees simultaneously with a single command.

```
# Compute the mean of every column of the trees data frame
apply(trees,2,mean)

## Girth Height Volume
## 13.24839 76.00000 30.17097

To compute the mean of each row, we can change the margin argument MARGIN from 2 (columns) to 1 (rows).
# Compute the mean of every row of the trees data frame
```

```
apply(trees,1,mean)

## [1] 29.53333 27.96667 27.33333 32.96667 36.83333 37.83333 30.86667

## [8] 34.73333 37.90000 35.36667 38.16667 36.13333 36.26667 34.00000

## [15] 35.36667 36.36667 43.90000 42.23333 36.80000 34.23333 42.16667
```

## [22] 41.96667 41.60000 42.10000 45.30000 51.23333 51.73333 52.06667

## [29] 49.83333 49.66667 61.53333

Side Note: Technically, the mean of every row or column in a matrix or data frame can also be computed using the rowMeans() and colMeans() functions, but apply() works more generally, since apply() allows us to apply any function, not just mean().

Caution: Use caution when using apply() to a data frame. Ideally, the columns of the data frame should all be of the same type. The apply() function is intended for matrices (and arrays, which are higher dimensional versions of matrices). Using apply() on a data frame will first coerce the data frame into a matrix with as.matrix() before applying the function in the FUN argument.

Question: What does apply(parks.df,2,mean) output? Why does this command not give the results we intended? How can we find the mean of each of the numeric columns in parks.df using apply()?

Note: If the applied function in the FUN argument of apply() returns a single value, the output of the apply() function will be a vector. If the applied function returns a vector (with more than one element), then the output of apply() will be a matrix.

```
# Compute the range (min and max) of every column of the trees data frame apply(trees,2,range)
```

```
## Girth Height Volume
## [1,] 8.3 63 10.2
## [2,] 20.6 87 77.0
```

Question: How is summary(trees) different from apply(trees,2,summary)?

**Note**: The FUN argument does not have to be a built-in function. We can also write our own function and apply it to each row or column.

As an example, suppose we want to compute the squared deviations from the mean for each variable in the trees data frame.

```
squared.devs <- function(x){
    # This function inputs a vector and computes the squared deviations away from the mean.
    (x - mean(x))^2
}
# Apply the squared.devs() function to every column of the trees data frame
trees.sdevs <- apply(trees,2,squared.devs)
# Print the first few rows of the output
head(trees.sdevs)</pre>
```

```
##
            Girth Height
                           Volume
## [1,] 24.486535
                      36 394.8554
## [2,] 21.607503
                     121 394.8554
## [3,] 19.788148
                     169 398.8396
## [4,]
        7.553632
                      16 189.6396
                      25 129.2989
## [5,]
         6.494277
## [6,]
        5.994599
                      49 109.6412
```

The function can also be written directly into the FUN argument without having to save it as a separate object.

```
# Creates the same object as apply(trees,2,squared.devs)
trees.sdevs <- apply(trees,2,function(x){(x - mean(x))^2})
# Print the first few rows of the output
head(trees.sdevs)</pre>
```

```
##
            Girth Height
                            Volume
## [1,] 24.486535
                      36 394.8554
## [2,] 21.607503
                     121 394.8554
## [3,] 19.788148
                     169 398.8396
## [4,]
        7.553632
                      16 189.6396
## [5,]
         6.494277
                      25 129.2989
## [6,]
        5.994599
                      49 109.6412
```

#### 2.2 The lapply() Function

The lapply() function is used to apply a function to each component of a list (lapply is short for "list apply"). The output of lapply() will be a list.

The syntax of lapply() is lapply(X,FUN,...), where the arguments are:

• X: A list

## \$Volume ## [1] 11

- FUN: The function to be applied.
- ...: Any optional arguments to be passed to the FUN function.

Note that there is no margin argument like in apply(), as lists have a single index.

```
# Return the class of each component in the L list
lapply(L,class)
## [[1]]
## [1] "integer"
##
## [[2]]
## [1] "matrix"
## [[3]]
## [1] "data.frame"
##
## [[4]]
## [1] "list"
Note: Since data frames are (stored as) lists, lapply() also works for data frames.
# Compute the range (min and max) of every column of the trees data frame
lapply(trees,range)
## $Girth
## [1] 8.3 20.6
##
## $Height
## [1] 63 87
## $Volume
## [1] 10.2 77.0
Question: How is apply(trees,2,range) different from lapply(trees,range)?
The list output from lapply() is particularly useful when the result from each component may have a
different length (or even a different dimension or class).
which.median <- function(x){</pre>
  # This function inputs a vector and
  # returns the indices for values that attain the median.
  which(x == median(x))
lapply(trees, which.median)
## $Girth
## [1] 16 17
## $Height
## [1] 12 13
##
```

#### 2.3 The sapply() Function

The output that is returned from lapply() is always a list, with the same number of components as the input list. In many cases, the output could be simplified to a vector or matrix.

The sapply() function is a wrapper for lapply(), so sapply() also applies a function to each component of a list. The only difference is that sapply() will try to simplify the output from lapply() whenever possible (sapply is short for "simplified [l]apply"). In particular:

- If the result is a list where every component is a vector of length 1 (i.e., a scalar), then sapply() will return a vector.
- If the result is a list where every component is a vector of the same length (greater than 1), then sapply() will return a matrix.
- If the result is a list where every component is not a vector of the same length, then sapply() will return a list (i.e., the same output as from lapply().)

By using lapply(), we found the class of each component of list L. Notice the difference when using sapply().

```
sapply(L,class)
```

```
## [1] "integer" "matrix" "data.frame" "list"
```

The output of each application of the class() function is a single character value, so sapply() returns a vector.

Just like lapply(), sapply() also works for data frames.

```
sapply(trees,range)
```

```
## Girth Height Volume
## [1,] 8.3 63 10.2
## [2,] 20.6 87 77.0
```

Note: Notice that sapply(trees,range) gives the same output as apply(trees,2,range). Since a data frame is stored as a list with the column vectors as its components, sapply() applies functions to the components of trees as a list, and apply() with MARGIN=2 applies functions to the columns of the coerced matrix version of trees (as.matrix(trees)). The output is the same in this case. However, since lapply() and sapply() do not coerce data frames into having columns of the same type, certain functions may produce different results.

Question: How is apply(parks.df,2,mean) different from sapply(parks.df,mean)?

#### 2.4 The tapply() Function

The tapply() function is used to apply a function to subsets of a vector.

The syntax of tapply() is tapply(X,INDEX,FUN,...,simplify=TRUE), where the arguments are:

- X: A numeric or logical vector
- INDEX: A factor or list of factors that identifies the subsets. Non-factors will be coerced into factors.
- FUN: The function to be applied.
- ...: Any optional arguments to be passed to the FUN function.
- simplify: Logical value that specifies whether to simplify the output to a vector or matrix.

The tapply() function splits the values of the vector X into groups, each group corresponding to a level of the INDEX factor, then applies the function in FUN to each group.

As an example, we will consider the diamonds data in the ggplot2 package.

```
library(ggplot2) # Load the ggplot2 package
data(diamonds) # Load the diamonds data into the workspace
diamonds
```

```
## # A tibble: 53,940 × 10
##
      carat
                   cut color clarity depth table price
                                                               x
                                                                            z
##
      <dbl>
                 <ord> <ord>
                                <ord> <dbl> <int> <dbl>
                                                                 <dbl>
## 1
       0.23
                 Ideal
                            Ε
                                   SI2
                                        61.5
                                                 55
                                                      326
                                                            3.95
                                                                  3.98
                                                                         2.43
## 2
                            Ε
                                   SI1
                                        59.8
                                                      326
                                                            3.89
                                                                  3.84
       0.21
               Premium
                                                 61
                                                                         2.31
                                                            4.05
                                   VS1
## 3
       0.23
                  Good
                            Ε
                                        56.9
                                                 65
                                                      327
                                                                  4.07
                                                                         2.31
                            Ι
                                        62.4
                                                            4.20
## 4
       0.29
               Premium
                                   VS2
                                                 58
                                                      334
                                                                  4.23
                                                                         2.63
                                                                  4.35
## 5
       0.31
                  Good
                            J
                                   SI2
                                        63.3
                                                 58
                                                      335
                                                            4.34
                                                                         2.75
## 6
       0.24 Very Good
                            J
                                  VVS2
                                        62.8
                                                 57
                                                      336
                                                            3.94
                                                                  3.96
                                                                         2.48
## 7
       0.24 Very Good
                            Ι
                                  VVS1
                                        62.3
                                                 57
                                                                         2.47
                                                      336
                                                            3.95
                                                                  3.98
                            Η
## 8
       0.26 Very Good
                                   SI1
                                        61.9
                                                 55
                                                      337
                                                            4.07
                                                                  4.11
                                                                         2.53
                            Ε
## 9
       0.22
                                   VS2
                                        65.1
                  Fair
                                                 61
                                                       337
                                                            3.87
                                                                  3.78
                                                                         2.49
## 10 0.23 Very Good
                            Η
                                   VS1
                                        59.4
                                                 61
                                                      338
                                                            4.00
                                                                  4.05
                                                                         2.39
## # ... with 53,930 more rows
```

Side Note: The diamonds data is an example of a tibble object, which is a "trimmed down" version of a data frame. Tibbles are data frames with the added class tbl\_df that makes the way R prints the object more readable (e.g., typing diamonds did not print the entire data frame with 53940 rows). Since tibbles are still data frames, all syntax and functions for data frames work for tibbles, so we will treat them the same way. Tibbles are one of the main objects used in the tidyverse packages. Documentation for tibbles can be found in the tibble package.

Suppose we are interested in whether the mean price of a diamond differs by the color of the diamond. The tapply() function can split the prices based on the color rating and compute the mean of each subset.

```
# Compute the mean of price of diamonds, grouped by color
tapply(diamonds$price, diamonds$color, mean)
## D E F G H I J
```

From the output, we see that the mean price of diamonds is higher for higher color ratings.

## 3169.954 3076.752 3724.886 3999.136 4486.669 5091.875 5323.818

Suppose we want to know the mean price for each color/cut combination. The tapply() function can also group values based on combinations of levels from multiple factors. When using multiple factors in the INDEX argument, the factors need to be put into a list.

```
# Compute the mean price of diamonds for each color/cut combination
with(diamonds,tapply(price,list(color,cut),mean))
```

```
##
         Fair
                  Good Very Good Premium
                                              Ideal
## D 4291.061 3405.382
                        3470.467 3631.293 2629.095
## E 3682.312 3423.644
                        3214.652 3538.914 2597.550
                        3778.820 4324.890 3374.939
## F 3827.003 3495.750
                        3872.754 4500.742 3720.706
## G 4239.255 4123.482
## H 5135.683 4276.255
                        4535.390 5216.707 3889.335
## I 4685.446 5078.533
                        5255.880 5946.181 4451.970
## J 4975.655 4574.173
                        5103.513 6294.592 4918.186
```

Question: How would you find out how many observations are in each category (or combination of categories)?