CompStat/R - Paper 2

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Part I: Functions

Functions I

Below we define a function dropNa which, given an atomic vector x as an argument, returns x after removing missing values.

```
dropNa <- function(x) {
    # Expects an atomic vector as an argument and returns it without missing
    # values
    #
    # Args:
    # x: atomic vector
#

# Returns:
    # The atomic vector x without missing values

# To remove the NAs, we use simple logical subsetting
    y <- x[!is.na(x)]

# Return y
    y
}</pre>
```

Let's test our implementation with the following line of code:

```
all.equal(dropNa(c(1, 2, 3, NA, 1, 2, 3)), c(1, 2, 3, 1, 2, 3))
```

[1] TRUE

As we can see from this positive test, our implementation was successful.

Functions II

Part I

Below we define a function meanVarSdSe which, given a numeric vector \mathbf{x} as an argument, returns the mean, the variance, the standard deviation and the standard error of \mathbf{x} .

```
meanVarSdSe <- function(x) {
    # Expects a numeric vector as an argument and returns the mean,
    # the variance, the standard deviation and the standard error
    #
    # Args:
    # x: numeric vector</pre>
```

```
# Returns:
      a numerical vector containing mean, variance, standard deviation
     and standard error of x
  \# We check if x is numeric vector
  # If not: stop and throw error
  if( !is.numeric(x) ) {
    stop("Argument needs to be numeric.")
  # Create vector object
  y <- vector()
  # Calculate mean, variance, standard deviation and standard error
  # and save it in y
  y[1] \leftarrow mean(x)
  y[2] \leftarrow var(x)
  y[3] \leftarrow sd(x)
  y[4] \leftarrow y[3]/sqrt(length(x))
  # Set names to vector entries
  names(y) <- c("mean", "var", "sd", "se")</pre>
  # Return the numeric vector y
}
```

To test the function, we define a numeric vector, which contains numbers from 1 to 100, and use it as an argument for our function meanVarSdSe:

```
x <- 1:100
meanVarSdSe(x)

## mean var sd se</pre>
```

2.901149

Finally we can confirm that the result is of class numeric:

50.500000 841.666667 29.011492

```
class(meanVarSdSe(x))
```

```
## [1] "numeric"
```

Part II

Now we will have a look at the case below. We would expect that the function will return a vector with NAs:

```
x \leftarrow c(NA, 1:100)
meanVarSdSe(x)
```

```
## mean var sd se
## NA NA NA NA
```

The reason for the result is that the functions mean(), var() and sd() use na.rm = FALSE as default, which means that missing values are not removed. If the vector x contains a missing value, the mean() function (as well as var() and sd()) will just return NA to inform about missing values. In the case of calculating standard error we use the result from our sd() function and calculate an NA value with some other numeric values, which will ultimately result in NA again.

To solve the problem, we can add na.rm = TRUE to these three functions. To make this optional, we will improve the meanVarSdSe function from above as follows:

```
meanVarSdSe <- function(x, ...) {</pre>
  # Expects a numeric vector and flag to handle missing values as an argument
  # and returns the mean, the variance, the standard deviation
  # and the standard error
  # Args:
  #
      x: numeric vector, na.rm: boolean
  # Returns:
      a numerical vector containing mean, variance, standard deviation
      and standard error of x
  # We check if x is numeric vector
  # If not: stop and throw error
  if( !is.numeric(x) ) {
    stop("Argument needs to be numeric.")
  }
  # Create vector object
  y <- vector()
  # Calculate mean, variance, standard deviation and standard error
  # and save it in y
  y[1] \leftarrow mean(x, ...)
  y[2] \leftarrow var(x, ...)
  y[3] < - sd(x, ...)
  y[4] \leftarrow y[3]/sqrt(length(x) - sum(is.na(x)))
  # Set names to vector entries
  names(y) <- c("mean", "var", "sd", "se")</pre>
  # Return the numeric vector y
}
```

We define the function with an ellipse Our function can now receive multiple arguments after the first input x. These arguments are used in mean(), var() and sd(). If we want to remove missing values in all of these functions (to get a result in the case of missing values), we can pass na.rm = TRUE as another argument, such as here: meanVarSdSe(x, na.rm = TRUE). We just have to be aware of length(x) in this case. If we want to have the same result as above we have to remove the sum of NA values from the length of x. Otherwise the function will calculate a different result than in Part I, because then lentgh differs.

Let's confirm the result:

```
meanVarSdSe(c(x, NA), na.rm = TRUE)
```

```
## mean var sd se
## 50.500000 841.666667 29.011492 2.901149
```

Part III

Now we will use the function dropNa from Functions I to deal with missing values in meanVarSdSe.

```
meanVarSdSe <- function(x) {</pre>
  # Expects a numeric vector as an argument and returns the mean,
  # the variance, the standard deviation and the standard error
  \# it also removes missing values if x contains some
  # Arqs:
  #
     x: numeric vector
  # Returns:
  # a numerical vector containing mean, variance, standard deviation
    and standard error of x
  \# We check if x is numeric vector
  # If not: stop and throw error
  if( !is.numeric(x) ) {
    stop("Argument needs to be numeric.")
  }
  \# We check if x contains missing values
  # If so: remove missing values using dropNA
  if( sum(is.na(x)) > 0 ) {
    x <- dropNa(x)
  # Create vector object
  y <- vector()
  # Calculate mean, variance, standard deviation and standard error
  # and save it in y
  y[1] \leftarrow mean(x)
  y[2] \leftarrow var(x)
 y[3] \leftarrow sd(x)
 y[4] \leftarrow y[3]/sqrt(length(x))
  # Set names to vector entries
  names(y) <- c("mean", "var", "sd", "se")</pre>
  # Return the numeric vector y
  У
}
```

We used the function from Part I and added a condition which checks if we have missing values in x, using is.na. If the sum of NA values is greater than 0 (i.e., if there is one or more missing value), we use the function

dropNA from the first exercise to remove all missing values. The remaining code of the function can remain as above in Part I.

We can confirm the result:

```
meanVarSdSe(c(x, NA))

## mean var sd se
## 50.500000 841.666667 29.011492 2.901149
```

Functions III

In this section we define an infix function %or%. This function should behave like the logical operator |.

First we check if we have logical vectors. If a and/or b are not logical, we leave the function and throw an error. Otherwise we can calculate the or operation using the ifelse function and return the result directly after calculation. Inside of the ifelse function, the first argument checks the condition if the sum of the values a and b are greater than or equal to 1, where TRUE is equal to 1 and FALSE is equal to 0.

To confirm the function, we test an example:

```
c(TRUE, FALSE, TRUE, FALSE) %or% c(TRUE, TRUE, FALSE, FALSE)
```

Part II: Scoping and related topics

TRUE FALSE

TRUE

Scoping I

[1]

TRUE

The main concept behind this exercise is that of the Search Path, which R uses to locate objects when called upon. In order for R to carry out a command or calculation, it seeks the necessary information according to a hierarchical path of "environments". Each environment has a "parent" environment, to which R moves if the required information is not yet found. The R workspace is known as the Global Environment and also has a parent, which is either the most recently loaded package or the base package (package:base), also known as the "end" of the search path.

Below we can observe the importance of the search path with a simple example:

```
# We assign numeric values to the vectors x and y in the workspace
# which we call the global environment
x <- 5
y <- 7

f <- function() x * y
    # With no specified argument inputs, the function f follows the search path
    # and locates values for x and y in the global environment
g <- function(x = 2, y = x) x * y
    # A new environment is created within the function g, where arguments for x and y
# are clearly defined</pre>
```

Although both functions f and g depend on values for x and y, they return different results when called:

```
f()
## [1] 35
g()
```

[1] 4

Beginning with function f, if we follow the search path we begin in the environment within the function itself. Since there is no information regarding the values of x and y, R moves to the parent environment, which is the global environment in this case. In the global environment, x takes the value of 5 and y takes the value of 7. Thus, the function returns 5*7=35.

For function g the search path also begins in the environment within the function itself. However, in this case there is a defined value for x, as well as an expression defining a value for y based on x. The search path ends and the function returns 2 * 2 = 4.

By manipulating the arguments of a function, it is also possible to alter the original search path. We see this when calling the following function:

```
g(y = x)
```

[1] 10

Looking back at the code for function g, we see the two arguments x and y. When calling g(y = x) however, we are ignoring the first argument, which then defaults to the value 2. When we simply call g(), the y = x argument also defaults to a value dependent on x. But by inputing y = x manually, we send the search path to the global environment where x takes the value of 5. Thus the function returns 2*5=10.

Scoping II

Scoping III

Dynamic lookup