problem_set_4a.R

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```
# PROBLEM SET 4a
# This assignment is done in a group of three
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#Part.1
#Question 1
#1
calculate_na_proportion <- function(x) {</pre>
  mean(is.na(x))
# This function calculated the proportion of NA values in {\it x}
calculate_na_proportion(c(10,20,NA,40,50,NA,60,70))
## [1] 0.25
standardize_vector <- function(x,na.rm = TRUE ) {</pre>
  x / sum(x, na.rm = na.rm)
#This function standardize the vector. It returns the normalize version where each element
#is divided by the sum of all the elements in the vector so that it the sums to one.
#The na.rm is used to handle the NA elements in the vector.
#If na.rm = FALSE, it returns NA and if na.rm = TRUE, ir drops NA
#when na.rm = TRUE
standardize_vector(c(10,20,NA,40,50,NA,60,70))
## [1] 0.04 0.08
                   NA 0.16 0.20
                                 NA 0.24 0.28
#when na.rm = FALSE
standardize_vector(c(10,20,NA,40,50,NA,60,70),na.rm = FALSE)
## [1] NA NA NA NA NA NA NA
calculate_coefficient_of_variation <- function(x) {</pre>
  sd(x, na.rm = TRUE) / mean(x, na.rm = TRUE)
#This function calculates the coefficient of variation by dividing standard deviation with mean.
#Here both mean and standard deviation are calculated by dropping NA values since na.rm = TRUE
calculate_coefficient_of_variation(c(10,20,NA,40,50,NA,60,70))
```

```
both_na <- function(x,y) {
 if(length(x) == length(y)) {
   result <- is.na(x) & is.na(y)</pre>
 return(result)
both_na(c(10,20,NA,40), c(NA,60,NA,NA))
## [1] FALSE FALSE TRUE FALSE
is_directory <- function(x) file.info(x)$isdir</pre>
is_readable <- function(x) file.access(x, 4) == 0</pre>
#The function is_directory checks whether the path in x is a directory.
#t uses the file.info function to retrieve file information and then extracts the "isdir" field from th
#The function is_readqble checks whether the file is readable which means
#that the file in the given path exists and the user has permission to open it.
#These two functions are useful while accessing the files, to check for the correct directory and
#user permissions to access the file.
#4
f1 <- function(string, prefix) {</pre>
 substr(string, 1, nchar(prefix)) == prefix
#This function checks if each element in the vector statrs with the given prefix.
#the fucntion can be renamed as starts_with_prefix
starts_with_prefix <- function(string, prefix) {</pre>
 substr(string, 1, nchar(prefix)) == prefix
starts_with_prefix(c("abort", "adapt", "absorb", "abroad", "append"), "ab")
## [1] TRUE FALSE TRUE TRUE FALSE
f2 <- function(x) {
 if (length(x) <= 1) return(NULL)</pre>
 x[-length(x)]
#This function drops the last element of the vector. If the vector has only one element or if the vecto
#empty, it returns NULL.the fucntion can be renamed as remove_last_element
remove_last_element <- function(x) {</pre>
 if (length(x) <= 1) return(NULL)</pre>
 x[-length(x)]
}
remove_last_element(c(1,2,3,4,5))
```

```
## [1] 1 2 3 4
remove_last_element(c(1))
## NULL
remove_last_element(c())
## NULL
f3 <- function(x, y) {
 rep(y, length.out = length(x))
\#This function repeats the element y element in the vector of length x.
#the fucntion can be renamed as repeat_element
repeat_element <- function(x, y) {</pre>
 rep(y, length.out = length(x))
}
repeat_element(c(1,2,3,4,5),3)
## [1] 3 3 3 3 3
#5
fizzbuzz <- function(x) {</pre>
 if((x \% 3 == 0) \&\& (x \% 5 == 0)){
   "fizzbuzz"
 else if (x %% 3 == 0){
   "fizz"
 else if (x \% 5 == 0){
   "buzz"
 }else {
   Х
fizzbuzz(15)
## [1] "fizzbuzz"
fizzbuzz(3)
## [1] "fizz"
fizzbuzz(10)
## [1] "buzz"
```

```
fizzbuzz(11)
## [1] 11
#6
temp <- 40
cut(temp, breaks = c(-Inf, 5, 10, 20, 30, Inf),
                     labels = c("freezing", "cold", "cool", "warm", "hot"),
                     right = TRUE)
## [1] hot
## Levels: freezing cold cool warm hot
#To use '<' instead of '<=' while working with cut, we can change right = FALSE
#chief advantage of using cut is that it can handle multiple values which means that it can work with v
#which cannot happen while using if statement which takes single value. Also the handling the intervals
#temperature values, we need to change the operators in case of if statement whereas with cut, we just
#change right = False.
#using cut also makes the code simple and readable.
x <-'e'
switch_out <- switch(x,</pre>
                 b = "ab"
                c = ,
                 d = "cd"
)
switch_out
## NULL
#The switch function returns the first non missing values it encounters while matching the arguments.
#According to the above code, it will return 'ab' for 'a', 'ab' for 'b', 'cd' for 'c', 'cd' for 'd.'
#If given 'e', it returns NULL since 'e' is not there in the given arguments.
#Question 2
#1
#is.vector() checks if it is a specific type of vector with no attributes other than names. That is
#is.vector() checks for a specific type of vector as defined by the specified mode and imposes
#the constraint that the vector should have no attributes other than names.
#is.atomic() specifically checks if an object belongs to the atomic modes: "logical", "integer",
#"numeric" (synonym "double"), "complex", "character", or "raw". If an object is of any of these
#modes and has no attributes other than names, is.atomic() returns TRUE.
#2
x \leftarrow c(5,10,15,20,25,30)
#returns last element
```

last_ele <- function(x) {</pre>

```
x[length(x)]
last_ele(x)
## [1] 30
#elements at even numbered positions
even_positions <- function(x){</pre>
 x[seq(from = 2, length(x), by = 2)]
even_positions(x)
## [1] 10 20 30
#Every element except the last value.
except_last_value <- function(x){</pre>
 x[1:length(x)-1]
except_last_value(x)
## [1] 5 10 15 20 25
#Only even numbers
even_numbers <- function(x){</pre>
 x[x \% 2 == 0]
even_numbers(x)
## [1] 10 20 30
#3
\#x[-which(x > 0)] and x[x \le 0] gives similar results. However, the difference comes while handling nul
\#x[-which(x > 0)] will ignore `NA`'s and leave them as it is and
# x[x \le 0] returns any value that cannot be comparable as NA
x \leftarrow c(1,2,3,4,5,NaN,NA)
x[-which(x > 0)]
## [1] NaN NA
x[x \le 0]
## [1] NA NA
#4
#when we subset with a positive integer that's bigger than the length of the vector, It returns NA
x \leftarrow c(2,4,6,8)
x[5]
## [1] NA
```

```
#when we subset with a name that doesn't exist, still it returns NA
x \leftarrow c(a = 1, b = 2, c = 3)
x["d"]
## <NA>
##
   NA
res = list('a', 'b', list('c', 'd'), list('e', 'f'))
res1 = list(list(list(list(list('a'))))))
#Question 3
#1
#he mean of every column in mtcars
library(tidyverse)
## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.4.0 v purrr 0.3.5
## v tibble 3.1.8
                  v dplyr 1.0.10
## v tidyr 1.2.1
                  v stringr 1.5.0
## v readr 2.1.3 v forcats 0.5.2
## Warning: package 'ggplot2' was built under R version 4.2.2
## Warning: package 'stringr' was built under R version 4.2.2
## -- Conflicts ----- tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
mtcars_means <- vector("double", ncol(mtcars))</pre>
names(mtcars_means) = names(mtcars)
for (i in names(mtcars)) {
 mtcars_means[i] <- mean(mtcars[, i])</pre>
mtcars_means
                                           drat
                 cyl
                         disp
                                   hp
                                                      wt
                                                             qsec
        mpg
## 20.090625 6.187500 230.721875 146.687500
                                        3.596563 3.217250 17.848750
##
        VS
                         gear
                                  carb
                 am
  0.437500 0.406250 3.687500 2.812500
#type of each column in nycflights13::flights
library(nycflights13)
flights_types <- vector("list", ncol(flights))</pre>
names(flights_types) <- names(flights)</pre>
for (i in names(flights)){
 flights_types[[i]] <- class(flights[[i]])</pre>
flights_types
```

```
## $year
## [1] "integer"
##
## $month
## [1] "integer"
##
## $day
## [1] "integer"
##
## $dep_time
## [1] "integer"
## $sched_dep_time
## [1] "integer"
##
## $dep_delay
## [1] "numeric"
##
## $arr_time
## [1] "integer"
##
## $sched_arr_time
## [1] "integer"
## $arr_delay
## [1] "numeric"
##
## $carrier
## [1] "character"
##
## $flight
## [1] "integer"
##
## $tailnum
## [1] "character"
## $origin
## [1] "character"
##
## $dest
## [1] "character"
##
## $air_time
## [1] "numeric"
## $distance
## [1] "numeric"
##
## $hour
## [1] "numeric"
##
## $minute
## [1] "numeric"
##
```

```
## $time hour
## [1] "POSIXct" "POSIXt"
#number of unique values in each column of iris
data(iris)
iris_unique_counts <- vector("double", ncol(iris))</pre>
names(iris_unique_counts) <- names(iris)</pre>
for (i in names(iris)) {
  iris_unique_counts[i] <- length(unique(iris[[i]]))</pre>
iris_unique_counts
## Sepal.Length Sepal.Width Petal.Length Petal.Width
                                                         Species
                        23
                                     43
#10 random normals for each of mu = -10, 0, 10, and 100
mu_values \leftarrow c(-10, 0, 10, 100)
random_normals <- vector("list", length(mu_values))</pre>
for (i in seq_along(random_normals)) {
 random_normals[[i]] <- rnorm(10, mean = mu_values[i])</pre>
random_normals
## [[1]]
   [1] -11.924662 -10.755108 -8.877296 -10.201195 -8.625040 -10.027645
## [7] -9.088611 -9.266832 -10.476637 -10.988767
## [[2]]
## [1] -0.8192203 1.7567165 -0.7367987 0.2884779 0.5098887 0.6157295
## [7] 0.1437723 0.6244447 0.9215035 0.3620429
##
## [[3]]
## [1] 10.426648 13.674756 8.071447 9.062279 10.391058 10.824349 9.531927
## [8] 9.741673 9.575444 9.475149
##
## [[4]]
## [1] 101.87441 100.79778 100.43538 100.44907 99.49232 98.61299 99.07646
## [8] 100.86896 100.80077 100.24302
#2
out <- ""
for (x in letters) {
  out <- str_c(out, x)
}
out
## [1] "abcdefghijklmnopqrstuvwxyz"
\#str\_c works with vectors. Hence we can use str\_c() with the collapse to return single argument
str_c(letters, collapse = "")
```

```
## [1] "abcdefghijklmnopqrstuvwxyz"
```

```
x \leftarrow sample(100)
sd <- 0
for (i in seq_along(x)) {
 sd \leftarrow sd + (x[i] - mean(x))^2
}
sd <- sqrt(sd / (length(x) - 1))</pre>
## [1] 29.01149
#we have inbuilt function for standard deviation and we can use that
sd(x)
## [1] 29.01149
x \leftarrow runif(100)
out <- vector("numeric", length(x))</pre>
out [1] \leftarrow x[1]
for (i in 2:length(x)) {
  out[i] \leftarrow out[i-1] + x[i]
}
out
##
     [1] \quad 0.6356822 \quad 0.6516984 \quad 1.0968228 \quad 1.7936585 \quad 2.5832804 \quad 2.8784087
##
     [7] 3.3058126 4.0942250 4.5260372 5.4884799 6.3560634 7.2616213
## [13] 7.9767410 8.2520153 8.6657439 8.7889136 9.5864055 10.0293412
##
    [19] 10.4938538 11.3910198 12.3187792 12.3375007 12.7834097 13.7641795
## [25] 14.4773917 14.6695184 15.6598150 16.4199068 16.8991088 17.6548045
## [31] 18.2766297 18.3009445 18.7838471 19.7821069 20.7311701 21.4798825
## [37] 21.9937426 22.5850064 23.4521036 23.8622832 24.1271329 25.1165413
   [43] 26.0940669 26.4197936 26.5224378 26.7833610 27.0767097 27.5342914
## [49] 28.5001404 29.0139957 30.0099129 30.9274816 31.1272090 31.2412186
## [55] 32.1326147 32.8127133 33.4492759 33.9806792 34.4748673 35.3848277
## [61] 36.0794068 36.5693499 36.6946817 36.7092654 36.7213446 36.7372823
## [67] 37.1091941 37.6289677 38.2858735 38.8395104 39.2138672 39.4605764
## [73] 39.8628801 40.7758210 41.7616143 42.4265713 42.9014504 43.7740011
## [79] 44.2919272 44.7406753 45.0949241 45.7056694 46.4610271 46.8067474
## [85] 47.7445954 48.0877819 48.7200890 48.8653614 49.3034205 49.9143285
   [91] 49.9645368 50.3482669 50.7922232 50.8350444 51.3113823 51.6528141
## [97] 52.6344343 52.8019273 53.0151463 53.5625808
#The code is calculating cumulative sum which can be done using cumsum()
all.equal(cumsum(x), out)
```

[1] TRUE

```
for (i in seq_along(x)) {
 output <- c(output, lengths(x[[i]]))</pre>
}
output
##
    ##
   #To discuss the performance effect, we are definig two functions and use the microbenchmark package
#to compare the time
add_to_vector <- function(n) {</pre>
 output <- vector("integer", 0)</pre>
 for (i in seq_len(n)) {
   output <- c(output, i)</pre>
 output
}
add_to_vector_2 <- function(n) {</pre>
 output <- vector("integer", n)</pre>
 for (i in seq_len(n)) {
   output[[i]] <- i</pre>
 output
library(microbenchmark)
## Warning: package 'microbenchmark' was built under R version 4.2.3
timings <- microbenchmark(add_to_vector(10000), add_to_vector_2(10000), times = 3)</pre>
timings
## Unit: microseconds
##
                 expr
                        min
                                       mean
                                             median
                                 lq
                                                       uq
    add to vector(10000) 99724.5 103697.55 105769.4 107670.6 108791.9 109913.2
##
  add_to_vector_2(10000)
                       207.6
                              212.85
                                      986.8
                                              218.1
                                                    1376.4
##
  neval
      3
##
      3
##
#From the results we can say that the pre-allocated vector performs much faster. However, the longer th
#and the bigger the objects, the more that pre-allocation will outperform appending
#Part2
```

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#allowing for flexibility and customization.

#which makes beneficial while working in teams.

#RMarkdown is widely appreciated for its ability to integrate code, text, and visualizations in a singl #I find it valuable for creating reproducible and dynamic reports, especially in statistical modeling p

#Once missing feature which I came across is that the feacture facilitating collaborative writing and e

#I also appreciate the ability to render the document in various formats (e.g., HTML, Word, PDF)