Regular Expressions based programs

Yi Chua

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A small project creating software programs found in everyday life using Regular Expression to match character combinations in strings!

```
library(stringr)
```

Counting Vowels

```
#' @title Counting Vowels
#' @description computes the number of vowels in a character string
#' @param string (character)
#' @return a vector count of the number of vowels
count_vowels <- function(string){</pre>
  extract <- str_extract_all(str_to_lower(string), regex('[aeiou]'))</pre>
 list_vowels <- c('a','e','i','o','u')
  count <- rep(0,length(list_vowels))</pre>
 for (v in (1:length(list_vowels))){
    cum_sum <- cumsum(extract[[1]] == list_vowels[v])</pre>
    count[v] <- cum_sum[length(cum_sum)]</pre>
 names(count) <- list_vowels</pre>
 return(count)
}
#test cases
count_vowels("FIAT LUX")
## a e i o u
## 1 0 1 0 1
count_vowels("MaY tHe FoRce Be wIth yOu")
## a e i o u
## 1 3 1 2 1
count_vowels("a fox jumps over the lazy dog")
## a e i o u
## 2 2 0 3 1
```

Counting Consonants

```
#' @title Counting Consonants
#' @description computes the number of consonants in a character string
#' @param string (character)
#' Oreturn table count of the number of consonants (vector)
count_consonants <- function(string){</pre>
  extract <- str_extract_all(str_to_lower(string), regex('[^aeiou]'))</pre>
 list_consonants <- c('b','c','d','f','g','h','j','k','l','m','n','p',
                        'q','r','s','t','v','w','x','y','z')
 count <- rep(0, length(list_consonants))</pre>
 for (c in (1:length(list consonants))){
   cum_sum <- cumsum(extract[[1]] == list_consonants[c])</pre>
   count[c] <- cum_sum[length(cum_sum)]</pre>
 names(count) <- list_consonants</pre>
 return(count)
#test cases
count_consonants("FIAT LUX")
## b c d f g h j k l m n p q r s t v w x y z
## 0 0 0 1 0 0 0 0 1 0 0 0 0 0 0 1 0 0 1 0 0
count_consonants("MaY tHe FoRce Be wIth yOu")
## b c d f g h j k l m n p q r s t v w x y z
## 1 1 0 1 0 2 0 0 0 1 0 0 0 1 0 2 0 1 0 2 0
count_consonants("a fox jumps over the lazy dog")
## bcdfghjklmnpqrstvwxyz
## 0 0 1 1 1 1 1 0 1 1 0 1 0 1 1 1 1 0 1 1 1
```

Hex verifier

```
#' @title Hexadecimal checker
#' @description checks whether the string is a valid color in hexadecimal notation
#' @param string (character)
#' @return boolean value (Logic)
is_hex <- function(string){</pre>
  return(str_detect(str_to_upper(string), regex('^#[A-F0-9]{6}$')))
#test case
is_hex("#ff0000")
## [1] TRUE
is_hex("#123456")
## [1] TRUE
is_hex("#12Fb56")
## [1] TRUE
is_hex("#1234GF")
## [1] FALSE
is_hex("#1234567")
## [1] FALSE
is_hex("blue")
## [1] FALSE
#' @title Hexadecimal w/ Alpha Transparency checker
#' @description checks whether the string is a valid hex color with alpha transparency
#' @param string (character)
#' @return boolean value (Logic)
is_hex_alpha <- function(string){</pre>
 return(str_detect(str_to_upper(string), regex('^#[A-F0-9]{8}$')))
}
#test cases
is_hex_alpha("#FF000078")
```

[1] TRUE

is_hex_alpha("#FF0000")

[1] FALSE

Hexadecimal-RGB Converter

```
#' @title Hexadecimal to RGB Values
#' @description takes a hex-color and returns a named vector with the values of the RGB as well as the
#' @param string (character)
#' @return RGB values (vector)
hex_values <- function(string){</pre>
 list_rgba <- c("red", "green", "blue", "alpha") #will be used to label the vector with the values of
 rgb <- rep("0",3)
  if (!is_hex(string) & !is_hex_alpha(string)){
  stop("\ninput is not a valid hexadecimal color")
  } else{
   # set pattern depending if it is hex with alpha or without alpha
   if(is_hex_alpha(string)){
   } else{
     pattern <- '^#([A-F0-9]{2})([A-Fa-f0-9]{2})([A-F0-9]{2})$' #pattern for just hex
   match_group <- str_match(string,regex(pattern)) #pattern matched</pre>
   for (m in 2:(length(match_group))){
     rgb[m-1] <- match_group[m]</pre>
     names(rgb)[m-1] <- list_rgba[m-1]</pre>
 }
 return(rgb)
#test cases
hex values("#12345655")
   red green blue alpha
## "12" "34" "56" "55"
hex_values("#12Fb56")
   red green blue
## "12" "Fb" "56"
hex_values("#1234GF")
## Error in hex_values("#1234GF"):
## input is not a valid hexadecimal color
hex values("#1234567")
## Error in hex_values("#1234567"):
## input is not a valid hexadecimal color
```

hex_values("blue")

```
## Error in hex_values("blue"):
## input is not a valid hexadecimal color
```

Password verifier

Checks if the string is a "valid" password

We consider a string to be a valid password if it:

- Has a minimum of 8 characters
- Contains at least 1 lowercase letter
- Contains at least 1 uppercase letter
- Contains at least 1 number
- Contains at least 1 of the permitted special characters -!@#\$%^&*
- Should not contain pattern hello or 123

```
#' @title Password Checker
#' @description checks if a string is a 'valid' password
#' @param string (character)
#' @return boolean value (logic)
check_pswd <- function(passwords){
return(!(str_detect(passwords, regex("hello"))) & !str_detect(passwords, regex("123")) &str_detect(passwords)}

passwords <- c("Hello12!", "Hell@125!", "Hel7o123", "Hel7o123!", "L@R!6373", "Fanny*1", "L@ttie#@*!", "]
check_pswd(passwords)</pre>
```

[1] TRUE TRUE FALSE FALSE FALSE FALSE FALSE

Currency converter

```
#' @title Currency converter
#' @description converts from one currency to another
#' @param amount numeric value (double)
#' @param from type of currency: euro, pound, yen, yuan, rupee, peso, bitcoin (character)
#' @param to type of currency: euro, pound, yen, yuan, rupee, peso, bitcoin (character)
#' Creturn the amount based on the type of currency one desired (double)
exchange <- function(amount = 10, from = 'us', to = 'euro'){</pre>
  conversion_table <- c("us" = 1,"euro" = 0.87, "pound" = 0.68, "yen" = 106.43, "yuan" = 6.47, "rupee"
  conversion_table
  if ((!(str_extract(from, regex("\\w+"))) %in% names(conversion_table)) | !(str_extract(to, regex("\\w
    stop("currency type must be the following:\nus, euro, pound, yen, yuan, rupee, peso, or bitcoin")
  if (amount < 0){</pre>
    stop("\n amount must be nonnegative")
  } else {
    #extract the numerical values that will be used as the currency exchange rate
   f = conversion_table[which(names(conversion_table) == (str_extract(from, regex("\\w+"))))]
   t = conversion_table[which(names(conversion_table) == (str_extract(to, regex("\\w+"))))]
    converted <- (t / f) * amount</pre>
   names(converted) <- NULL</pre>
   return(converted)
 }
}
#test cases
exchange(amount = 1, from = 'us', to = 'euro')
## [1] 0.87
exchange(amount = 10, from = 'us', to = 'euro')
## [1] 8.7
exchange(amount = 10, from = 'us', to = 'bitcoin')
## [1] 0.022
exchange(amount = 10, from = 'us', to = 'pound')
## [1] 6.8
#negative amount
exchange(amount = -50, from = 'us', to = 'euro')
## Error in exchange(amount = -50, from = "us", to = "euro"):
## amount must be nonnegative
```

```
#incorrect currency type/nonexistent
exchange(amount = 10, from = 'us', to = 'ringgit')
```

Error in exchange(amount = 10, from = "us", to = "ringgit"): currency type must be the following:
us, euro, pound, yen, yuan, rupee, peso, or bitcoin

Dataframe Cleaner

6

7.9

```
#load messy data csv file
dat <- read.csv('USA_crime.csv', header = FALSE, stringsAsFactors = FALSE)</pre>
##
## 1 Alabama >>+13.2&&236$^58
     Alaska??10.0 263>
## 3 Arizona+ 8.1:; 294
## 4 Arkansas- 8.8* 190 50
## 5 California? 9.0 276= 91
     Colorado 7.9= 204> 78
#' @title Data cleaner
#' @description cleans messy data and return a clean data frame containing information
#' Oparam df takes in a data frame (data frame)
#' @return a cleaned up data frame (data frame)
clean_data <- function(df){</pre>
 matched <- str_match(df[[1]],regex(pattern))</pre>
 #create empty matrix
 clean_df <- matrix(data= NA, nrow =nrow(matched), ncol = ncol(matched) - 1)</pre>
 for (i in 1:nrow(matched)){
   row_df <- matched[i,]</pre>
   for (j in 2:ncol(matched)){
     clean_df[i,j-1] <- row_df[j]</pre>
   }
 }
 #convert matrix to data frame
 clean_df <- as.data.frame(clean_df,header = FALSE, stringsAsFactors = FALSE)</pre>
 names(clean_df) <- c("State", "Murder", "Assault", "UrbanPop")</pre>
 return(clean_df)
clean_data(dat)
##
         State Murder Assault UrbanPop
## 1
       Alabama
                13.2
                         236
                                   58
## 2
        Alaska
                10.0
                         263
                                   48
                         294
## 3
       Arizona
                 8.1
                                   80
                                   50
## 4
      Arkansas
                 8.8
                         190
                         276
                                   91
## 5 California
               9.0
      Colorado
                         204
                                   78
```