Week 2 Exercises

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March 26, 2023

Please complete all exercises below. You may use stringr, lubridate, or the forcats library.

Place this at the top of your script: library(stringr) library(lubridate) library(forcats)

Exercise 1

Read the sales_pipe.txt file into an R data frame as sales.

Exercise 2

You can extract a vector of columns names from a data frame using the colnames() function. Notice the first column has some odd characters. Change the column name for the FIRST column in the sales date frame to Row.ID.

Note: You will need to assign the first element of colnames to a single character.

```
#rename column 1 header to Row.ID
colnames(sales_pipe_df)[1] <- "Row.ID"</pre>
```

Exercise 3

Convert both Order.ID and Order.Date to date vectors within the sales data frame. What is the number of days between the most recent order and the oldest order? How many years is that? How many weeks?

Note: Use lubridate

```
#convert Order.Date to vector of Dates
sales_pipe_df$Order.Date.Format <- mdy(sales_pipe_df$Order.Date)</pre>
#create variables for minimum and maximum Order dates
order_max_date <- max(sales_pipe_df$Order.Date.Format)</pre>
order_min_date <- min(sales_pipe_df$Order.Date.Format)</pre>
#then subtract min from max to get the difference between the two dates
difference_dates <- order_max_date - order_min_date</pre>
#then convert date difference to number of days, save in variable and print.
#Repeat for weeks and years
days_between_all_orders<- time_length(difference_dates, "day")</pre>
days_between_all_orders
## [1] 1457
weeks_between_all_orders <- time_length(difference_dates, "week")</pre>
weeks_between_all_orders
## [1] 208.1429
years_between_all_orders <- time_length(difference_dates, "year")</pre>
years_between_all_orders
## [1] 3.989049
```

Exercise 4

What is the average number of days it takes to ship an order?

[1] 3.908482

Exercise 5

How many customers have the first name Bill? You will need to split the customer name into first and last name segments and then use a regular expression to match the first name bill. Use the length() function to determine the number of customers with the first name Bill in the sales data.

```
#split first and last name column into 2 columns, then paste them to exisitng
#df and rename them.
customer_name <- str_split_fixed(string=sales_pipe_df$Customer.Name, pattern=" ", n=2)
sales_pipe_df$Customer.First.Name <- paste(customer_name[,1])</pre>
```

```
sales_pipe_df$Customer.Last.Name <- paste(customer_name[,2])

#Then use regex to determine if first name is Bill.
#Then use logical indexing to find all the "Bill"s then use
#length to determine the number of elements in the vector.
matched_names <- str_match(sales_pipe_df$Customer.First.Name, pattern= "Bill")
length(which(matched_names == "Bill"))</pre>
```

[1] 37

Exercise 6

How many mentions of the word 'table' are there in the Product.Name column? Note you can do this in one line of code

```
# find "table" in column using reg expression matching, and take length of
# which matches
length(which(str_match(sales_pipe_df$Product.Name, pattern = "table")== "table"))
## [1] 197
```

Exercise 7

##

Create a table of counts for each state in the sales data. The counts table should be ordered alphabetically from A to Z.

```
#create a table of states and their counts, and sort alphabetically by name
table(sort(sales_pipe_df$State, decreasing = FALSE))
```

Alabama	Arizona	Arkansas
28	119	22
California	Colorado	Connecticut
993	90	50
Delaware	District of Columbia	Florida
47	1	186
Georgia	Idaho	Illinois
79	9	286
Indiana	Iowa	Kansas
74	11	16
Kentucky	Louisiana	Maine
64	18	4
Maryland	Massachusetts	Michigan
63	71	142
Minnesota	Mississippi	Missouri
41	27	37
Montana	Nebraska	Nevada
2	26	24
New Hampshire	New Jersey	New Mexico
9	58	11
New York	North Carolina	North Dakota
555	117	7
Ohio	Oklahoma	Oregon
211	38	56
	California 993 Delaware 47 Georgia 79 Indiana 74 Kentucky 64 Maryland 63 Minnesota 41 Montana 2 New Hampshire 9 New York 555 Ohio	28 119 California Colorado 993 90 Delaware District of Columbia 47 1 Georgia Idaho 79 9 Indiana Iowa 74 11 Kentucky Louisiana 64 18 Maryland Massachusetts 63 71 Minnesota Mississippi 41 27 Montana Nebraska 2 26 New Hampshire New Jersey 9 58 New York North Carolina 555 117 Ohio Oklahoma

##	Pennsylvania	Rhode Island	South Carolina
##	312	25	28
##	South Dakota	Tennessee	Texas
##	9	88	460
##	Utah	Vermont	Virginia
##	27	10	80
##	Washington	West Virginia	Wisconsin
##	254	4	38
##	Wyoming		
##	1		

Exercise 8

Create an alphabetically ordered barplot for each sales Category in the State of Texas.

```
#create a df for sales in Texas, then a table for the categories listed
#alphabetically, and a barplot of the table
sales_catergoies_texas_df <- sales_pipe_df[(sales_pipe_df$State== "Texas") ,]
barplot(table(sort(sales_catergoies_texas_df$Category, decreasing = FALSE)))</pre>
```



Exercise 9

Find the average profit by region. Note: You will need to use the aggregate() function to do this. To understand how the function works type ?aggregate in the console.

```
#aggregate the Profit column, listing by Region, using the mean function
aggregate(sales_pipe_df$Profit, list(sales_pipe_df$Region), FUN=mean)
```

```
## Group.1 x
## 1 Central 20.46822
## 2 East 29.91937
## 3 South 11.27720
## 4 West 32.77000
```

Exercise 10

Find the average profit by order year. Note: You will need to use the aggregate() function to do this. To understand how the function works type ?aggregate in the console.

```
#extract year from order date and add year to df as new column
Order.Year <- str_split_fixed(string=sales_pipe_df$Order.Date.Format, pattern="-", n=3)
sales_pipe_df$Order.Year <- paste(Order.Year[,1])
#then aggregate profits by year, using the mean function
aggregate(sales_pipe_df$Profit, list(sales_pipe_df$Order.Year), FUN=mean)</pre>
```

```
## Group.1 x
## 1 2014 32.24582
## 2 2015 21.58676
## 3 2016 30.10960
## 4 2017 21.31825
```