

Challenge-4

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Questions

Load the “CommQuest2023.csv” dataset using the `read_csv()` command and assign it to a variable named “comm_data.”

```
# Enter code here
library(tidyverse)

## — Attaching core tidyverse packages — tidyverse
## 2.0.0 —
## ✓ dplyr      1.1.2      ✓ readr      2.1.4
## ✓ forcats    1.0.0      ✓ stringr    1.5.0
## ✓ ggplot2     3.4.3      ✓ tibble     3.2.1
## ✓ lubridate  1.9.2      ✓ tidyr      1.3.0
## ✓ purrr      1.0.2
## — Conflicts —
tidyverse_conflicts() —
## ✗ dplyr::filter() masks stats::filter()
## ✗ dplyr::lag()     masks stats::lag()
## ⓘ Use the conflicted package (<http://conflicted.r-lib.org/>) to force
all conflicts to become errors

comm_data <- read.csv("CommQuest2023_Larger.csv")
```

Question-1: Communication Chronicles

Using the `select` command, create a new dataframe containing only the “date,” “channel,” and “message” columns from the “comm_data” dataset.

Solution:

```
# Enter code here
select(comm_data, date, channel, message)
```

Question-2: Channel Selection

Use the filter command to create a new dataframe that includes messages sent through the “Twitter” channel on August 2nd.

Solution:

```
# Enter code here
comm_data %>%
  filter(channel == "Twitter" , date == "2023-08-02") %>%
  select(date,channel)

##           date channel
## 1  2023-08-02 Twitter
## 2  2023-08-02 Twitter
## 3  2023-08-02 Twitter
## 4  2023-08-02 Twitter
## 5  2023-08-02 Twitter
## 6  2023-08-02 Twitter
## 7  2023-08-02 Twitter
## 8  2023-08-02 Twitter
## 9  2023-08-02 Twitter
## 10 2023-08-02 Twitter
## 11 2023-08-02 Twitter
## 12 2023-08-02 Twitter
## 13 2023-08-02 Twitter
## 14 2023-08-02 Twitter
## 15 2023-08-02 Twitter
```

Question-3: Chronological Order

Utilizing the arrange command, arrange the “comm_data” dataframe in ascending order based on the “date” column.

Solution:

```
# Enter code here
comm_data %>% arrange(date)
```

Question-4: Distinct Discovery

Apply the distinct command to find the unique senders in the “comm_data” dataframe.

Solution:

```
# Enter code here
comm_data %>% distinct(sender)
```

```
##          sender
## 1 dave@example
## 2  @bob_tweets
## 3  @frank_chat
## 4  @erin_tweets
## 5 alice@example
## 6  carol_slack
```

Question-5: Sender Stats

Employ the count and group_by commands to generate a summary table that shows the count of messages sent by each sender in the “comm_data” dataframe.

Solution:

```
# Enter code here
comm_data %>% count(sender)

##          sender    n
## 1  @bob_tweets  179
## 2  @erin_tweets  171
## 3  @frank_chat  174
## 4 alice@example  180
## 5  carol_slack  141
## 6 dave@example  155
```

Question-6: Channel Chatter Insights

Using the group_by and count commands, create a summary table that displays the count of messages sent through each communication channel in the “comm_data” dataframe.

Solution:

```
# Enter code here
comm_data %>% group_by(channel) %>%
  summarise(count = n())

## # A tibble: 3 × 2
##   channel count
##   <chr>   <int>
## 1 Email     331
## 2 Slack     320
## 3 Twitter   349
```

Question-7: Positive Pioneers

Utilize the filter, select, and arrange commands to identify the top three senders with the highest average positive sentiment scores. Display their usernames and corresponding sentiment averages.

Solution:

```
# Enter code here
comm_data %>%
  group_by (sender) %>%
  summarise(average_sentiment = mean(sentiment)) %>%
  filter(average_sentiment > 0) %>%
  arrange(desc(average_sentiment)) %>%
  slice(1:3)

## # A tibble: 3 × 2
##   sender          average_sentiment
##   <chr>              <dbl>
## 1 carol_slack         0.118
## 2 alice@example      0.0570
## 3 dave@example       0.00687
```

Question-8: Message Mood Over Time

With the group_by, summarise, and arrange commands, calculate the average sentiment score for each day in the “comm_data” dataframe.

Solution:

```
# Enter code here
comm_data %>%
  group_by (date) %>%
  summarise(average_sentiment = mean(sentiment)) %>%
  arrange(date)

## # A tibble: 20 × 2
##   date          average_sentiment
##   <chr>              <dbl>
## 1 2023-08-01        -0.0616
## 2 2023-08-02         0.136
## 3 2023-08-03         0.107
## 4 2023-08-04        -0.0510
## 5 2023-08-05         0.193
## 6 2023-08-06        -0.0144
## 7 2023-08-07         0.0364
## 8 2023-08-08         0.0666
## 9 2023-08-09         0.0997
## 10 2023-08-10        -0.0254
## 11 2023-08-11        -0.0340
```

```
## 12 2023-08-12      0.0668
## 13 2023-08-13     -0.0604
## 14 2023-08-14     -0.0692
## 15 2023-08-15      0.0617
## 16 2023-08-16     -0.0220
## 17 2023-08-17     -0.0191
## 18 2023-08-18     -0.0760
## 19 2023-08-19      0.0551
## 20 2023-08-20      0.0608
```

Question-9: Selective Sentiments

Use the filter and select commands to extract messages with a negative sentiment score (less than 0) and create a new dataframe.

Solution:

```
# Enter code here
negative_sentiment <- comm_data %>%
  filter(sentiment < 0) %>%
  select(message,sentiment)
```

Question-10: Enhancing Engagement

Apply the mutate command to add a new column to the “comm_data” dataframe, representing a sentiment label: “Positive,” “Neutral,” or “Negative,” based on the sentiment score.

Solution:

```
# Enter code here
comm_data %>%
  mutate(sentiment_label = ifelse(sentiment > 0, "Positive",
                                ifelse(sentiment == 0, "Neutral",
                                       "Negative"))) %>%
  select(message, sentiment ,sentiment_label)
```

Question-11: Message Impact

Create a new dataframe using the mutate and arrange commands that calculates the product of the sentiment score and the length of each message. Arrange the results in descending order.

Solution:

```
# Enter code here
product_sentiment <- comm_data %>%
  mutate(message_impact = sentiment * nchar(message)) %>%
  select(message_impact) %>%
  arrange(desc(message_impact))
```

Question-12: Daily Message Challenge

Use the `group_by`, `summarise`, and `arrange` commands to find the day with the highest total number of characters sent across all messages in the “comm_data” dataframe.

Solution:

```
# Enter code here
comm_data %>% group_by(date) %>%
  summarise(highest = sum(nchar(message))) %>%
  arrange(desc(highest)) %>%
  slice(1)

## # A tibble: 1 × 2
##   date      highest
##   <chr>      <int>
## 1 2023-08-10      875
```

Question-13: Untidy data

Can you list at least two reasons why the dataset illustrated in slide 10 is non-tidy? How can it be made Tidy?

Solution: There are multiple variables in one column. For example, ‘Subject’ mixes variables like ‘Employment Status’, ‘Civilian labour’, ‘Females 16 years and above’ and other employment statuses as well. Separate columns containing ‘Employment Status’ or ‘Demographic Categories’. // Additionally some header columns contain data as well, scattering them within the data rows rather than the consistently being at the top. // Lastly, there are (X) used as placeholders which can represent multiple meanings - for example, it can either be missing or uncalculated percentages. Rather than using placeholders, use a Null value, which would automatically indicate it as a missing value.