

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.”

```
library(tidyverse)
```

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

```
comm_data <- read_csv('CommQuest2023_Larger.csv')
```

```
## Rows: 1000 Columns: 5
## -- Column specification -----
## Delimiter: ","
## chr  (3): channel, sender, message
## dbl  (1): sentiment
## date (1): date
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

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:

```
new<-
comm_data %>%
  select(date,channel,message)
new
```

```
## # A tibble: 1,000 x 3
##   date      channel message
##   <date>    <chr>    <chr>
## 1 2023-08-11 Twitter Fun weekend!
## 2 2023-08-11 Email  Hello everyone!
## 3 2023-08-11 Slack  Hello everyone!
## 4 2023-08-18 Email  Fun weekend!
## 5 2023-08-14 Slack  Need assistance
## 6 2023-08-04 Email  Need assistance
## 7 2023-08-10 Twitter Hello everyone!
## 8 2023-08-04 Slack  Hello everyone!
## 9 2023-08-20 Email  Team meeting
## 10 2023-08-09 Slack  Hello everyone!
## # i 990 more rows
```

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:

```
new %>%
  filter(
    channel == 'Twitter' , date == '2023-08-02'
  )
```

```
## # A tibble: 15 x 3
##   date      channel message
##   <date>    <chr>    <chr>
## 1 2023-08-02 Twitter Team meeting
## 2 2023-08-02 Twitter Exciting news!
## 3 2023-08-02 Twitter Exciting news!
## 4 2023-08-02 Twitter Exciting news!
## 5 2023-08-02 Twitter Exciting news!
## 6 2023-08-02 Twitter Team meeting
## 7 2023-08-02 Twitter Great work!
## 8 2023-08-02 Twitter Hello everyone!
## 9 2023-08-02 Twitter Hello everyone!
## 10 2023-08-02 Twitter Need assistance
## 11 2023-08-02 Twitter Need assistance
## 12 2023-08-02 Twitter Need assistance
## 13 2023-08-02 Twitter Exciting news!
## 14 2023-08-02 Twitter Need assistance
## 15 2023-08-02 Twitter Need assistance
```

Question-3: Chronological Order Utilizing the arrange command, arrange the “comm_data” dataframe in ascending order based on the “date” column.

Solution:

```
comm_data %>%
  arrange(date)
```

```
## # A tibble: 1,000 x 5
```

```
##   date      channel sender      message      sentiment
##   <date>    <chr>   <chr>      <chr>          <dbl>
## 1 2023-08-01 Twitter alice@example Need assistance 0.677
## 2 2023-08-01 Twitter @bob_tweets  Need assistance 0.148
## 3 2023-08-01 Twitter @frank_chat  Need assistance 0.599
## 4 2023-08-01 Twitter @frank_chat  Exciting news! -0.823
## 5 2023-08-01 Slack  @frank_chat  Team meeting   -0.202
## 6 2023-08-01 Slack  @bob_tweets  Exciting news! 0.146
## 7 2023-08-01 Slack  @erin_tweets Great work!    0.244
## 8 2023-08-01 Twitter @frank_chat  Team meeting   -0.526
## 9 2023-08-01 Twitter @frank_chat  Exciting news! -0.399
## 10 2023-08-01 Slack  @frank_chat  Need assistance 0.602
## # i 990 more rows
```

Question-4: Distinct Discovery Apply the distinct command to find the unique senders in the “comm_data” dataframe.

Solution:

```
comm_data %>%
  distinct(sender)
```

```
## # A tibble: 6 x 1
##   sender
##   <chr>
## 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:

```
comm_data %>%
  group_by(sender) %>%
  summarise(count = n())
```

```
## # A tibble: 6 x 2
##   sender      count
##   <chr>      <int>
## 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:

```
comm_data %>%
  group_by(channel) %>%
  summarise(count = n())
```

```
## # A tibble: 3 x 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:

```
comm_data %>%
  arrange(desc(sentiment)) %>%
  filter(sender >= 0) %>%
  select(sender, sentiment) %>%
  slice(1:3)
```

```
## # A tibble: 3 x 2
##   sender      sentiment
##   <chr>         <dbl>
## 1 carol_slack    0.991
## 2 dave@example  0.987
## 3 dave@example  0.985
```

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:

```
comm_data %>%
  group_by(date) %>%
  summarise(mean_sentiment = mean(sentiment)) %>%
  arrange(date)
```

```
## # A tibble: 20 x 2
##   date      mean_sentiment
##   <date>         <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:

```
new_dtf <-
  comm_data %>%
  filter(
    sentiment < 0
  ) %>%
  select(message,sentiment)
new_dtf
```

```
## # A tibble: 487 x 2
##   message      sentiment
##   <chr>         <dbl>
## 1 Hello everyone! -0.143
## 2 Need assistance -0.108
## 3 Hello everyone! -0.741
## 4 Hello everyone! -0.188
## 5 Hello everyone! -0.933
## 6 Need assistance -0.879
## 7 Great work!     -0.752
## 8 Team meeting    -0.787
## 9 Fun weekend!     -0.539
## 10 Exciting news! -0.142
## # i 477 more rows
```

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:

```
comm_data %>%
  mutate(sentiment_label = case_when(
```

```

sentiment > 0 ~ "Positive",
sentiment < 0 ~ "Negative",
TRUE ~ "Neutral" # Everything else is Neutral
))

```

```

## # A tibble: 1,000 x 6
##   date      channel sender      message      sentiment sentiment_label
##   <date>    <chr>  <chr>      <chr>      <dbl> <chr>
## 1 2023-08-11 Twitter dave@example Fun weekend!    0.824 Positive
## 2 2023-08-11 Email  @bob_tweets Hello everyone! 0.662 Positive
## 3 2023-08-11 Slack  @frank_chat Hello everyone! -0.143 Negative
## 4 2023-08-18 Email  @frank_chat Fun weekend!    0.380 Positive
## 5 2023-08-14 Slack  @frank_chat Need assistance 0.188 Positive
## 6 2023-08-04 Email  @erin_tweets Need assistance -0.108 Negative
## 7 2023-08-10 Twitter @frank_chat Hello everyone! -0.741 Negative
## 8 2023-08-04 Slack  alice@example Hello everyone! -0.188 Negative
## 9 2023-08-20 Email  dave@example Team meeting    0.618 Positive
## 10 2023-08-09 Slack  @erin_tweets Hello everyone! -0.933 Negative
## # i 990 more rows

```

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:

```

new<-
comm_data%>%
  mutate(product_sentiment_length=sentiment*nchar(message))%>%
  arrange(desc(product_sentiment_length))
new

```

```

## # A tibble: 1,000 x 6
##   date      channel sender      message      sentiment product_sentiment_le-1
##   <date>    <chr>  <chr>      <chr>      <dbl> <dbl>
## 1 2023-08-16 Email  @frank_chat Hello every~    0.998      15.0
## 2 2023-08-14 Slack  @erin_tweets Hello every~    0.988      14.8
## 3 2023-08-18 Email  dave@example Hello every~    0.978      14.7
## 4 2023-08-17 Email  dave@example Hello every~    0.977      14.7
## 5 2023-08-07 Slack  carol_slack Hello every~    0.973      14.6
## 6 2023-08-06 Slack  dave@example Hello every~    0.968      14.5
## 7 2023-08-08 Slack  @frank_chat Need assist~    0.964      14.5
## 8 2023-08-09 Email  @erin_tweets Need assist~    0.953      14.3
## 9 2023-08-17 Twitter @frank_chat Hello every~    0.952      14.3
## 10 2023-08-12 Email  carol_slack Need assist~    0.938      14.1
## # i 990 more rows
## # i abbreviated name: 1: product_sentiment_length

```

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:

```
comm_data %>%
  group_by(date) %>%
  summarise(num_char = sum(nchar(message))) %>%
  arrange(desc(num_char)) %>%
  slice(1)
```

```
## # A tibble: 1 x 2
##   date      num_char
##   <date>      <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: 1. Each observation has different units of measurement across each attribute, hence they cannot be measured and compared equally among the observation.

2. There is a mix of data types in each cell of the dataset, hence it will be difficult to manipulate the variables.

It can be tidy by removing units, and maintain that each variable has the same units of measurement.