Solutions

Overview

This script makes some plots from Q10, which is about what solutions participants would find most useful.

Import packages and utilities

```
project_root <- here::here() # requires that you be somewhere in the
# project directory (not above it)
# packages
suppressMessages(source(file.path(project_root, "scripts/packages.R")))
# functions and objects used across scripts
suppressMessages(source(file.path(project_root, "scripts/utils.R")))</pre>
```

Load data

```
data <- load_qualtrics_data("deidentified_no_qual.tsv")</pre>
```

Wrangle data

```
solutions <- data %>%
  select(
    starts_with("solution_offerings")
)
```

head(solutions)

	solution_offerings_1		solution_offerings_2			solution_offerings_3		
1	V	ery useful		Very	useful	Vei	y us	eful
2		Useful		Very	useful	Vei	y us	eful
3	V	ery useful		Very	useful	Vei	y us	eful
4	Not v	ery useful			Useful		Us	eful
5		Useful	Not	very	useful		Us	eful
6								
	solution_offerings_4 solution_offerings_5 solution_offerings_						gs_6	
1	V	ery useful		Very	useful	Vei	y us	eful
2	Not v	ery useful			Useful	Non-ar	plic	able
3	V	ery useful			Useful		Us	eful
4	V	ery useful	Not	very	useful		Us	eful
5	V	ery useful	Not	very	useful	Not ve	y us	eful
6								
	solution_offerings_7 solution_offerings_8 solution_offerings_9							
1	V	ery useful		Very	useful	Vei	y us	eful
2	V	ery useful		Very	useful	Vei	y us	eful
3		Useful	Not	very	useful	Vei	y us	eful
4	Not v	ery useful	Not	very	useful		Us	eful
5		Useful		Very	useful		Us	eful
6								
	solution_o	fferings_1	o solution	n_offe	rings_1	1 solution_o	ffer	ings_12
1		Very useful	l	Ver	y usefı	11	Very	useful
2		Usefu	l	Ver	y usefı	11		Useful
3		Very useful	l	Ver	y usefı	11	Very	useful
4	Not	very useful	l	Ver	y usefı	11	Very	useful
5		Usefu	l	Ver	y usefi	ıl		Useful
6					•			

STOP!! Presumably, "solution_offerings_1" corresponds to the first option, "solution_offerings_2" corresponds to the second option, etc., but we still need to check. I am manually comparing the answers in this data frame to those in the Qualtrics interface, which shows the whole response, i.e. "Access to free, feature-rich computing environments", not just "solution_offerings_1". To be extra confident that I am comparing the same rows between the two tables, I am looking at responses associated with a particular email. After this code chunk, I go back to using the data frame that doesn't contain the emails.

Since this code only needed to be run once, I've commented it out.

```
# pii <- load_qualtrics_data("pii.tsv")
# emails <- pii %>%
# select(starts_with("stay_in_touch_email"))

# t <- cbind(emails, solutions)

# Next, I run this line repeatedly with different emails,
# to make sure that this person's response to "solution_offerings_1"
# # matches their response to "Access to free, feature-rich computing environments", etc.
# subset(t, startsWith(stay_in_touch_email, "PERSON_NAME_HERE"))</pre>
```

My assumption above was correct; the options are ordered as expected. Let's rename the columns accordingly.

```
codes <- c(
    "Computing environments" = "solution_offerings_1",
    "Publicity" = "solution_offerings_2",
    "Containerization" = "solution_offerings_3",
    "Documentation help" = "solution_offerings_4",
    "A learning community" = "solution_offerings_5",
    "Event planning" = "solution_offerings_6",
    "Mentoring programs" = "solution_offerings_7",
    "Education" = "solution_offerings_8",
    "Legal support" = "solution_offerings_9",
    "Industry partnerships" = "solution_offerings_10",
    "Sustainability grants" = "solution_offerings_11",
    "Help finding funding" = "solution_offerings_12"
)
solutions <- rename(solutions, any_of(codes))</pre>
```

Next, remove empty rows, i.e. rows from respondents who didn't receive this question. As with many questions in this survey, we can cut some corners in the code because the question was mandatory. For example, no need to worry about incomplete answers.

```
nrow(solutions)
```

[1] 332

```
solutions <- exclude_empty_rows(solutions) # from scripts/utils.R
nrow(solutions)</pre>
```

[1] 233

Let's reshape the data from wide to long format for easier plotting later.

```
long_data <- solutions %>%
  pivot_longer(
   cols = everything(),
    names_to = "solution",
    values_to = "utility"
long_data <- long_data %>%
  mutate(
   utility_score = recode(
      utility,
      "Non-applicable" = OL,
      "Not very useful" = OL,
      "Useful" = 1L,
      "Very useful" = 2L
    )
  )
# Using interger literals OL, 1L, etc., ensures that
# the new column will be integers, not doubles.
long_data
```

```
# A tibble: 2,796 x 3
   solution
                          utility
                                       utility_score
   <chr>
                                              <int>
                          <chr>>
1 Computing environments Very useful
                                                   2
2 Publicity
                          Very useful
                                                   2
                                                   2
3 Containerization
                          Very useful
4 Documentation help
                          Very useful
                                                   2
5 A learning community
                                                   2
                          Very useful
                                                   2
6 Event planning
                          Very useful
                                                   2
7 Mentoring programs
                          Very useful
8 Education
                          Very useful
                                                   2
                                                   2
9 Legal support
                          Very useful
10 Industry partnerships Very useful
                                                   2
# i 2,786 more rows
```

Next, let's calculate some simple descriptive statistics. I will choose: * The total "score", that is, the total number of "points" a solution received (see scoring scheme in previous code chunk) * The mean (which might be misleading if 0s drag it down, and also, who's to say what a 1.5 really means? Are the distances between the Likert points equal? We don't know.) * The mode * The standard deviation

```
# Helper to compute the (numeric) mode
get_mode <- function(x) {</pre>
  ux <- unique(x)</pre>
  ux[which.max(tabulate(match(x, ux)))]
}
summary_df <- long_data %>%
  group_by(solution) %>%
  summarise(
    total = sum(utility_score),
           = mean(utility_score, na.rm = TRUE),
    mean
           = get_mode(utility_score),
    st_dev = sd(utility_score, na.rm = TRUE)
  ) %>%
  ungroup()
# Order by highest total "score"
summary_df <- summary_df %>%
    arrange(desc(total))
summary_df
```

A tibble: 12 x 5 solution total mean mode st_dev <chr> <int> <dbl> <int> <dbl> 1 Sustainability grants 353 1.52 2 0.732 2 Help finding funding 2 0.764 316 1.36 3 Computing environments 301 1.29 2 0.783 4 A learning community 251 1.08 1 0.733 5 Documentation help 1 0.788 248 1.06 6 Legal support 242 1.04 1 0.762 7 Education 236 1.01 1 0.801 8 Industry partnerships 232 0.996 0 0.838 9 Publicity 232 0.996 1 0.817 10 Mentoring programs 216 0.927 1 0.776 0 0.820 11 Containerization 203 0.871

Cool. It looks like sustainability grants are by far the most popular, with assistance identifying funding sources and free computing environments in second and third place. These were the only three solutions that had a mode of 2.

Out of curiosity, how does it look when we order by variability?

```
sd_df <- summary_df %>%
    arrange(desc(st_dev))
sd_df
```

```
# A tibble: 12 x 5
  solution
                         total mean mode st_dev
  <chr>
                         <int> <dbl> <int>
                                            <dbl>
1 Industry partnerships
                           232 0.996
                                         0 0.838
2 Containerization
                           203 0.871
                                         0 0.820
3 Publicity
                           232 0.996
                                         1 0.817
4 Event planning
                           190 0.815
                                         0 0.807
5 Education
                           236 1.01
                                         1 0.801
6 Documentation help
                           248 1.06
                                           0.788
7 Computing environments
                           301 1.29
                                         2 0.783
8 Mentoring programs
                           216 0.927
                                         1 0.776
9 Help finding funding
                           316 1.36
                                         2 0.764
10 Legal support
                           242 1.04
                                         1 0.762
11 A learning community
                           251 1.08
                                         1 0.733
12 Sustainability grants
                           353 1.52
                                         2
                                            0.732
```

This analysis doesn't seem as interesting as it was for the challenges. Industy partnerships, Containerization, and Publicity all show high variance/stdev. These were also somewhat less popular.

Out of curiosity, how many people said they would all be very useful?

```
t <- solutions %>%
    filter(if_all(.cols=everything(), ~ . == "Very useful"))
nrow(t)
```

[1] 14

Ah, ok. Not that many.

Plot the distributions

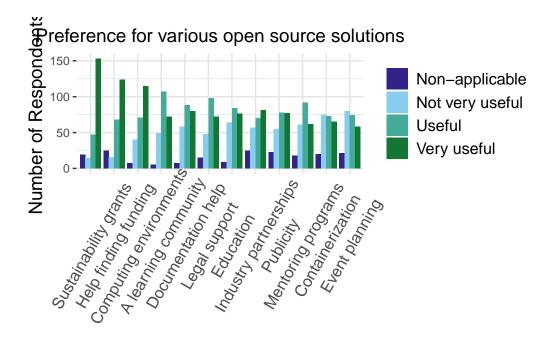
Prepare data for plotting.

```
ordered_levels <- (summary_df %>%
    arrange(desc(total)))$solution

long_data$solution <- factor(long_data$solution, levels = ordered_levels)</pre>
```

Grouped bar chart showing the distributions of answers.

```
grouped_plot <- grouped_bar_chart(
    df = long_data,
    x_var = "solution",
    fill_var = "utility",
    title = "Preference for various open source solutions"
)
grouped_plot</pre>
```



Save the plot if you wish.

```
save_plot("fave_solutions.tiff", 10, 6, p=grouped_plot)
```

Now let's make a simpler bar plot from the next question, which asked participants to choose their favorite solution.

```
favorites <- data.frame(data$favorite_solution)</pre>
favorites <- exclude_empty_rows(favorites) # from scripts/utils.R
codes2 <- c(
  "Access to" = "Computing environments",
  "Assistance promoting" = "Publicity",
  "Assistance creating" = "Containerization",
  "Assistance writing" = "Documentation help",
  "An open source" = "A learning community",
  "Assistance with" = "Event planning",
  "A mentor" = "Mentoring programs",
  "Educational" = "Education",
  "Legal" = "Legal support",
  "Assistance building" = "Industry partnerships",
  "Dedicated" = "Sustainability grants",
  "Assistance identifying" = "Help finding funding"
)
favorites <- shorten_long_responses(favorites, codes2)</pre>
fav_to_plot <- data.frame(table(favorites[,1]))</pre>
# from scripts/utils.R
fav_to_plot <- reorder_factor_by_column(</pre>
  df = fav_to_plot,
  factor_col = Var1,
  value_col = Freq,
  descending = FALSE
```

```
faves_plot <- basic_bar_chart(
    df = fav_to_plot,
    x_var = "Var1",
    y_var = "Freq",
    title = "Participant's favorite solution",
    show_axis_title_y = FALSE,
    ylabel = "Number of Respondents",
    show_bar_labels = TRUE,</pre>
```

```
color_index = 7,
horizontal = TRUE
)
faves_plot
```

Participant's favorite s



Save the plot if you wish.

```
save_plot("fave_solutions.tiff", 10, 6, p=faves_plot)
```