Importance of open source

Overview

This script creates bar plots from question 2 on the survey, which is about the perceived importance of open source for different job categories and different tasks. My favorite plot is the one in the final section, "Percent more than moderately important".

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>
```

Define functions

get_percent_more_than_moderate

- Arguments:
 - df: A data frame with a column named importance_level. Should contain only rows that you want to count. Extra columns are okay.
- Details:
 - A simple function to count what percent of rows in a data frame have an importance_level of more than Moderately important. Checks that there are no extra rows with an unexpected value in the importance_level column.

- Outputs:
 - A scalar value representing the percentage of respondents who selected an importance_level of more than Moderately important, rounded to 2 decimal places.

```
get percent more than moderate <- function(df) {
  # check that df has the required column
 if (!"importance_level" %in% names(df)) {
    stop("`df` must have a column named 'importance_level'.")
 }
 high_importance_cats <- c(
    "Very important",
    "Important"
 low_importance_cats <- c(</pre>
   "Moderately important",
    "Slightly important",
    "Not at all important"
 )
 n_high_rows <- nrow(df %>% filter(importance_level %in% high_importance_cats))
 n_low_rows <- nrow(df %>% filter(importance_level %in% low_importance_cats))
 total <- nrow(df)
 if (n_high_rows + n_low_rows != total) {
    stop("'df' has at least one unexpected value in 'importance_level'.")
 pct <- round(n_high_rows / total * 100, 2)</pre>
 return(pct)
```

Load data

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

Wrangle data

Let's extract the columns we care about.

```
importance_and_job <- data %>%
  select(
    starts_with("importance_opensrc") | starts_with("job_category")
  )
head(importance_and_job)
```

```
importance_opensrc_1 importance_opensrc_2 importance_opensrc_3
1
                             Very important
                                                   Very important
        Very important
2
        Very important Moderately important
                                                         Important
3
        Very important
                             Very important
                                                   Very important
4
        Very important
                         Slightly important
                                                         Important
5
        Very important
                                                   Very important
                                   Important
6
        Very important
                             Non-applicable
                                                         Important
  importance_opensrc_4 importance_opensrc_5
1
        Very important
                              Very important
2
             Important
                             Non-applicable
        Very important
3
                             Non-applicable
4
                             Non-applicable
             Important
5
        Very important
                             Non-applicable
6
             Important
                             Non-applicable
                                                                   job_category
1
                                                                        Faculty
2
                                                                       Post-Doc
3 Other research staff (e.g., research scientist, research software engineer)
4
                                                                        Faculty
5
                                                                        Faculty
6 Other research staff (e.g., research scientist, research software engineer)
```

Let's reshape the data from wide to long format.

```
long_data <- importance_and_job %>%
  pivot_longer(
    cols = starts_with("importance_opensrc"),
    names_to = "importance_area",
    values_to = "importance_level"
)

long_data <- long_data %>%
  mutate(
  importance_area = recode(
  importance_area,
```

```
"importance_opensrc_1" = "Research",
    "importance_opensrc_2" = "Teaching",
    "importance_opensrc_3" = "Learning",
    "importance_opensrc_4" = "Professional Development",
    "importance_opensrc_5" = "Job"
    )
    )
long_data
```

```
# A tibble: 1,660 x 3
   job_category importance_area
                                          importance_level
   <chr>
                <chr>>
                                          <chr>
1 Faculty
                Research
                                          Very important
2 Faculty
                Teaching
                                          Very important
3 Faculty
                Learning
                                          Very important
4 Faculty
                Professional Development Very important
5 Faculty
                Job
                                          Very important
6 Post-Doc
                Research
                                          Very important
7 Post-Doc
                Teaching
                                          Moderately important
8 Post-Doc
                                          Important
                Learning
9 Post-Doc
                Professional Development Important
10 Post-Doc
                Job
                                          Non-applicable
# i 1,650 more rows
```

STOP!!! At this point, I manually compared this data frame to the results table in Qualtrics to make sure the columns (e.g. importance_opensrc_1) correspond to the options I expect (e.g. "Research"). I had to use peoples' email addresses to make sure I was comparing the same rows in each table. I assumed that the variables were ordered by their order on the survey, but you never know. In this case, my assumption was correct. I've commented out the code for this because it only needed to be done once.

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

# t <- cbind(emails, importance_and_job)
# subset(t, startsWith(stay_in_touch_email, "PERSONNAMEHERE"))</pre>
```

Back to data wrangling.

Here, I removed all rows that contain an empty string in any column. Since both questions were mandatory, I'm actually only removing people who never saw the demographic questions:

people who are not affiliated with UC (2) + people who are neither past nor future open source contributors (36). (2+36)*5 importance areas = 190 rows removed.

```
dim(long_data)
```

[1] 1660 3

```
long_data <- long_data %>%
  filter(!if_any(everything(), ~ . == ""))
dim(long_data)
```

[1] 1470 3

Shorten this one long category name. Other research staff (e.g., research scientist, research software engineer) becomes simply Other.

```
long_data$job_category <- gsub(
    "^Other.*",
    "Research Staff",
    long_data$job_category
)</pre>
```

Reorder factor levels for plotting.

```
long_data$importance_level <- factor(
  long_data$importance_level,
  levels = c(
    "Very important",
    "Important",
    "Moderately important",
    "Slightly important",
    "Not at all important",
    "Non-applicable"
  ),
   ordered = TRUE
)
long_data</pre>
```

```
# A tibble: 1,470 \times 3
   job_category importance_area
                                          importance_level
   <chr>
                <chr>>
                                          <ord>
1 Faculty
                                          Very important
                Research
                                          Very important
2 Faculty
                Teaching
3 Faculty
                                          Very important
                Learning
4 Faculty
                Professional Development Very important
5 Faculty
                Job
                                          Very important
6 Post-Doc
                Research
                                          Very important
7 Post-Doc
                                          Moderately important
                Teaching
8 Post-Doc
                                          Important
                Learning
9 Post-Doc
                Professional Development Important
10 Post-Doc
                                          Non-applicable
# i 1,460 more rows
```

Bar plots

Simple bar plot for teachers

Now let's start making some bar plots. Let's start by making a bar plot showing how teachers rate the importance of open source for their teaching. Since we didn't ask people "Do you teach?", and since there was a "Non-applicable" option, we will simply assume that if they gave an answer for the "Teaching" option, they must be a teacher.

```
teaching <- long_data %>%
  filter(
    importance_area == "Teaching"
) %>%
  filter(
    importance_level != "Non-applicable"
)

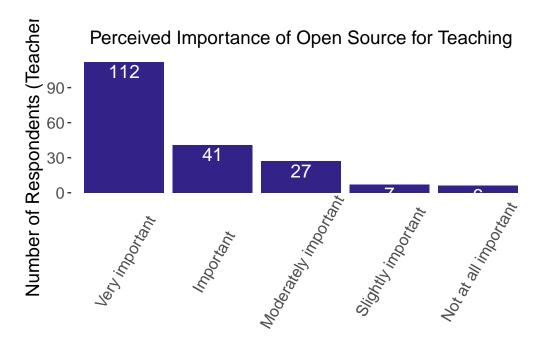
# For our bar plot, we only care about how many times each 'importance level' was selected.
teaching_to_plot <- teaching %>% select(-c(job_category, importance_area))

teaching_to_plot <- teaching_to_plot %>%
    count(importance_level, name = "Counts")

# By sheer luck, the columns are already ordered by response rates.
teaching_to_plot
```

Now let's make that bar chart using a function that lives in my utilities script (scripts/utils.R).

```
basic_bar_chart(
  teaching_to_plot,
  x_var = "importance_level",
  y_var = "Counts",
  title = "Perceived Importance of Open Source for Teaching",
  ylabel = "Number of Respondents (Teachers Only)",
  show_bar_labels = TRUE
)
```



Save the plot using a function that lives in my utilities script (scripts/utils.R).

```
#save_plot("importance_teachers.tiff", 8, 5)
```

Grouped bar plot for researchers

Now let's look at researchers, and the importance categories that apply to all researchers. The importance categories again are:

Research
Teaching -> Does not apply
Learning

Professional Development

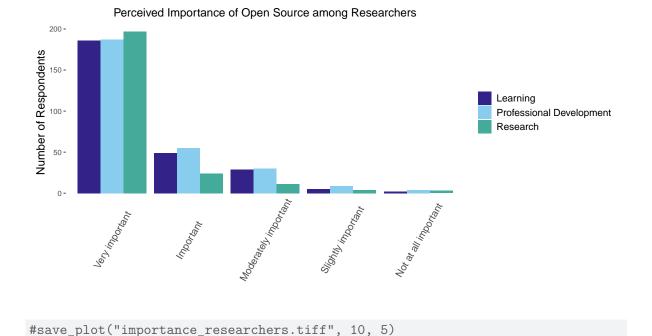
Job (For non-research staff) -> Does not apply

So we'll make a bar plot with just those three categories that apply to all researchers. As with teachers above, we will assume that if they didn't select "Non-applicable", they must be a researcher.

The grouped_bar_chart function, like the basic_bar_chart function, lives in my utility script.

```
research_learning_pd <- long_data %>%
  filter(
   importance_area == "Research" |
    importance_area == "Learning" |
    importance_area == "Professional Development"
) %>%
  filter(importance_level != "Non-applicable")

grouped_bar_chart(
  df = research_learning_pd,
  x_var = "importance_level",
  fill_var = "importance_area",
  title = "Perceived Importance of Open Source among Researchers"
)
```

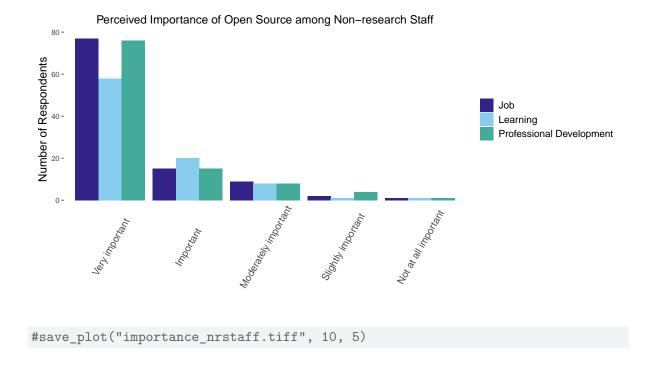


Grouped bar plot for non-research staff

This is very similar to what I did above, except the three applicable categories have changed.

```
job_learning_pd <- long_data %>%
    filter(
    importance_area == "Job" |
        importance_area == "Learning" |
        importance_area == "Professional Development"
    ) %>%
    filter(job_category == "Non-research Staff") %>%
    filter(importance_level != "Non-applicable")

grouped_bar_chart(
    df = job_learning_pd,
    x_var = "importance_level",
    fill_var = "importance_area",
    title = "Perceived Importance of Open Source among Non-research Staff"
)
```



Percent more than moderately important

Renata suggested I try to combine all these data into one figure that summarizes the question at a glance. Here's my attempt.

I think a useful "statistic" is the percent of a particular group that said OS is more than moderately important for a particular area of work. Let's make a dataframe with those percentages. I'd ultimately like to turn this into a bar plot where the color or design of the bars corresponds to the five job categories, and the x-axis shows four groups that I think are most relevant: teachers, researchers, non-research staff, and students. However, not all these groups were explicit survey categories, and not all 5 importance areas apply to all groups, so we'll need to do a fair amount of data wrangling.

To start, let's get the percent of teachers who said teaching was more than moderately important.

```
teaching <- long_data %>%
  filter(
   importance_area == "Teaching"
) %>%
  filter(
  importance_level != "Non-applicable"
```

```
more_than_mod <- data.frame(
   job_category = "Teachers",
   importance_area = "Teaching",
   pct = get_percent_more_than_moderate(teaching)
)
more_than_mod</pre>
```

```
job_category importance_area pct
1 Teachers Teaching 79.27
```

The code is basically the same for researchers and researching: as with teaching, we will assume that anyone who gave an answer for research (i.e, didn't select "N/A") is a researcher.

```
research <- long_data %>%
    filter(
    importance_area == "Research"
) %>%
    filter(
    importance_level != "Non-applicable"
)

more_than_mod <- rbind(
    more_than_mod,
    list(
        "Researchers",
        "Research",
        get_percent_more_than_moderate(research)
)
)</pre>
```

The code for non-research staff and "Job" is slightly different. Our intention was that only non-research staff would answer this question, but there were some people who answered this but did not select "non-research staff" as their job category. So let's just ensure that we're only looking at responses from non-research staff by filtering for non-research staff using the job_category column.

```
nrstaff <- long_data %>%
  filter(
    job_category == "Non-research Staff"
  ) %>%
  filter(
    importance_area == "Job"
  ) %>%
  filter(
    importance_level != "Non-applicable"
more_than_mod <- rbind(</pre>
  more_than_mod,
  list(
    "Non-research Staff",
    "Job",
    get_percent_more_than_moderate(nrstaff)
)
```

Next, I want to look at two importance areas, Learning and Professional Development, for all four job categories: Teachers, Researchers, Non-research staff, and Students. We'll have to determine teachers and researchers based on who answered the teaching question and who answered the research question, respectively. Meanwhile, for "Student", we'll have to combine grad students and undergrads into one group.

We will have to go back to an earlier data frame and redo some of the data wrangling. (We want to filter for people who answered e.g. teaching, but look at their answers for the other questions. This information was lost when we rearranged from wide to long format.)

```
# Rename this one long job category
importance_and_job$job_category <- gsub(
    "^Other.*",
    "Research Staff",
    importance_and_job$job_category
)

#Rename columns for readability
importance_and_job <- importance_and_job %>%
    rename(
    Research = importance_opensrc_1,
    Teaching = importance_opensrc_2,
```

```
Learning = importance_opensrc_3,
    `Professional Development` = importance_opensrc_4,
    Job = importance_opensrc_5
)

# Remove rows that contain any empty strings
importance_and_job <- importance_and_job %>%
    filter(!if_any(everything(), ~ . == ""))
```

Let's keep rows from teachers, but keep columns for Learning and Professional Development. Then we change the job_category column to "Teacher".

```
teachers_learn_pd <- importance_and_job %>%
  filter(Teaching != "Non-applicable") %>%
  select(Learning, `Professional Development`, job_category)

teachers_learn_pd$job_category <- "Teacher"
head(teachers_learn_pd)</pre>
```

```
Learning Professional Development job_category
1
       Very important
                                Very important
                                                     Teacher
2
            Important
                                      Important
                                                     Teacher
       Very important
3
                                Very important
                                                     Teacher
                                                     Teacher
4
            Important
                                      Important
5
       Very important
                                 Very important
                                                     Teacher
6 Moderately important
                           Moderately important
                                                     Teacher
```

Now we can add two more rows to more_than_mod.

```
teachers_learning <- teachers_learn_pd %>%
    select(Learning, job_category) %>%
    #unlikely but you never know
    filter(Learning != "Non-applicable") %>%
    # Change the column name because our function expects it
    rename(importance_level = Learning)

more_than_mod <- rbind(
    more_than_mod,
    list(
        "Teachers",
        "Learning",</pre>
```

```
get_percent_more_than_moderate(teachers_learning)
  )
)
teachers_pd <- teachers_learn_pd %>%
  select(`Professional Development`, job_category) %>%
  #unlikely but you never know
  filter(`Professional Development` != "Non-applicable") %>%
  # Change the column name because our function expects it
  rename(importance_level = `Professional Development`)
more_than_mod <- rbind(</pre>
  more_than_mod,
  list(
    "Teachers",
    "Professional Development",
    get_percent_more_than_moderate(teachers_pd)
)
more_than_mod
```

```
job_category importance_area pct
1 Teachers Teaching 79.27
2 Researchers Research 92.47
3 Non-research Staff Job 88.46
4 Teachers Learning 84.29
5 Teachers Professional Development 81.48
```

And let's do the same for researchers.

```
researchers_learn_pd <- importance_and_job %>%
  filter(Research != "Non-applicable") %>%
  select(Learning, `Professional Development`, job_category)

researchers_learn_pd$job_category <- "Researcher"

researchers_learning <- researchers_learn_pd %>%
  select(Learning, job_category) %>%
  #unlikely but you never know
  filter(Learning != "Non-applicable") %>%
  # Change the column name because our function expects it
```

```
rename(importance_level = Learning)
more_than_mod <- rbind(</pre>
  more_than_mod,
  list(
    "Researchers",
    "Learning",
    get_percent_more_than_moderate(researchers_learning)
  )
researchers_pd <- researchers_learn_pd %>%
  select(`Professional Development`, job_category) %>%
  #unlikely but you never know
  filter(`Professional Development` != "Non-applicable") %>%
  # Change the column name because our function expects it
  rename(importance_level = `Professional Development`)
more_than_mod <- rbind(</pre>
  more_than_mod,
  list(
    "Researchers",
    "Professional Development",
    get_percent_more_than_moderate(researchers_pd)
  )
more_than_mod
```

```
importance_area
        job_category
                                                 pct
1
            Teachers
                                     Teaching 79.27
2
         Researchers
                                     Research 92.47
3 Non-research Staff
                                           Job 88.46
                                     Learning 84.29
            Teachers
5
            Teachers Professional Development 81.48
6
         Researchers
                                     Learning 85.90
7
         Researchers Professional Development 83.98
```

Now get percentages for non-research staff. This is straightforward since it was a survey category.

```
nrstaff_learn_pd <- importance_and_job %>%
  filter(job_category == "Non-research Staff") %>%
  select(Learning, `Professional Development`, job_category)
nrstaff_learning <- nrstaff_learn_pd %>%
  select(Learning, job_category) %>%
  #unlikely but you never know
  filter(Learning != "Non-applicable") %>%
  # Change the column name because our function expects it
  rename(importance_level = Learning)
more_than_mod <- rbind(</pre>
  more_than_mod,
  list(
    "Non-research Staff",
    "Learning",
    get_percent_more_than_moderate(nrstaff_learning)
  )
)
nrstaff_pd <- nrstaff_learn_pd %>%
  select(`Professional Development`, job_category) %>%
  #unlikely but you never know
  filter(`Professional Development` != "Non-applicable") %>%
  # Change the column name because our function expects it
  rename(importance_level = `Professional Development`)
more_than_mod <- rbind(</pre>
  more_than_mod,
  list(
    "Non-research Staff",
    "Professional Development",
    get_percent_more_than_moderate(nrstaff_pd)
  )
```

Finally, let's get students.

```
students_learn_pd <- importance_and_job %>%
filter(job_category == "Undergraduate" | job_category == "Grad Student") %>%
select(Learning, `Professional Development`, job_category)
```

```
students_learn_pd$job_category <- "Student"
students_learning <- students_learn_pd %>%
  select(Learning, job_category) %>%
  #unlikely but you never know
  filter(Learning != "Non-applicable") %>%
  # Change the column name because our function expects it
  rename(importance_level = Learning)
more_than_mod <- rbind(</pre>
  more_than_mod,
  list(
    "Students",
    "Learning",
    get_percent_more_than_moderate(students_learning)
  )
)
students_pd <- students_learn_pd %>%
  select(`Professional Development`, job_category) %>%
  #unlikely but you never know
  filter(`Professional Development` != "Non-applicable") %>%
  # Change the column name because our function expects it
  rename(importance_level = `Professional Development`)
more_than_mod <- rbind(</pre>
  more_than_mod,
  list(
    "Students",
    "Professional Development",
    get_percent_more_than_moderate(students_pd)
  )
)
```

FINALLY, let's plot it!

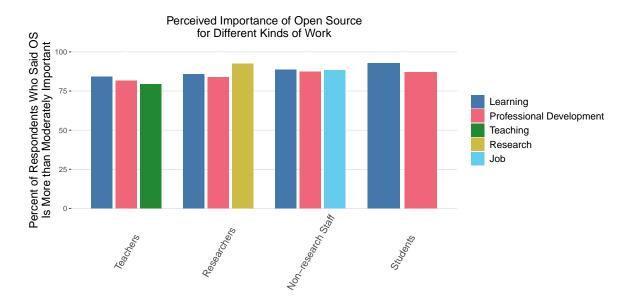
```
more_than_mod$job_category <- factor(
  more_than_mod$job_category,
  levels = c(
    "Teachers",
    "Researchers",
    "Non-research Staff",</pre>
```

```
"Students"
)

more_than_mod$importance_area <- factor(
  more_than_mod$importance_area,
  levels = c(
    "Learning",
    "Professional Development",
    "Teaching",
    "Research",
    "Job"
)</pre>
```

I'm not using my grouped_bar_chart function in scripts/utils.R because I have precomputed the bar heights, and that function counts rows. Since I'm currently only creating this kind of bar chart once, I'm not bothering to create a new function (or incorporate this option into the grouped_bar_chart function).

```
p <- ggplot(</pre>
  more_than_mod,
  aes(
    x = job_category,
    y = pct,
    fill = importance_area
  )
) +
  geom_col(position = position_dodge(width = 0.8), width = 0.7) +
  ggtitle("Perceived Importance of Open Source\nfor Different Kinds of Work") +
  labs(
    y = "Percent of Respondents Who Said OS\nIs More than Moderately Important"
  ) +
  ylim(0, 100) +
  theme(
    axis.title.x = element_blank(),
    axis.title.y = element_text(size = 14),
    axis.text.x = element_text(angle = 60, vjust = 0.6, size = 12),
    axis.ticks.x = element_blank(),
    legend.title = element_blank(),
    legend.text = element_text(size = 12),
    panel.background = element_blank(),
```



```
save_plot("importance_all_pct.tiff", 10, 5)
```