lab2

Lab 2

```
library(opendatatoronto)
library(tidyverse)
library(stringr)
library(skimr) # EDA
library(visdat) # EDA
library(janitor)
library(lubridate)
library(ggrepel)
```

1.

```
all_data <- list_packages(limit = 500)
#head(all_data)

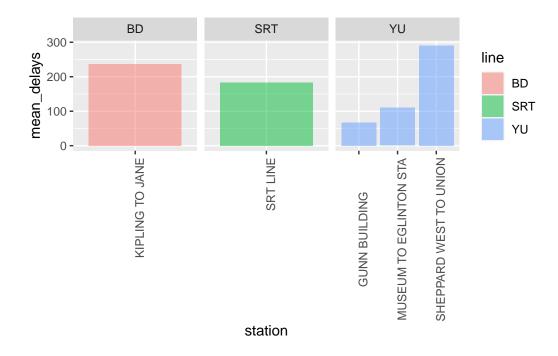
res <- list_package_resources("996cfe8d-fb35-40ce-b569-698d51fc683b") # obtained code from
res <- res %>% mutate(year = str_extract(name, "202.?"))
delay_2022_ids <- res %>% filter(year==2022) %>%
    select(id) %>% pull()

delay_2022 <- get_resource(delay_2022_ids)

# make the column names nicer to work with
delay_2022 <- clean_names(delay_2022)

delay_2022 %>%
    filter(min_delay>0) %>%
    group_by(station, line) %>%
```

```
summarise(mean_delays = mean(min_delay, na.rm =T)) %>%
distinct() %>% arrange(desc(mean_delays)) %>%
head(5) %>%
ggplot(aes(x = station, y = mean_delays)) +
geom_col(aes(fill = line), alpha = .5) +
facet_wrap(~line, scales = "free_x") +
theme(axis.text.x = element_text(angle = 90))
```



2.

```
all_data <- list_packages(limit = 500)
#head(all_data)

res2 <- list_package_resources("f6651a40-2f52-46fc-9e04-b760c16edd5c") # obtained code fro
campaign_2014 <- get_resource("5b230e92-0a22-4a15-9572-0b19cc222985")[[2]]</pre>
```

```
colnames(campaign_2014) <- campaign_2014[1,]</pre>
  campaign_2014 <- campaign_2014[-1,] %>% clean_names()
  head(campaign_2014)
# A tibble: 6 x 13
  contributors~1 contr~2 contr~3 contr~4 contr~5 goods~6 contr~7 relat~8 presi~9
                 <chr>
                          <chr>
                                  <chr>
                                          <chr>>
                                                   <chr>
                                                           <chr>
                                                                   <chr>
                                                                            <chr>
1 A D'Angelo, T~ <NA>
                          M6A 1P5 300
                                          Moneta~ <NA>
                                                           Indivi~ <NA>
                                                                            <NA>
2 A Strazar, Ma~ <NA>
                         M2M 3B8 300
                                          Moneta~ <NA>
                                                           Indivi~ <NA>
                                                                            <NA>
3 A'Court, K Su~ <NA>
                         M4M 2J8 36
                                          Moneta~ <NA>
                                                           Indivi~ <NA>
                                                                            <NA>
4 A'Court, K Su~ <NA>
                         M4M 2J8 100
                                          Moneta~ <NA>
                                                           Indivi~ <NA>
                                                                            <NA>
5 A'Court, K Su~ <NA>
                         M4M 2J8 100
                                          Moneta~ <NA>
                                                           Indivi~ <NA>
                                                                            <NA>
6 Aaron, Robert~ <NA>
                         M6B 1H7 250
                                          Moneta~ <NA>
                                                           Indivi~ <NA>
                                                                            <NA>
```

- # ... with 4 more variables: authorized_representative <chr>, candidate <chr>,
- # office <chr>, ward <chr>, and abbreviated variable names
- # 1: contributors_name, 2: contributors_address, 3: contributors_postal_code,
- # 4: contribution_amount, 5: contribution_type_desc,
- # 6: goods_or_service_desc, 7: contributor_type_desc,
- # 8: relationship_to_candidate, 9: president_business_manager

4.

There are several variables with a large number of missing values (e.g., contributors address, goods or service, relationship to candidate,...). While some of these variables may not be important (e.g., contributors address), other variables such as relationship to candidate could be necessary to answer research questions such as 'which factors drive contributions?'.

The contribution amount variable which was in character format was changed to numeric.

```
skim(campaign_2014)
```

Table 1: Data summary

Name	$campaign_2014$
Number of rows	10199
Number of columns	13
Column type frequency:	
character	13

Table 1: Data summary

Group variables	None

Variable type: character

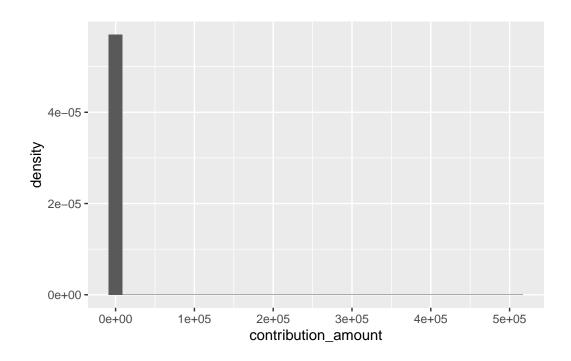
skim_variable	n_missing	$complete_{-}$	_rate	e min	max	empty	n_unique	whitespace
contributors_name	0		1	4	31	0	7545	0
contributors_address	10197		0	24	26	0	2	0
contributors_postal_code	0		1	7	7	0	5284	0
contribution_amount	0		1	1	18	0	209	0
contribution_type_desc	0		1	8	14	0	2	0
goods_or_service_desc	10188		0	11	40	0	9	0
contributor_type_desc	0		1	10	11	0	2	0
relationship_to_candidate	e 10166		0	6	9	0	2	0
president_business_mana	ger 10197		0	13	16	0	2	0
authorized_representative	10197		0	13	16	0	2	0
candidate	0		1	9	18	0	27	0
office	0		1	5	5	0	1	0
ward	10199		0	NA	NA	0	0	0

campaign_2014\$contribution_amount <- as.numeric(campaign_2014\$contribution_amount)</pre>

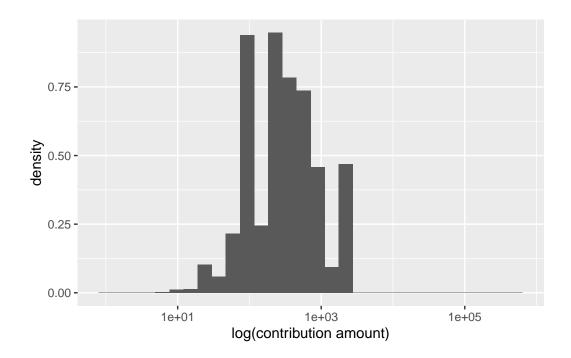
5.

There is one notable outlier with contribution amount 508224.73. However, contributions that exceed 10000 can also be considered as potential outliers. These outliers are mostly monetary contributions from Doug and Rob Ford for their election campaigns.

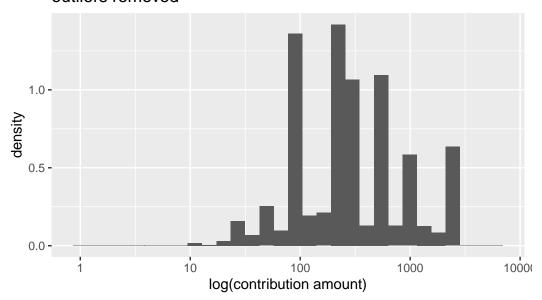
```
campaign_2014 %>%
  ggplot(aes(x = contribution_amount, y = ..density..)) +
  geom_histogram()
```



```
campaign_2014 %>%
  ggplot(aes(x = contribution_amount, y = ..density..)) +
  geom_histogram() + scale_x_log10() +
  labs(x= "log(contribution amount)")
```



Distribution of contributions with potential outliers removed



6.

• Highest mean contributions

- # A tibble: 5 x 1
 contributors_name
 <chr>
- 1 Ford, Doug
- 2 Ford, Rob
- 3 Goldkind, Ari
- 4 Di Paola, Rocco
- 5 kindred's Muze
 - Highest total contributions

```
df %>% slice_max(total_cont, n = 5)%>%
    select(contributors_name)
# A tibble: 5 x 1
  contributors_name
  <chr>
1 Ford, Doug
2 Ford, Rob
3 Goldkind, Ari
4 Thomson, Sarah
5 Pappalardo, Victor
  • Highest number of contributions
  df \%% slice_max(n_cont, n = 5)%>%
    select(contributors_name)
# A tibble: 6 x 1
  contributors_name
  <chr>
1 Italiano, Rob
2 Cranston, Jacqueline
3 Henery, Marjorie
4 Martin, Martha
5 Quin, Derek
6 Stewart, Carol
7.
  • Highest mean contributions
  df2 <- campaign_2014 %>% filter(relationship_to_candidate != "Candidate") %>%
    group_by(contributors_name) %>%
    summarise(ave_cont = mean(contribution_amount),
              total_cont = sum(contribution_amount),
              n_{cont} = n()
```

df2 %>% slice_max(ave_cont, n = 5) %>%

select(contributors_name)

```
df2 %>% slice_max(total_cont, n = 5)%>%
  select(contributors_name)
```

- # A tibble: 3 x 1
 contributors_name
 <chr>
- 1 Hackett, Barbara
- 2 Yan, Flora
- 3 Johnson, Suzanne
 - Highest number of contributions

```
df2 %>% slice_max(n_cont, n = 5)%>%
  select(contributors_name)
```

- # A tibble: 3 x 1
 contributors_name
 <chr>
- 1 Hackett, Barbara
- 2 Johnson, Suzanne
- 3 Yan, Flora

8.

```
campaign_2014 %>% select(contributors_name, candidate) %>%
  distinct() %>% group_by(contributors_name) %>%
  summarise(ncand = n()) %>% filter(ncand >1) %>%
  nrow()
```

[1] 184