

# lab2

Yiyu Wang

```
library(opendatatoronto)
library(tidyverse)
```

```
-- Attaching packages ----- tidyverse 1.3.2 --
v ggplot2 3.4.0      v purrr   0.3.5
v tibble  3.1.8      v dplyr   1.0.10
v tidyr   1.2.1      v stringr 1.5.0
v readr   2.1.3      v forcats 0.5.2
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()     masks stats::lag()
```

```
library(stringr)
library(skimr) # EDA
library(visdat) # EDA
library(janitor)
```

Attaching package: 'janitor'

The following objects are masked from 'package:stats':

chisq.test, fisher.test

```
library(lubridate)
```

Loading required package: timechange

Attaching package: 'lubridate'

The following objects are masked from 'package:base':

date, intersect, setdiff, union

```
library(ggrepel)
library(dplyr)
```

```
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)
```

## Question 1

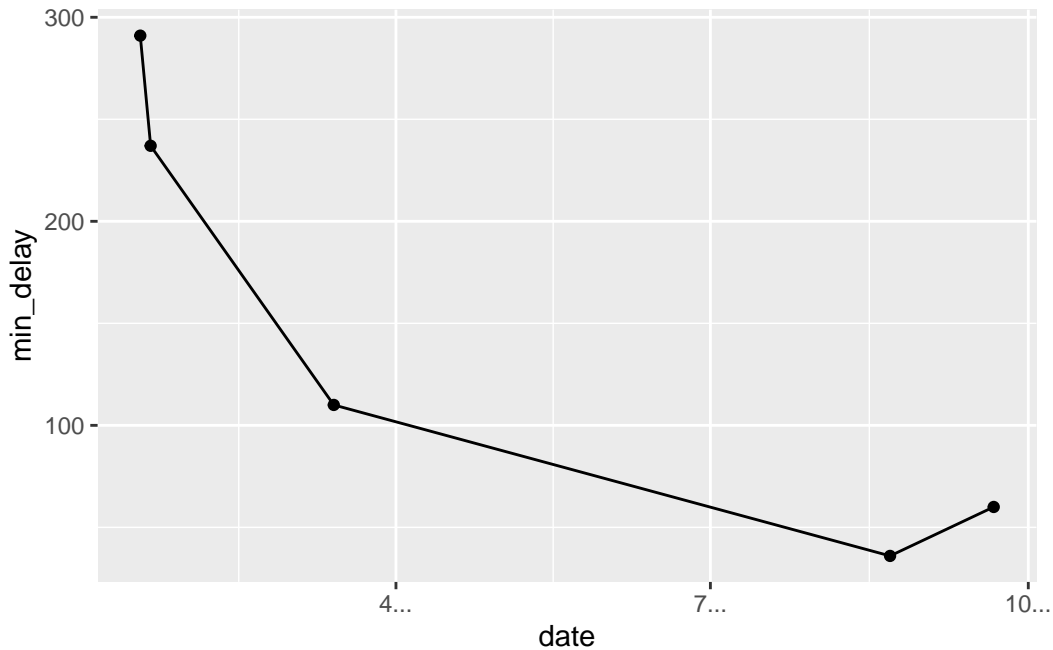
```
delay_2022_mean_delay <- delay_2022 |>
  group_by(station)|>
  summarise(mean(min_delay))
delay_2022_mean_delay <- clean_names(delay_2022_mean_delay)
head(arrange(delay_2022_mean_delay,desc(mean_min_delay),n=5))
```

# A tibble: 6 x 2

	station <chr>	mean_min_delay <dbl>
1	SHEPPARD WEST TO UNION	291
2	KIPLING TO JANE	237
3	MUSEUM TO EGLINTON STA	110
4	WILSON YARD HOSTLER 2	60
5	VIADUCT	36
6	MAIN TO VICTORIA PARK	31

the five stations with the highest mean delays is SHEPPARD WEST TO UNION, Kipling to Jane, Museum to Eglinton Sta, Wilson Yard Hostler 2 and Viaduct.

```
delay_2022|>
  filter(station=='SHEPPARD WEST TO UNION'|station=='KIPLING TO JANE'|station=='MUSEUM TO EGLINTON STA'|station=='WILSON YARD HOSTLER 2'|station=='VIADUCT')
  ggplot(aes(x=date,y=min_delay))+geom_line()+geom_point()
```



## Question 2

```
all_data <- list_packages(limit = 500)
all_data|>
  filter(str_detect(title,pattern = "Campaign"))
```

# A tibble: 5 x 11

	title	id	topics	civic~1	publi~2	excerpt	datas~3	num_r~4	formats	refre~5
	<chr>	<chr>	<chr>	<chr>	<chr>	<chr>	<chr>	<int>	<chr>	<chr>
1	Civic Is~	7d0d~	City ~	Afford~	Inform~	"The 0~	Table	5	XML,JS~	As ava~
2	Election~	67d2~	Finan~	<NA>	City C~	"This ~	Docume~	2	ZIP,XL~	As ava~
3	Election~	f665~	City ~	<NA>	City C~	"This ~	Docume~	2	ZIP,XLS	As ava~
4	Election~	28e5~	City ~	<NA>	City C~	"This ~	Docume~	2	ZIP,XLS	As ava~

```

5 Election~ 2ee8~ City ~ <NA>      City C~ "This ~ Docume~      2 ZIP,XLS As ava~
# ... with 1 more variable: last_refreshed <date>, and abbreviated variable
#   names 1: civic_issues, 2: publisher, 3: dataset_category, 4: num_resources,
#   5: refresh_rate

```

The ID for ‘Elections - Campaign Contributions - 2014 to 2017’ is f6651a40-2f52-46fc-9e04-b760c16edd5c

```

may <- list_package_resources("f6651a40-2f52-46fc-9e04-b760c16edd5c")
may

```

```

# A tibble: 2 x 4
  name                id                format last_mod~1
  <chr>              <chr>              <chr>  <date>
1 campaign-contributions-2014-data  5b230e92-0a22-4a15-9~ ZIP    2019-07-23
2 campaign-contributions-2014-readme-xls aaf736f4-7468-4bda-9~ XLS    2019-07-23
# ... with abbreviated variable name 1: last_modified

```

```

cap <- get_resource("5b230e92-0a22-4a15-9572-0b19cc222985")

```

```

New names:
New names:
New names:
New names:
New names:
New names:
New names:
* `` -> `...2`
* `` -> `...3`

```

```

may_con_2014 <- cap[["2_Mayor_Contributions_2014_election.xls"]]

```

### Question 3

```
colnames(may_con_2014)=may_con_2014[1,]  
may_con_2014<-may_con_2014[-1,]  
may_con_2014<-clean_names(may_con_2014)  
may_con_2014$contribution_amount<- as.numeric(may_con_2014$contribution_amount)
```

### Question 4

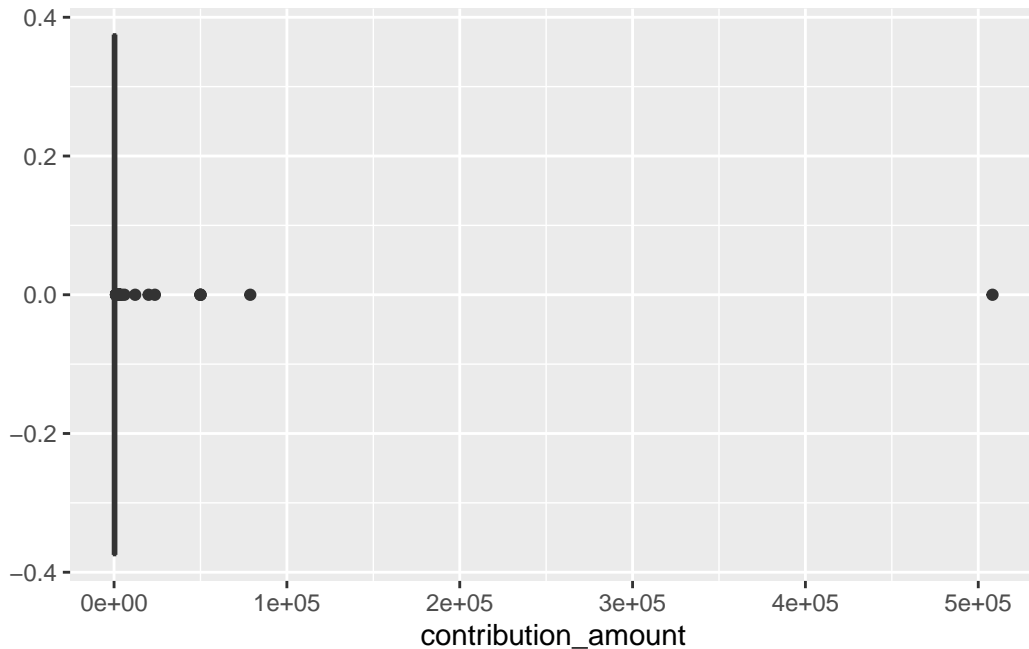
```
may_con_2014 |>  
  summarize(across(everything(), ~ sum(is.na(.x))))
```

```
# A tibble: 1 x 13  
  contributors~1 contr~2 contr~3 contr~4 contr~5 goods~6 contr~7 relat~8 presi~9  
      <int>    <int>    <int>    <int>    <int>    <int>    <int>    <int>    <int>  
1         0   10197         0         0         0   10188         0   10166   10197  
# ... with 4 more variables: authorized_representative <int>, candidate <int>,  
#   office <int>, ward <int>, 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
```

There are a lot of missing variables in the dataset, but they are belong to contributors\_address, good/service description, relationship\_to\_candidate, president\_business\_manager, authorized\_representative and ward, so we do not need to be worried about them. Every variable in the format it should be.

### Question 5

```
ggplot(may_con_2014,aes(x=contribution_amount))+geom_boxplot()
```

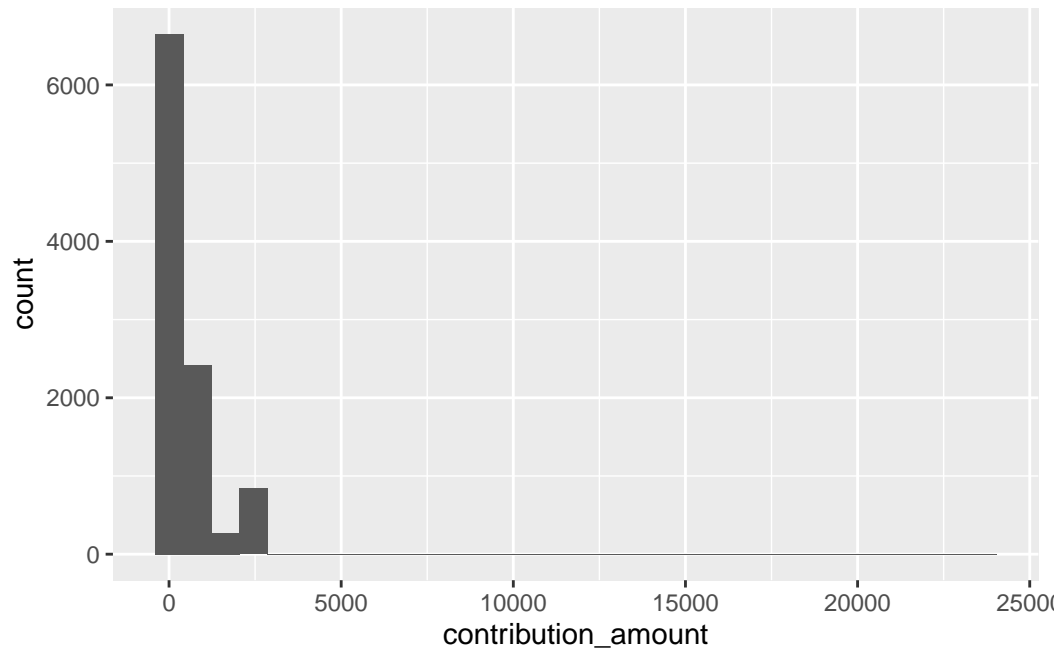


```
may_con_2014|>
  filter(contribution_amount>49999)
```

```
# A tibble: 5 x 13
  contributors~1 contr~2 contr~3 contr~4 contr~5 goods~6 contr~7 relat~8 presi~9
  <chr>          <chr>   <chr>    <dbl> <chr>   <chr>   <chr>   <chr>   <chr>
1 Ford, Doug    <NA>    M9A 2C3 508225. Moneta~ <NA>    Indivi~ Candid~ <NA>
2 Ford, Doug    <NA>    M9A 2C3 50000 Moneta~ <NA>    Indivi~ Candid~ <NA>
3 Ford, Rob     <NA>    M9A 3G9 50000 Moneta~ <NA>    Indivi~ Candid~ <NA>
4 Ford, Rob     <NA>    M9A 3G9 50000 Moneta~ <NA>    Indivi~ Candid~ <NA>
5 Ford, Rob     <NA>    M9A 3G9 78805. Moneta~ <NA>    Indivi~ Candid~ <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
```

```
may_con_2014_no <- may_con_2014|>
  filter(contribution_amount<49999)
ggplot(may_con_2014_no, aes(x=contribution_amount)) + geom_histogram()
```

``stat_bin()`` using ``bins = 30``. Pick better value with ``binwidth``.



There is 5 notable outliers. The outlier shows similar characteristic which is they are all come from Ford family to fund their own campaign. After removing the notable outlier, we can find that the majority of the contribution amount is below 2500.

## Question 6

### total contributions

```
may_con_2014_total <- may_con_2014|>
  group_by(contributors_name)|>
  summarise(sum_contribution_amount)
may_con_2014_total<-clean_names(may_con_2014_total)
head(arrange(may_con_2014_total,desc(sum_contribution_amount)),n=5)
```

```
# A tibble: 5 x 2
  contributors_name  sum_contribution_amount
  <chr>              <dbl>
```

1	Ford, Doug	561225.
2	Ford, Rob	213139.
3	Goldkind, Ari	23624.
4	Thomson, Sarah	6926.
5	Pappalardo, Victor	6300

## mean contribution

```
may_con_2014_mean <- may_con_2014|>
  group_by(contributors_name)|>
  summarise(mean(contribution_amount))
may_con_2014_mean<-clean_names(may_con_2014_mean)
head(arrange(may_con_2014_mean,desc(mean_contribution_amount)),n=5)
```

```
# A tibble: 5 x 2
  contributors_name mean_contribution_amount
  <chr>              <dbl>
1 Ford, Doug        140306.
2 Ford, Rob         30448.
3 Goldkind, Ari     23624.
4 Di Paola, Rocco   6000
5 kindred's Muze    3660
```

## number of contributions

```
may_con_2014_count <- may_con_2014|>
  group_by(contributors_name)|>
  count(contributors_name)
head(arrange(may_con_2014_count,desc(n)),n=5)
```

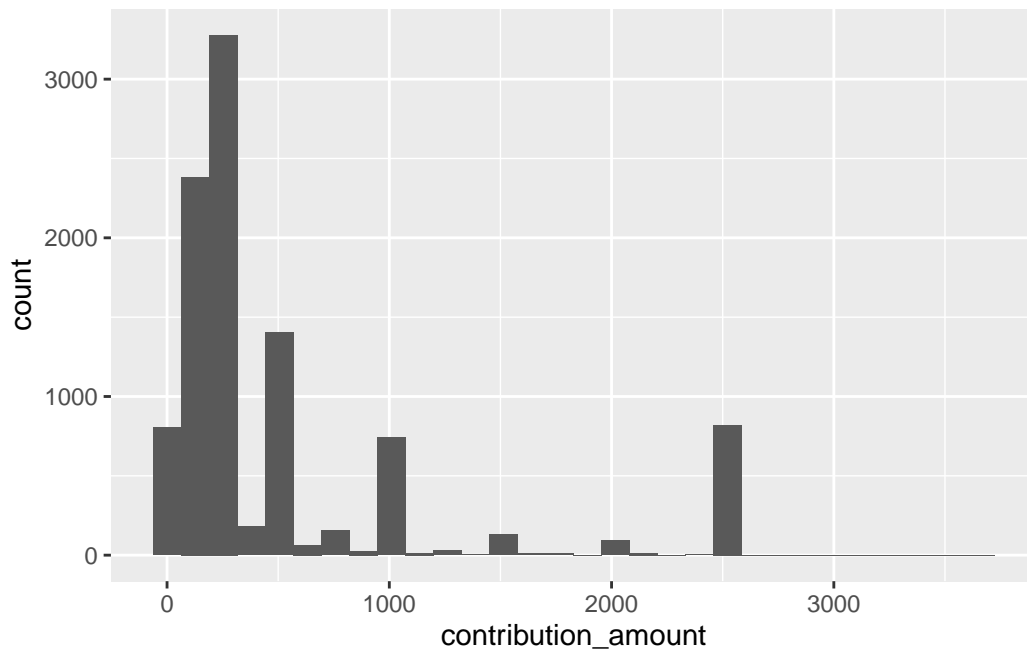
```
# A tibble: 5 x 2
# Groups:   contributors_name [5]
  contributors_name      n
  <chr>                <int>
1 Italiano, Rob        12
2 Cranston, Jacqueline 10
3 Henery, Marjorie     8
4 Martin, Martha       8
5 Quin, Derek          8
```



## Question 7

```
may_con_2014|>  
  filter(!contributors_name==candidate)|>  
  ggplot(aes(x=contribution_amount)) +geom_histogram()
```

`stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.



## Question 8

```
contributor_name <-may_con_2014|>  
  group_by(contributors_name)|>  
  count(candidate)|>  
  get_dupes(contributors_name)  
length(unique(contributor_name$contributors_name))
```

[1] 184