Gender bias in NYC employment

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Introduction

The gender pay gap is an often studied issue in the United States where men are routinely paid more than women year over year. It is a difficult issue to study because it is explained varyingly by sexism, education access, tenure, job preferences, and unequal home duties. We were interested in the following two questions:

- 1. Where is the gender pay gap the worst in NYC?
- 2. Is the gender pay gap partially explained by an agency's male/female ratio?

Exploration/Analysis

```
# Load and convert to RDS
library(tidyverse)
x <- readRDS("payrolldata.rds")</pre>
```

Fix date issues

```
library(lubridate)
x$Agency.Start.Date <- mdy(x$Agency.Start.Date)
x <- x[x$Agency.Start.Date < ymd("2017-01-01"),]
x <- x[x$Agency.Start.Date > ymd("1940-01-01"),]
```

Remove unreliable data

Remove unreliable data

```
x <- x %>%
filter(Fiscal.Year == 2017 &
Regular.Gross.Paid >= 0 &
Total.OT.Paid >= 0 &
Total.Other.Pay >= 0 &
Regular.Gross.Paid >= 0)
```

Bin together agencies by group

Estimate gender based on first name & census

```
x$female <- NA
x$male <- NA
x$female[x$gender == "female"] <- 1
x$female[x$gender == "male"] <- 0
x$male[x$gender == "male"] <- 1
x$male[x$gender == "female"] <- 0</pre>
```

Wrangle

Create a dataframe of male and female salaary difference by grouped agency in terms of regular gross paid.

```
gender.group <- x %>%
  group by (Agency. Name, gender) %>%
  mutate(salary = mean(Regular.Gross.Paid)) %>%
  ungroup() %>%
  group_by(Agency.Name) %>%
  summarize(female=sum(female, na.rm=T),
            male=sum(male, na.rm=T),
            salary = mean(salary)) %>%
  mutate(perc = female/(male+female)) %>%
  select(Agency.Name, perc)
```

Continued

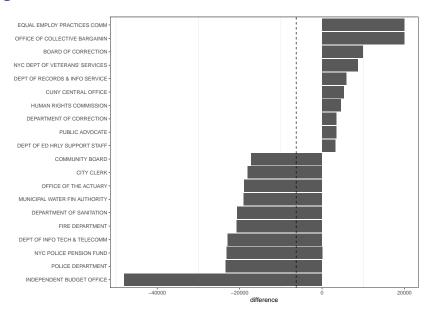
Calculate the actual pay difference in dollars

```
diff <- data.frame(Agency = x2$Agency.Name[x2$gender == "fe
                           difference = x2$salary[x2$gender
                             x2$salary[x2$gender == "male"]]
top <- diff %>%
  top_n(10, difference)
bottom <- diff %>%
  arrange(difference)
bottom <- bottom[1:10.]
topdiffs <- rbind(top, bottom)</pre>
```

Code for Plot 1 - agency vs difference

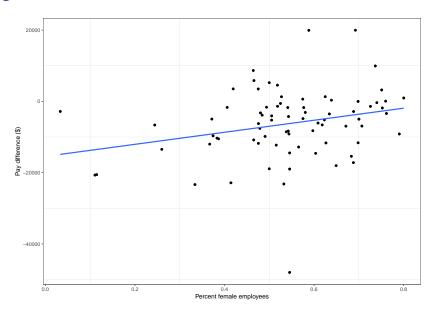
```
p1 <- ggplot(data=topdiffs,
             aes(x=reorder(Agency, difference),
                                 v = difference))+
  geom col()+
  coord flip()+
  geom_hline(yintercept=mean(diff$difference),
             linetype="dashed")+
  theme bw()+
  theme(panel.grid.major=element_blank())+
  xlab("")
```

Figure 1



Code for plot 2 - difference vs. male/female ratio

Figure 2



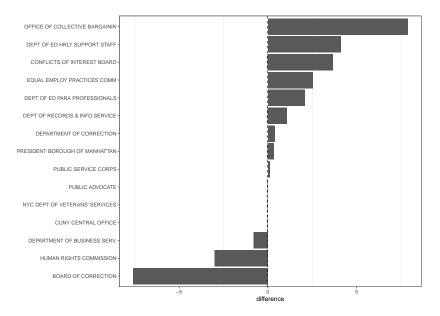
Tenure

```
tenure.group <- x %>%
  group_by(Agency.Name, gender) %>%
  mutate(tenure2 = mean(tenure)) %>%
  ungroup() %>%
  group_by(Agency.Name) %>%
  summarize(female=sum(female, na.rm=T),
            male=sum(male, na.rm=T),
            tenure = mean(tenure)) %>%
  mutate(perc = female/(male+female)) %>%
  select(Agency.Name, perc)
```

Top 15 jobs in which women are paid better than men

```
p11 <- ggplot(data=tenuretop,
              aes(x=reorder(Agency, difference),
                                   v = difference))+
 geom col()+
 coord flip()+
 geom_hline(yintercept=mean(tenure.diff$difference),
            linetype="dashed")+
 theme bw()+
 theme(panel.grid.major=element blank())+
 xlab("")
```

Top 15 jobs in which women are paid better than men



Top 15 jobs in which women are paid less than men

```
p12 <- ggplot(data=tenurebottom,
              aes(x=reorder(Agency, difference),
                                      y = difference))+
 geom col()+
 coord flip()+
 geom_hline(yintercept=mean(tenure.diff$difference),
            linetype="dashed")+
 theme bw()+
 theme(panel.grid.major=element blank())+
 xlab("")
```

Top 15 jobs in which women are paid less than men

