

Sta303- Professional Development

```
library("tidytuesdayR")

## Warning in system("timedatectl", intern = TRUE): running command 'timedatectl'
## had status 1

library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.0 --
## v ggplot2 3.3.3      v purrr  0.3.4
## v tibble  3.1.0      v dplyr  1.0.5
## v tidyr   1.1.3      v stringr 1.4.0
## v readr   1.4.0      v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

library(scales)

##
## Attaching package: 'scales'
##
## The following object is masked from 'package:purrr':
##
##   discard
##
## The following object is masked from 'package:readr':
##
##   col_factor

tuesdata <- tidytuesdayR::tt_load('2021-02-23')

## --- Compiling #TidyTuesday Information for 2021-02-23 ----
## --- There are 2 files available ---
## --- Starting Download ---
##
##   Downloading file 1 of 2: `earn.csv`
##   Downloading file 2 of 2: `employed.csv`
## --- Download complete ---

earn<- tuesdata$earn
employed <- tuesdata$employed

employed %>%
  count(race_gender, sort = TRUE)

## # A tibble: 6 x 2
##   race_gender          n
##   <chr>          <int>
```

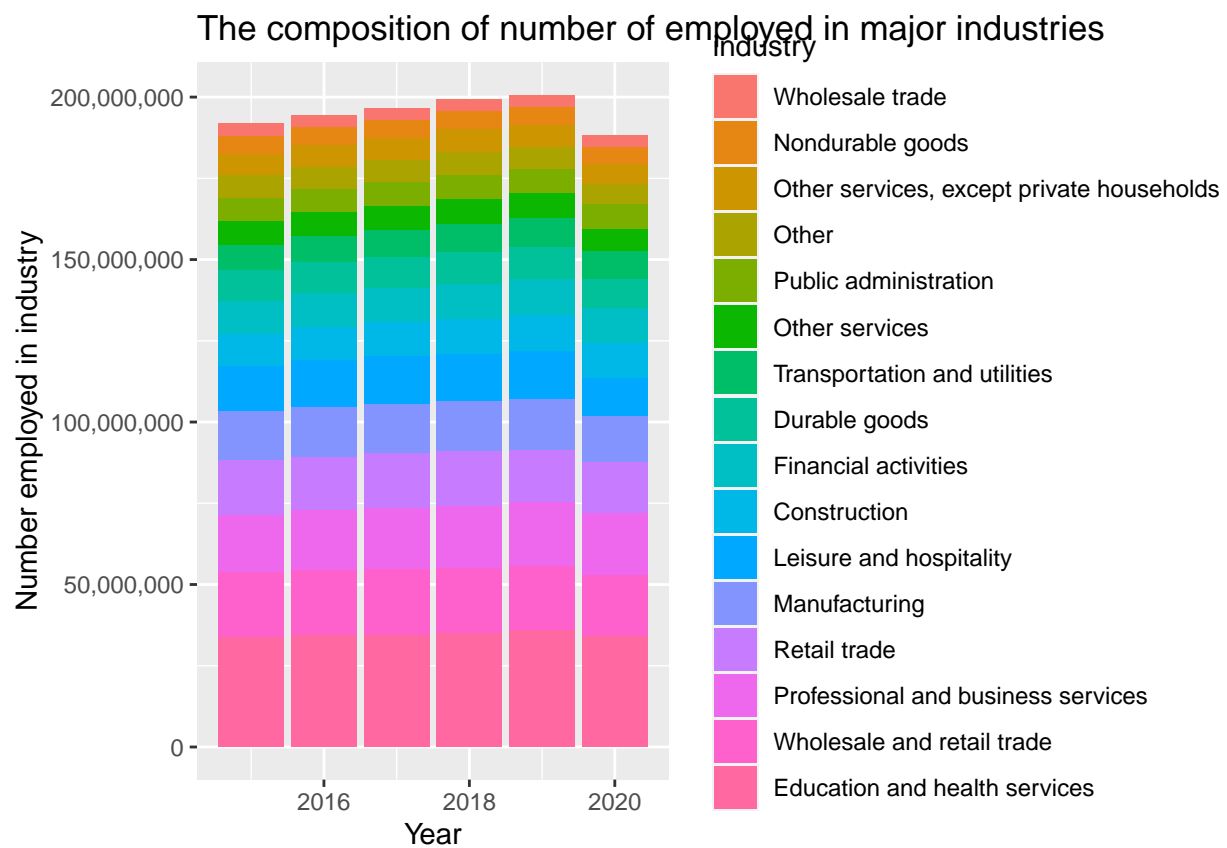
```
## 1 Black or African American 1386
## 2 Men 1386
## 3 TOTAL 1386
## 4 White 1386
## 5 Women 1386
## 6 Asian 1254

# The original data does not categorize gender and race
# We need to specify the categories

employed<- tuesdata$employed %>%
  mutate(dimension = case_when(
    race_gender == "TOTAL" ~ "Total",
    race_gender %in% c("Men", "Women") ~ "Gender",
    TRUE ~ "Race"
  ))
```

Bar plot that shows the composition of number of employed in major industries.

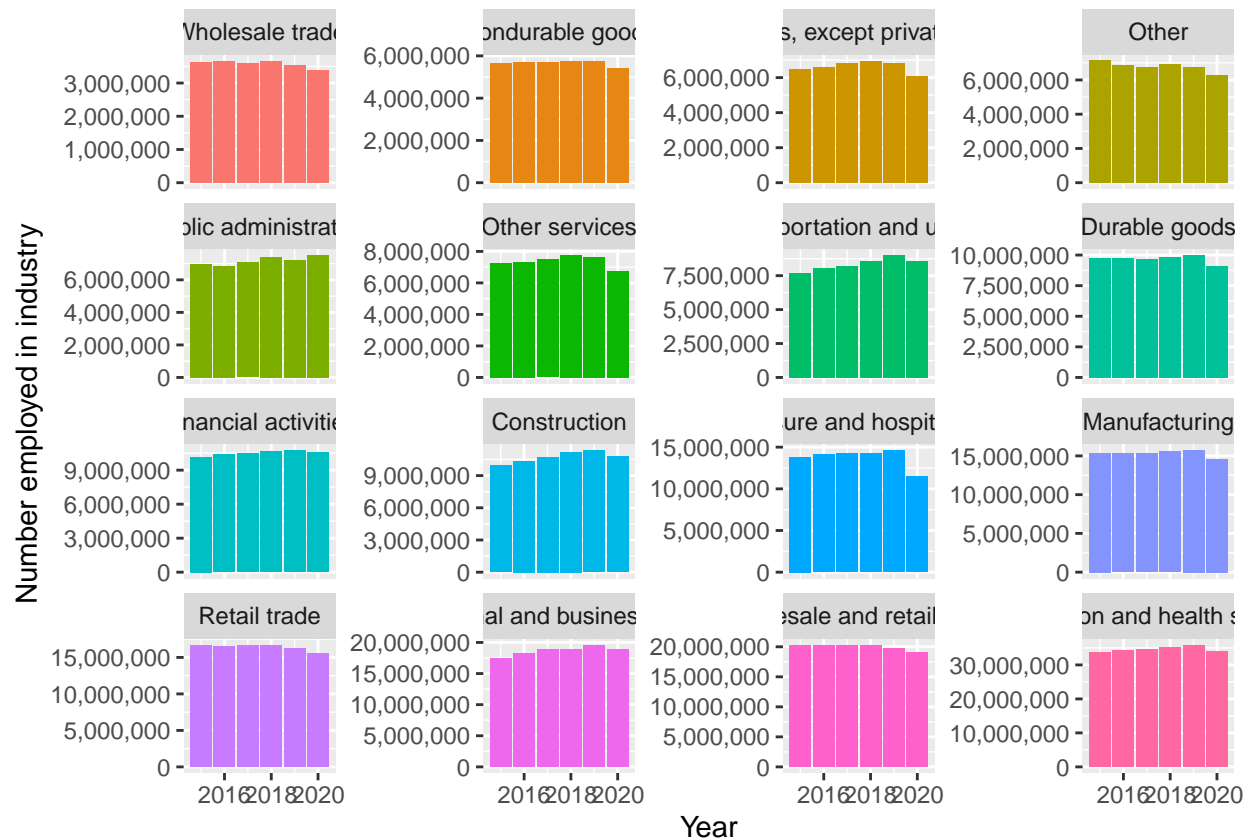
```
employed_clean<- employed %>%
  filter(!is.na(employ_n)) %>%
  mutate(industry = fct_lump(industry, 15, w = employ_n), # The top 15 industries with largest workers.
         industry = fct_reorder(industry, employ_n, sum)) # Aggregate the workers in each industries.
# Bar plot
employed_clean %>%
  filter(dimension == "Total") %>%
  ggplot(aes(x = year, y = employ_n, fill = industry)) +
  geom_col() +
  scale_y_continuous(labels = comma) +
  labs(y = "Number employed in industry",
       x = "Year",
       title = "The composition of number of employed in major industries")
```



Facet bar plots that show the total number of workers in each industry for each year.

Facet bar plots that show the total number of workers in each industry for each year.

```
employed_clean %>%
  filter(dimension == "Total") %>%
  ggplot(aes(x = year, y = employ_n, fill = industry)) +
  geom_col() +
  facet_wrap(~industry, scales = "free_y") +
  theme(legend.position = "none") +
  scale_y_continuous(labels = comma) +
  labs(y = "Number employed in industry",
       x = "Year")
```

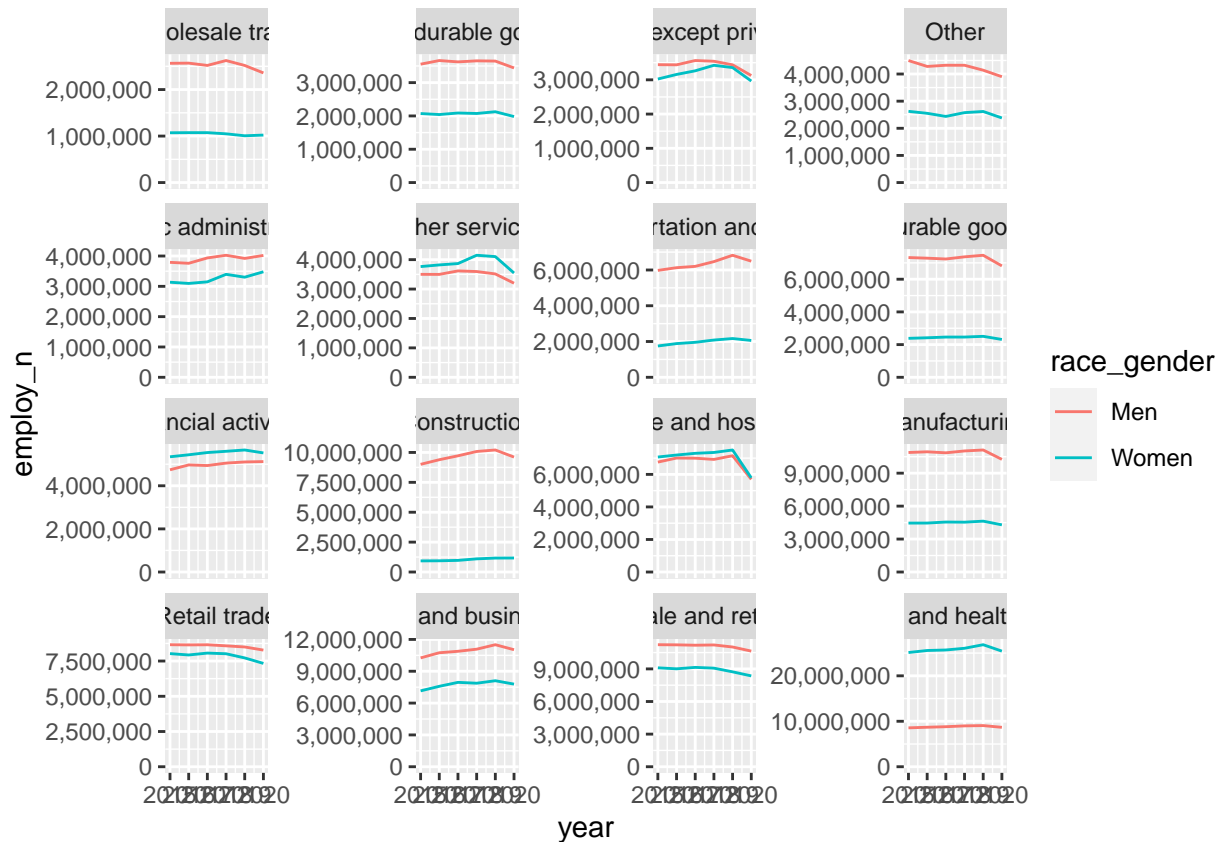


```

employed_gender <- employed_clean %>%
  filter(dimension == "Gender") %>%
  filter(!is.na(employ_n)) %>%
  group_by(industry, year, race_gender)%>%
  summarise(employ_n = sum(employ_n), .groups = "drop")

employed_gender %>%
  ggplot(aes(x = year, y = employ_n, color = race_gender)) +
  geom_line() +
  facet_wrap(~industry, scales = "free_y") +
  scale_y_continuous(labels = comma) +
  expand_limits(y = 0)

```



```

labs(y = "Number employed in industry",
     x = "Year")

```

```

## $y
## [1] "Number employed in industry"
##
## $x
## [1] "Year"
##
## attr(,"class")
## [1] "labels"

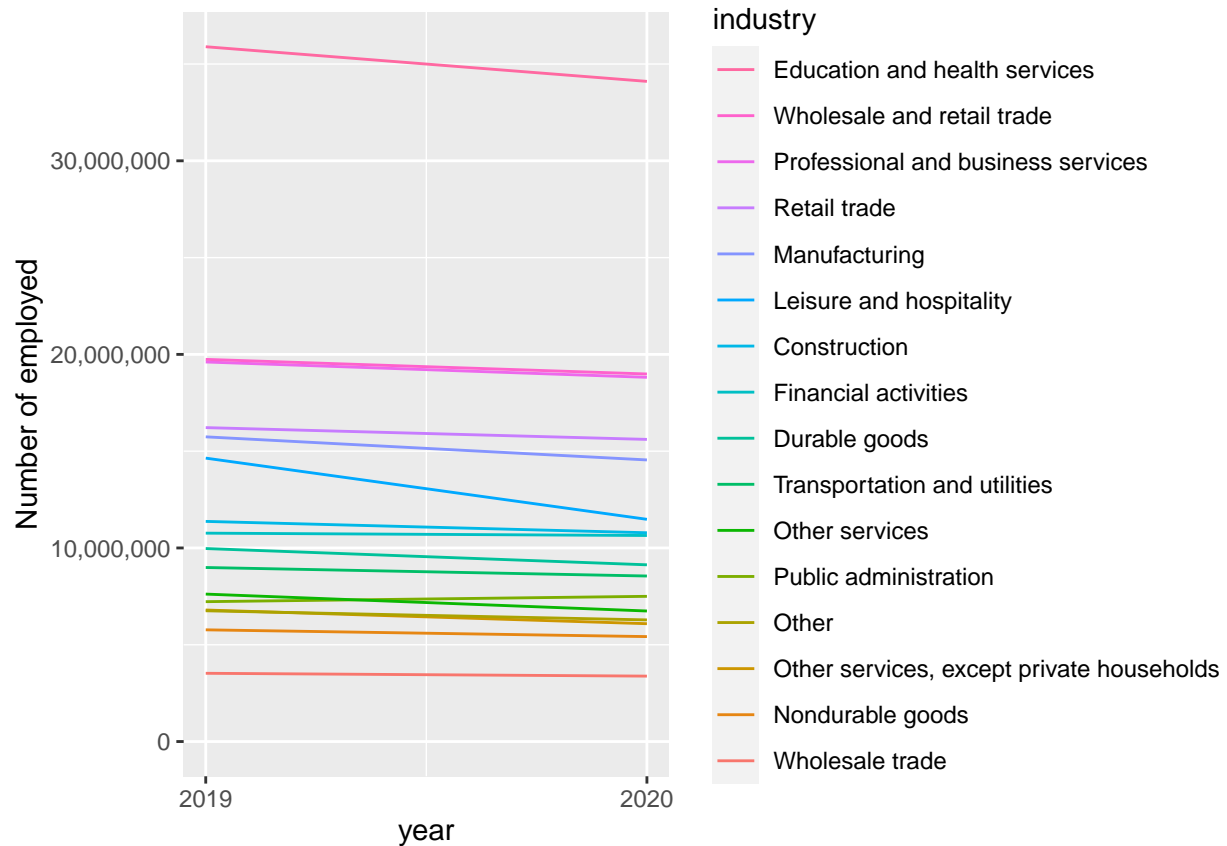
```

Since there is not so much variation in employment between male and female from 2015 to 2019. Let's look at 2019 to 2020 specifically.

```
industry_2019_2020 <- employed_clean %>%
  filter(year %in% c(2019, 2020)) %>%
  group_by(year, industry, dimension, race_gender) %>%
  summarise(employ_n = sum(employ_n), .groups = "drop")
```

A line plot that shows the number of employment from 2019 to 2020 in each industry.

```
industry_2019_2020 %>%
  filter(dimension == "Total") %>%
  ggplot(aes(x = year, y = employ_n, color = industry)) +
  geom_line() +
  expand_limits(y = 0) +
  scale_y_continuous(labels = comma) +
  scale_x_continuous(breaks = 2019:2020) +
  scale_color_discrete(guide = guide_legend(reverse = TRUE)) +
  labs(x = "year",
       y = "Number of employed")
```

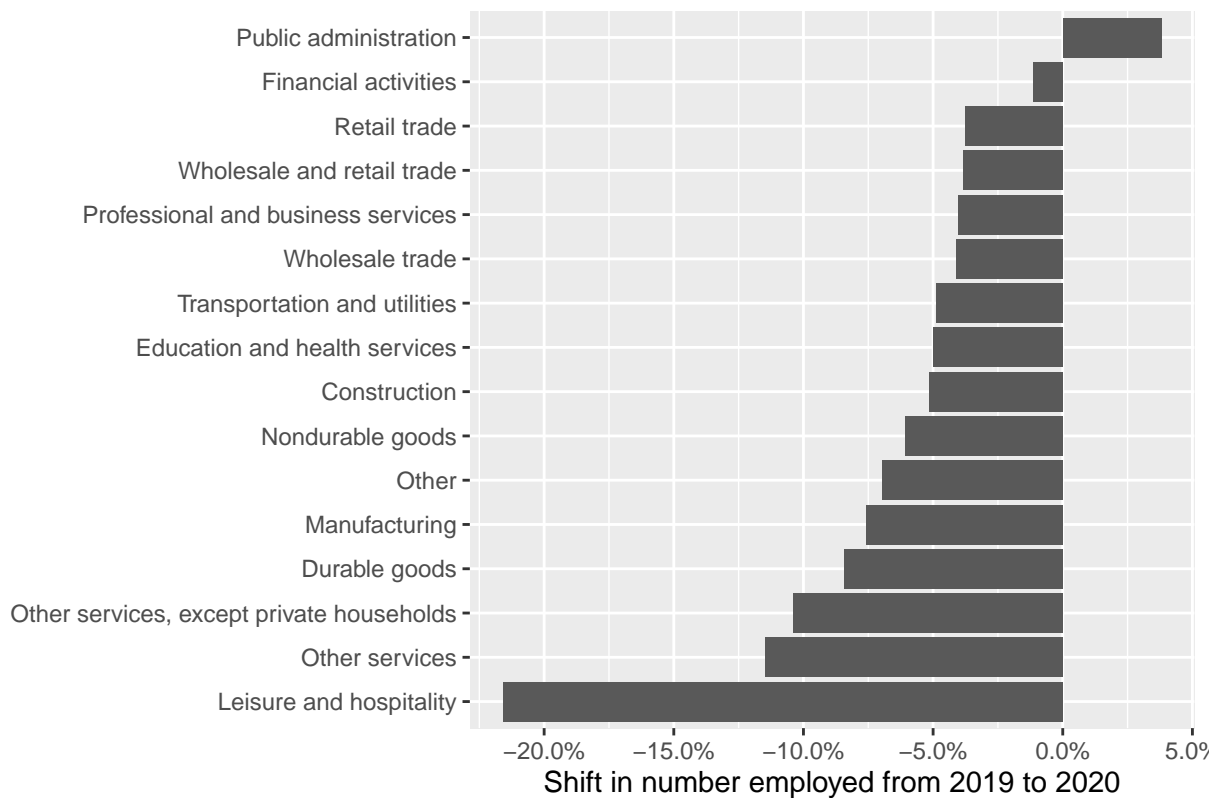


Estimate the effect of pandemic on employment for each industry

```
compare_2019_2020 <- industry_2019_2020 %>%
  arrange(year) %>%
  group_by(industry, dimension, race_gender) %>%
  summarize(ratio = last(employ_n) / first(employ_n),
            change = ratio - 1,
            employed_2019 = first(employ_n), .groups = "drop") %>%
  ungroup()

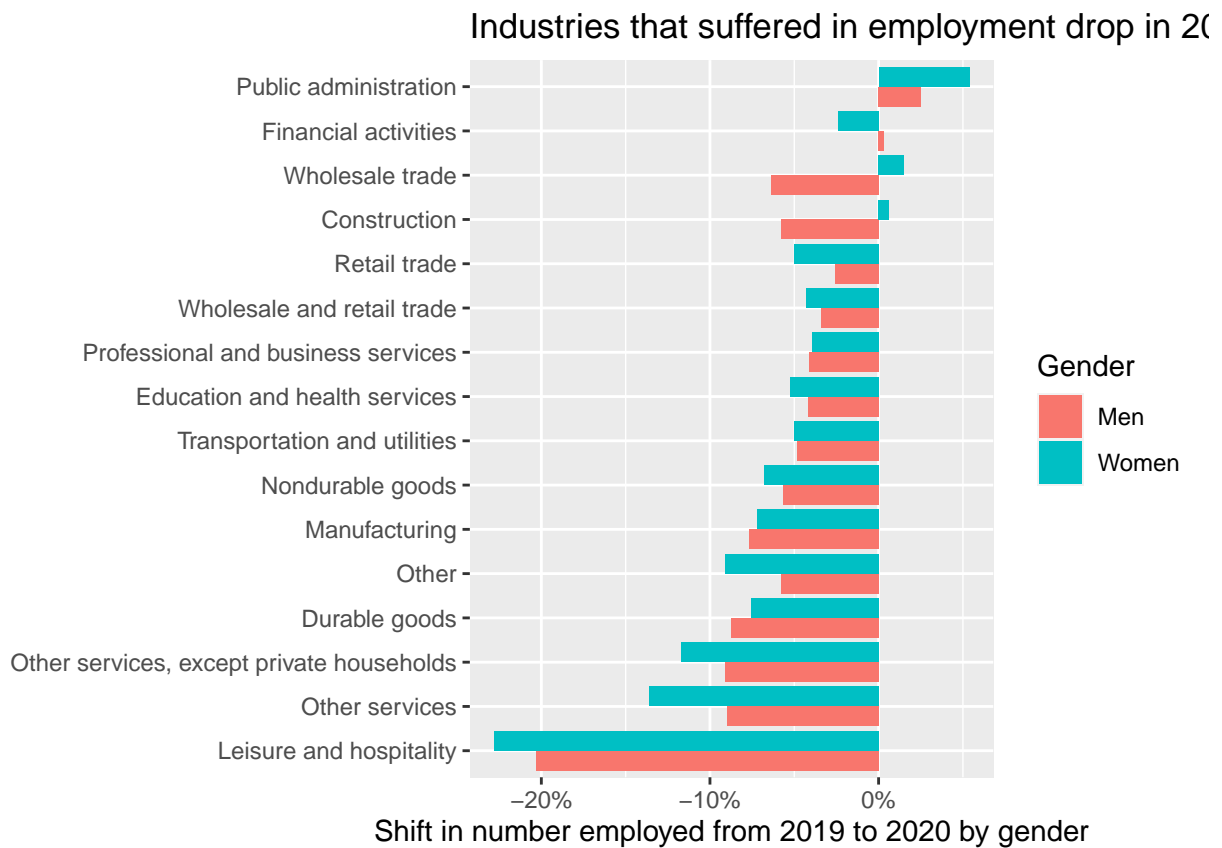
compare_2019_2020 %>%
  filter(dimension == "Total") %>%
  mutate(industry = fct_reorder(industry, change)) %>%
  ggplot(aes(x = change, y = industry)) +
  geom_col() +
  scale_x_continuous(labels = percent) +
  labs(title = "Industries that suffered in employment drop in 2020",
       x = "Shift in number employed from 2019 to 2020",
       y = "")
```

Industries that suffered in employment drop in 2020



Estimate the effect of pandemic on employment for each industry on gender.

```
compare_2019_2020 %>%
  filter(dimension == "Gender") %>%
  mutate(industry = fct_reorder(industry, change)) %>%
  ggplot(aes(x = change, y = industry, fill = race_gender)) +
  geom_col(position = "dodge") +
  scale_x_continuous(labels = percent) +
  labs(title = "Industries that suffered in employment drop in 2020",
       x = "Shift in number employed from 2019 to 2020 by gender",
       y = "",
       fill = "Gender")
```



Estimate the effect of pandemic on employment for each industry on different races.

```
compare_2019_2020 %>%
  filter(dimension == "Race") %>%
  mutate(industry = fct_reorder(industry, change)) %>%
  ggplot(aes(x = change, y = industry, fill = race_gender)) +
  geom_col(position = "dodge") +
  scale_x_continuous(labels = percent) +
  labs(title = "Industries that suffered in employment drop in 2020",
       x = "Shift in number employed from 2019 to 2020 by gender",
       y = "",
       fill = "Race")
```

