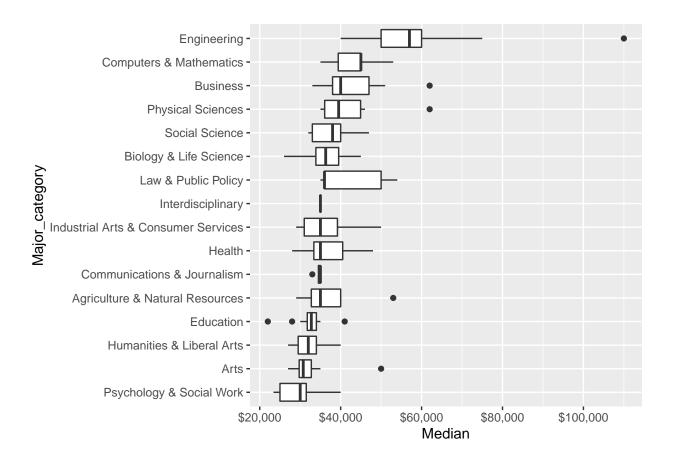
# R Notebook

```
library(tidyverse)
## -- Attaching packages ------
                                                                         ----- tidyverse 1.3
## v ggplot2 3.3.2 v purrr 0.3.4

## v tibble 3.0.3 v dplyr 1.0.2

## v tidyr 1.1.2 v stringr 1.4.0

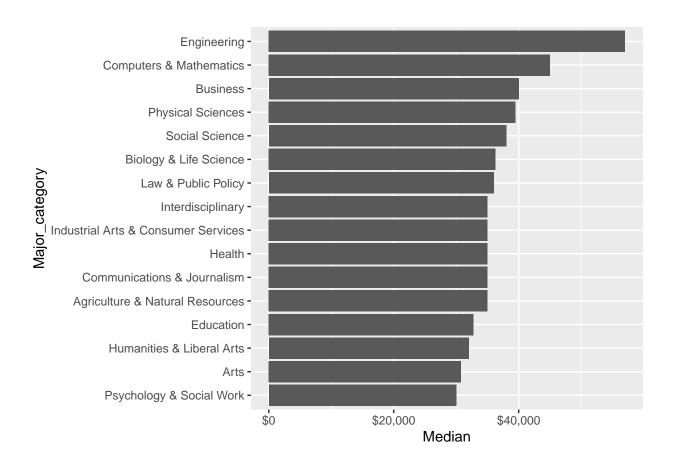
## v readr 1.3.1 v forcats 0.5.0
## -- Conflicts ----- tidyverse_conflicts
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
college_grad <- read_csv("https://raw.githubusercontent.com/fivethirtyeight/data/master/college-majors/</pre>
## Parsed with column specification:
## cols(
##
    .default = col_double(),
## Major = col_character(),
   Major_category = col_character()
## )
## See spec(...) for full column specifications.
view(college_grad)
college_grad %>%
    mutate(Major_category = fct_reorder(Major_category, Median)) %>%
    ggplot(aes(Major_category, Median)) +
    geom_boxplot() +
    scale_y_continuous(labels = scales::dollar_format()) +
    coord_flip()
```



#### Here is the categories that make a lot money upon graduation

```
college_grad %>%
  group_by(Major_category) %>%
  summarize(Median = median(Median)) %>%
  mutate(Major_category = fct_reorder(Major_category, Median)) %>%
  ggplot(aes(Major_category, Median)) +
  geom_col() +
  scale_y_continuous(labels = scales::dollar_format()) +
  coord_flip()
```

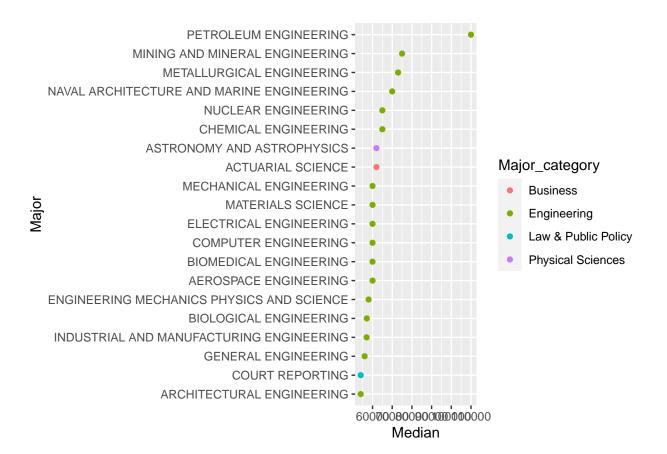
## `summarise()` ungrouping output (override with `.groups` argument)



# Here, we will find the highest top earning majors

```
Majors <- college_grad %>%
    arrange(desc(Median)) %>%
    select(Major, Major_category, Median, P25th, P75th, Sample_size) %>%
    mutate(Major = fct_reorder(Major, Median))

Majors %>% head(20) %>%
    ggplot(aes(Major, Median,, color = Major_category)) +
    geom_point() +
    coord_flip()
```



#### The lowest earning Majors

```
college_grad %>%
   arrange(desc(Median)) %>%
   select(Major, Major_category, Median, P25th, P75th) %>%
   tail(20) %>%

mutate(Major = fct_reorder(Major, Median)) %>%
   ggplot(aes(Major, Median,, color = Major_category)) +
   geom_jitter() +
   coord_flip()
```

```
VISUAL AND PERFORMING ARTS -
TEACHER EDUCATION: MULTIPLE LEVELS -
MISCELLANEOUS PSYCHOLOGY -
HUMANITIES -
HUMAN SERVICES AND COMMUNITY ORGANIZATION -
THEOLOGY AND RELIGIOUS VOCATIONS -
```

- MISCELLANEOUS AGRICULTURE -
- COSMETOLOGY SERVICES AND CULINARY ARTS
- EARLY CHILDHOOD EDUCATION COMMUNICATION DISORDERS SCIENCES AND SERVICES -
  - ANTHROPOLOGY AND ARCHEOLOGY -
    - OTHER FOREIGN LANGUAGES +
      - DRAMA AND THEATER ARTS
    - COMPOSITION AND RHETORIC -
      - ZOOLOGY -

STUDIO ARTS -

- EDUCATIONAL PSYCHOLOGY -
  - CLINICAL PSYCHOLOGY -
- COUNSELING PSYCHOLOGY -
  - LIBRARY SCIENCE -

# 20000 Median

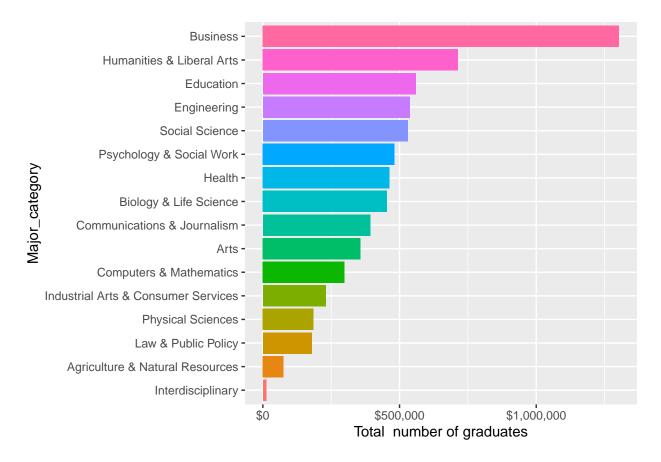
## Major\_category

- Agriculture & Natural Resources
- Arts
- Biology & Life Science
- Education
- Health
- Humanities & Liberal Arts
- Industrial Arts & Consumer Services
- Psychology & Social Work

```
# Majors %>%
      ggplot(aes(Sample_size, Median)) +
      geom_point() +
      scale_x_log10()
# install.packages("tm")
                                   # for text mining
# install.packages("SnowballC")
                                 # for text stemming
                                 # word-cloud generator
# install.packages("wordcloud")
# install.packages("RColorBrewer") # color palettes
# Load the packages
# library("tm")
# library("SnowballC")
# library("wordcloud")
# library("RColorBrewer")
# wordcloud(words = Majors$Major category,
#
            freq = Majors$Median,
#
           min.freq = 1,
#
            max.words =200,
            random.order = TRUE,
#
#
            rot.per = 0.35,
            colors = brewer.pal(8, "Dark2"))
```

#### Most common majors:

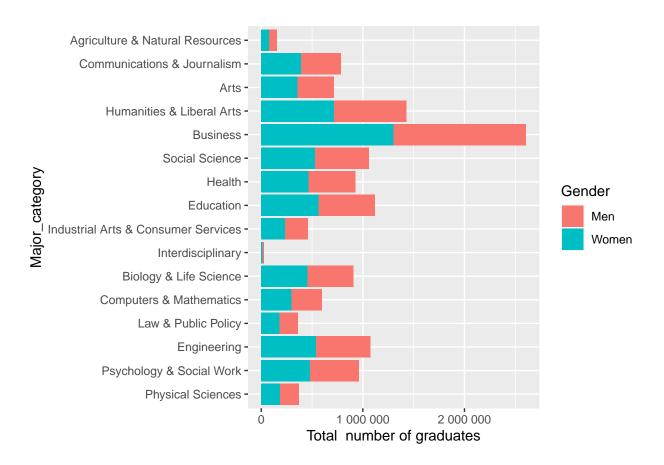
```
college_grad %>%
  count(Major_category, wt = Total, sort = TRUE) %>%
  mutate(Major_category = fct_reorder(Major_category,n)) %>%
  ggplot(aes(Major_category, n, fill = Major_category)) +
  theme(legend.position = "none") +
  geom_col() +
  coord_flip() +
  scale_y_continuous(labels = scales::dollar_format()) +
  labs(y = "Total number of graduates")
```



```
college_grad %>%
  mutate(Major_category = fct_reorder(Major_category, Total)) %>%
  #count(Major_category, wt = Total, sort = TRUE) %>%
  gather(Gender, Number, Men, Women) %>%
  arrange(desc(Major_category)) %>%

ggplot(aes(Major_category, Total, fill = Gender)) +
  #theme(legend.position = "none") +
  geom_col() +
  coord_flip() +
  scale_y_continuous(labels = scales::number_format()) +
  labs(y = "Total number of graduates")
```

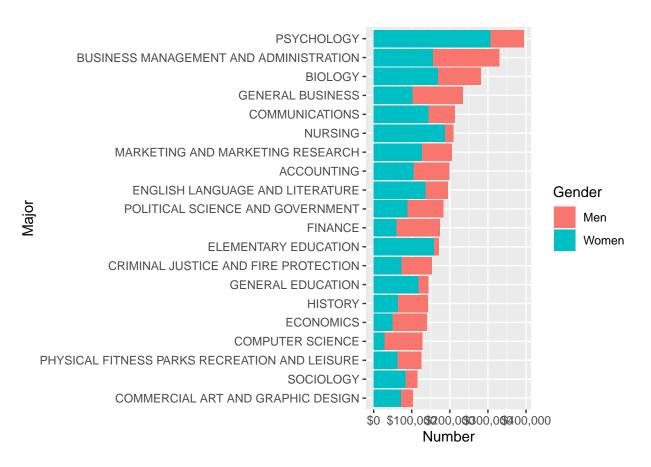
## Warning: Removed 2 rows containing missing values (position\_stack).



### which gender earn more money based on top Major

```
college_grad %>%
   arrange(desc(Total)) %>%
   head(20) %>%
   mutate(Major = fct_reorder(Major, Total)) %>%
   gather(Gender, Number, Men, Women) %>%
   select(Major, Gender, Number) %>%

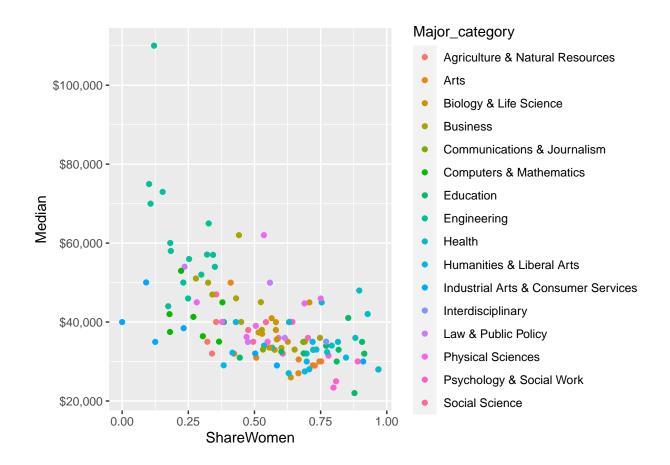
ggplot(aes(Major, Number, fill = Gender)) +
   geom_col() +
   scale_y_continuous(labels = scales::dollar_format()) +
   coord_flip()
```



```
college_grad %>%
  group_by(Major_category, Median) %>%
  summarize_at(vars(Total, Men, Women), sum, na.rm=TRUE) %>%
  mutate(ShareWomen = Women/Total) %>%
  arrange(desc(ShareWomen)) %>%

ggplot(aes(ShareWomen, Median, color = Major_category)) +
  geom_jitter() +
  scale_y_continuous(labels = scales::dollar_format())
```

## Warning: Removed 1 rows containing missing values (geom\_point).



### Total Major in each major's category

```
college_grad %>%
   select(Major, Major_category, Total, ShareWomen, Sample_size, Median) %>%
   add_count(Major_category) %>%
   filter(n >= 10) %>%
   count(Major_category) %>%
   arrange(desc(n))
```

```
## # A tibble: 9 x 2
##
     Major_category
                                          n
     <chr>
##
                                      <int>
## 1 Engineering
                                          29
## 2 Education
                                          16
## 3 Humanities & Liberal Arts
                                          15
## 4 Biology & Life Science
                                          14
## 5 Business
                                          13
## 6 Health
                                          12
## 7 Computers & Mathematics
                                          11
## 8 Agriculture & Natural Resources
                                         10
## 9 Physical Sciences
                                          10
```

### Predict the median of ShareWomen