

R Notebook

Principles of Data Visualization and Introduction to ggplot2

I have provided you with data about the 5,000 fastest growing companies in the US, as compiled by Inc. magazine. lets read this in:

```
inc <- read.csv("https://raw.githubusercontent.com/charleyferrari/CUNY_DATA_608/master/module1/Data/inc.csv")
```

And lets preview this data:

```
head(inc)
```

```
##      Rank      Name Growth_Rate  Revenue
## 1      1      Fuhu      421.48 1.179e+08
## 2      2 FederalConference.com 248.31 4.960e+07
## 3      3      The HCI Group 245.45 2.550e+07
## 4      4      Bridger      233.08 1.900e+09
## 5      5      DataXu      213.37 8.700e+07
## 6      6 MileStone Community Builders 179.38 4.570e+07
##
##      Industry Employees      City State
## 1 Consumer Products & Services      104 El Segundo CA
## 2      Government Services      51 Dumfries VA
## 3      Health      132 Jacksonville FL
## 4      Energy      50 Addison TX
## 5 Advertising & Marketing      220 Boston MA
## 6      Real Estate      63 Austin TX
```

```
summary(inc)
```

```
##      Rank      Name      Growth_Rate      Revenue
## Min.   : 1 Length:5001 Min.   : 0.340 Min.   :2.000e+06
## 1st Qu.:1252 Class :character 1st Qu.: 0.770 1st Qu.:5.100e+06
## Median :2502 Mode  :character Median : 1.420 Median :1.090e+07
## Mean   :2502 Mean   : 4.612 Mean   :4.822e+07
## 3rd Qu.:3751 3rd Qu.: 3.290 3rd Qu.:2.860e+07
## Max.   :5000 Max.   :421.480 Max.   :1.010e+10
##
##      Industry      Employees      City      State
## Length:5001 Min.   : 1.0 Length:5001 Length:5001
## Class :character 1st Qu.: 25.0 Class :character Class :character
## Mode :character Median : 53.0 Mode :character Mode :character
## Mean   : 232.7
## 3rd Qu.: 132.0
## Max.   :66803.0
## NA's   :12
```

Think a bit on what these summaries mean. Use the space below to add some more relevant non-visual exploratory information you think helps you understand this data:

```
# I used the describe feautre from Hmisc to get another look at the data.
library(Hmisc)
```

```
## Warning: package 'Hmisc' was built under R version 4.0.3
```

```
## Loading required package: lattice
```

```
## Loading required package: survival
```

```
## Loading required package: Formula
```

```
## Warning: package 'Formula' was built under R version 4.0.3
```

```
## Loading required package: ggplot2
```

```
##
```

```
## Attaching package: 'Hmisc'
```

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

```
##
```

```
##      format.pval, units
```

```
describe(inc)
```

```
## inc
```

```
##
```

```
## 8 Variables      5001 Observations
```

```
## -----
```

```
## Rank
```

```
##      n missing distinct      Info      Mean      Gmd      .05      .10
```

```
##    5001         0     4999         1     2502     1667     252     502
```

```
##      .25      .50      .75      .90      .95
```

```
##    1252     2502     3751     4501     4751
```

```
##
```

```
## lowest :      1      2      3      4      5, highest: 4996 4997 4998 4999 5000
```

```
## -----
```

```
## Name
```

```
##      n missing distinct
```

```
##    5001         0     5001
```

```
##
```

```
## lowest : (Add)ventures
```

```
@Properties
```

```
1-Stop Transl
```

```
## highest: Zoup!
```

```
ZT Wealth and Altus Group of Companies Zumasy
```

```
## -----
```

```
## Growth_Rate
```

```
##      n missing distinct      Info      Mean      Gmd      .05      .10
```

```
##    5001         0     1147         1     4.612     6.493     0.43     0.50
```

```
##      .25      .50      .75      .90      .95
```

```
##     0.77     1.42     3.29     9.12    17.16
```

```
##
## lowest : 0.34 0.35 0.36 0.37 0.38, highest: 213.37 233.08 245.45 248.31 421.48
## -----
## Revenue
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    5001      0      1069      1 48222535 75111227 2400000 3000000
##      .25      .50      .75      .90      .95
## 5100000 10900000 28600000 76900000 155600000
##
## lowest : 2.00e+06 2.10e+06 2.20e+06 2.30e+06 2.40e+06
## highest: 3.80e+09 4.50e+09 4.60e+09 4.70e+09 1.01e+10
## -----
## Industry
##      n missing distinct
##    5001      0      25
##
## lowest : Advertising & Marketing      Business Products & Services Computer Hardware      Cons
## highest: Retail      Security      Software      Tele
## -----
## Employees
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    4989      12      691      1 232.7 365.6 10.0 14.0
##      .25      .50      .75      .90      .95
##    25.0 53.0 132.0 351.2 688.0
##
## lowest : 1 2 3 4 5, highest: 17057 18887 20000 32000 66803
## -----
## City
##      n missing distinct
##    5001      0      1519
##
## lowest : Acton      Addison      Adrian      Agoura Hills Aiea
## highest: Worthington Wyomissing Yonkers      Youngsville Zumbrota
## -----
## State
##      n missing distinct
##    5001      0      52
##
## lowest : AK AL AR AZ CA, highest: VT WA WI WV WY
## -----
```

Question 1

Create a graph that shows the distribution of companies in the dataset by State (ie how many are in each state). There are a lot of States, so consider which axis you should use. This visualization is ultimately going to be consumed on a 'portrait' oriented screen (ie taller than wide), which should further guide your layout choices.

```
library(ggplot2)
library(ggthemes)
```

```
## Warning: package 'ggthemes' was built under R version 4.0.3
```

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.0 --
```

```
## 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
## v purrr   0.3.4
```

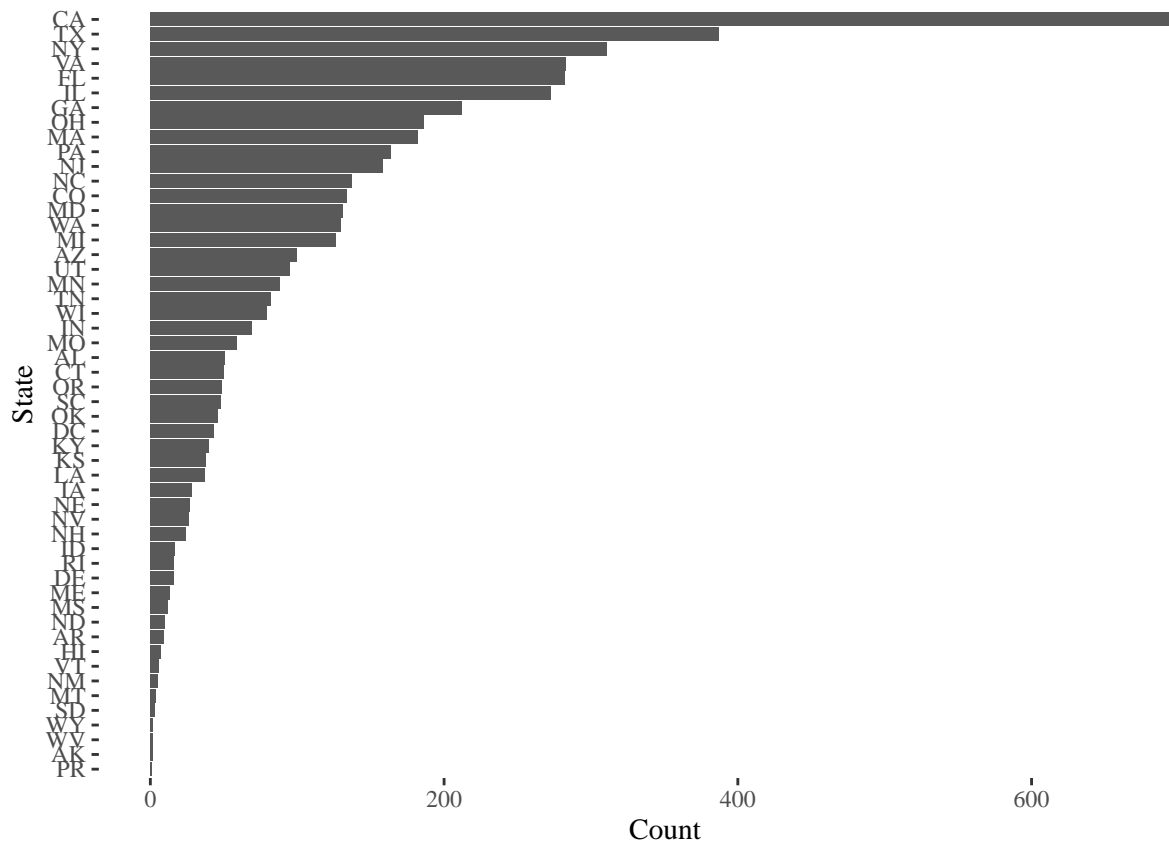
```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter()      masks stats::filter()
## x dplyr::lag()         masks stats::lag()
## x dplyr::src()         masks Hmisc::src()
## x dplyr::summarize()   masks Hmisc::summarize()
```

```
by_state <- inc %>%
  group_by(State) %>%
  summarise(count = n())
```

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

```
ggplot(by_state, aes(x = reorder(State, count), y = count)) +
  geom_col() +
  labs(x = "State",
       y = "Count") +
  coord_flip() +
  theme_tufte()
```



Question 2

Lets dig in on the state with the 3rd most companies in the data set. Imagine you work for the state and are interested in how many people are employed by companies in different industries. Create a plot that shows the average and/or median employment by industry for companies in this state (only use cases with full data, use R's `complete.cases()` function.) In addition to this, your graph should show how variable the ranges are, and you should deal with outliers.

```
# First I filtered down to just complete cases in NY.
ny <- filter(inc, State == "NY")
ny_complete <- ny[complete.cases(ny), ]

# Next, I looked at the data raw. I grouped the data by industry and created a boxplot to check for dis
ny_by_ave_empl <- ny_complete %>%
  group_by(Industry) %>%
  mutate(ave_empl = mean(Employees))

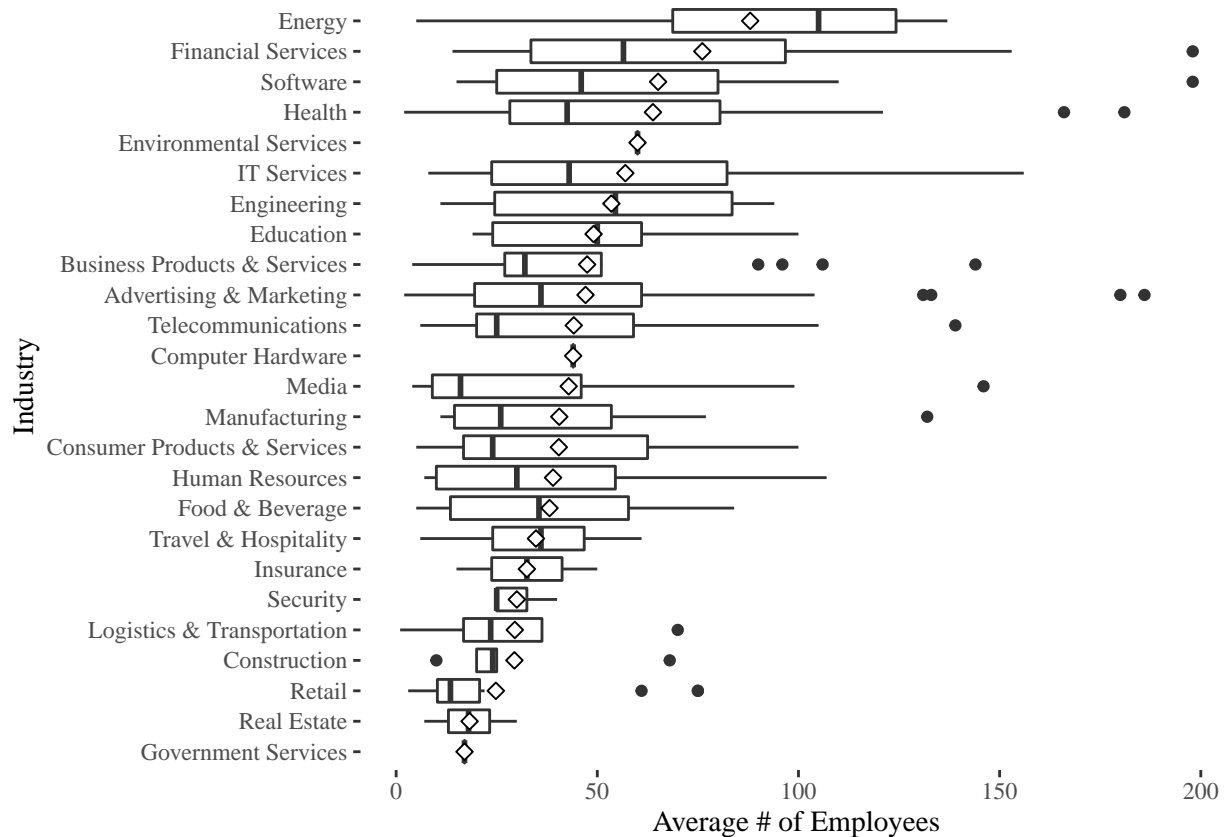
ggplot(ny_by_ave_empl, aes(x = reorder(Industry, ave_empl), y = Employees)) +
  geom_boxplot() +
  labs(x = "Industry",
       y = "Average # of Employees") +
  coord_flip() +
  theme_tufte()
```



```
# Most companies had employees numbering in the hundreds, so the thousands were clear outliers. I removed them.
ny_by_ave_empl <- ny_complete %>%
  filter(Employees < 200) %>%
  group_by(Industry) %>%
  mutate(ave_empl = mean(Employees))

ggplot(ny_by_ave_empl, aes(x = reorder(Industry, ave_empl), y = Employees)) +
  geom_boxplot() +
  labs(x = "Industry",
       y = "Average # of Employees") +
  coord_flip() +
  stat_summary(fun.y = "mean", geom = "point", shape = 23, size = 2, fill = "white") +
  theme_tufte()
```

```
## Warning: 'fun.y' is deprecated. Use 'fun' instead.
```



Question 3

Now imagine you work for an investor and want to see which industries generate the most revenue per employee. Create a chart that makes this information clear. Once again, the distribution per industry should be shown.

```
ny_by_rev_per_empl <- ny_complete %>%
  group_by(Industry) %>%
  mutate(per_empl = (mean(Revenue)/1000)/mean(Employees)) %>%
  summarise(rev_per_empl = mean(per_empl))
```

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

```
ggplot(ny_by_rev_per_empl, aes(x = reorder(Industry, rev_per_empl), y = rev_per_empl)) +
  geom_col() +
  labs(x = "Industry",
       y = "$1000 Revenue Per Employee") +
  coord_flip() +
  theme_tufte()
```

