

DATA 608 - Project 1

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02/08/2018

```
# Load packages
packages <- c("tidyverse")
invisible(lapply(packages, library, character.only = T))
```

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
## Min.   : 1 (Add)ventures : 1 Min.   : 0.340
## 1st Qu.:1252 @Properties   : 1 1st Qu.: 0.770
## Median :2502 1-Stop Translation USA: 1 Median : 1.420
## Mean   :2502 110 Consulting   : 1 Mean   : 4.612
## 3rd Qu.:3751 11thStreetCoffee.com : 1 3rd Qu.: 3.290
## Max.   :5000 123 Exteriors    : 1 Max.   :421.480
##      (Other) :4995
##
##      Revenue      Industry      Employees
## Min.   :2.000e+06 IT Services : 733 Min.   : 1.0
## 1st Qu.:5.100e+06 Business Products & Services: 482 1st Qu.: 25.0
## Median :1.090e+07 Advertising & Marketing : 471 Median : 53.0
## Mean   :4.822e+07 Health : 355 Mean   : 232.7
## 3rd Qu.:2.860e+07 Software : 342 3rd Qu.: 132.0
## Max.   :1.010e+10 Financial Services : 260 Max.   :66803.0
##      (Other) :2358 NA's :12
##
##      City      State
## New York : 160 CA : 701
```

```
## Chicago      : 90 TX      : 387
## Austin       : 88 NY      : 311
## Houston      : 76 VA      : 283
## San Francisco: 75 FL      : 282
## Atlanta      : 74 IL      : 273
## (Other)      :4438 (Other):2764
```

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:

```
glimpse(inc) # View number of rows and columns, variable types
```

```
## Observations: 5,001
## Variables: 8
## $ Rank      <int> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15,...
## $ Name      <fct> Fuhu, FederalConference.com, The HCI Group, Bridge...
## $ Growth_Rate <dbl> 421.48, 248.31, 245.45, 233.08, 213.37, 179.38, 17...
## $ Revenue    <dbl> 1.179e+08, 4.960e+07, 2.550e+07, 1.900e+09, 8.700e...
## $ Industry   <fct> Consumer Products & Services, Government Services,...
## $ Employees  <int> 104, 51, 132, 50, 220, 63, 27, 75, 97, 15, 149, 16...
## $ City       <fct> El Segundo, Dumfries, Jacksonville, Addison, Bosto...
## $ State      <fct> CA, VA, FL, TX, MA, TX, TN, CA, UT, RI, VA, CA, FL...
```

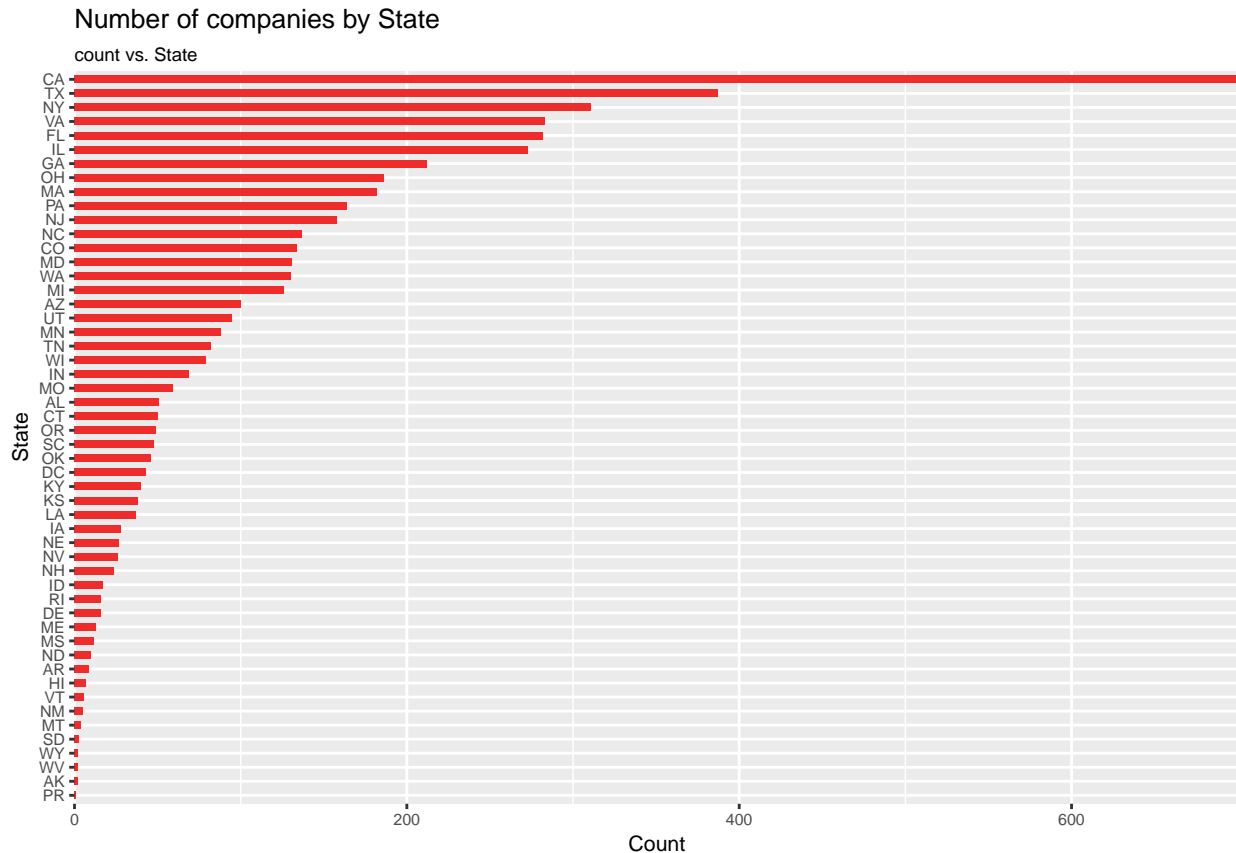
Question 1

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.

Since we’ll be displaying the output on a portrait screen, we want to flip the coordinates, and use the y axis.

```
# Group by state, and take the count
state.count <- inc %>%
  count(State)

ggplot(state.count, aes(x=reorder(State, n), y=n)) +
  geom_bar(stat="identity", fill="firebrick2", width=0.5) +
  coord_flip() +
  labs(title="Number of companies by State",
       subtitle="count vs. State",
       x = "State",
       y = "Count") +
  theme_grey(base_size = 8) +
  scale_y_continuous(expand=c(0,0))
```



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.

```
inc <- inc[complete.cases(inc),]

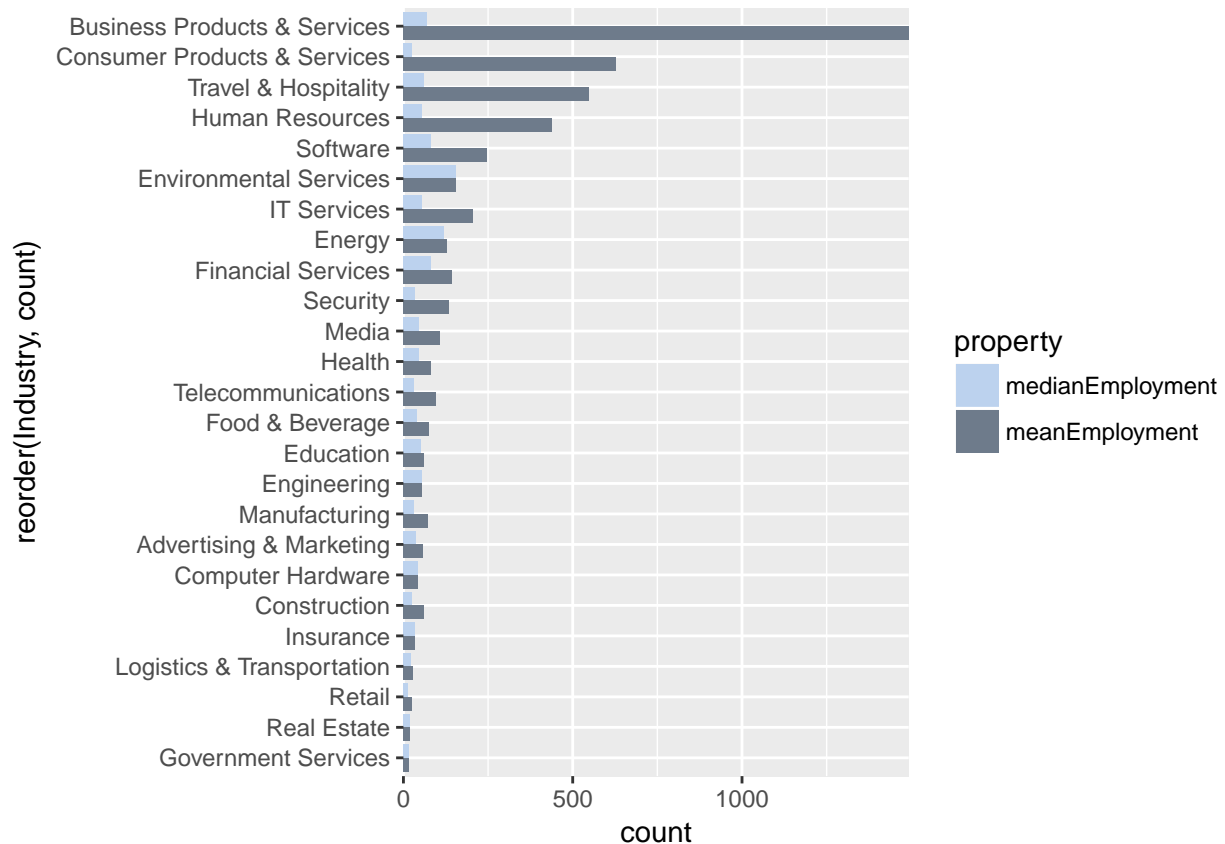
find.third.state <- state.count %>%
  arrange(desc(n))
find.third.state <- find.third.state$State[[3]]

third.state <- filter(inc, State == find.third.state) %>%
  filter(complete.cases(.))

third.state.table <- group_by(third.state, Industry) %>%
  summarize(meanEmployment = mean(Employees),
            medianEmployment = median(Employees))
) %>%
gather(property, count, meanEmployment, medianEmployment)

ggplot(third.state.table, aes(x=reorder(Industry, count), y=count)) +
  geom_bar(stat="identity", position="dodge", aes(fill=property)) +
```

```
coord_flip() +
scale_fill_manual(values=c("lightsteelblue4", "lightsteelblue2"), guide=guide_legend(reverse=T)) +
scale_y_continuous(expand=c(0,0))
```



Business Products & Services appears to be an outlier. If we check the average difference between industry means, we can somewhat regulate the outlier.

```
state.outlier <- third.state.table %>%
  filter(property == "meanEmployment")
state.outlier <- state.outlier[-c(2),] # drop the outlier
mean(diff(sort(state.outlier$count))) # calculate the average difference
```

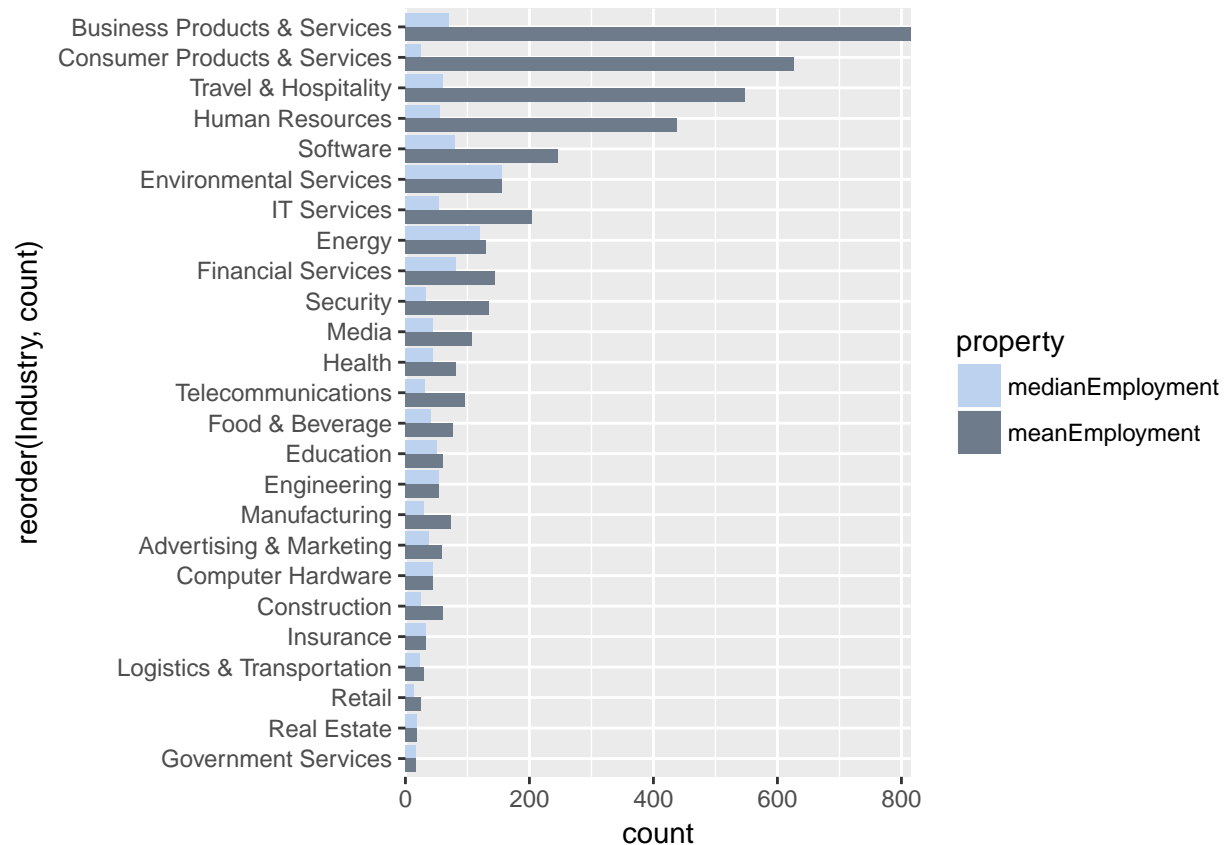
```
## [1] 26.49105
```

```
max(diff(sort(state.outlier$count)))
```

```
## [1] 191.6224
```

We see the average difference between consecutive leading industries is 26.49105, with a max of 191.6224. With this in mind, I think we can cap the outlier at ~200 more than the second-highest.

```
third.state.edited <- third.state.table
third.state.edited[2,3] <- 815
ggplot(third.state.edited, aes(x=reorder(Industry, count), y=count)) +
  geom_bar(stat="identity", position="dodge", aes(fill=property)) +
  coord_flip() +
  scale_fill_manual(values=c("lightsteelblue4", "lightsteelblue2"), guide=guide_legend(reverse=T)) +
  scale_y_continuous(expand=c(0,0))
```

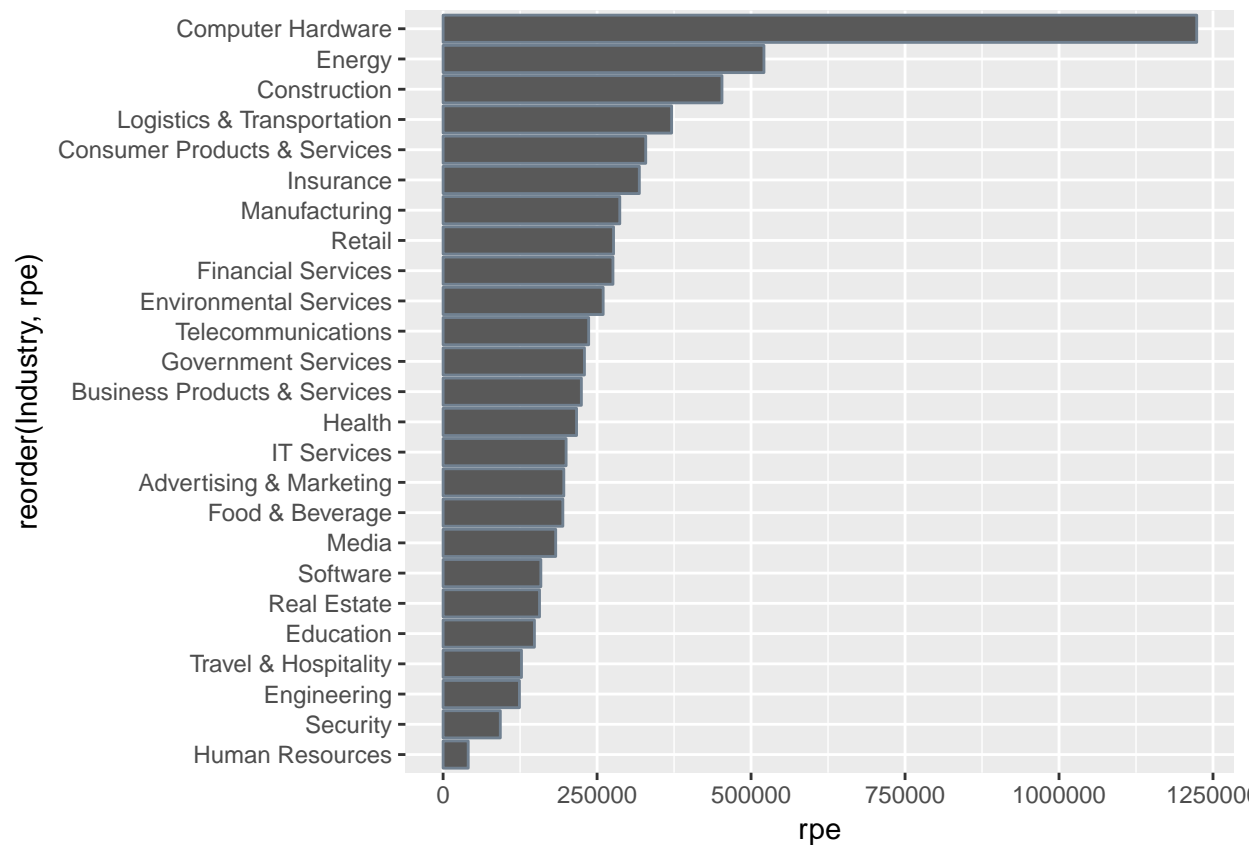


Question 3

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.

```
most.profitable <- inc %>%
  group_by(Industry) %>%
  summarize(rpe = sum(Revenue)/sum(Employees))

ggplot(most.profitable, aes(x=reorder(Industry, rpe), y=rpe)) +
  geom_bar(stat="identity", colour="slategrey") +
  coord_flip()
```



Once again, we have an outlier; in this case, it's **Computer Hardware**. Using the same method as in problem 2, we'll cap the outlier.

```
mp.edited <- most.profitable
mean(diff(sort(mp.edited$rpe)))
```

```
## [1] 49284.53
```

```
max(diff(sort(mp.edited$rpe)))
```

```
## [1] 702642.5
```

```
mp.edited[3,2] <- 600000
```

```
ggplot(mp.edited, aes(x=reorder(Industry, rpe), y=rpe)) +
  geom_bar(stat="identity", colour="slategrey") +
  coord_flip()
```

