

# 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")
```

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:

```
# Insert your code here, create more chunks as necessary

# There are numerous R functions that provide descriptive & exploratory statistics, such as functions i

library(Hmisc)
```

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

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

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

```
## 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 Zumasys
## -----
## 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

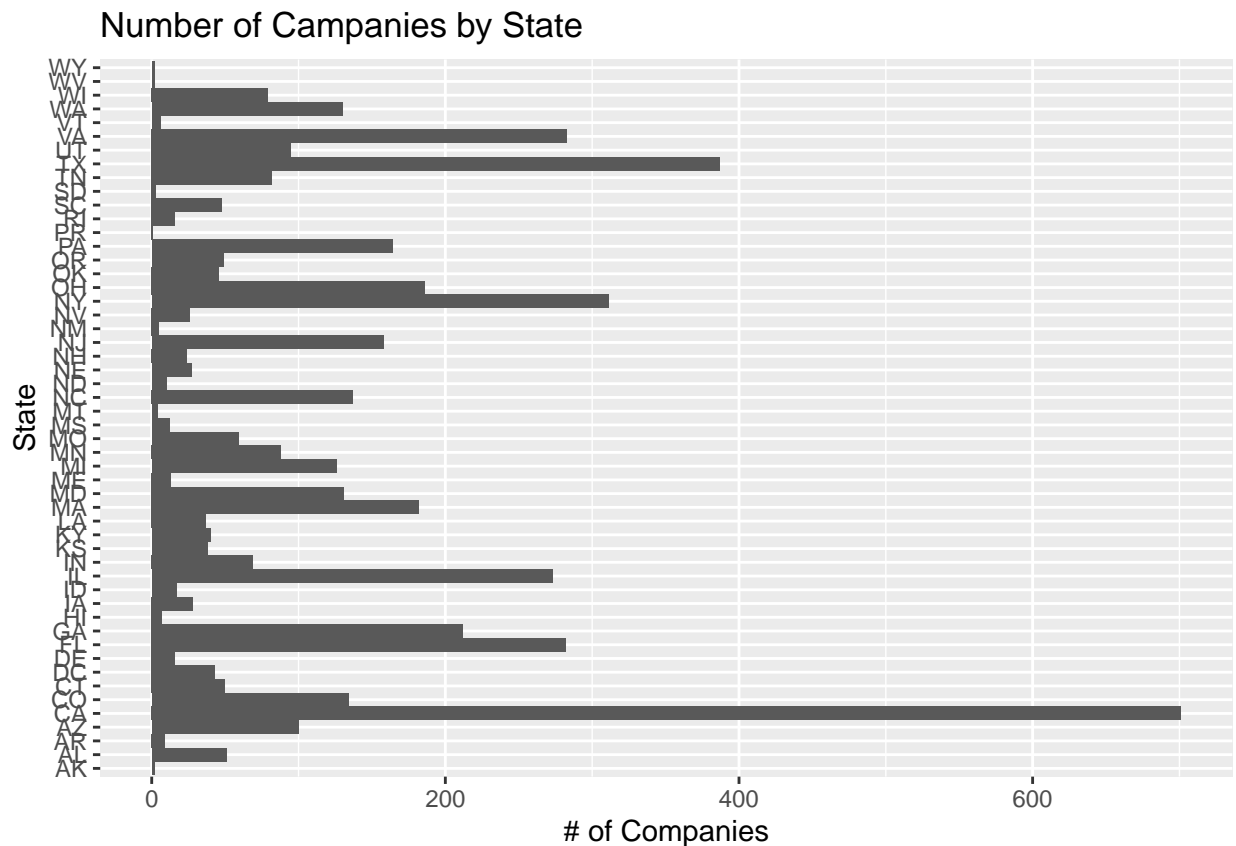
going to be consumed on a 'portrait' oriented screen (ie taller than wide), which should further guide your layout choices.

```
# Answer Question 1 here
```

```
# To represent the distribution of companies by state, I used histogram to represent the number of coun
```

```
library(ggplot2)
ggplot(inc, aes(x=State)) + geom_histogram(stat = "count", width = 1) +
  ggtitle("Number of Companies by State") +
  xlab("State") + ylab("# of Companies") +
  coord_flip()
```

```
## Warning: Ignoring unknown parameters: binwidth, bins, pad
```



## 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.

```
# Answer Question 2 here
```

```
library(dplyr)
```

```
##  
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:Hmisc':  
##  
##      src, summarize
```

```
## The following objects are masked from 'package:stats':  
##  
##      filter, lag
```

```
## The following objects are masked from 'package:base':  
##  
##      intersect, setdiff, setequal, union
```

```
# From the summary() function in the previous section, we can see that the state with the third most co  
# Apply complete.cases() function.
```

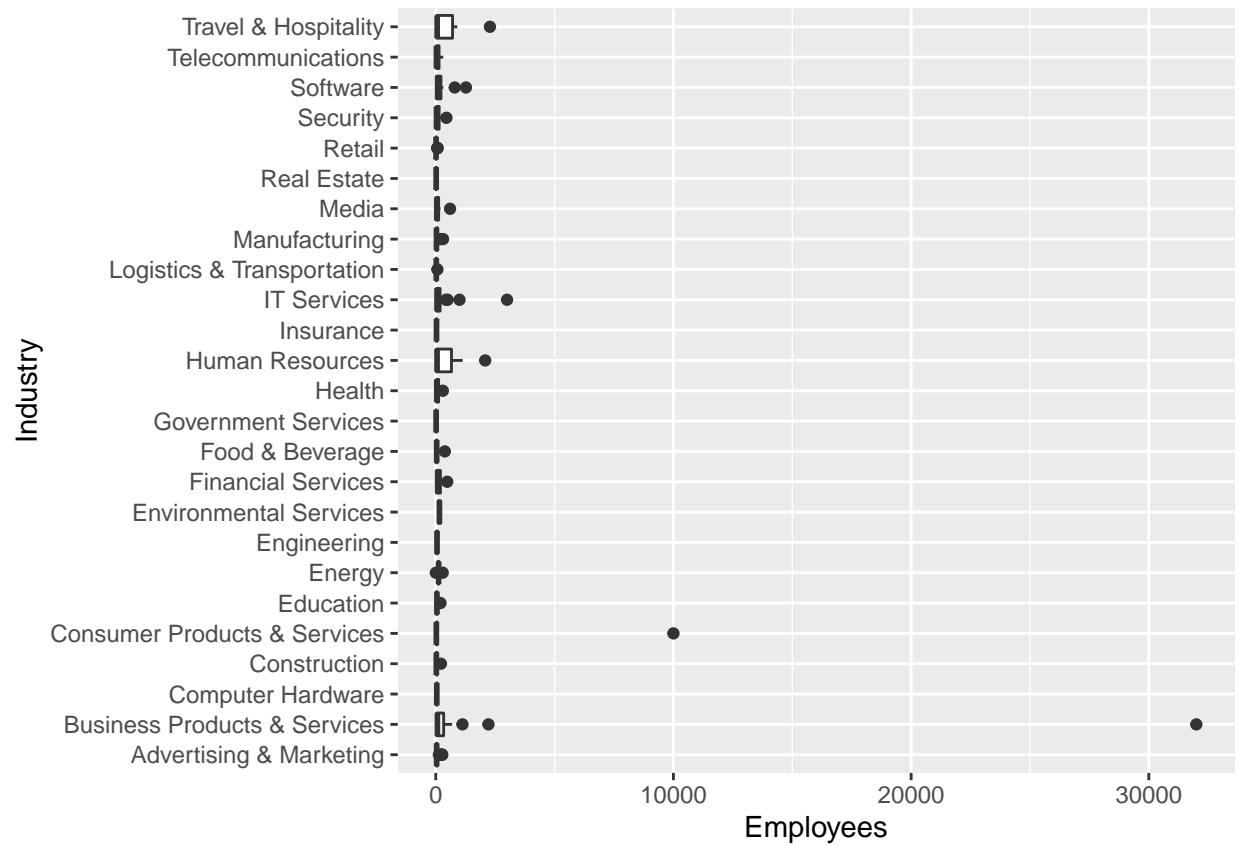
```
inc_complete <- inc[complete.cases(inc),]
```

```
# Prepare dataset.
```

```
NY <- filter(inc_complete, State == "NY")  
#head(NY)
```

```
# Create box plot WITH outliers for initial data exploration.
```

```
chart_initial <- ggplot(NY, aes(Industry, Employees)) + geom_boxplot() + coord_flip()  
chart_initial
```

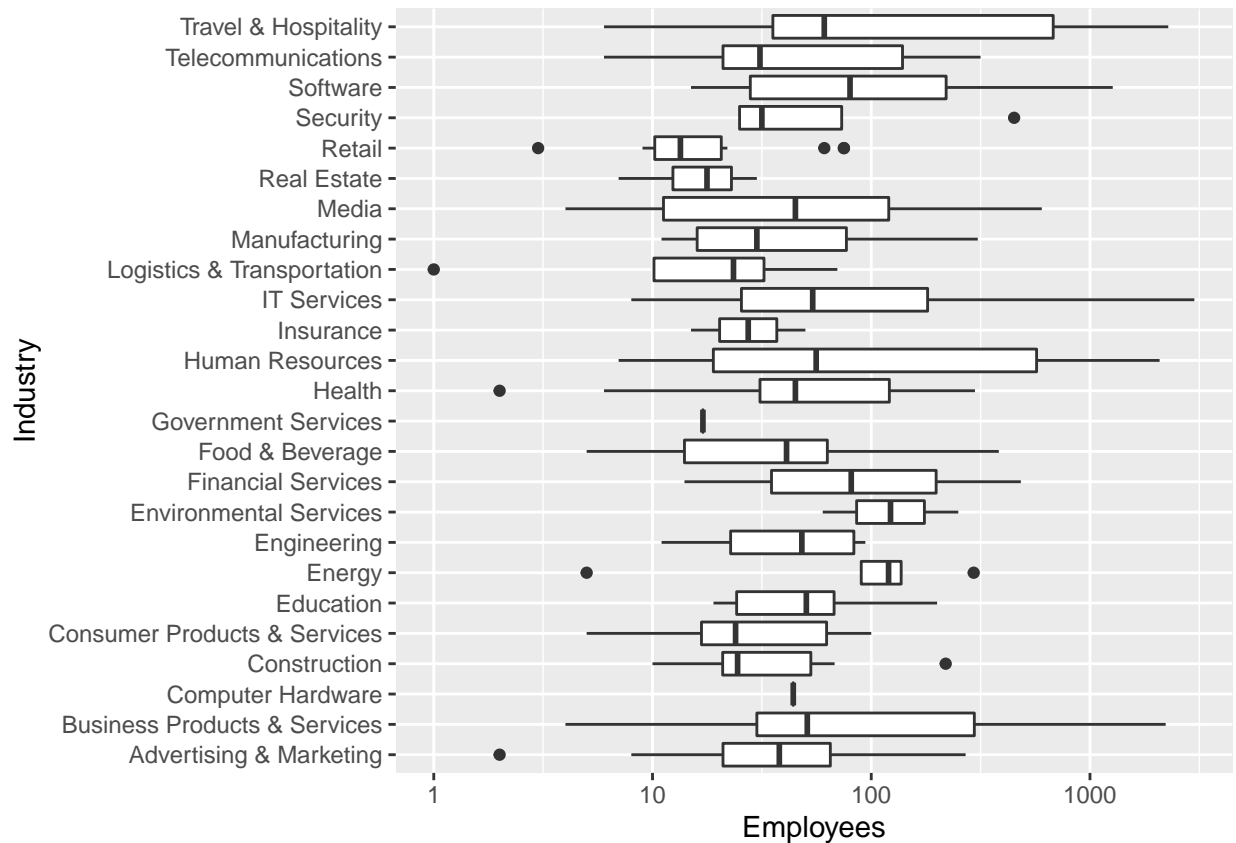


```
# Find and remove outliers - remove the two biggest outliers based on the original plot.

new_NY <- subset(NY, Employees < 10000)

# Apply log transformation to further normalize the data.
chart_log_transformed <- ggplot(new_NY, aes(Industry, Employees)) +
  geom_boxplot() +
  scale_y_log10() +
  coord_flip()

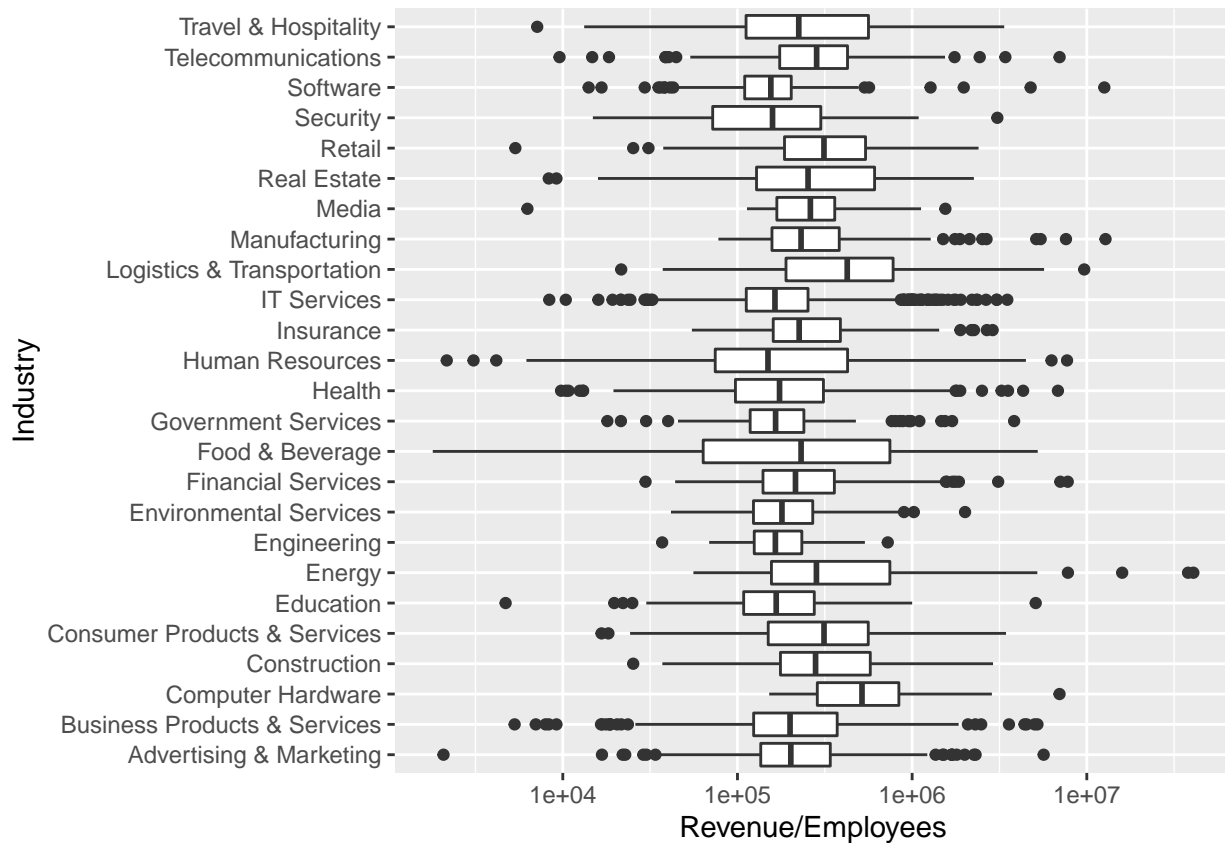
chart_log_transformed
```



### 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.

```
# Answer Question 3 here
# Plot boxplot.
ggplot(inc_complete, aes(Industry, Revenue/Employees)) +
  geom_boxplot() +
  scale_y_log10() +
  coord_flip()
```



*# The boxplot looks chaotic and does not show any trend without extensive data cleaning. I decided to u*

*# Add new column to dataframe.*

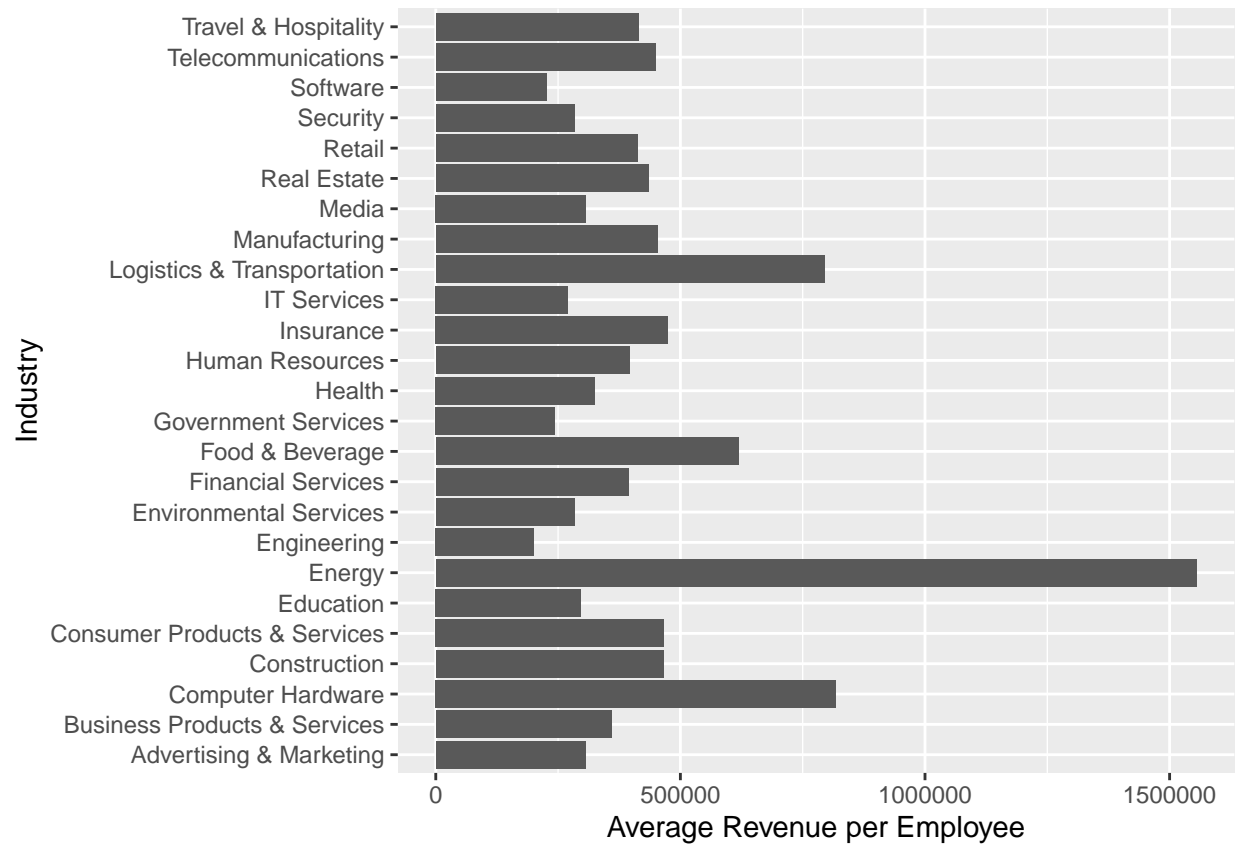
```
RevenuePerEmployee = inc_complete$Revenue / inc_complete$Employees
inc_complete <- cbind(inc_complete, RevenuePerEmployee)
```

*# Create bar charts for revenue per employee by industry.*

```
ggplot(inc_complete, aes(x=Industry, y=RevenuePerEmployee)) +
  stat_summary(fun.y="mean", geom="bar")+
  xlab("Industry") + ylab("Average Revenue per Employee") + coord_flip()
```

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





*# The bar chart shows a much clearer trend, with the Energy industry generating the most revenue per employee.*

```
library(knitr)
opts_chunk$set(tidy.opts=list(width.cutoff=60),tidy=TRUE)
```