

# R Notebook

Code ▼

## Neil Shah HW1: DATA 608

### 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:

Hide

```
inc <- read.csv("https://raw.githubusercontent.com/charleyferrari/CUNY_DATA_608/master/module1/Data/inc5000_data.csv", header= TRUE)
```

And lets preview this data:

```
head(inc)
```

Rank	Name	Growth_Rate	Revenue	Industry	Employees
1	Fuhu	421.48	1.179e+08	Consumer Products & Services	104
2	FederalConference.com	248.31	4.960e+07	Government Services	51
3	The HCI Group	245.45	2.550e+07	Health	132
4	Bridger	233.08	1.900e+09	Energy	50
5	DataXu	213.37	8.700e+07	Advertising & Marketing	220
6	MileStone Community Builders	179.38	4.570e+07	Real Estate	63

```
summary(inc)
```

Rank	Name	Growth_Rate	Revenue
Min. : 1	(Add)ventures	Min. : 0.340	Min. : 2.000e+06
1st Qu.: 1252	@Properties	1st Qu.: 0.770	1st Qu.: 5.100e+06
Median : 2502	1-Stop Translation USA:	Median : 1.420	Median : 1.090e+07
Mean : 2502	110 Consulting	Mean : 4.612	Mean : 4.822e+07
3rd Qu.: 3751	11thStreetCoffee.com	3rd Qu.: 3.290	3rd Qu.: 2.860e+07
Max. : 5000	123 Exteriors	Max. : 421.480	Max. : 1.010e+10
	(Other)	: 4995	

Industry	Employees	City	State
IT Services	: 733	Min. : 1.0	New York : 160 CA : 701
Business Products & Services	: 482	1st Qu.: 25.0	Chicago : 90 TX : 387
Advertising & Marketing	: 471	Median : 53.0	Austin : 88 NY : 311
Health	: 355	Mean : 232.7	Houston : 76 VA : 283
Software	: 342	3rd Qu.: 132.0	San Francisco: 75 FL : 282
Financial Services	: 260	Max. : 66803.0	Atlanta : 74 IL : 273
(Other)	: 2358	NA's : 12	(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:

These summaries provide robust statistics on the overall columns in the data-set. They give us a cursory view of the entire data-set and alert us to overall trends, outliers and serve as a baseline to start out analysis.

As a financial profession—I really like to look at skewness to give me an idea of how a distribution might lean.

I found this package called Performance Analytics here (<https://rviews.rstudio.com/2017/12/13/introduction-to-skewness/>) and used it's skew function.

Let's apply this to the numerical categories in inc.

```
library('PerformanceAnalytics')
```

```
> skewness(inc$Growth_Rate)
```

```
[1] 12.55327
```

```
>
```

```
> skewness(inc$Revenue)
```

```
[1] 22.1811
```

```
> skewness(inc$Employees)
```

```
[1] 29.81938
```

Notice that all of these values have a positive skew—meaning that they have tails to the right—this is interesting and might point to possible outliers. From a robust statistics side—we might need to look at median instead of mean to get an idea of variability.

## 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

First lets get some imports

```
library(ggplot2)
library(zeallot)
```

Now let's make a table and group by State

```
state <- inc %>% group_by(State) %>% summarize(Count = n())
```

Take a look at it

```
head(state)

> head(state)
# A tibble: 6 x 2
  State Count
  <fct> <int>
1 AK      2
2 AL     51
3 AR      9
4 AZ    100
5 CA    701
6 CO    134
```

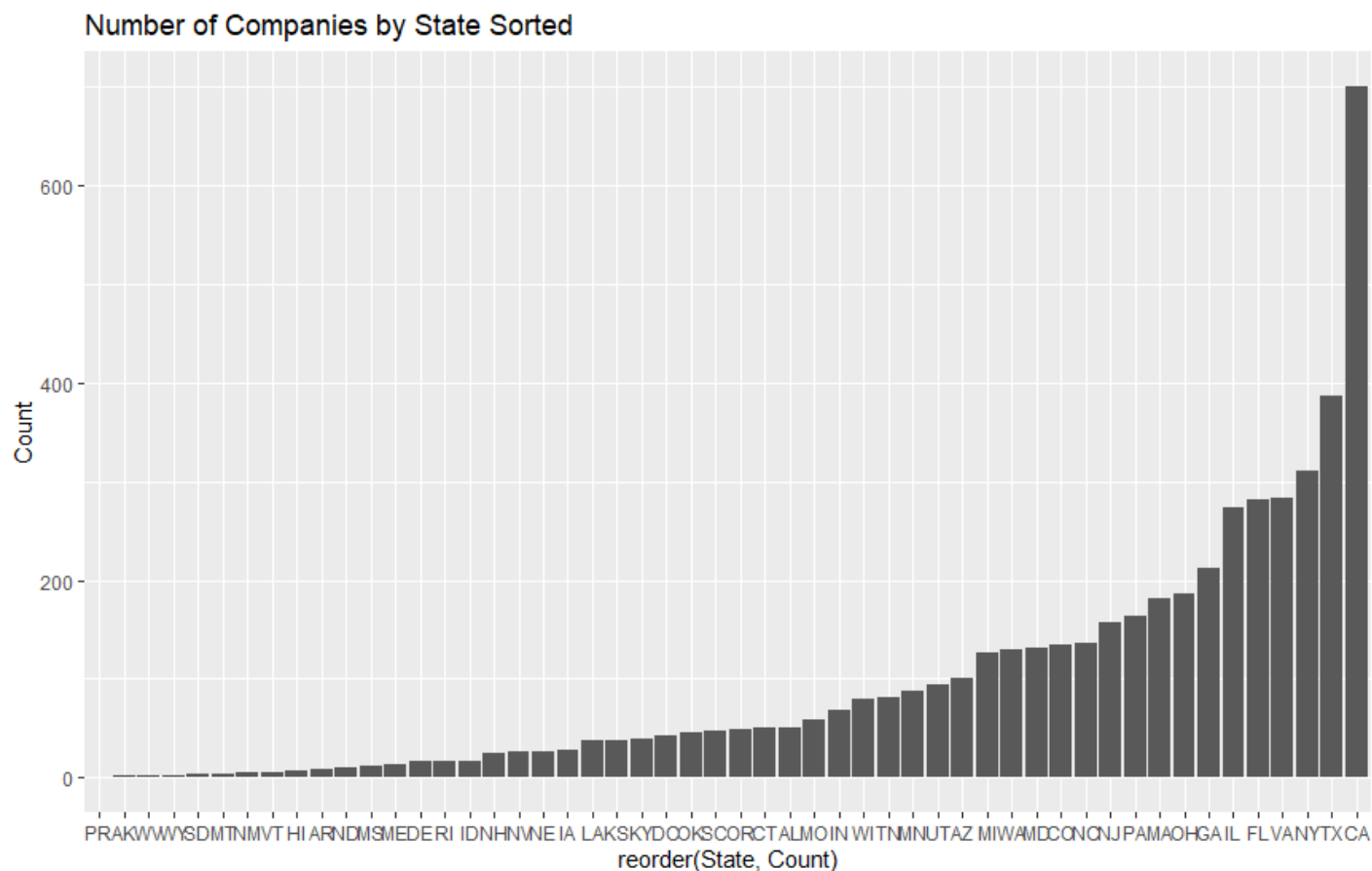
Ok let's sort this out

```
> state %>% arrange(desc(Count))
# A tibble: 52 x 2
  State Count
  <fct> <int>
1 CA    701
2 TX    387
3 NY    311
4 VA    283
5 FL    282
6 IL    273
7 GA    212
8 OH    186
9 MA    182
10 PA    164

state <- state %>% arrange(desc(Count))
```

Now let's plot it

```
>ggplot(state, aes(x = reorder(State, Count), y = Count)) +geom_bar(stat = "identity") +ggtitle('Number of Companies by State Sorted')
```

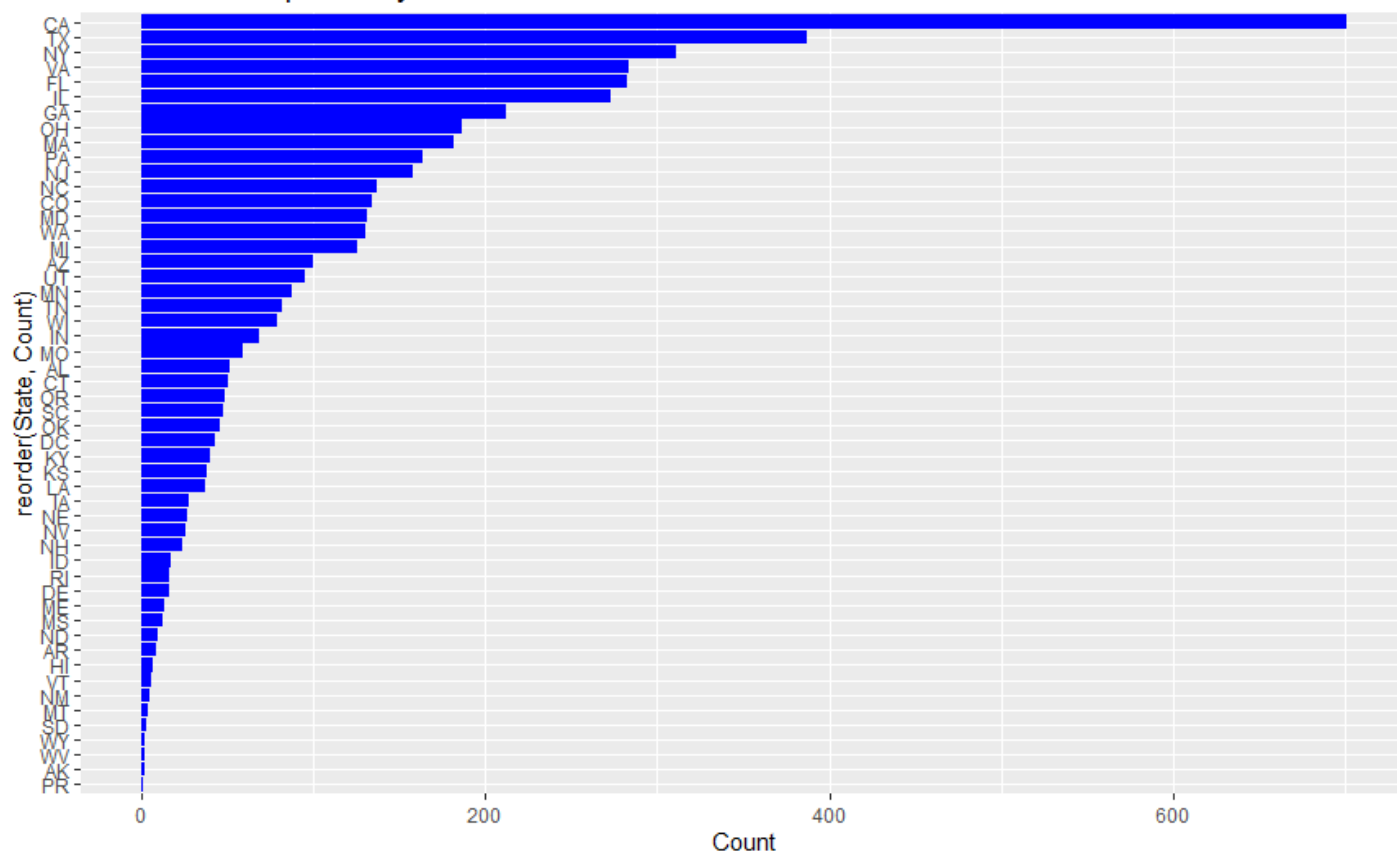


### Number of Companies by State Plot

Alright—now let's clean it up, add some color and fix the axis so we can see the labels.

```
ggplot(state, aes(x = reorder(State, Count), y = Count)) +geom_bar(stat = "identity", fill='blue') +ggtitle('Number of Companies by State Sorted')+coord_flip()
```

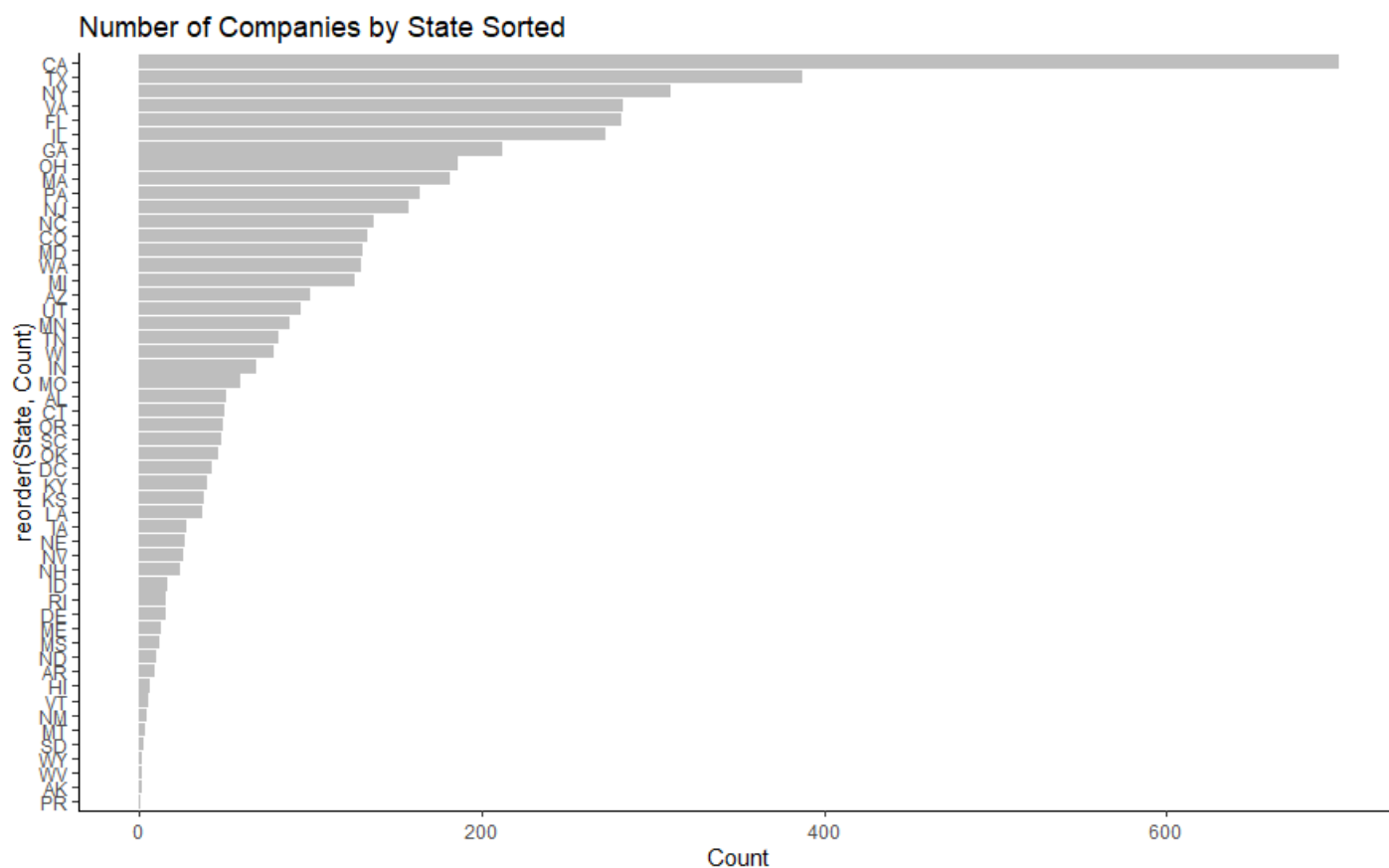
## Number of Companies by State Sorted



## Number of Companies by State Plot

Since I just learned about data-ink ratio—let's try to apply it here. I am referencing the methods from Felix Fans Reference Site (<https://felixfan.github.io/ggplot2-remove-grid-background-margin/>)

```
> ggplot(state, aes(x = reorder(State, Count), y = Count)) +geom_bar(stat = "identity", fill='grey') +ggtitle('Number of Companies by State Sorted')+coord_flip() + theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
+ panel.background = element_blank(), axis.line = element_line(colour = "black"))
```

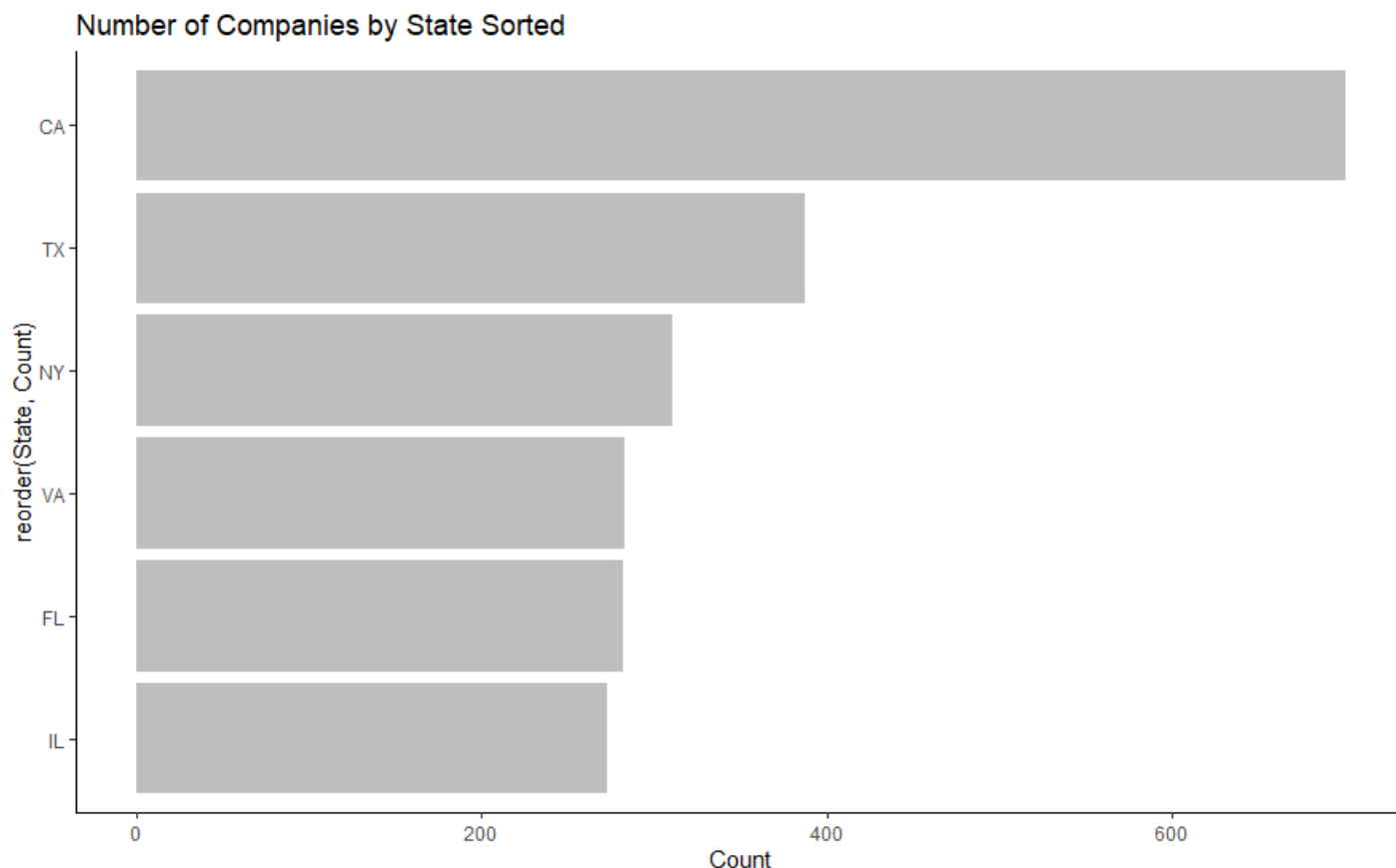


### Number of Companies by State Plot

There—much cleaner! So it seems that the top states are CA, TX and then NY.

Let's zoom in on the top values to focus on them

```
> ggplot(head(state), aes(x = reorder(State, Count), y = Count)) +geom_bar(stat = "identity", fill='grey') +ggtitle('Number of Companies by State Sorted')+coord_flip() + theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
+ panel.background = element_blank(), axis.line = element_line(colour = "black"))
>
```



Top by State Plot

This makes some sense to me given that these states have the highest populations.

## 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.

So based on our data-set we are looking at good ole NY

First let's sort out our dataset for NY only and by industry

```

> inc %>% filter(State=='NY') %>%filter(complete.cases(.)) %>% group_by(Industry)
# A tibble: 311 x 8
# Groups:   Industry [25]
  Rank Name Growth_Rate Revenue Industry Employees C
ity State
  <int> <fct> <dbl> <dbl> <fct> <int> <fct>
1 26 BeenVerified 84.4 13700000 Consumer Products & Services 17 N
ew York NY
2 30 Sailthru 73.2 8100000 Advertising & Marketing 79 N
ew York NY
3 37 YellowHammer 67.4 18000000 Advertising & Marketing 27 N
ew York NY
4 38 Conductor 67.0 7100000 Advertising & Marketing 89 N
ew York NY
5 48 Cinium Financial Services 53.6 5900000 Financial Services 32 R
ock Hill NY
6 70 33Across 45.0 27900000 Advertising & Marketing 75 N
ew York NY
7 71 LiveIntent 44.8 6900000 Advertising & Marketing 42 N
ew York NY
8 124 Quantum Networks 29.4 11500000 Telecommunications 28 N
ew York NY
9 126 Renegade Furniture Group 29.3 9800000 Retail 17 H
ewlett NY
10 153 Regal Wings 25.1 15400000 Travel & Hospitality 42 B
rooklyn NY
# ... with 301 more rows``

```

I like to do a quick summary statistics to explore the data-set



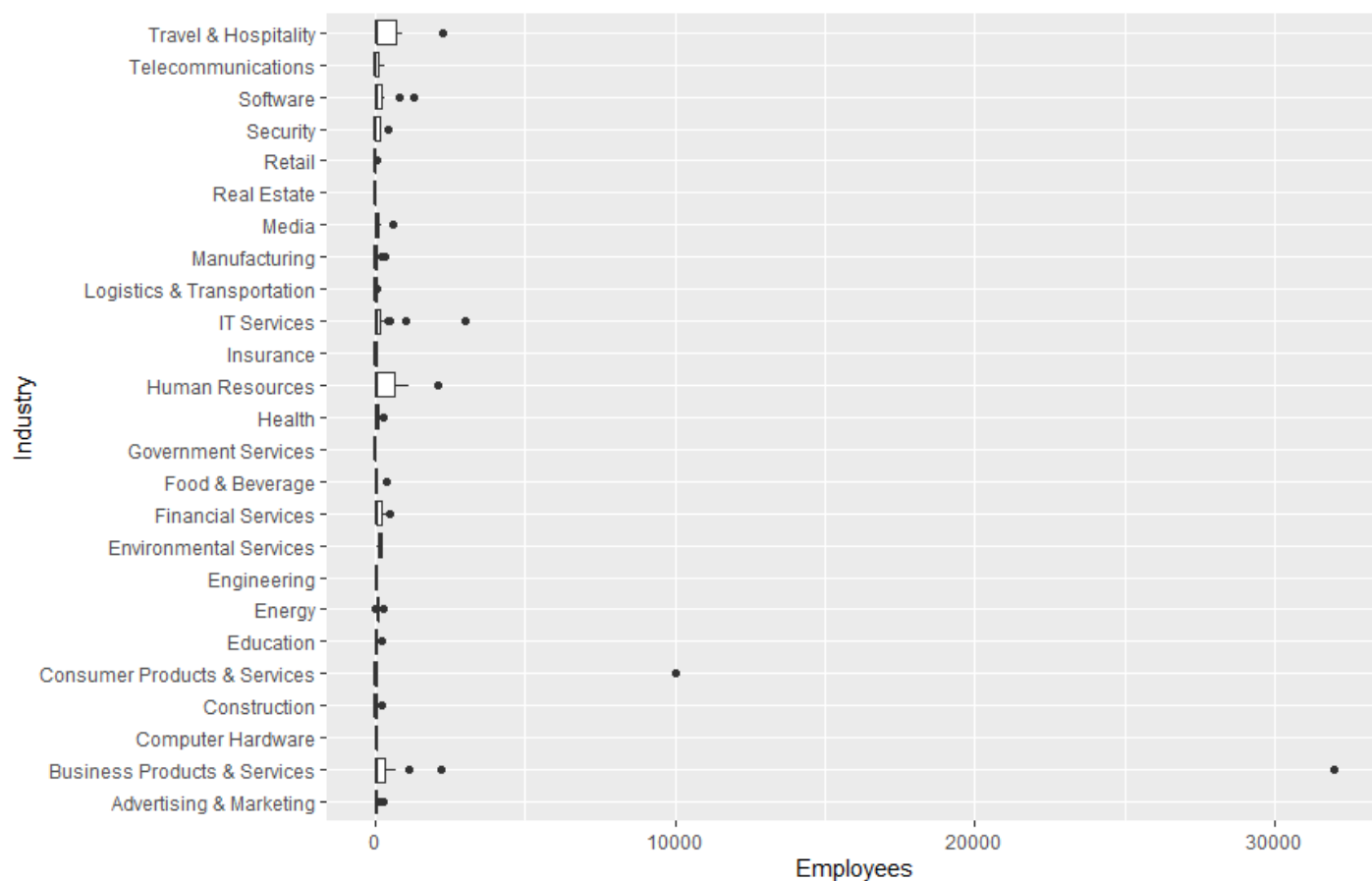
```
> summary(inc %>% filter(State=='NY') %>%filter(complete.cases(.)) %>% group_by(Industry))
```

Industry	Rank	Name	Growth_Rate	Revenue
Min. : 26	1st Equity	: 1	Min. : 0.350	Min. :2.000e+06
Advertising & Marketing : 57				
1st Qu.:1186	33Across	: 1	1st Qu.: 0.670	1st Qu.:4.300e+06
IT Services : 43				
Median :2702	5Linx Enterprises	: 1	Median : 1.310	Median :8.800e+06
Business Products & Services: 26				
Mean :2612	Access Display Group:	1	Mean : 4.371	Mean :5.872e+07
Consumer Products & Services: 17				
3rd Qu.:4005	Adafruit	: 1	3rd Qu.: 3.580	3rd Qu.:2.570e+07
Telecommunications : 17				
Max. :4981	AdCorp Media Group	: 1	Max. :84.430	Max. :4.600e+09
Education : 14				
(Other) :305				(Other)
:137				
Employees	City	State		
Min. : 1.0	New York :160	NY :311		
1st Qu.: 21.0	Brooklyn : 15	AK : 0		
Median : 45.0	Rochester: 9	AL : 0		
Mean : 271.3	Buffalo : 5	AR : 0		
3rd Qu.: 105.5	Fairport : 5	AZ : 0		
Max. :32000.0	new york : 5	CA : 0		
(Other) :112	(Other): 0			

Two things that I want to point out–1) State only has values for NY, which is good! That means my filter by NY worked out and 2) looking at the statistical summary of Employees–the max is 32000 which is well above the IQR ranges; we definitely are going to have outliers!

The easiest way to display variance, median and spread is a boxplot; let's do that. To make things easier I'll save the modified dataframe.

```
ny <- inc %>% filter(State=='NY') %>%filter(complete.cases(.)) %>% group_by(Industry)
ggplot(ny,aes(x=Industry,y=Employees))+geom_boxplot()+coord_flip()
```

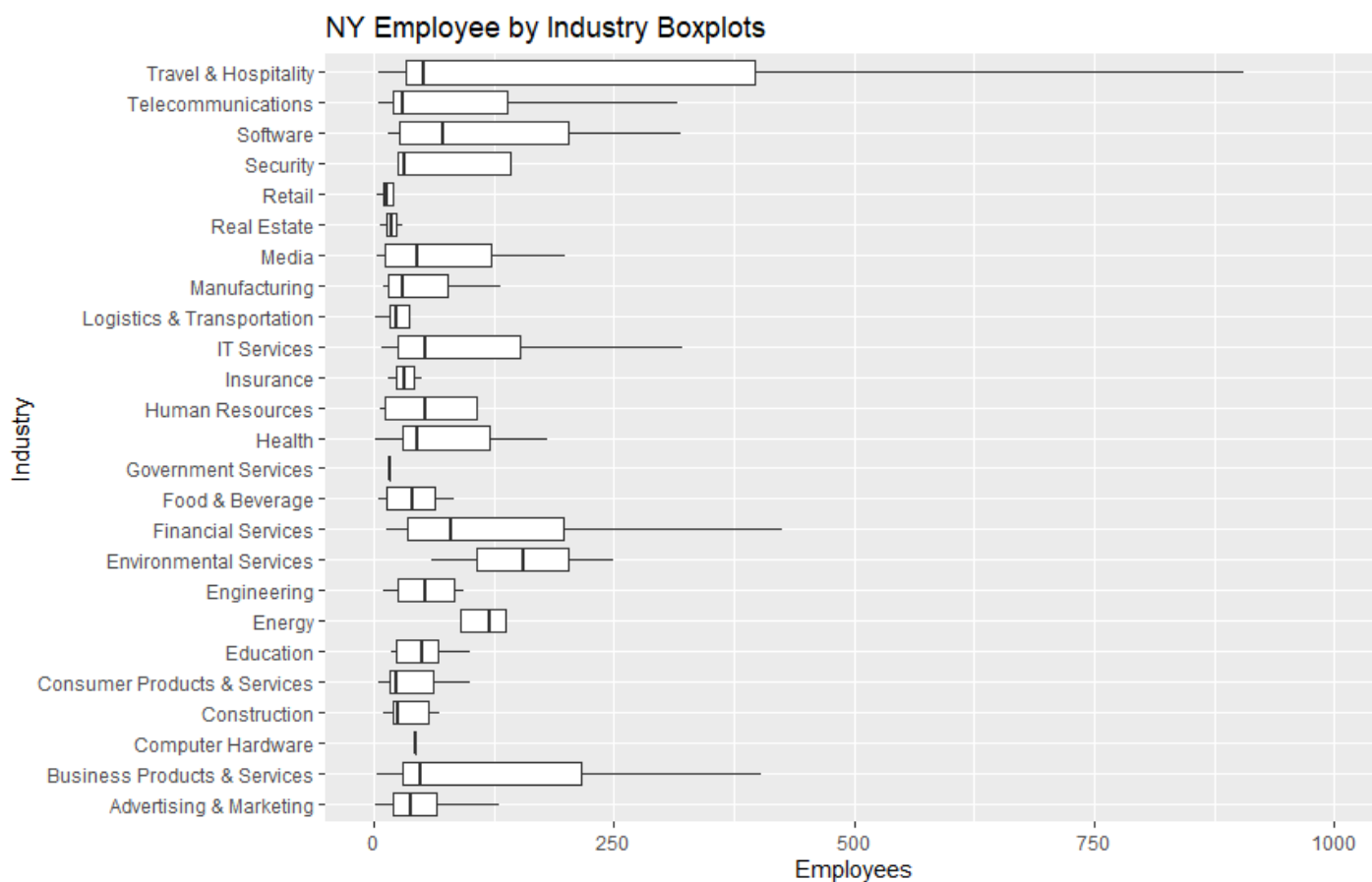


### Box Plot

Looks like we have some serious outliers!

So I could extract the outliers and remove them but what is easier is just to cut my axis and hide the outliers.

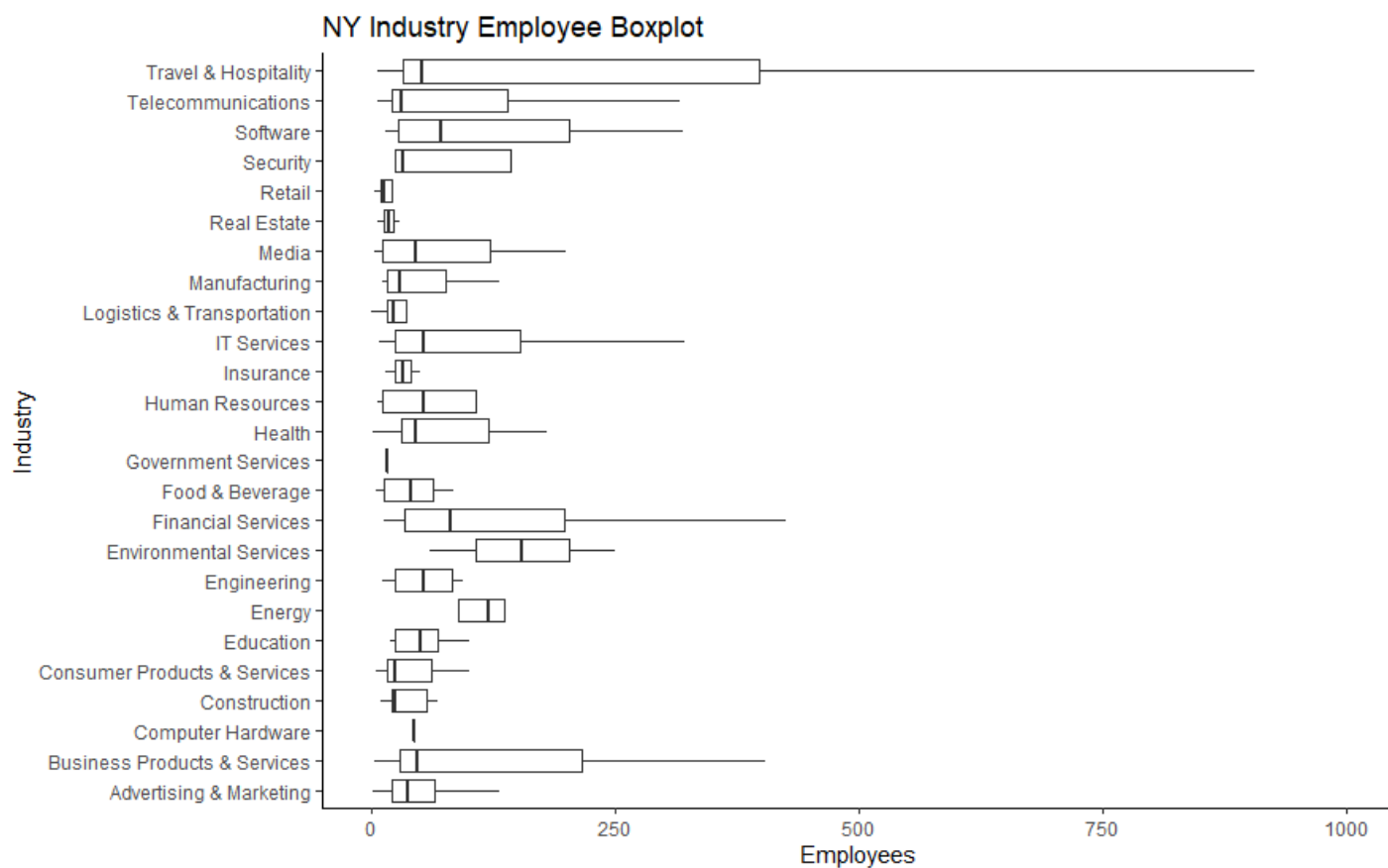
```
> ggplot(ny,aes(x=Industry,y=Employees))+geom_boxplot(outlier.shape=NA)+ ggtitle('NY Industry Employee Boxplot') + coord_flip()+ylim(0,1000)
```



### Box Plot

Now let's combine everything like we did before and make the plot readable via our Data to Ink ratio method.

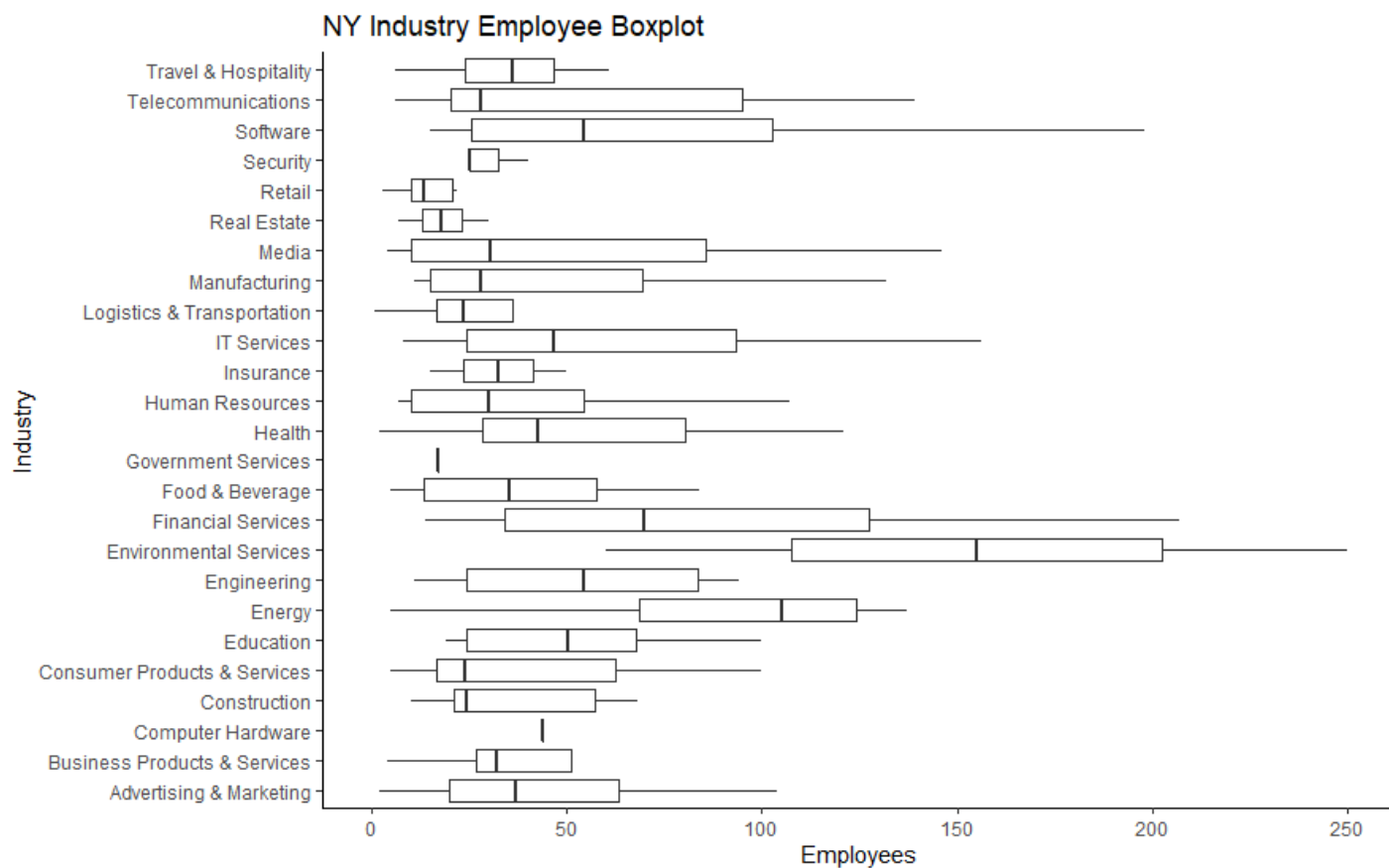
```
> ggplot(ny,aes(x=Industry,y=Employees))+geom_boxplot(outlier.shape=NA)+ ggtitle('NY Industry Employee Boxplot') + coord_flip()+ylim(0,1000) + theme_bw() + theme(panel.border = element_blank(), panel.grid.major = element_blank(), panel.grid.minor = element_blank(), axis.line = element_line(colour = "black"))
```



Clean Box Plot

Much nicer!

I'm going to include one more zoomed in plot just to see more granuarity.



## Clean Box Plot

So looking over the data just some quick observations:

The Travel and Hospitality industry has the largest spread/variability, given the whisker range/IQR range.

Computer hardware and Government Services have the most narrow spreads

The median for the NY industries are all below 250

Government services has the lowest median employees.

Environmental Services has the highest median employees.

Fascinating—this would be a cool study to dig down further.

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## 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.

First let's define a metric called medianemp defined by Revenue/Employees.

```
subset(inc, complete.cases(inc)) %>%  
  mutate(medianemp = Revenue/Employees)
```

I am going to just store this as a new dataframe to make my life easier

```
inc_investor <- subset(inc, complete.cases(inc)) %>%  
  mutate(medianemp = Revenue/Employees)
```

Let's quickly look at summary statistics

```
> summary(inc_case)
```

Rank	Name	Growth_Rate	Revenue
Min. : 1	(Add)ventures	Min. : 0.340	Min. : 2.000e+06
1st Qu.: 1252	@Properties	1st Qu.: 0.770	1st Qu.: 5.100e+06
Median : 2502	1-Stop Translation USA:	Median : 1.420	Median : 1.090e+07
Mean : 2501	110 Consulting	Mean : 4.615	Mean : 4.825e+07
3rd Qu.: 3750	11thStreetCoffee.com	3rd Qu.: 3.290	3rd Qu.: 2.860e+07
Max. : 5000	123 Exteriors	Max. : 421.480	Max. : 1.010e+10
	(Other)	: 4983	

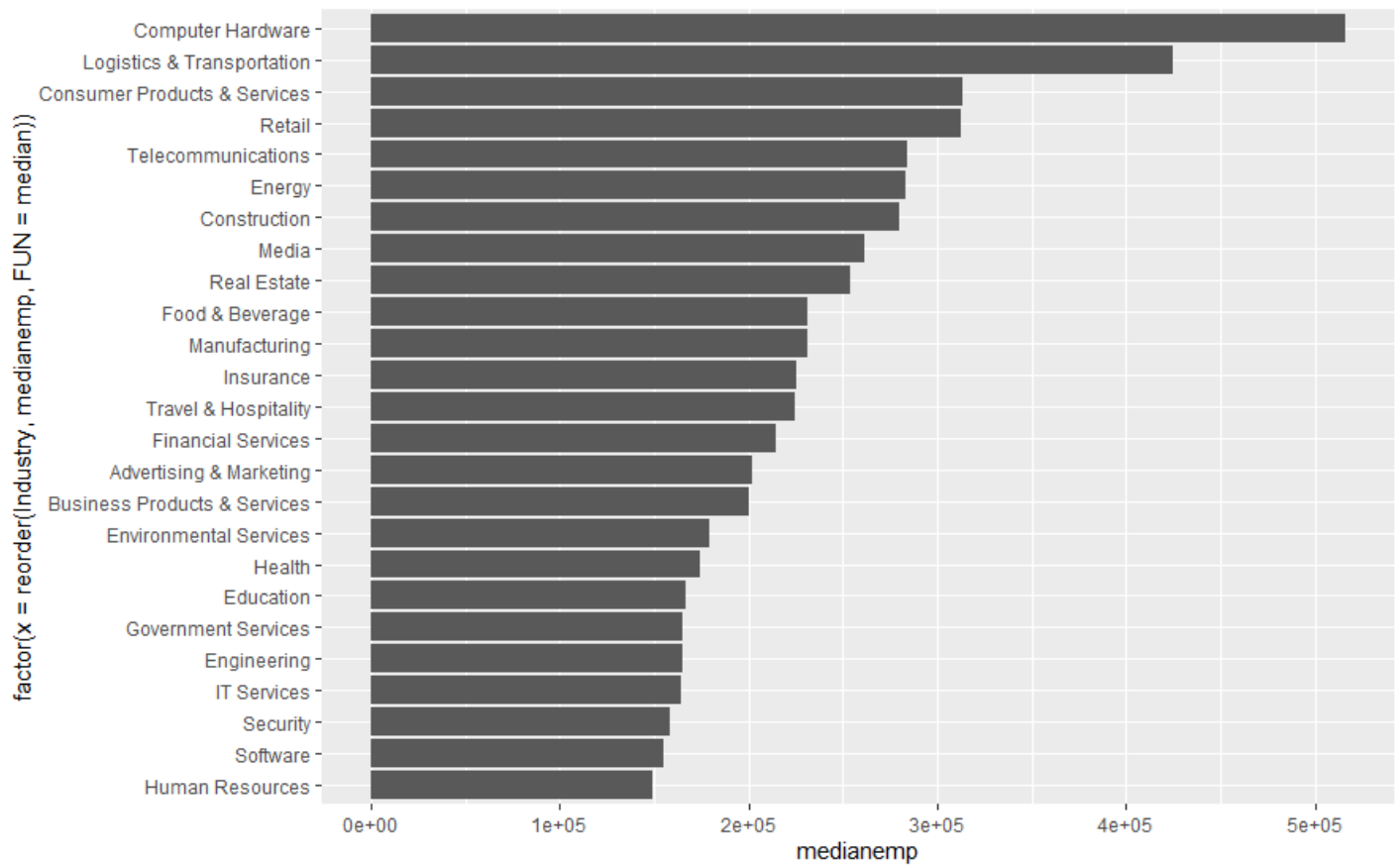
  

Industry	Employees	City	State	medianemp
IT Services	: 732	Min. : 1.0	New York	: 160
CA	: 700	Min. : 1801		
Business Products & Services:	480	1st Qu.: 25.0	Chicago	: 90
TX	: 386	1st Qu.: 125000		
Advertising & Marketing	: 471	Median : 53.0	Austin	: 88
NY	: 311	Median : 198658		
Health	: 354	Mean : 232.7	Houston	: 76
VA	: 283	Mean : 393613		
Software	: 341	3rd Qu.: 132.0	San Francisco:	74
FL	: 282	3rd Qu.: 375000		
Financial Services	: 260	Max. : 66803.0	Atlanta	: 73
IL	: 272	Max. : 40740000		
(Other)	: 2351		(Other)	: 4428
			(Other):	2755

Focusing in on medianemp—the metric i defined, it appears that the median is around \$200,000 per employee and most of the distribution is under \$400,000—however look at that outlier! Let's investigate.

Let's plot this out

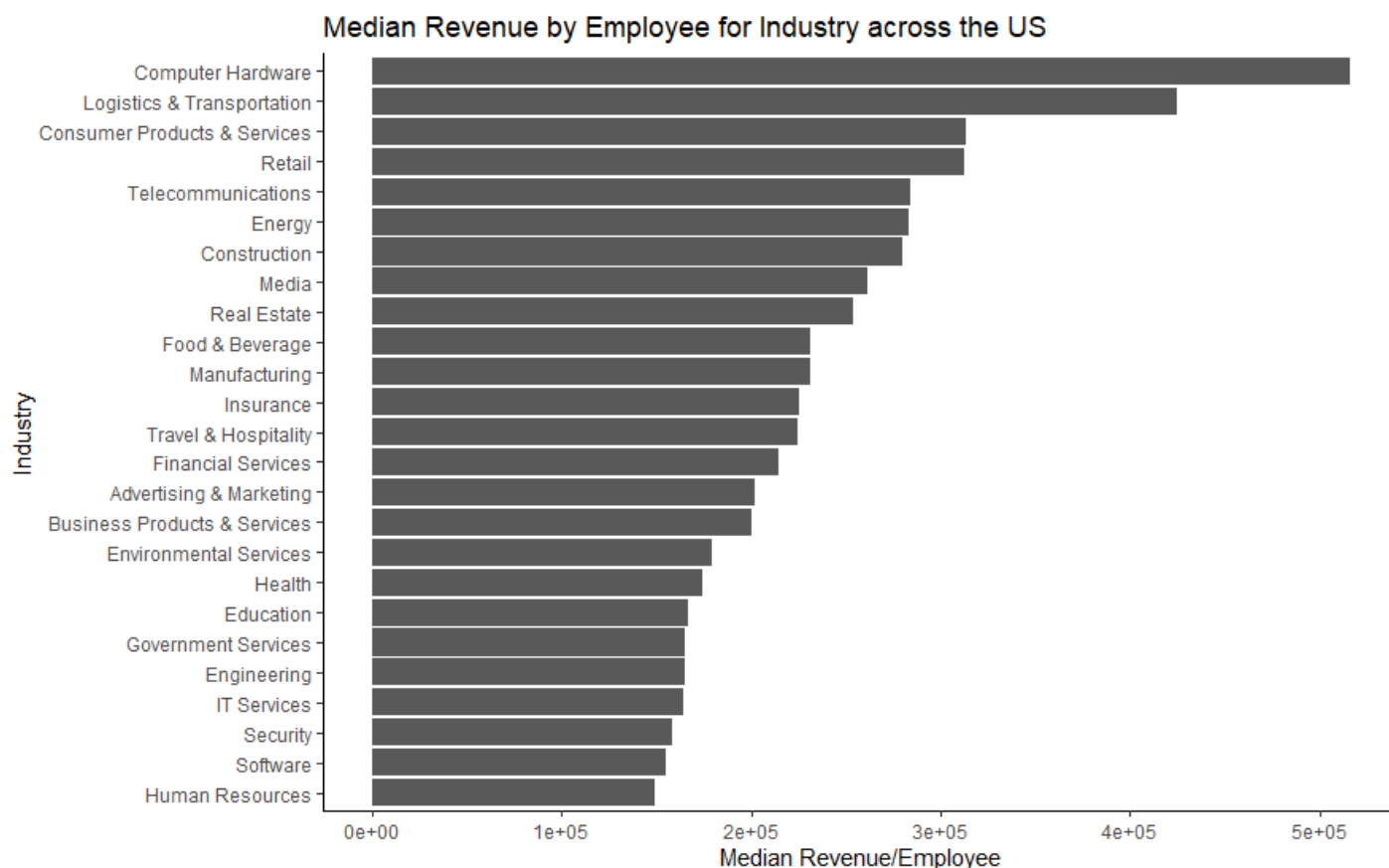
```
> ggplot(inc_investor, aes(factor(x = reorder(Industry, medianemp, FUN = median)))) +
+ stat_summary_bin(aes(y = medianemp), fun.y = "median", geom = "bar") +
+ coord_flip()
```



### Median Revenue/Employee by Industry

Now combining our plotting methods from before

```
> ggplot(inc_investor, aes(factor(x = reorder(Industry, medianemp, FUN = median)))) +
+ stat_summary_bin(aes(y = medianemp), fun.y = "median", geom = "bar") + coord_flip() + xlab("In
dustry") + ylab("Median Revenue/Employee")+ ggtitle('Median Revenue by Employee for Industry acr
oss the US') + theme_bw() + theme(panel.border = element_blank(), panel.grid.major = element_bla
nk(),
+ panel.grid.minor = element_blank(), axis.line = element_line(colour = "black"))
```



Median Revenue/Employee by Industry

Much cleaner and clearer!

If we recall from the original summary table—the median revenue per employee was around \$200,000 but it seems that two industries are outliers (on the high end)—Computer Hardware and Logistics and Transport. As an investor—I don't want average performance since we have to beat the market; Computer Hardware and Logistics Transports seem two industries we should further analyze for investment opportunities.

This provides a good starting ground for an investment thesis!

## Conclusions

In this assignment I was able to load a dataframe, manipulate it through filtering, produce summary statistics and plot the data in a clear/easy to read fashion.

These skills—while basic—are powerful and will serve as the foundation for my Data Analysis.