

**Yuhao Chen**  
**CS112 Assignment 2**  
**Regression and Bootstrapping**  
**Fall 2018**

## Link to The Code:

<https://drive.google.com/open?id=17xi4CBW85kS1DzvMigDur622h3c39zrO>

## Question 1

1. Your original data-generating equation

```
x=10*runif(99)
```

```
y=10+60*x+rnorm(99)
```

2. Regression results for the original 99 (copy/paste the “summary” output)

```
> summary(lm1)

Call:
lm(formula = y ~ x)

Residuals:
    Min       1Q   Median       3Q      Max
-2.3889 -0.6502 -0.0875  0.5821  2.0903

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 10.28959    0.17920   57.42  <2e-16 ***
x           59.95985    0.03121 1921.07  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.885 on 97 degrees of freedom
Multiple R-squared:  1,    Adjusted R-squared:  1
F-statistic: 3.691e+06 on 1 and 97 DF,  p-value: < 2.2e-16
```

3. Regression results with the outlier included (copy/paste “summary” output)

```
> summary(lm2)

Call:
lm(formula = y2 ~ x2)

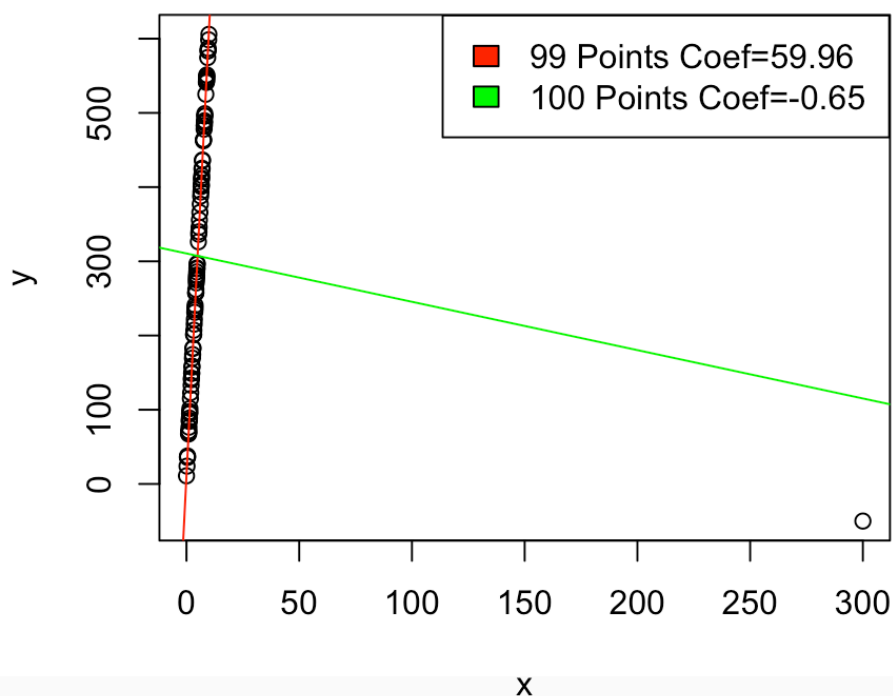
Residuals:
    Min       1Q   Median       3Q      Max
-299.75 -153.85  -19.95   157.90   301.58

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 310.7262    18.0631  17.202  <2e-16 ***
x2          -0.6518     0.5915  -1.102   0.273
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 174.4 on 98 degrees of freedom
Multiple R-squared:  0.01224,    Adjusted R-squared:  0.002159
F-statistic: 1.214 on 1 and 98 DF,  p-value: 0.2732
```

4. A properly-labeled data visualization that shows a single scatterplot, the regression line based on the original 99 points, and another differentiated regression line based on 100 points.

**Regression Models of 99 and 100 Points**



5. No more than 3 sentences that would serve as a caption for your figure if it were to be included in an econometrics textbook to illustrate the dangers of extrapolation.

Answer: When there are outliers appearing, the model can be largely influenced. Therefore, when doing extrapolation with a trained model, we need to analyze the effectivity of the model first, focusing on whether or not to consider the outliers. If the outliers are ridiculous and appear due to the wrong observation or type-in, we need to ignore them; if the outlier are in the acceptable range, we cannot ignore them.

## Question 2

1. A table with the relevant point estimates (e.g., the bounds of the prediction intervals of  $y$  for the different ages, and the medians of the other predictors)

	[,14]	[,15]	[,16]	[,17]	[,18]	[,19]	[,20]	[,21]
2.5%	NA	NA	NA	-6662.36	-6839.577	-6735.086	-6807.791	-6577.317
97.5%	NA	NA	NA	15196.25	15098.285	14872.080	15068.488	14937.930
	[,22]	[,23]	[,24]	[,25]	[,26]	[,27]	[,28]	
2.5%	-7005.02	-6631.218	-6628.67	-6712.292	-6764.641	-6651.488	-6677.396	
97.5%	14987.50	15121.140	15321.73	15120.787	15095.833	15045.038	14917.211	
	[,29]	[,30]	[,31]	[,32]	[,33]	[,34]	[,35]	
2.5%	-6562.701	-6717.553	-6689.57	-6590.607	-6574.434	-6663.529	-7084.381	
97.5%	15015.273	15274.418	15211.22	15031.679	15310.179	15019.754	14883.526	
	[,36]	[,37]	[,38]	[,39]	[,40]	[,41]	[,42]	
2.5%	-6731.937	-6695.371	-6793.294	-6717.843	-6764.099	-6959.599	-6639.081	
97.5%	15129.263	15220.951	15052.206	15330.769	15153.220	15233.153	15177.797	
	[,43]	[,44]	[,45]	[,46]	[,47]	[,48]	[,49]	
2.5%	-6870.343	-6724.43	-7043.224	-6962.92	-7083.499	-7051.442	-6816.198	
97.5%	15020.593	15475.12	15509.989	15423.25	15115.893	15208.657	15252.912	
	[,50]	[,51]	[,52]	[,53]	[,54]	[,55]		
2.5%	-6936.933	-6916.489	-6856.463	-7176.667	-7013.379	-7255.463		
97.5%	15293.982	15414.855	15264.109	15364.337	15661.866	15537.266		

*Table of Confidence Intervals for Predicted Revenues in 1978(Fixed Median)*

	[,14]	[,15]	[,16]	[,17]	[,18]	[,19]	[,20]	[,21]
2.5%	NA	NA	NA	-5217.841	-4994.808	-5359.806	-5240.083	-4771.26
97.5%	NA	NA	NA	17314.702	17276.037	17342.326	17161.841	17211.84
	[,22]	[,23]	[,24]	[,25]	[,26]	[,27]	[,28]	
2.5%	-5310.052	-4992.704	-5049.981	-4881.293	-5122.958	-4848.93	-5147.146	
97.5%	17363.814	17322.154	17231.646	17192.657	17134.799	17321.13	17153.340	
	[,29]	[,30]	[,31]	[,32]	[,33]	[,34]	[,35]	
2.5%	-4793.92	-5061.749	-4852.703	-5226.298	-5246.929	-5272.722	-5152.235	
97.5%	17416.65	17269.258	17538.646	17552.330	17348.530	17437.959	17301.378	
	[,36]	[,37]	[,38]	[,39]	[,40]	[,41]	[,42]	
2.5%	-5396.73	-5398.916	-5226.554	-5199.616	-5275.308	-5218.445	-5217.166	
97.5%	17169.64	17394.870	17423.655	17644.714	17498.856	17343.481	17272.252	
	[,43]	[,44]	[,45]	[,46]	[,47]	[,48]	[,49]	
2.5%	-5323.718	-5497.535	-5482.606	-5621.285	-5861.223	-5828.517	-5897.163	
97.5%	17549.730	17423.249	17721.412	17901.274	18080.399	17964.429	18252.224	
	[,50]	[,51]	[,52]	[,53]	[,54]	[,55]		
2.5%	-6208.732	-6240.545	-5983.171	-5930.477	-6452.278	-5994.13		
97.5%	17980.199	18012.895	17992.053	18319.592	18577.951	18319.67		

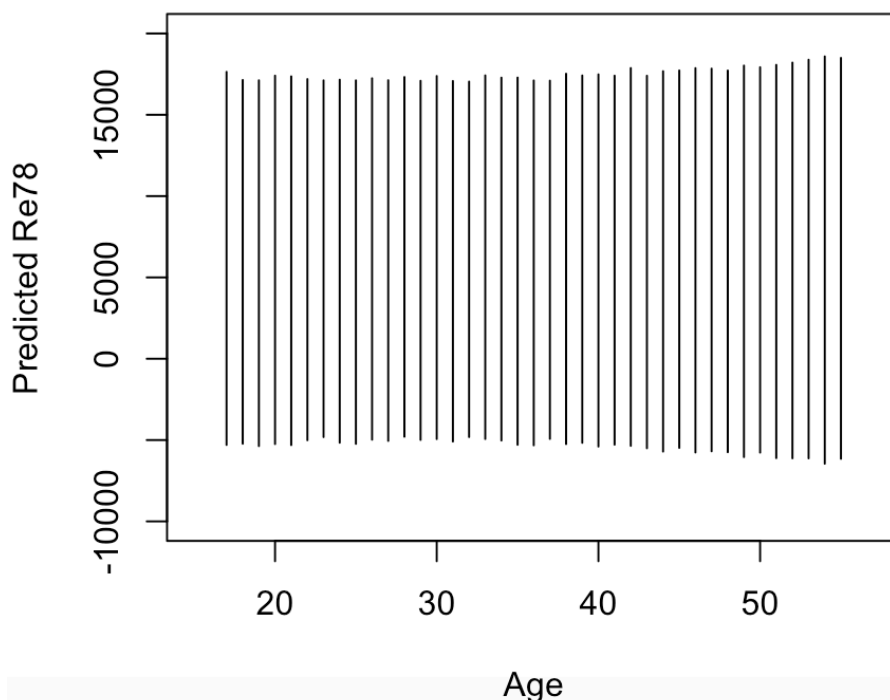
*Table of Confidence Intervals for Predicted Revenues in 1978(Fixed Quantile)*

	Value
Median of Educ	10
Median of Re74	0
Median of Re75	0
90% Quantile of Educ	12
90% Quantile of Re74	7628.052
90% Quantile of Re75	4492.998

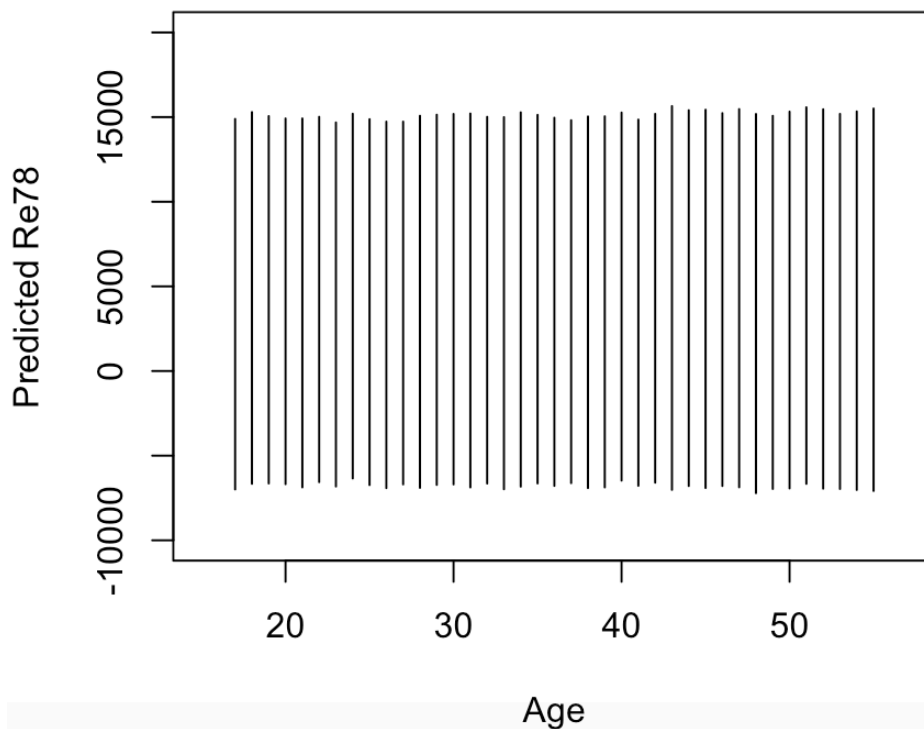
*Table of Medians and 90% Quantiles of The Other Predictors*

2. 2 figures showing the scatterplots (one for the analysis holding predictors at their medians, and other for the analysis holding predictors at their 90% quantiles). The “scatterplots” don’t have to show the original data--all I am interested in are the prediction intervals for each age. Each of these figures should show how the prediction intervals’ change over time (i.e., over the range of ages in the data set). Be sure to label your plot’s features (axis, title, etc.).

### **Predicted Revenues in 1978(Fixed Quantile)**



## Predicted Revenues in 1978(Fixed Median)



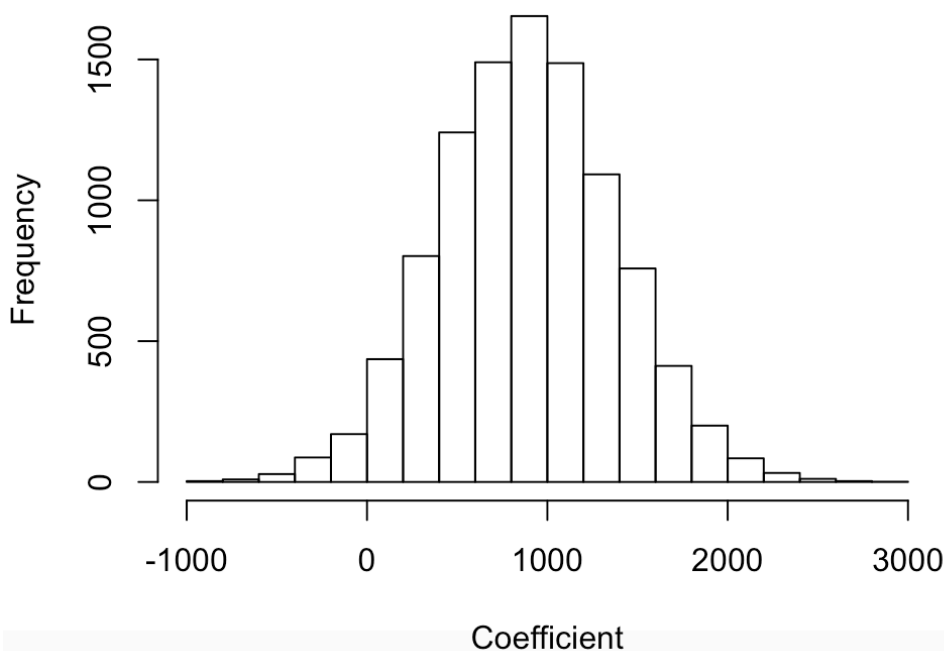
### Question 3

1. A table with the relevant results (bounds on the 2 confidence intervals).

Type	2.5%	97.5%
Analytical Confidence Interval of Coefficient	-42.48831	1866.237
Bootstrap Confidence Interval of Coefficient	-40.52635	1813.134

2. 1 histogram (properly labeled) showing your bootstrap-sample results. How you do this one is up to you.

## Frequency of Coefficients with Bootstrapping



3. No more than 3 sentences summarizing the results and drawing any conclusions you find relevant and interesting.

From the analytical and bootstrapping methods, we get the 95% confidence intervals of coefficients:  $(-42.5, 1866.2)$  and  $(-40.5, 1813.1)$ , which are quite similar. This activity makes me understand how to generate a large amount of samples with bootstrapping, and how bootstrapping decrease the error and generate normal distribution as the graph presents.

### Question 4

Write a function (5 lines max) that takes  $Y_s$  and predicted  $Y_s$  as inputs, and outputs  $R^2$ . Copy/paste an example using the `nsw.dta` data (from #3 above) that shows it working.

*The function:*

```
r_squared=function(y,predicted_y){  
  mean_y=mean(y)  
  return (sum((predicted_y-mean_y)**2)/sum((y-mean_y)**2))  
}
```

*The results we get by using our function to calculate the R-squared of our predicted re78:*

```
> r_squared(re78,predict_re78)  
[1] 0.004871571
```

*The result of R-squared we get from summary(lm) function:*

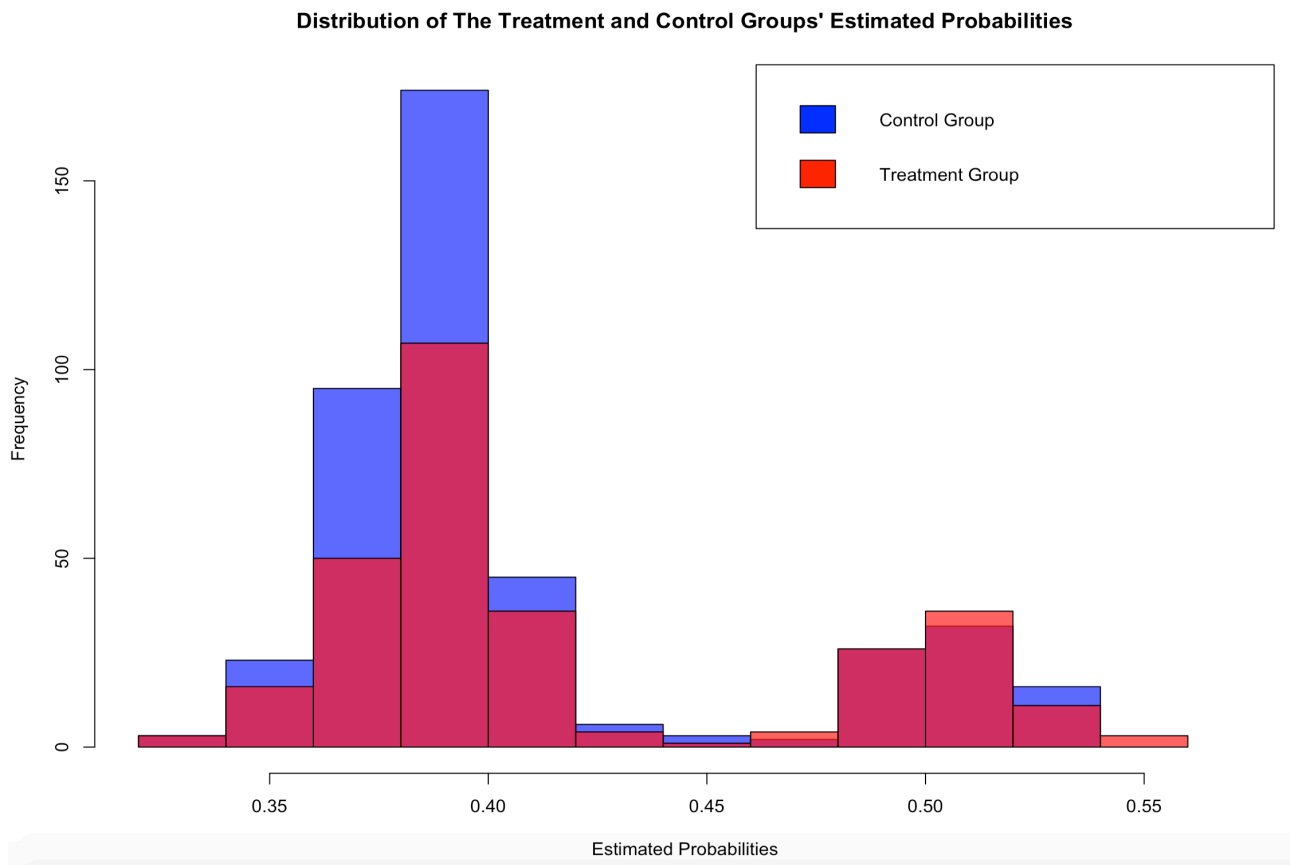
Multiple R-squared: 0.004872

*We can find that the results are the same.*

## Question 5

1. Two properly labeled histograms: one in red (showing the distribution of the treatment group's estimated probabilities) and one in blue (showing the distribution of the control group's estimated probabilities). Extra credit for a legend in the plot.





2. No more than 3 sentences summarizing the differences between the two distributions of estimated probabilities, and whether/not your results are surprising and/or intuitive.

Answer: From the graph, we can find that the overall shapes of the histograms for both control and treatment groups are similar(both lean to left), while the histogram for control group leans left more. This observation tells us that although most of the data in both control and treatment groups are predicted to be a data in control group, the data in control group is predicted more likely to be assigned to a data in control group. The annotation suggests that the treatment does influence the performance of the experimental objects, but the influence is not much obvious.