

# World Rank of a University

*Rohit Kamat*

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The code below was used to determine if alumni and Awards account for a significantly greater proportion of variance in the total score for determining the world rank of a university after controlling Nature and Science Score.

```
library(SDSRegressionR)
```

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

```
## Loading required package: grid
```

```
## Loading required package: plyr
```

```
## Loading required package: readr
```

```
## Loading required package: stringr
```

```
## Loading required package: tibble
```

```
library(psych)
```

```
##  
## Attaching package: 'psych'
```

```
## The following objects are masked from 'package:ggplot2':  
##  
## %+%, alpha
```

```
#Import the College Data
college <- read.csv("/Volumes/USB20FD/Applied Regression Analysis/Data/shanghaiData.csv")

#Run the Initial Model
college_lm <- lm(total_score ~ ns + pub + alumni + award, college)
summary(college_lm)
```

```
##
## Call:
## lm(formula = total_score ~ ns + pub + alumni + award, data = college)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.2165 -1.2321 -0.0233  1.1807  5.6281
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.334557   0.251810    5.30 1.4e-07 ***
## ns           0.415853   0.006447   64.50 < 2e-16 ***
## pub          0.235631   0.006215   37.91 < 2e-16 ***
## alumni       0.105988   0.004485   23.63 < 2e-16 ***
## award        0.228947   0.004508   50.79 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.742 on 1096 degrees of freedom
## (3796 observations deleted due to missingness)
## Multiple R-squared:  0.9835, Adjusted R-squared:  0.9835
## F-statistic: 1.638e+04 on 4 and 1096 DF, p-value: < 2.2e-16
```

```
#Remove the Outliers
library(car)
```

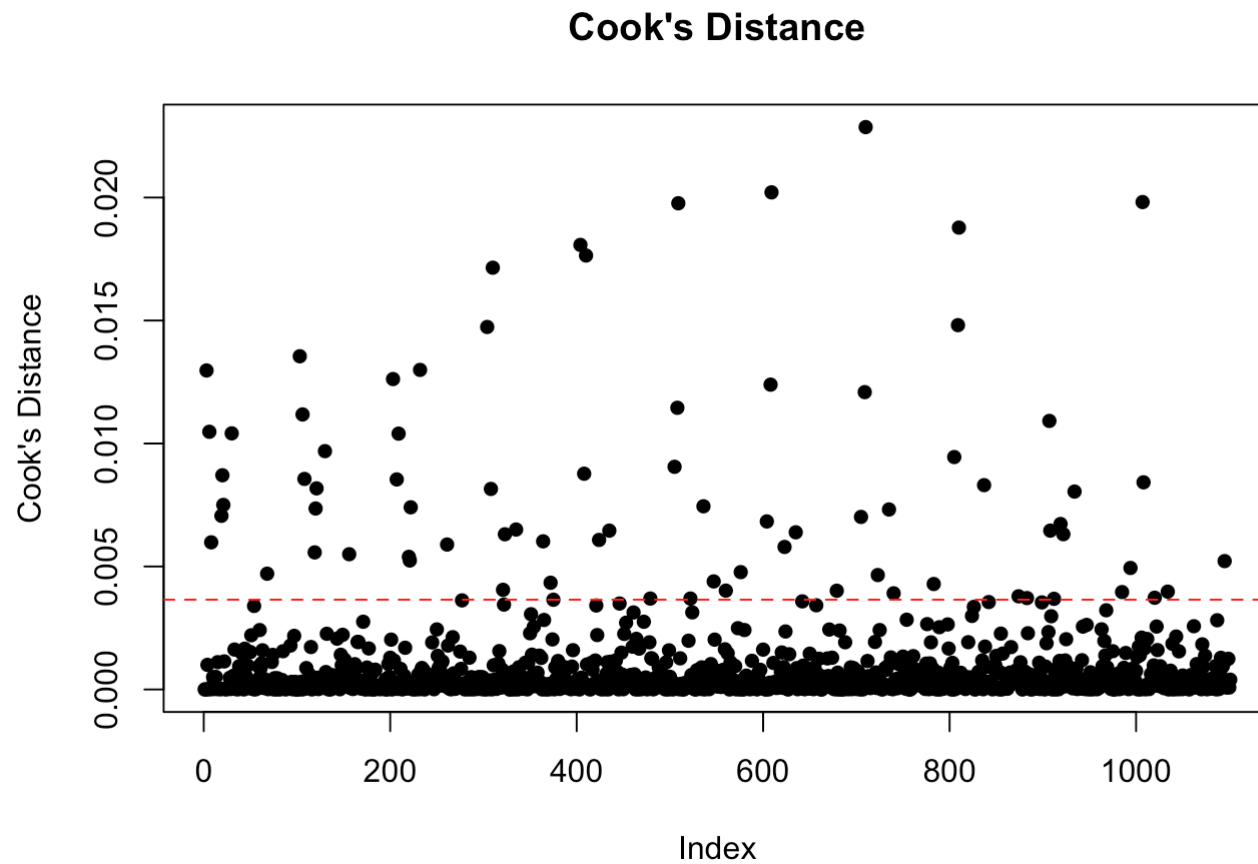
```
##
## Attaching package: 'car'
```

```
## The following object is masked from 'package:psych':  
##  
##      logit
```

```
vif(college_lm)
```

```
##          ns          pub    alumni    award  
## 3.399103 2.111479 2.536606 3.572054
```

```
cooksPlot(college_lm, print.obs = TRUE, sort.obs = TRUE, save.cutoff=TRUE)
```



##	row.names	total_score	ns	pub	alumni	award	Predicted_Y	Cooks_Distance
## 1	3521	62.1	44.4	44.5	52.3	91.3	56.73005	0.022863947
## 2	3021	61.2	43.8	43.4	56.7	87.1	55.72611	0.020210418
## 3	4403	61.0	43.0	42.4	53.3	93.4	56.23980	0.019816110
## 4	2521	60.8	43.3	44.3	56.4	84.8	55.17188	0.019767151
## 5	3805	61.9	45.8	44.0	52.9	89.2	56.77723	0.018777527
## 6	2015	73.1	67.3	70.1	39.0	78.7	67.99089	0.018073866
## 7	2021	60.2	41.5	45.7	57.8	85.2	54.99319	0.017643562
## 8	1518	58.9	40.5	44.8	59.3	80.4	53.42531	0.017147104
## 9	3804	62.9	62.0	45.2	47.8	67.2	58.21945	0.014811998
## 10	1512	73.7	68.9	71.6	40.0	78.7	69.11569	0.014737328
## 11	503	72.5	70.0	71.4	39.7	70.7	67.66263	0.013546931
## 12	1031	38.0	45.6	23.2	21.2	58.6	41.42734	0.012992334
## 13	3	73.4	70.9	72.3	41.1	72.2	68.74077	0.012969569
## 14	1002	73.7	69.6	70.3	42.0	78.7	69.31244	0.012620382
## 15	3020	64.7	65.0	46.5	52.6	68.8	60.64839	0.012393265
## 16	3520	64.1	64.8	46.2	48.5	68.8	60.05998	0.012088715
## 17	2520	64.4	64.8	46.9	50.3	68.8	60.41570	0.011451925
## 18	506	66.0	64.5	50.1	57.1	69.1	61.83436	0.011181237
## 19	3903	60.7	46.2	44.2	52.1	88.5	56.74565	0.010917931
## 20	6	67.1	65.8	52.5	59.2	68.6	63.04859	0.010477425
## 21	30	38.2	44.1	24.0	22.6	59.8	41.41518	0.010411889
## 22	1008	59.5	42.9	46.5	62.3	80.4	55.14189	0.010400583
## 23	530	38.3	44.8	24.1	21.8	58.6	41.37032	0.009689830
## 24	3800	72.6	68.7	69.4	40.0	80.7	68.97202	0.009448971
## 25	2517	72.1	68.4	69.7	40.2	78.4	68.41257	0.009052810
## 26	2019	64.8	66.2	47.7	51.5	69.1	61.38227	0.008771454
## 27	20	46.7	52.1	86.5	36.0	14.4	50.49504	0.008708106
## 28	508	58.6	43.5	47.3	61.1	75.3	54.28510	0.008559419
## 29	1006	66.4	67.6	50.3	55.5	69.1	63.00106	0.008539657
## 30	4404	59.6	56.4	44.0	49.5	66.7	55.67362	0.008421593
## 31	3832	37.1	43.1	20.4	17.7	60.2	39.72329	0.008306657
## 32	521	44.5	41.6	76.9	41.5	0.0	41.15262	0.008176183
## 33	1516	65.4	66.1	49.7	52.8	69.1	61.94973	0.008153455
## 34	3930	37.6	44.2	20.8	17.5	59.8	40.16221	0.008045879
## 35	21	44.9	43.0	76.5	43.0	0.0	41.79954	0.007503650
## 36	2548	36.7	42.3	21.0	19.2	58.4	39.27887	0.007447260
## 37	1021	44.0	40.8	77.1	40.3	0.0	40.73988	0.007405944
## 38	520	46.6	52.2	68.8	49.5	27.8	50.86467	0.007357272
## 39	3546	37.0	41.8	20.2	18.0	61.6	39.48788	0.007317678

## 40	19	46.9	52.2	67.7	51.4	28.3	50.92133	0.007060851
## 41	3516	72.8	71.6	69.6	38.0	79.7	69.78420	0.007015222
## 42	3016	72.6	70.2	70.3	41.2	78.4	69.40848	0.006831439
## 43	3915	45.2	56.5	59.9	0.0	39.9	48.07957	0.006727258
## 44	1543	37.2	42.2	22.3	20.2	58.6	39.69538	0.006502601
## 45	3904	60.5	60.0	44.9	48.5	66.7	57.27677	0.006457894
## 46	2046	36.9	41.9	21.6	19.7	58.6	39.35269	0.006457195
## 47	3047	37.1	42.5	21.5	19.5	58.4	39.51165	0.006389639
## 48	3918	43.2	52.6	73.0	31.6	14.1	46.98691	0.006311711
## 49	1531	44.2	41.7	78.1	38.4	0.0	41.14840	0.006309402
## 50	2035	43.8	41.4	77.3	37.5	0.0	40.73975	0.006080014
## 51	1572	29.3	11.3	35.7	31.2	32.8	25.26202	0.006022076
## 52	8	60.9	48.7	48.5	63.4	76.8	57.31749	0.005982028
## 53	1060	30.0	12.0	36.7	32.8	32.8	25.95833	0.005892113
## 54	3035	44.6	52.7	75.8	35.3	14.1	48.08042	0.005794232
## 55	519	46.7	51.5	85.5	34.8	14.1	49.81403	0.005572485
## 56	556	30.5	12.7	37.5	33.7	32.8	26.53332	0.005498645
## 57	1019	46.1	50.9	67.9	48.1	27.8	49.96361	0.005391649
## 58	1020	45.9	52.7	80.9	33.8	14.1	49.12316	0.005243776
## 59	4491	24.5	33.9	29.0	0.0	0.0	22.26530	0.005219562
## 60	3990	24.7	34.3	29.4	0.0	0.0	22.52589	0.004940279
## 61	2588	27.9	9.6	52.4	46.8	34.1	30.44116	0.004770622
## 62	68	28.4	7.5	54.0	51.4	34.9	30.61558	0.004706284
## 63	3534	43.8	52.0	73.7	32.5	14.6	47.11220	0.004651984
## 64	2559	32.3	20.8	49.9	31.7	46.0	35.63368	0.004392186
## 65	1580	28.1	8.3	53.2	49.1	34.2	30.35572	0.004339080
## 66	3594	26.3	9.1	48.1	43.7	34.1	28.89145	0.004287559
## 67	1529	46.4	51.9	83.3	32.2	14.1	49.18641	0.004050786
## 68	2572	30.2	15.2	34.2	32.9	32.7	26.68769	0.004018168
## 69	3091	26.9	10.1	46.9	47.4	34.1	29.41670	0.004014160
## 70	4430	36.5	40.9	19.0	17.0	59.8	38.31276	0.003977517
## 71	3981	26.1	9.7	47.9	42.4	33.0	28.70421	0.003960026
## 72	3551	34.9	20.0	48.8	32.1	55.9	37.35077	0.003908825
## 73	3869	28.5	16.7	26.7	51.0	28.2	26.43235	0.003787892
## 74	4416	44.5	53.5	59.5	0.0	39.9	46.73775	0.003728944
## 75	3878	26.1	8.8	48.1	43.1	33.3	28.51995	0.003716277
## 76	2534	45.9	52.0	80.4	33.3	14.1	48.66125	0.003695739
## 77	2090	26.9	8.2	48.7	47.9	34.2	29.12661	0.003695530
## 78	3908	55.2	61.5	62.9	48.8	50.4	58.44188	0.003685612

```
threeOuts(college_lm)
```

##	row.names	total_score	ns	pub	alumni	award	Predicted_Y
## 1	3521	62.1	44.4	44.5	52.3	91.3	56.73005
## 2	3021	61.2	43.8	43.4	56.7	87.1	55.72611
## 3	4403	61.0	43.0	42.4	53.3	93.4	56.23980
## 4	2521	60.8	43.3	44.3	56.4	84.8	55.17188
## 5	3805	61.9	45.8	44.0	52.9	89.2	56.77723
## 6	2015	73.1	67.3	70.1	39.0	78.7	67.99089
## 7	2021	60.2	41.5	45.7	57.8	85.2	54.99319
## 8	3804	62.9	62.0	45.2	47.8	67.2	58.21945
## 9	1512	73.7	68.9	71.6	40.0	78.7	69.11569
## 10	1002	73.7	69.6	70.3	42.0	78.7	69.31244
## 11	3020	64.7	65.0	46.5	52.6	68.8	60.64839
## 12	3520	64.1	64.8	46.2	48.5	68.8	60.05998
## 13	2520	64.4	64.8	46.9	50.3	68.8	60.41570
## 14	506	66.0	64.5	50.1	57.1	69.1	61.83436
## 15	3903	60.7	46.2	44.2	52.1	88.5	56.74565
## 16	6	67.1	65.8	52.5	59.2	68.6	63.04859
## 17	3800	72.6	68.7	69.4	40.0	80.7	68.97202
## 18	2517	72.1	68.4	69.7	40.2	78.4	68.41257
## 19	1518	58.9	40.5	44.8	59.3	80.4	53.42531
## 20	503	72.5	70.0	71.4	39.7	70.7	67.66263
## 21	1031	38.0	45.6	23.2	21.2	58.6	41.42734
## 22	3	73.4	70.9	72.3	41.1	72.2	68.74077
## 23	30	38.2	44.1	24.0	22.6	59.8	41.41518
## 24	1008	59.5	42.9	46.5	62.3	80.4	55.14189
## 25	530	38.3	44.8	24.1	21.8	58.6	41.37032
## 26	2019	64.8	66.2	47.7	51.5	69.1	61.38227
## 27	20	46.7	52.1	86.5	36.0	14.4	50.49504
## 28	508	58.6	43.5	47.3	61.1	75.3	54.28510
## 29	1006	66.4	67.6	50.3	55.5	69.1	63.00106
## 30	4404	59.6	56.4	44.0	49.5	66.7	55.67362
## 31	3832	37.1	43.1	20.4	17.7	60.2	39.72329
## 32	521	44.5	41.6	76.9	41.5	0.0	41.15262
## 33	1516	65.4	66.1	49.7	52.8	69.1	61.94973
## 34	3930	37.6	44.2	20.8	17.5	59.8	40.16221
## 35	21	44.9	43.0	76.5	43.0	0.0	41.79954
## 36	2548	36.7	42.3	21.0	19.2	58.4	39.27887
## 37	1021	44.0	40.8	77.1	40.3	0.0	40.73988
## 38	520	46.6	52.2	68.8	49.5	27.8	50.86467
## 39	3546	37.0	41.8	20.2	18.0	61.6	39.48788

## 40	19	46.9	52.2	67.7	51.4	28.3	50.92133
## 41	3516	72.8	71.6	69.6	38.0	79.7	69.78420
## 42	3016	72.6	70.2	70.3	41.2	78.4	69.40848
## 43	3915	45.2	56.5	59.9	0.0	39.9	48.07957
## 44	1543	37.2	42.2	22.3	20.2	58.6	39.69538
## 45	3904	60.5	60.0	44.9	48.5	66.7	57.27677
## 46	2046	36.9	41.9	21.6	19.7	58.6	39.35269
## 47	3047	37.1	42.5	21.5	19.5	58.4	39.51165
## 48	3918	43.2	52.6	73.0	31.6	14.1	46.98691
## 49	1531	44.2	41.7	78.1	38.4	0.0	41.14840
## 50	2035	43.8	41.4	77.3	37.5	0.0	40.73975
## 51	1572	29.3	11.3	35.7	31.2	32.8	25.26202
## 52	8	60.9	48.7	48.5	63.4	76.8	57.31749
## 53	1060	30.0	12.0	36.7	32.8	32.8	25.95833
## 54	3035	44.6	52.7	75.8	35.3	14.1	48.08042
## 55	519	46.7	51.5	85.5	34.8	14.1	49.81403
## 56	556	30.5	12.7	37.5	33.7	32.8	26.53332
## 57	1019	46.1	50.9	67.9	48.1	27.8	49.96361
## 58	1020	45.9	52.7	80.9	33.8	14.1	49.12316
## 59	4491	24.5	33.9	29.0	0.0	0.0	22.26530
## 60	3990	24.7	34.3	29.4	0.0	0.0	22.52589
## 61	2588	27.9	9.6	52.4	46.8	34.1	30.44116
## 62	68	28.4	7.5	54.0	51.4	34.9	30.61558
## 63	3534	43.8	52.0	73.7	32.5	14.6	47.11220
## 64	2559	32.3	20.8	49.9	31.7	46.0	35.63368
## 65	1580	28.1	8.3	53.2	49.1	34.2	30.35572
## 66	3594	26.3	9.1	48.1	43.7	34.1	28.89145
## 67	1529	46.4	51.9	83.3	32.2	14.1	49.18641
## 68	2572	30.2	15.2	34.2	32.9	32.7	26.68769
## 69	3091	26.9	10.1	46.9	47.4	34.1	29.41670
## 70	4430	36.5	40.9	19.0	17.0	59.8	38.31276
## 71	3981	26.1	9.7	47.9	42.4	33.0	28.70421
## 72	3551	34.9	20.0	48.8	32.1	55.9	37.35077
## 73	3869	28.5	16.7	26.7	51.0	28.2	26.43235
## 74	4416	44.5	53.5	59.5	0.0	39.9	46.73775
## 75	3878	26.1	8.8	48.1	43.1	33.3	28.51995
## 76	2534	45.9	52.0	80.4	33.3	14.1	48.66125
## 77	2090	26.9	8.2	48.7	47.9	34.2	29.12661
## 78	3908	55.2	61.5	62.9	48.8	50.4	58.44188
## 79	51	31.4	30.4	51.9	37.1	21.1	34.96870
## 80	60	29.9	28.7	49.1	25.9	32.9	35.11649



## 81	85	25.6	22.7	52.2	25.1	15.8	29.35205
## 82	97	24.5	21.8	54.6	16.6	15.8	28.64240
## 83	565	29.3	24.6	49.3	25.0	32.2	33.20296
## 84	601	23.5	19.4	53.3	16.0	15.5	27.20575
## 85	1066	28.8	23.6	49.1	24.3	32.2	32.66579
## 86	1604	24.5	24.5	48.2	13.7	18.9	28.65953
## 87	2110	24.2	24.0	47.6	13.4	18.9	28.27843
## 88	2612	24.2	23.5	48.4	13.1	18.9	28.22721
## 89	3100	25.4	24.3	49.3	13.2	22.1	29.51519
## 90	3600	25.5	23.9	50.8	12.1	22.8	29.74597
## 91	3879	26.0	25.6	50.9	12.0	22.2	30.32852
## 92	3971	27.4	26.7	51.9	11.8	22.1	30.97749
## 93	1	100.0	100.0	100.0	100.0	100.0	99.97651
## 94	2	73.6	56.6	70.9	99.8	93.4	73.53935
## 95	7	62.3	54.2	69.5	79.4	60.6	62.53981
## 96	9	60.1	44.7	56.4	75.6	81.9	59.97624
## 97	18	47.8	57.9	58.8	0.0	37.6	47.87599
## 98	93	24.6	18.1	27.2	47.8	25.0	26.06057
## 99	501	100.0	100.0	100.0	100.0	100.0	99.97651
## 100	502	72.6	59.5	67.1	96.3	91.5	73.04397
## 101	509	58.6	43.7	54.1	72.9	80.2	58.34306
## 102	518	47.7	54.8	61.1	0.0	36.8	46.94564
## 103	571	27.9	5.6	54.3	49.5	34.2	29.53451
## 104	599	23.6	14.8	27.3	46.1	24.5	24.41717
## 105	1001	100.0	100.0	100.0	100.0	100.0	99.97651
## 106	1004	71.6	58.2	65.4	93.6	91.5	71.81662
## 107	1005	70.0	68.4	61.7	74.6	80.6	70.67719
## 108	1018	46.8	53.7	59.8	0.0	36.8	46.18188
## 109	1076	27.1	7.5	48.5	51.6	34.2	29.18054
## 110	1084	25.5	16.8	25.8	48.5	31.6	26.77532
## 111	1511	100.0	100.0	100.0	100.0	100.0	99.97651
## 112	1514	70.4	56.0	64.1	90.3	91.5	70.24566
## 113	1519	57.1	39.5	51.9	67.4	81.9	55.88436
## 114	1528	46.6	51.5	60.8	0.0	36.8	45.50263
## 115	1583	27.7	17.2	25.9	53.4	24.5	25.85904
## 116	2014	100.0	100.0	100.0	100.0	100.0	99.97651
## 117	2017	70.2	53.9	65.4	89.4	91.5	69.58330
## 118	2022	57.0	38.6	51.6	65.8	84.3	55.81929
## 119	2031	45.9	49.7	59.9	0.0	36.8	44.54203
## 120	2083	28.1	17.9	27.6	52.1	24.5	26.41293
## 121	2515	100.0	100.0	100.0	100.0	100.0	99.97651

## 122	2518	71.4	70.1	61.4	70.5	80.3	70.81022
## 123	2519	69.6	54.3	65.7	88.5	92.6	69.97678
## 124	2523	57.3	39.8	50.5	65.5	83.9	55.93575
## 125	2533	46.0	51.8	60.7	0.0	40.1	46.35935
## 126	2536	44.2	43.4	79.8	36.5	0.0	42.05455
## 127	2585	28.3	18.7	27.9	50.8	24.4	26.65562
## 128	3015	100.0	100.0	100.0	100.0	100.0	99.97651
## 129	3017	72.0	70.6	60.6	72.8	81.9	71.43973
## 130	3019	70.0	54.1	65.1	87.1	96.7	70.54253
## 131	3023	57.5	40.4	50.9	65.0	83.9	56.22652
## 132	3031	46.7	54.2	62.0	0.0	40.1	47.66372
## 133	3036	44.1	45.0	77.7	36.7	0.0	42.24628
## 134	3083	28.7	19.4	27.4	54.2	24.4	27.18926
## 135	3515	100.0	100.0	100.0	100.0	100.0	99.97651
## 136	3519	69.8	55.0	65.9	80.3	97.2	70.49906
## 137	3523	57.2	40.6	49.2	61.8	85.3	55.89048
## 138	3532	46.6	54.1	60.6	0.0	41.2	47.54409
## 139	3536	43.4	44.6	76.2	33.9	0.0	41.42973
## 140	3587	27.6	17.4	27.3	50.0	24.4	25.88884
## 141	3609	24.6	32.0	28.4	13.3	16.8	26.58974
## 142	3799	100.0	100.0	100.0	100.0	100.0	99.97651
## 143	3803	69.6	54.0	66.2	79.1	97.3	70.04960
## 144	3816	46.2	54.1	59.3	0.0	40.2	47.00882
## 145	3837	34.5	19.7	48.7	31.6	54.7	36.87472
## 146	3894	24.9	33.5	28.4	13.1	16.4	27.10075
## 147	3898	100.0	100.0	100.0	100.0	100.0	99.97651
## 148	3899	72.1	71.1	70.9	41.8	82.8	70.99508
## 149	3900	70.5	73.6	61.5	68.4	80.7	72.15827
## 150	3902	69.2	56.2	66.5	79.1	96.6	70.87491
## 151	3906	57.4	43.0	50.7	61.4	86.3	57.42853
## 152	3920	42.3	42.4	77.3	35.3	0.0	40.92243
## 153	3939	34.2	18.5	48.4	31.1	54.3	36.16044
## 154	3964	28.3	16.8	26.8	50.2	28.0	26.36692
## 155	4398	100.0	100.0	100.0	100.0	100.0	99.97651
## 156	4399	73.3	70.1	70.6	40.7	89.6	71.94879
## 157	4400	70.4	73.1	61.1	68.2	80.7	71.83490
## 158	4402	68.8	55.6	66.4	77.1	96.6	70.38986
## 159	4406	57.1	42.9	49.8	59.8	86.3	57.00529
## 160	4439	33.6	17.4	47.3	30.3	54.3	35.35901
## 161	4469	27.6	19.5	26.4	48.9	28.0	27.25769
## 162	4483	25.3	7.7	46.4	41.4	33.0	27.41307

##	Student_Resid	Hat_Values	Cooks_Distance	inThree
## 1	3.100839	0.011749781	0.022863947	1
## 2	3.158109	0.010030279	0.020210418	1
## 3	2.750381	0.012928594	0.019816110	1
## 4	3.245885	0.009293790	0.019767151	1
## 5	2.956426	0.010627575	0.018777527	1
## 6	2.948045	0.010291067	0.018073866	1
## 7	3.003496	0.009684466	0.017643562	1
## 8	2.700433	0.010053760	0.014811998	1
## 9	2.645398	0.010419761	0.014737328	1
## 10	2.531010	0.009754335	0.012620382	1
## 11	2.338923	0.011200363	0.012393265	1
## 12	2.331988	0.010992515	0.012088715	1
## 13	2.299500	0.010712827	0.011451925	1
## 14	2.402792	0.009590526	0.011181237	1
## 15	2.281826	0.010375658	0.010917931	1
## 16	2.336800	0.009502431	0.010477425	1
## 17	2.093798	0.010661785	0.009448971	1
## 18	2.127290	0.009903248	0.009052810	1
## 19	3.156188	NA	0.017147104	NA
## 20	2.788909	NA	0.013546931	NA
## 21	NA	0.016241834	0.012992334	NA
## 22	2.686573	NA	0.012969569	NA
## 23	NA	0.014832804	0.010411889	NA
## 24	2.512015	NA	0.010400583	NA
## 25	NA	0.015128439	0.009689830	NA
## 26	NA	0.011141635	0.008771454	NA
## 27	-2.188385	NA	0.008708106	NA
## 28	2.485492	NA	0.008559419	NA
## 29	NA	0.010971237	0.008539657	NA
## 30	2.263146	NA	0.008421593	NA
## 31	NA	0.017673767	0.008306657	NA
## 32	NA	0.010833398	0.008176183	NA
## 33	NA	0.010181961	0.008153455	NA
## 34	NA	0.017935349	0.008045879	NA
## 35	NA	0.011571697	0.007503650	NA
## 36	NA	0.016437194	0.007447260	NA
## 37	NA	0.010355156	0.007405944	NA
## 38	-2.455549	NA	0.007357272	NA
## 39	NA	0.017322931	0.007317678	NA
## 40	-2.315988	NA	0.007060851	NA

## 41	NA	0.011437466	0.007015222	NA
## 42	NA	0.009974578	0.006831439	NA
## 43	NA	0.012016253	0.006727258	NA
## 44	NA	0.015362139	0.006502601	NA
## 45	NA	0.009257964	0.006457894	NA
## 46	NA	0.015777239	0.006457195	NA
## 47	NA	0.016136249	0.006389639	NA
## 48	-2.181037	NA	0.006311711	NA
## 49	NA	0.010074495	0.006309402	NA
## 50	NA	0.009661470	0.006080014	NA
## 51	2.324413	NA	0.006022076	NA
## 52	2.063716	NA	0.005982028	NA
## 53	2.326388	NA	0.005892113	NA
## 54	-2.005086	NA	0.005794232	NA
## 55	NA	NA	0.005572485	NA
## 56	2.283031	NA	0.005498645	NA
## 57	-2.223904	NA	0.005391649	NA
## 58	NA	NA	0.005243776	NA
## 59	NA	0.015375236	0.005219562	NA
## 60	NA	0.015375042	0.004940279	NA
## 61	NA	0.010965293	0.004770622	NA
## 62	NA	0.014139064	0.004706284	NA
## 63	NA	NA	0.004651984	NA
## 64	NA	NA	0.004392186	NA
## 65	NA	0.012614949	0.004339080	NA
## 66	NA	0.009504170	0.004287559	NA
## 67	NA	NA	0.004050786	NA
## 68	2.021161	NA	0.004018168	NA
## 69	NA	0.009435846	0.004014160	NA
## 70	NA	0.017720884	0.003977517	NA
## 71	NA	NA	0.003960026	NA
## 72	NA	0.009684410	0.003908825	NA
## 73	NA	0.013094290	0.003787892	NA
## 74	NA	0.011050827	0.003728944	NA
## 75	NA	0.009447926	0.003716277	NA
## 76	NA	NA	0.003695739	NA
## 77	NA	0.011061476	0.003695530	NA
## 78	NA	NA	0.003685612	NA
## 79	-2.051260	NA	NA	NA
## 80	-2.996486	NA	NA	NA
## 81	-2.155626	NA	NA	NA

## 82	-2.380187	NA	NA	NA
## 83	-2.242610	NA	NA	NA
## 84	-2.129623	NA	NA	NA
## 85	-2.221508	NA	NA	NA
## 86	-2.389405	NA	NA	NA
## 87	-2.342877	NA	NA	NA
## 88	-2.313528	NA	NA	NA
## 89	-2.364309	NA	NA	NA
## 90	-2.439946	NA	NA	NA
## 91	-2.487036	NA	NA	NA
## 92	-2.055514	NA	NA	NA
## 93	NA 0.023221422		NA	NA
## 94	NA 0.017024779		NA	NA
## 95	NA 0.009275730		NA	NA
## 96	NA 0.009853669		NA	NA
## 97	NA 0.012306754		NA	NA
## 98	NA 0.012385277		NA	NA
## 99	NA 0.023221422		NA	NA
## 100	NA 0.014449533		NA	NA
## 101	NA 0.009156171		NA	NA
## 102	NA 0.010924078		NA	NA
## 103	NA 0.015167586		NA	NA
## 104	NA 0.011482145		NA	NA
## 105	NA 0.023221422		NA	NA
## 106	NA 0.013637753		NA	NA
## 107	NA 0.009272689		NA	NA
## 108	NA 0.010554320		NA	NA
## 109	NA 0.012383647		NA	NA
## 110	NA 0.011781255		NA	NA
## 111	NA 0.023221422		NA	NA
## 112	NA 0.012928097		NA	NA
## 113	NA 0.009888073		NA	NA
## 114	NA 0.009974237		NA	NA
## 115	NA 0.015819747		NA	NA
## 116	NA 0.023221422		NA	NA
## 117	NA 0.013303782		NA	NA
## 118	NA 0.010815899		NA	NA
## 119	NA 0.009503949		NA	NA
## 120	NA 0.014261188		NA	NA
## 121	NA 0.023221422		NA	NA
## 122	NA 0.009152110		NA	NA

## 123	NA 0.013219638	NA	NA
## 124	NA 0.010107051	NA	NA
## 125	NA 0.010742211	NA	NA
## 126	NA 0.010121336	NA	NA
## 127	NA 0.013602039	NA	NA
## 128	NA 0.023221422	NA	NA
## 129	NA 0.009865075	NA	NA
## 130	NA 0.013841708	NA	NA
## 131	NA 0.009896603	NA	NA
## 132	NA 0.011416006	NA	NA
## 133	NA 0.010197702	NA	NA
## 134	NA 0.015696100	NA	NA
## 135	NA 0.023221422	NA	NA
## 136	NA 0.012865811	NA	NA
## 137	NA 0.010044074	NA	NA
## 138	NA 0.011536562	NA	NA
## 139	NA 0.009270332	NA	NA
## 140	NA 0.013399988	NA	NA
## 141	NA 0.009932443	NA	NA
## 142	NA 0.023221422	NA	NA
## 143	NA 0.013219808	NA	NA
## 144	NA 0.011281568	NA	NA
## 145	NA 0.009382628	NA	NA
## 146	NA 0.010850212	NA	NA
## 147	NA 0.023221422	NA	NA
## 148	NA 0.011323650	NA	NA
## 149	NA 0.010155751	NA	NA
## 150	NA 0.012314599	NA	NA
## 151	NA 0.009737918	NA	NA
## 152	NA 0.009258767	NA	NA
## 153	NA 0.009895269	NA	NA
## 154	NA 0.012743077	NA	NA
## 155	NA 0.023221422	NA	NA
## 156	NA 0.014098074	NA	NA
## 157	NA 0.010033907	NA	NA
## 158	NA 0.012335791	NA	NA
## 159	NA 0.009723516	NA	NA
## 160	NA 0.010332913	NA	NA
## 161	NA 0.012685466	NA	NA
## 162	NA 0.009382125	NA	NA

```
# Three Outs Outliers = 18
g_college <- college[!row.names(college) %in%c(3521,3021,4403,2521,3805,2015,2021,3804,1512,1002,3020,3520,2520,506,3903,6,3800,2517),]
```

```
#Rerun the Final Model
new_college_lm<-lm(total_score ~ ns + pub + alumni + award, g_college)
summary(new_college_lm)
```

```
##
## Call:
## lm(formula = total_score ~ ns + pub + alumni + award, data = g_college)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.0611 -1.1353  0.0595  1.1224  6.0705
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.244374   0.238944   5.208 2.29e-07 ***
## ns           0.410574   0.006180  66.436 < 2e-16 ***
## pub          0.242120   0.005943  40.741 < 2e-16 ***
## alumni       0.109702   0.004260  25.753 < 2e-16 ***
## award        0.218962   0.004365  50.166 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.638 on 1078 degrees of freedom
## (3796 observations deleted due to missingness)
## Multiple R-squared:  0.9845, Adjusted R-squared:  0.9844
## F-statistic: 1.706e+04 on 4 and 1078 DF, p-value: < 2.2e-16
```

```
#Tag the Observation in the model. Keep the observation in the model
g_college$in_new_college_lm<- tagObs(new_college_lm)
g_college_full<- g_college[which(g_college$in_new_college_lm == 1), ]
sum(g_college_full$in_new_college_lm) #Double Check
```

```
## [1] 1083
```

```
#Now I will perform the sequential regression. First we will perform the model with just our nuisance variables.
m1_seq<- lm(total_score ~ ns + pub, g_college_full)
summary(m1_seq)
```

```
##
## Call:
## lm(formula = total_score ~ ns + pub, data = g_college_full)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -13.0084  -3.1556  -0.0978   2.4618  17.5806
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   5.32279     0.66173   8.044 2.28e-15 ***
## ns             0.76845     0.01287  59.703 < 2e-16 ***
## pub            0.10881     0.01541   7.062 2.93e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.772 on 1080 degrees of freedom
## Multiple R-squared:  0.8678, Adjusted R-squared:  0.8675
## F-statistic: 3544 on 2 and 1080 DF, p-value: < 2.2e-16
```

```
summary(m1_seq)$r.squared
```

```
## [1] 0.8677743
```

```
lmBeta(m1_seq)
```

```
##           ns           pub
## 0.8626399 0.1020410
```

```
pCorr(m1_seq)
```



```
##      Partial_Corr Partial_Corr_sq Part_Corr Part_Corr_sq
## ns      0.8760507      0.76746485 0.66060654  0.436401001
## pub      0.2101003      0.04414214 0.07814261  0.006106268
```

```
#Now we will perform the model with the variable of interest.
m2_seq <- lm(total_score ~ ns + pub + alumni + award , g_college_full)
summary(m2_seq)
```

```
##
## Call:
## lm(formula = total_score ~ ns + pub + alumni + award, data = g_college_full)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.0611 -1.1353  0.0595  1.1224  6.0705
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.244374   0.238944   5.208 2.29e-07 ***
## ns           0.410574   0.006180  66.436 < 2e-16 ***
## pub          0.242120   0.005943  40.741 < 2e-16 ***
## alumni       0.109702   0.004260  25.753 < 2e-16 ***
## award        0.218962   0.004365  50.166 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.638 on 1078 degrees of freedom
## Multiple R-squared:  0.9845, Adjusted R-squared:  0.9844
## F-statistic: 1.706e+04 on 4 and 1078 DF,  p-value: < 2.2e-16
```

```
summary(m2_seq)$r.squared
```

```
## [1] 0.9844529
```

```
summary(m2_seq)$r.squared - summary(m1_seq)$r.squared #Difference of Variance between the variables of interest and the nuisance variables.
```

```
## [1] 0.1166785
```

```
#Perform the anova to find the significance of variance change
anova(m1_seq, m2_seq)
```

```
## Analysis of Variance Table
##
## Model 1: total_score ~ ns + pub
## Model 2: total_score ~ ns + pub + alumni + award
##   Res.Df    RSS Df Sum of Sq    F    Pr(>F)
## 1    1080 24598.2
## 2    1078  2892.3  2    21706 4045.1 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
#Find the Best Predictor of total score for college ranking
lmBeta(m2_seq)
```

```
##           ns           pub      alumni      award
## 0.4609014 0.2270609 0.1549163 0.3523259
```

```
#How much unique variance of total college score
pCorr(m2_seq)
```

```
##           Partial_Corr Partial_Corr_sq Part_Corr Part_Corr_sq
## ns           0.8964970           0.8037068 0.25230249 0.063656547
## pub           0.7786279           0.6062615 0.15472173 0.023938813
## alumni        0.6171708           0.3808998 0.09780255 0.009565338
## award         0.8367233           0.7001059 0.19051238 0.036294966
```