

601Final

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Homework 2

Let's read in the data and see what it looks like:

```
«««< HEAD
```

```
data <- read.csv("C:/Users/wolpe/DACSS601August2021/_data/Public_School_Characteristics_2017-18.csv")
```

What's in the data?

This dataset looks at the characteristics of various public schools across the United States. Among the variables in the dataset are identifying characteristics such as the name of the school, its school district, and its location; there are also several quantitative variables such as the number of students in each grade, as well as the overall number of students broken down in categories such as race and gender.

The numbers of the data

The data has 79 total columns and just over 100,000 rows. It is unlikely that all of this info will be useful, so in the next section we can see if the data can be cleaned and subset to be more useful to the project.

Homework 3

Cleaning Data

The data is definitely unclean, let's filter out some stuff so that we have more complete data.

```
data_clean <- data %>%
  filter(!is.na(TOTAL) & TOTAL > 0 & !is.na(FTE) & FTE > 0 & STUTERATIO < 500) %>%
  filter(SCHOOL_TYPE_TEXT == "Regular school" & VIRTUAL == "Not a virtual school") %>%
  filter(SCHOOL_LEVEL == "High" | SCHOOL_LEVEL == "Middle" | SCHOOL_LEVEL == "Elementary") %>%
  filter((is.na(G13) | G13 == 0) & (is.na(PK) | PK == 0)) %>%
  filter(!is.na(STABR) & !is.na(SCH_NAME))
head(data_clean)
```

##	i..X	Y	OBJECTID	NCESSCH	NMCNTY	SURVYEAR
## 1	-151.0701	60.49144	3	20039000448	Kenai Peninsula Borough	2017-2018
## 2	-151.2791	60.56828	4	20039000463	Kenai Peninsula Borough	2017-2018

## 3	-166.5224	53.86895	15	20072000340	Aleutians West Census Area 2017-2018											
## 4	-166.5296	53.87267	16	20072000661	Aleutians West Census Area 2017-2018											
## 5	-161.7707	60.80436	18	20000100207	Bethel Census Area 2017-2018											
## 6	-161.7704	60.80258	19	20000100208	Bethel Census Area 2017-2018											
##	STABR	LEAID	ST_LEAID	LEA_NAME												
## 1	AK	200390	AK-24	Kenai Peninsula Borough School District												
## 2	AK	200390	AK-24	Kenai Peninsula Borough School District												
## 3	AK	200720	AK-47	Unalaska City School District												
## 4	AK	200720	AK-47	Unalaska City School District												
## 5	AK	200001	AK-31	Lower Kuskokwim School District												
## 6	AK	200001	AK-31	Lower Kuskokwim School District												
##				SCH_NAME	LSTREET1 LSTREET2											
## 1	Soldotna Montessori Charter School			158 E Park Ave												
## 2	Kaleidoscope School of Arts & Science			549 N Forest Dr												
## 3	Eagle's View Elementary School			503 East Broadway												
## 4	Unalaska Jr/Sr High School			55 E Broadway												
## 5	Gladys Jung Elementary			1007 Ron Edwards Memorial Dr												
## 6	Bethel Regional High School			1006 Ron Edwards Memorial Dr												
##	LSTREET3	LCITY	LSTATE	LZIP	LZIP4	PHONE	GSLO	GSHI								
## 1		Soldotna	AK	99669	NA	(907)260-9221	KG	06								
## 2		Kenai	AK	99611	NA	(907)283-0804	KG	05								
## 3		Unalaska	AK	99685	NA	(907)581-3979	PK	06								
## 4		Unalaska	AK	99685	NA	(907)581-1222	07	12								
## 5		Bethel	AK	99559	NA	(907)543-4440	03	06								
## 6		Bethel	AK	99559	NA	(907)543-3957	07	12								
##	VIRTUAL		TOTFRL	FRELCH	REDLCH	PK	KG	G01	G02	G03	G04	G05	G06	G07		
## 1	Not a virtual school	43	23	20	NA	23	23	27	22	25	28	19	NA			
## 2	Not a virtual school	69	50	19	NA	40	43	42	46	46	43	NA	NA			
## 3	Not a virtual school	53	35	18	0	32	30	36	33	31	26	29	NA			
## 4	Not a virtual school	38	27	11	NA	NA	NA	NA	NA	NA	NA	NA	30			
## 5	Not a virtual school	294	294	0	NA	NA	NA	NA	97	75	79	90	NA			
## 6	Not a virtual school	373	373	0	NA	NA	NA	NA	NA	NA	NA	NA	94			
##	G08	G09	G10	G11	G12	G13	TOTAL	MEMBER	AM	HI	BL	WH	HP	TR	FTE	LATCOD
## 1	NA	NA	NA	NA	NA	NA	167	167	8	5	0	136	0	15	10.35	60.49144
## 2	NA	NA	NA	NA	NA	NA	260	260	16	14	3	168	0	56	16.75	60.56828
## 3	NA	NA	NA	NA	NA	NA	217	217	23	30	2	56	13	3	13.50	53.86895
## 4	25	26	38	36	29	NA	184	184	24	21	0	38	8	0	14.50	53.87267
## 5	NA	NA	NA	NA	NA	NA	341	341	284	6	1	44	0	1	22.13	60.80436
## 6	90	106	52	63	70	NA	475	475	418	7	1	38	0	2	33.05	60.80258
##	LONCOD	ULOCAL			STUTERATIO			STITLEI			AMALM	AMALF	ASALM	ASALF		
## 1	-151.0702	33-Town: Remote			16.14			Not Applicable			4	4	0	3		
## 2	-151.2791	33-Town: Remote			15.52			Not Applicable			10	6	1	2		
## 3	-166.5225	43-Rural: Remote			16.07			Not Applicable			11	12	52	38		
## 4	-166.5296	43-Rural: Remote			12.69			Not Applicable			12	12	52	41		
## 5	-161.7707	41-Rural: Fringe			15.41			Yes			141	143	2	3		
## 6	-161.7704	41-Rural: Fringe			14.37			Yes			221	197	5	4		
##	HIALM	HIALF	BLALM	BLALF	WHALM	WHALF	HPALM	HPALF	TRALM	TRALF	TOTMENROL					
## 1	2	3	0	0	58	78	0	0	7	8	71					
## 2	6	8	3	0	82	86	0	0	26	30	128					
## 3	14	16	0	2	26	30	7	6	1	2	111					
## 4	12	9	0	0	23	15	4	4	0	0	103					
## 5	4	2	0	1	21	23	0	0	0	1	168					
## 6	4	3	0	1	20	18	0	0	1	1	251					
##	TOTFENROL	STATUS	UG	AE	SCHOOL_TYPE_TEXT	SY_STATUS_TEXT			SCHOOL_LEVEL							

```
## 1      96      1 NA NA Regular school Currently operational Elementary
## 2     132      1 NA NA Regular school Currently operational Elementary
## 3     106      1 NA NA Regular school Currently operational Elementary
## 4      81      1 NA NA Regular school Currently operational High
## 5     173      1 NA NA Regular school Currently operational Elementary
## 6     224      1 NA NA Regular school Currently operational High
## AS CHARTER_TEXT MAGNET_TEXT
## 1 3      Yes      No
## 2 3      Yes      No
## 3 90     No       No
## 4 93     No       No
## 5 5      No       No
## 6 9      No       No
```

With filtering, we now have a subset of the original data that will be much more useful for analysis. The schools were filtered to include only regular, non-virtual schools at the elementary, middle, and high school levels. It also removed any schools that had students younger than kindergarten (PK) or those past their senior year of high school (G13).

There is still an issue however with student to teacher ratios, so let's look at a distribution of that:

```
data_clean %>%
  summarise(quantile = c(0, 0.005, 0.025, 0.5, 0.975, 0.995, 1), quant.val = quantile(STUTERATIO, c(0, 0.005, 0.025, 0.5, 0.975, 0.995, 1)))

## quantile quant.val
## 1 0.000 0.1900
## 2 0.005 4.8200
## 3 0.025 8.1800
## 4 0.500 15.7700
## 5 0.975 27.0000
## 6 0.995 34.4274
## 7 1.000 485.0000
```

Looking at the distribution, we see that 99% of the dataset falls between a student to teacher ratio of 4.82 and 34.4274, which we can use as a boundary to remove outliers of this variable from the dataset.

```
data_clean <- data_clean %>%
  filter(STUTERATIO >= 4.82 & STUTERATIO <= 34.4274)
```

Subsetting Columns

Now let's subset the columns in the dataset to only include those that interest this project.

```
data_sub <- data_clean %>%
  select(SCH_NAME, STABR, GSLO, GSHI, G01, G02, G03, G04, G05, G06, G07, G08, G09, G10, G11, G12, TOTAL)
head(data_sub)
```

```
## SCH_NAME STABR GSLO GSHI G01 G02 G03 G04 G05 G06
## 1 Soldotna Montessori Charter School AK KG 06 23 27 22 25 28 19
## 2 Kaleidoscope School of Arts & Science AK KG 05 43 42 46 46 43 NA
## 3 Eagle's View Elementary School AK PK 06 30 36 33 31 26 29
## 4 Unalaska Jr/Sr High School AK 07 12 NA NA NA NA NA NA
```

```
## 5          Gladys Jung Elementary    AK    03    06 NA NA 97 75 79 90
## 6          Bethel Regional High School AK    07    12 NA NA NA NA NA NA
##   G07 G08 G09 G10 G11 G12 TOTAL  AM HI BL  WH HP TR   FTE STUTERATIO AMALM
## 1  NA  NA  NA  NA  NA  NA   167   8 5 0 136 0 15 10.35      16.14    4
## 2  NA  NA  NA  NA  NA  NA   260  16 14 3 168 0 56 16.75      15.52   10
## 3  NA  NA  NA  NA  NA  NA   217  23 30 2  56 13 3 13.50      16.07   11
## 4  30  25  26  38  36  29   184  24 21 0  38  8 0 14.50      12.69   12
## 5  NA  NA  NA  NA  NA  NA   341 284 6 1  44 0 1 22.13      15.41  141
## 6  94  90 106  52  63  70   475 418 7 1  38 0 2 33.05      14.37  221
##   AMALF ASALM ASALF HIALM HIALF BLALM BLALF WHALM WHALF HPALM HPALF TRALM TRALF
## 1     4     0     3     2     3     0     0    58     78     0     0     7     8
## 2     6     1     2     6     8     3     0    82     86     0     0    26    30
## 3    12    52    38    14    16     0     2    26    30     7     6     1     2
## 4    12    52    41    12     9     0     0    23    15     4     4     0     0
## 5   143     2     3     4     2     0     1    21    23     0     0     0     1
## 6   197     5     4     4     3     0     1    20    18     0     0     1     1
##   TOTMENROL TOTFENROL SCHOOL_LEVEL
## 1         71         96   Elementary
## 2        128        132   Elementary
## 3        111        106   Elementary
## 4        103         81      High
## 5        168        173   Elementary
## 6        251        224      High
```

Arranging Some Data

It may be useful to be the top several rows for certain columns, in this script, we look at the top 6 schools ordered based on their student to teacher ratio.

```
data_stuteratio <- data_sub %>%
  arrange(STUTERATIO, by_group = TRUE)
head(data_stuteratio)
```

```
##          SCH_NAME STABR GSLO GSHI G01 G02 G03 G04 G05 G06 G07
## 1      George Jr Republic HS    PA    09    12 NA NA NA NA NA NA NA
## 2      Medicine Bow Elementary    WY    KG    06    1  4  0  3  2  1 NA
## 3 LONE STAR UNDIVIDED HIGH SCHOOL    CO    09    12 NA NA NA NA NA NA NA
## 4      VERDIGRE MIDDLE SCHOOL    NE    07    08 NA NA NA NA NA NA  4
## 5      Judith Gap High School    MT    09    12 NA NA NA NA NA NA NA
## 6      Mobile Elementary School    AZ    KG    08    5  0  5  1  3  1  0
##   G08 G09 G10 G11 G12 TOTAL  AM HI  BL WH HP TR   FTE STUTERATIO AMALM AMALF
## 1  NA  54  54  88  83   279  1 36 135 80 0 26 57.94      4.82     1     0
## 2  NA  NA  NA  NA  NA    11  0 0  0 11 0 0  2.28      4.82     0     0
## 3  NA  16   5   4   4    29 NA 3  NA 26 NA NA  6.01      4.83    NA    NA
## 4  10  NA  NA  NA  NA    14  2 0  0 12 0 0  2.90      4.83     1     1
## 5  NA   4   3   3   3    13  0 1  0 12 0 0  2.68      4.85     0     0
## 6   2  NA  NA  NA  NA    19  0 4  5  9 0 1  3.90      4.87     0     0
##   ASALM ASALF HIALM HIALF BLALM BLALF WHALM WHALF HPALM HPALF TRALM TRALF
## 1     1     0     36     0   135     0    80     0     0     0    26     0
## 2     0     0     0     0     0     0     4     7     0     0     0     0
## 3    NA    NA     2     1    NA    NA    13    13    NA    NA    NA    NA
## 4     0     0     0     0     0     0     5     7     0     0     0     0
## 5     0     0     1     0     0     0     5     7     0     0     0     0
```

```
## 6      0      0      3      1      2      3      5      4      0      0      0      1
## TOTMENROL TOTFENROL SCHOOL_LEVEL
## 1      279      0      High
## 2      4      7      Elementary
## 3      15     14      High
## 4      6      8      Middle
## 5      6      7      High
## 6     10      9      Elementary
```

And now we look at the bottom 6 schools with the highest student-teacher ratio.

```
data_stuteratio <- data_sub %>%
  arrange(STUTERATIO)
tail(data_stuteratio)
```

```
##                                SCH_NAME STABR GSLO GSHI G01 G02
## 58631                College of So. NV HS East    NV   11   12  NA  NA
## 58632                Everest High School    OH   09   12  NA  NA
## 58633                Birch Grove Intermediate    CA   03   06  NA  NA
## 58634                John C. Fremont Elementary    CA   KG   08  85  76
## 58635 Aspire Benjamin Holt College Preparatory Academy    CA   09   12  NA  NA
## 58636                Hamburg Middle School    IA   06   08  NA  NA
##      G03 G04 G05 G06 G07 G08 G09 G10 G11 G12 TOTAL AM  HI BL WH HP TR  FTE
## 58631  NA  NA  NA  NA  NA  NA  NA  NA 102  68   170 NA  85 23 42  4  5  5.00
## 58632  NA  NA  NA  NA  NA  NA  20  24  17   7    68  0   4 37 24  0  3  2.00
## 58633 109 118 119 112  NA  NA  NA  NA  NA  NA   458  1 180 28 52  6 29 13.40
## 58634  77  95 115 113  91 102  NA  NA  NA  NA   867 13 759 26 33  4  6 25.24
## 58635  NA  NA  NA  NA  NA  NA 122 130  94  72   418  5 184 23 83  3 16 12.17
## 58636  NA  NA  NA  14  18  13  NA  NA  NA  NA    45 NA   7 NA 37  NA  1  1.31
##      STUTERATIO AMALM AMALF ASALM ASALF HIALM HIALF BLALM BLALF WHALM WHALF
## 58631      34.00    NA    NA    3    8    42    43    5    18    17    25
## 58632      34.00    0    0    0    0    1    3    27    10    15    9
## 58633      34.18    1   NA   93   69   95   85   12   16   30   22
## 58634      34.35    8    5   14   12  400  359   16   10   19   14
## 58635      34.35    1    4   55   49   87   97   10   13   35   48
## 58636      34.35   NA   NA   NA   NA    3    4   NA   NA   17   20
##      HPALM HPALF TRALM TRALF TOTMENROL TOTFENROL SCHOOL_LEVEL
## 58631    1    3    2    3        70        100      High
## 58632    0    0    1    2        44        24      High
## 58633    5    1   18   11       254       204  Elementary
## 58634    1    3    4    2       462       405  Elementary
## 58635    3   NA    4   12       195       223      High
## 58636   NA   NA   NA    1        20        25    Middle
```

Summary Data

Now let's take a look at some summary of the dataset in terms of its student-teacher ratio in each state.

```
data_stats <- data_sub %>%
  group_by(STABR) %>%
  summarise(STABR = STABR, AvgRatio = mean(STUTERATIO), SDRatio = sd(STUTERATIO))
```

'summarise()' has grouped output by 'STABR'. You can override using the '.groups' argument.

```
data_summary <- distinct(data_stats)
data_summary
```

```
## # A tibble: 53 x 3
## # Groups:   STABR [53]
##   STABR AvgRatio SDRatio
##   <chr>    <dbl>    <dbl>
## 1 AK      16.0      3.02
## 2 AL      18.1      2.64
## 3 AR      13.2      2.81
## 4 AZ      18.9      3.77
## 5 CA      23.4      3.47
## 6 CO      16.4      3.71
## 7 CT      12.0      1.92
## 8 DC      13.1      4.02
## 9 DE      14.9      2.63
## 10 FL     17.2      3.66
## # ... with 43 more rows
```

Homework 4

For this section we need to look deeper into two variables from the dataset, for this I want to look at school type and student to teacher ratio.

How This Data Was Collected

Taken from the NCES website: “The National Center for Education Statistics’ (NCES) Common Core of Data (CCD) program is an annual collection of basic administrative characteristics for all public schools, school districts, and state education agencies in the United States. These characteristics are reported by state education officials and include directory information, number of students, number of teachers, grade span, and other conditions. The NCES Education Demographic and Geographic Estimate (EDGE) program develops annually updated point locations (latitude and longitude) for public elementary and secondary schools included in the CCD. The NCES EDGE program collaborates with the U.S. Census Bureau’s Education Demographic, Geographic, and Economic Statistics (EDGE) Branch to develop point locations for schools and school district administrative offices based on reported physical addresses” (NCES).

School Type

Let’s take a look at all the different levels of schools from the original, uncleaned dataset.

```
data %>%
  group_by(SCHOOL_LEVEL) %>%
  summarise(SCHOOL_LEVEL = SCHOOL_LEVEL, count.type = n()) %>%
  distinct()
```

‘summarise()’ has grouped output by ‘SCHOOL_LEVEL’. You can override using the ‘.groups’ argument.

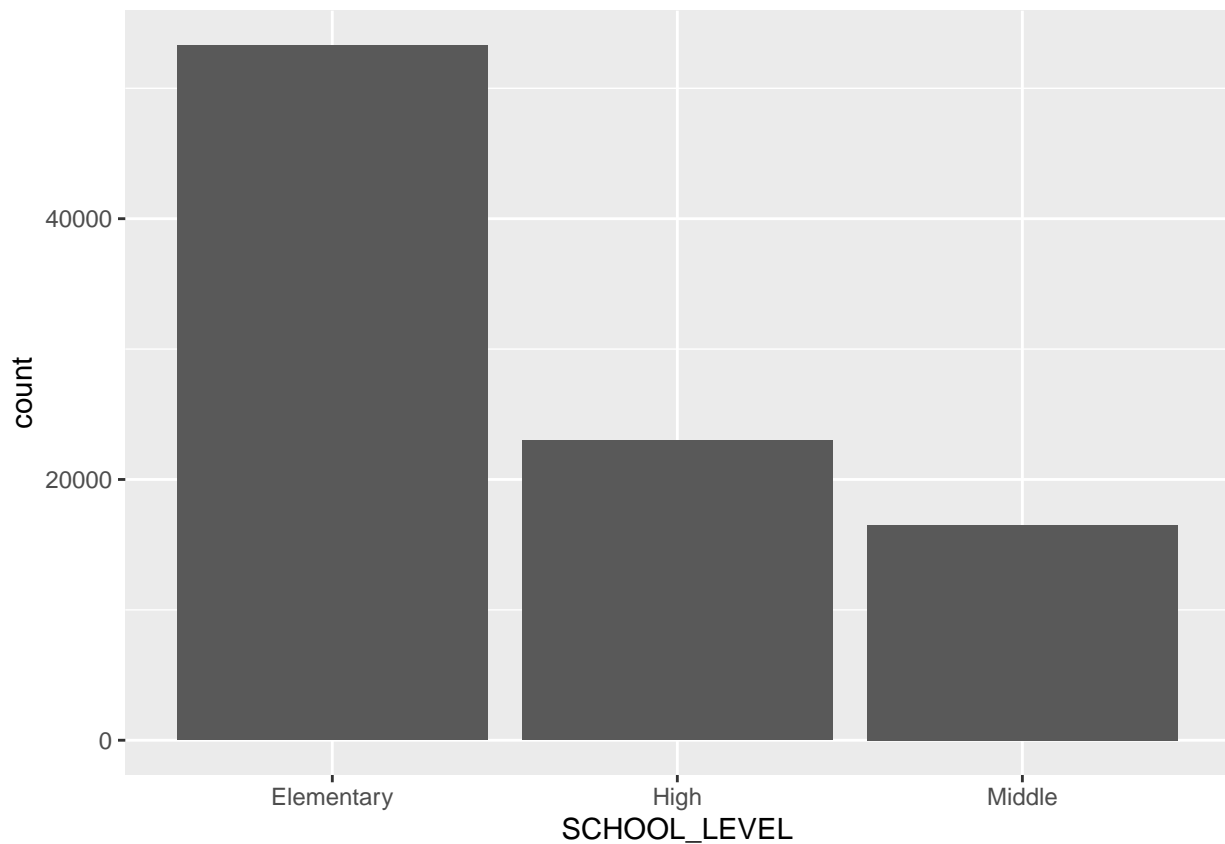
```
## # A tibble: 10 x 2
## # Groups:   SCHOOL_LEVEL [10]
```

```
## SCHOOL_LEVEL count.type
## <chr> <int>
## 1 Adult Education 28
## 2 Elementary 53287
## 3 High 22977
## 4 Middle 16506
## 5 Not Applicable 796
## 6 Not Reported 1113
## 7 Other 3824
## 8 Prekindergarten 1430
## 9 Secondary 602
## 10 Ungraded 166
```

For the dataset I wanted to focus on the most common types of school, which immediately meant filtering out adult education, secondary, and ungraded schools. This also left not applicable, not reported, and other in the dataset, which also should be removed. The last four remaining categories were elementary, high, middle, and prekindergarten. Prekindergarten schools were then removed to focus on the main three types of public schools as well as because of the number of fewer schools of this type.

Now let's filter out these unwanted categories and look at the breakdown of school type.

```
data %>%
  filter(SCHOOL_LEVEL == "High" | SCHOOL_LEVEL == "Middle" | SCHOOL_LEVEL == "Elementary") %>%
  ggplot(aes(x = SCHOOL_LEVEL)) + geom_bar()
```



This looks significantly cleaner and more organized than with all the other unused variables.

Student to Teacher Ratio

Let's start by looking at some summary statistics and a barplot of student to teacher ratio from the unclean dataset:

```
data %>%
  summarise(count = n(), mean.val = mean(STUTERATIO), sd.val = sd(STUTERATIO), median.val = median(STUTERATIO))

##   count mean.val sd.val median.val
## 1 100729      NA      NA         NA
```

We see that there are probably a large amount of NA values, so we first need to clean that.

```
data_stuteratio <- data %>%
  filter(!is.na(STUTERATIO))
data_stuteratio %>%
  summarise(count = n(), mean.val = mean(STUTERATIO), sd.val = sd(STUTERATIO), median.val = median(STUTERATIO))

##   count mean.val  sd.val median.val
## 1  93894 16.94477 85.73974      15.33
```

Now we are getting valid statistics, but the standard deviation appears to be very large, over 5 times the mean and median! Let's take a look at some quantiles to see what the breakdown is.

```
quantiles <- c(0, 0.005, 0.025, 0.05, 0.1, 0.25, 0.5, 0.75, 0.9, 0.95, 0.975, 0.995, 1)
data_stuteratio %>%
  summarise(quantiles = quantiles, value = quantile(STUTERATIO, quantiles))

##   quantiles      value
## 1      0.000    0.00000
## 2      0.005    0.00000
## 3      0.025    5.10325
## 4      0.050    8.02000
## 5      0.100   10.33000
## 6      0.250   12.85000
## 7      0.500   15.33000
## 8      0.750   18.18000
## 9      0.900   22.50000
## 10     0.950   25.11000
## 11     0.975   27.50000
## 12     0.995   50.30070
## 13     1.000 22350.00000
```

We see that our wide range of values that cause a large standard deviation is probably due to the top 0.5% of values for student to teacher ration, as the bottom 99.5% are all equal to or less than 50.3, with the top 0.5% ranging from that to over 20,000! To look at more meaningful data, we can select the middle 99% of the data to use.

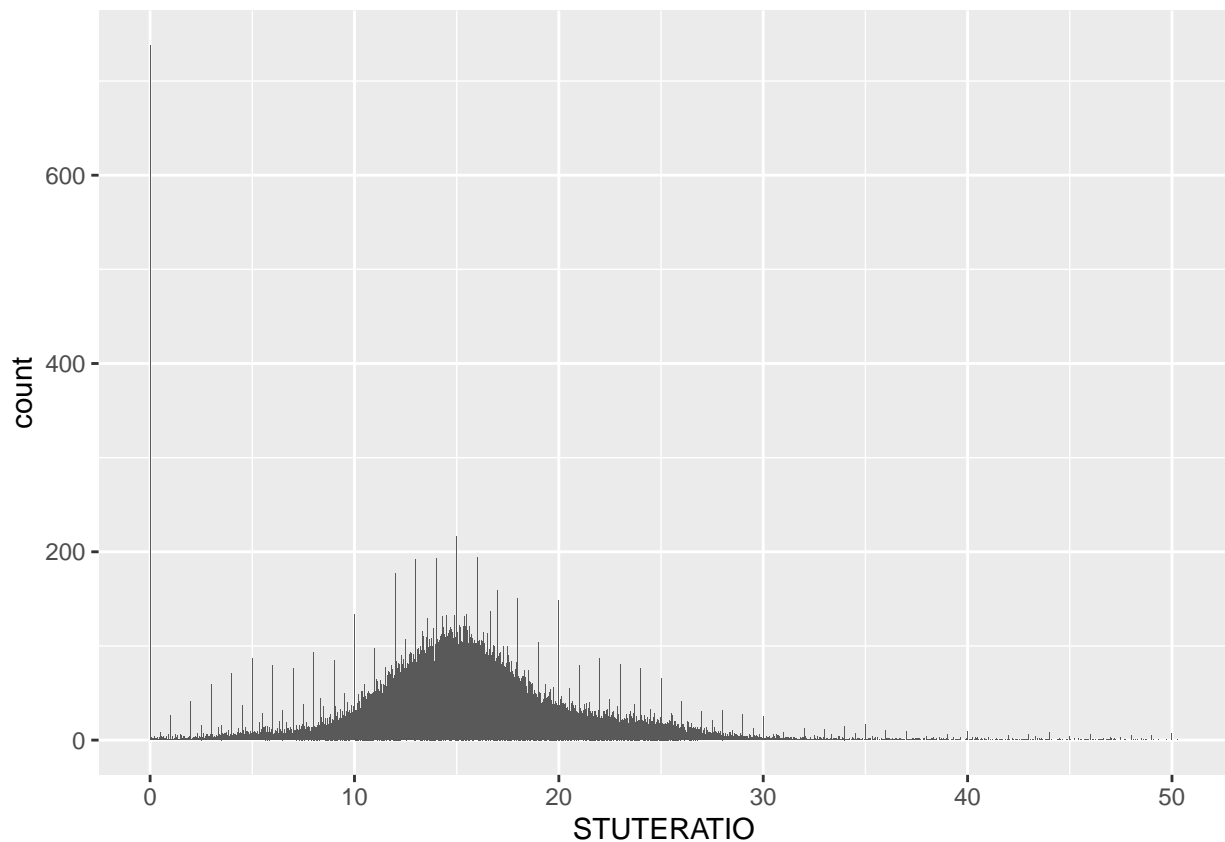
```
data_stuteratio <- data_stuteratio %>%
  filter(STUTERATIO <= 50.3007)
data_stuteratio %>%
  summarise(count = n(), mean.val = mean(STUTERATIO), sd.val = sd(STUTERATIO), median.val = median(STUTERATIO))
```



```
##   count mean.val sd.val median.val
## 1 93424 15.71938 5.2405      15.31
```

The standard deviation for this value now makes a lot more sense, and the median has moved closer to the mean. Let's take a look at a barplot to see the distribution.

```
data_stuteratio %>%
  ggplot(aes(x = STUTERATIO)) + geom_bar()
```



The data still appears somewhat skewed, but not in a way that will negatively affect our analysis.

NOTE: In the actual filtering of data the cutoffs used for STUTERATIO are different because of other filtering, in actuality the cutoffs of 4.82 and 34.4274 will be used.

Homework 5

Let's take a look at student to teacher ratio in relation to the percentage of black and hispanic students in the school.

```
quant <- c()
axis_val <- c()
count <- 1
heat.data <- data_sub %>%
  summarise(SCH_NAME, STUTERATIO, BLHI_PCT = (BL + HI)/TOTAL) %>%
```

```

filter(!is.na(BLHI_PCT))
while (count <= 20) {
  count2 <- 1
  while (count2 <= 20) {
    cur.data <- heat.data %>%
      filter(STUTERATIO >= quantile(STUTERATIO, prob = c((count2-1)*0.05)) & STUTERATIO < quantile(STUTERATIO, prob = c(count2*0.05)))
    quant <- append(quant, nrow(cur.data))
    count2 <- count2 + 1
  }
  axis_val <- append(axis_val, 0.05*count)
  count <- count + 1
}
data <- expand.grid(X=axis_val, Y=axis_val)
data$Z <- quant
print(axis_val)

```

```

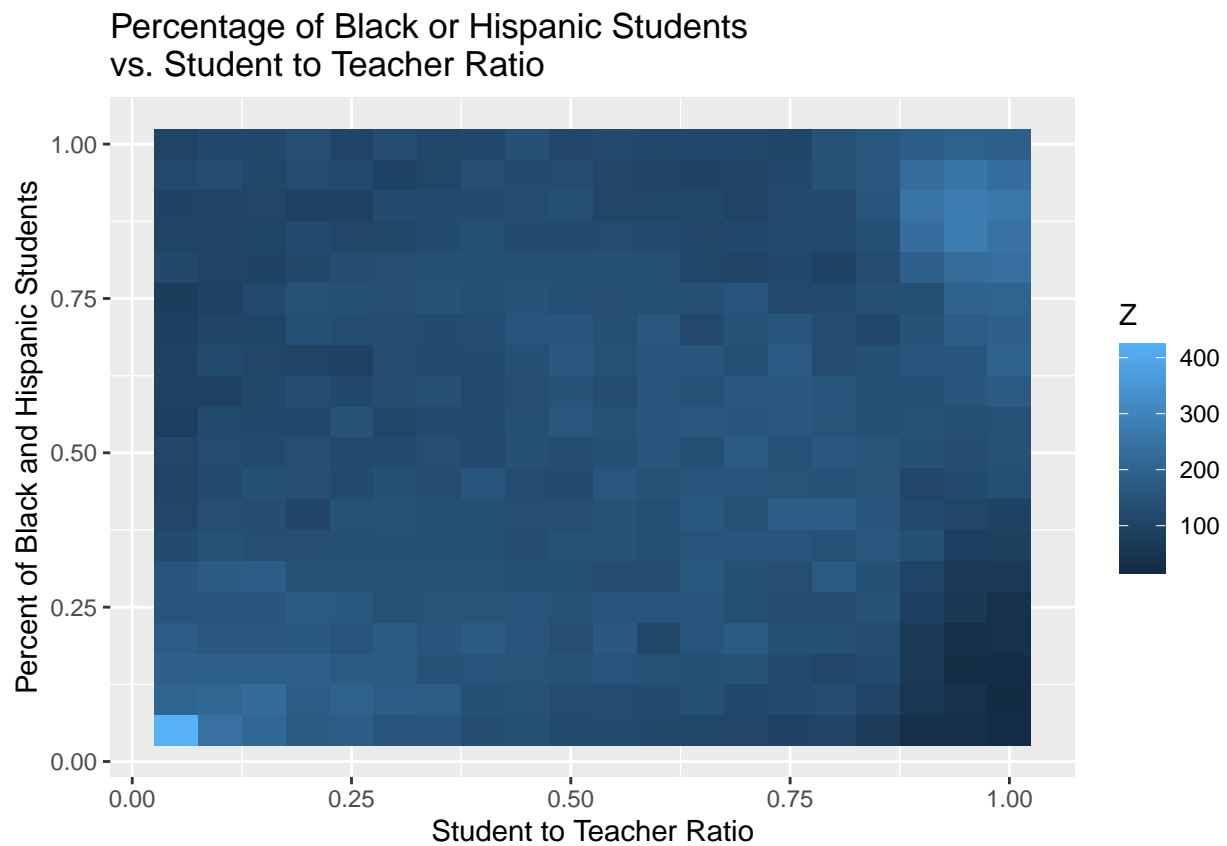
## [1] 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40 0.45 0.50 0.55 0.60 0.65 0.70 0.75
## [16] 0.80 0.85 0.90 0.95 1.00

```

```

ggplot(data, aes(X, Y, fill= Z)) +
  geom_tile() + labs(title = "Percentage of Black or Hispanic Students \nvs. Student to Teacher Ratio",

```



```
quant
```

```

## [1] 425 241 214 172 180 150 158 129 136 118 122 113 110 108 98 102 73 33
## [19] 31 20 201 209 222 180 192 176 175 143 144 127 127 124 141 115 122 128
## [37] 102 57 35 15 188 189 185 187 169 172 147 155 151 147 150 144 140 145
## [55] 124 106 122 62 24 26 176 162 165 166 150 171 156 169 150 131 165 113
## [73] 155 171 134 143 126 68 28 36 154 150 153 173 166 146 151 152 153 148
## [91] 150 156 154 137 130 129 147 88 59 35 153 170 179 145 149 144 145 146
## [109] 134 137 127 125 166 142 133 167 143 100 67 65 122 146 133 133 136 142
## [127] 135 141 128 144 148 136 153 159 155 146 161 142 85 89 108 131 128 102
## [145] 145 148 135 140 131 132 149 134 166 148 181 183 153 118 106 96 103 121
## [163] 143 131 123 142 130 155 127 122 161 149 158 153 151 147 151 107 124 135
## [181] 106 125 123 131 121 122 132 121 141 128 138 157 142 171 146 164 151 142
## [199] 129 145 85 118 116 114 145 110 120 123 140 161 147 158 163 159 164 152
## [217] 134 145 134 146 93 94 112 130 115 128 139 119 131 149 130 157 146 160
## [235] 164 155 141 140 159 172 93 118 108 101 93 129 118 122 138 164 146 153
## [253] 159 143 174 130 142 153 155 195 88 102 105 139 129 128 123 129 153 157
## [271] 140 160 120 147 154 126 115 148 179 190 78 94 119 144 138 133 148 136
## [289] 149 136 132 134 139 150 120 118 134 137 195 201 112 99 98 113 129 131
## [307] 138 142 142 141 138 139 109 104 117 97 126 190 225 235 99 100 105 120
## [325] 107 110 122 135 123 121 125 119 108 117 123 118 137 233 277 244 98 99
## [343] 107 93 93 124 124 123 125 135 106 109 110 99 123 119 158 252 275 261
## [361] 121 129 111 125 122 98 109 133 122 125 106 99 96 100 114 145 160 230
## [379] 256 233 99 115 113 131 102 127 109 114 138 107 121 111 114 112 105 144
## [397] 160 180 193 186

```

Looking at the generated heatmap, we can see that the bottom right, where schools have the highest student to teacher ratio and lowest percentage of black and hispanic are the least common. However, it appears that the most common combination is a high student to teacher ratio and a high percentage of black and hispanic students. The heat map was used to convey the number of schools whose student to teacher ratio and black and hispanic percentage matched the x and y axes for those values. It would have been nice if there was more robust racial statistics available for the schools as more specific heatmaps could be created.

Works Cited

NCES.PublicSchoolcharacteristics.2017–18.CommonCoreDataSet.<https://data-nces.opendata.arcgis.com/datasets/nce-public-school-characteristics-2017-18/about>